B-Human

Team Report and Code Release 2016

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1 Introduction
1.1 About Us .................................................. 9
1.2 About the Document ..................................... 9
1.3 Major Changes Since 2015 ............................. 10

2 Getting Started ............................................ 12
2.1 Download .................................................. 12
2.2 Components and Configurations ....................... 12
2.3 Building the Code ....................................... 13
   2.3.1 Project Generation ................................ 13
   2.3.2 Visual Studio on Windows ......................... 14
       2.3.2.1 Required Software ......................... 14
       2.3.2.2 Compiling ................................ 15
   2.3.3 Xcode on macOS .................................. 15
       2.3.3.1 Required Software ......................... 15
       2.3.3.2 Compiling ................................ 15
       2.3.3.3 Support for Xcode ......................... 16
   2.3.4 Linux ................................................. 16
       2.3.4.1 Required Software ......................... 16
       2.3.4.2 Compiling ................................ 17
2.4 Setting Up the NAO .................................... 17
   2.4.1 Requirements ..................................... 17
   2.4.2 Creating Robot Configuration Files for a NAO .... 17
   2.4.3 Managing Wireless Configurations ............... 18
   2.4.4 Setup .............................................. 19
2.5 Copying the Compiled Code .......................... 19
2.6 Working with the NAO ................................ 20
2.7 Starting SimRobot ..................................... 21
2.8 Calibrating the Robots ................................ 21
2.8.1 Overall physical calibration ........................................ 22
2.8.2 Joint Calibration ..................................................... 22
2.8.3 Camera Calibration ................................................... 24
2.8.4 Color Calibration ..................................................... 25
2.9 Configuration Files .................................................... 26

3 Architecture ............................................................. 28

3.1 Binding ....................................................................... 28
3.2 Processes .................................................................... 29
3.3 Modules and Representations ......................................... 30
  3.3.1 Blackboard ............................................................. 30
  3.3.2 Module Definition .................................................. 30
  3.3.3 Configuring Providers .............................................. 32
  3.3.4 Pseudo-Module default ........................................... 32
  3.3.5 Parameterizing Modules .......................................... 32
3.4 Serialization ................................................................ 33
  3.4.1 Streams ................................................................. 33
  3.4.2 Streaming Data ...................................................... 35
  3.4.3 Streamable Classes ............................................... 36
  3.4.4 Generating Streamable Classes ................................. 37
  3.4.5 Configuration Maps ............................................... 39
  3.4.6 Enumerations ......................................................... 40
    3.4.6.1 Iterating over Enumerations ............................... 41
    3.4.6.2 Enumerations as Array Indices ......................... 41
3.5 Communication .......................................................... 42
  3.5.1 Inter-process Communication .................................... 42
  3.5.2 Message Queues .................................................... 42
  3.5.3 Debug Communication ........................................... 43
  3.5.4 Team Communication ............................................ 43
3.6 Debugging Support ...................................................... 44
  3.6.1 Debug Requests ..................................................... 44
  3.6.2 Debug Images ....................................................... 45
  3.6.3 Debug Drawings .................................................... 46
  3.6.4 3-D Debug Drawings ............................................. 47
  3.6.5 Plots .................................................................... 48
  3.6.6 Modify ................................................................. 49
  3.6.7 Stopwatches .......................................................... 50
3.7 Logging ....................................................................... 50
3.7.1 Online Logging ................................................. 50
3.7.2 Configuring the Online Loggers .......................... 51
3.7.3 Remote Logging .............................................. 51
3.7.4 Log File Format .............................................. 52
3.7.5 Replaying Log Files ......................................... 53
3.7.6 Annotations .................................................. 53
3.7.7 Thumbnail Images ........................................... 54
3.7.8 Image Patches ............................................... 54

4 Perception .......................................................... 57
  4.1 Perception Infrastructure .................................... 57
    4.1.1 Using Both Cameras ...................................... 57
    4.1.2 Definition of Coordinate Systems ..................... 59
      4.1.2.1 Camera Matrix and Camera Calibration ............ 59
      4.1.2.2 Image Coordinate System .......................... 61
    4.1.3 Body Contour ............................................ 61
    4.1.4 Color Classification .................................... 61
    4.1.5 Segmentation and Region-Building .................... 63
    4.1.6 Detecting The Field Boundary ......................... 64
  4.2 Localization Features ....................................... 66
    4.2.1 Detecting Lines ......................................... 66
    4.2.2 Detecting the Center Circle ........................... 66
    4.2.3 Line Coincidence Detection ............................ 66
    4.2.4 Preprocessed Lines and Intersections ................ 68
    4.2.5 Penalty Mark Perception ............................... 68
    4.2.6 Field Features ......................................... 69
  4.3 Detecting the Black and White Ball ....................... 70
    4.3.1 Searching for Ball Candidates ....................... 70
    4.3.2 Fitting Ball Contours ................................. 71
    4.3.3 Filtering Ball Candidates ............................. 72
    4.3.4 Checking the Surface Pattern ......................... 72
  4.4 Ball Perception Statistics .................................. 72
  4.5 Detecting Other Robots and Obstacles ...................... 73

5 Modeling ........................................................... 75
  5.1 Self-Localization ............................................ 75
    5.1.1 Probabilistic State Estimation ....................... 75
      5.1.1.1 Perceptions Used for Self-Localization ........... 76
6 Behavior Control

6.1 CABSL

6.1.1 Options

6.1.2 Libraries

6.2 Setting Up a New Behavior

6.3 Behavior Used at RoboCup 2016

6.3.1 Roles and Tactic

6.3.2 Striker

6.3.3 Supporter

6.3.4 Defender

6.3.5 Keeper

6.3.6 Kickoff

6.3.7 Head Control

6.4 Penalty Shoot-out Behavior

6.5 Path Planner

6.5.1 Approach

6.5.2 Avoiding Oscillations
6.5.3 Overlapping Obstacle Circles .............................................. 105
6.5.4 Forbidden Areas ................................................................. 106
6.5.5 Avoiding Impossible Plans ..................................................... 107
6.6 Kick Pose Provider ................................................................. 107
6.7 Camera Control Engine ......................................................... 107
6.8 LED Handler .......................................................................... 108

7 Proprioception (Sensing) ............................................................... 111
7.1 Ground Contact Recognition ..................................................... 112
7.2 Robot Model Generation ........................................................... 112
7.3 Inertia Sensor Data Filtering ....................................................... 113
7.4 Torso Matrix ........................................................................... 113
7.5 Detecting a Fall ...................................................................... 114
7.6 Arm Contact Recognition .......................................................... 114

8 Motion Control ........................................................................... 117
8.1 Motion Selection ...................................................................... 117
8.2 Motion Combination ................................................................. 118
8.3 Walking .................................................................................... 119
  8.3.1 Load Balancing ............................................................... 119
  8.3.2 In-Walk Kicks .................................................................... 120
  8.3.3 Inverse Kinematic ............................................................. 121
8.4 Special Actions ........................................................................ 124
8.5 Get Up Motion ........................................................................ 125
  8.5.1 Modify Get Up Motions ....................................................... 126
8.6 Head Motions ......................................................................... 127
8.7 Arm Motions ........................................................................... 127
  8.7.1 Pointing at Something ....................................................... 128
  8.7.2 Key Frame Motion ............................................................. 128

9 Sub-Competitions and Technical Challenges .................................. 130
9.1 Drop-In Competition ............................................................... 130
  9.1.1 Reliability of Teammates ................................................... 130
  9.1.2 Behavior .......................................................................... 131
  9.1.3 Analysis and Results .......................................................... 132
9.2 Outdoor Competition ............................................................... 132
  9.2.1 Vision Adaptions ............................................................... 132
  9.2.2 Analysis and Results .......................................................... 132
9.3 Outdoor Challenge ............................................. 133
9.4 No Wi-Fi Challenge ........................................... 134
  9.4.1 Evaluated Communication Approaches ................. 134
  9.4.2 Transmitting and Detecting Sounds ................... 134
  9.4.3 Base 4 Hamming Encoding ............................... 135
  9.4.4 Analysis and Results .................................. 135

10 Tools .......................................................... 136
  10.1 SimRobot .................................................. 136
    10.1.1 Architecture .......................................... 136
    10.1.2 B-Human Toolbar ...................................... 136
    10.1.3 Scene View ............................................. 137
    10.1.4 Information Views ................................... 138
      10.1.4.1 Cognition ........................................... 138
      10.1.4.2 Behavior Control ................................. 142
      10.1.4.3 Sensing ............................................ 143
      10.1.4.4 Motion Control .................................... 143
      10.1.4.5 General Debugging Support ...................... 146
    10.1.5 Scene Description Files ............................... 149
    10.1.6 Console Commands .................................... 150
      10.1.6.1 Initialization Commands ......................... 150
      10.1.6.2 Global Commands .................................. 151
      10.1.6.3 Robot Commands .................................. 152
      10.1.6.4 Input Selection Dialog ......................... 158
    10.1.7 Recording a Remote Log File ........................ 159
  10.2 B-Human User Shell ....................................... 159
    10.2.1 Configuration ........................................ 159
    10.2.2 Commands .............................................. 159
    10.2.3 Deploying Code to the Robots ....................... 160
    10.2.4 Managing Multiple Wireless Configurations ......... 160
    10.2.5 Substituting Robots .................................. 161
    10.2.6 Monitoring Robots ................................... 161
  10.3 GameController ........................................... 162
    10.3.1 Architecture .......................................... 163
    10.3.2 UI Design ................................................ 165
    10.3.3 Game State Visualizer ................................. 166
    10.3.4 Log Analyzer ........................................... 167
  10.4 Team Communication Monitor ............................... 167
Chapter 1

Introduction

1.1 About Us

*B-Human* is a joint RoboCup team of the Universität Bremen and the German Research Center for Artificial Intelligence (DFKI). The team was founded in 2006 as a team in the Humanoid League, but switched to participating in the Standard Platform League in 2009. Since then, B-Human has won seven RoboCup German Open competitions, the RoboCup European Open 2016 competition, and has become RoboCup world champion five times.

After reaching the third place in 2014 and becoming runner-up world champion in 2015, we finally regained the world champion title in 2016. We won all six matches in the main indoor competition, achieving a goal difference of 43 : 1 (including three penalty shots in the final). Instead of the first indoor round, we participated in the *Outdoor Competition*, which we finished as vice champion by winning four games and losing only the final (total goal difference: 18 : 1). Furthermore, we also won the *Drop-In Competition* by reaching the maximum possible score of 300/300 and became third in the overall ranking of the technical challenges.

The 2016 team consisted of the following persons (most of them are shown in Fig. 1.1):

**Members:** Tristan Bruns, Yannick Bülter, Mathis Engelbart, Miguel Kasparick, Daniel Krause, Jonas Kuball, Lam Duy Le, André Lübken, Florian Maaß, André Mühlenbrock, Tim Müller, Lukas Post, Kameran Q.E. Abdi, Jesse Richter-Klug, Peter Schulz, Leonid Schwenke, René Schröder, Dennis Schürholz, Andreas Stolpmann, Alexander Stöwing, Kannan Thambiah, Felix Thielke, Alexis Tsogias.

**Associated Researchers:** Udo Frese, Judith Müller, Dennis Schüitche, Felix Wenk

**Leaders:** Tim Laue, Thomas Röfer.

1.2 About the Document

This document provides a survey of this year’s code release, continuing the tradition of annual releases that was started several years ago. A short description of the changes to our system compared to last year has already been given in our Team Description Paper for RoboCup 2016 [28]. This document is based on the code releases of the last years and aims to provide a complete description of the system that we used at RoboCup 2016. The major changes made to the system since last year are shortly enumerated in Section 1.3.
The remainder of this document is organized as follows: Chapter 2 gives a short introduction on how to build the code including the required software and how to run the NAO with our software. Chapter 3 describes our framework’s architecture. Our image processing approaches are presented in Chapter 4, followed by the state estimation approaches in Chapter 5. Chapter 6 explains the use of our behavior description language and gives an overview of the behavior used at RoboCup 2016. Surveys of the sensor reading and motion control parts of the system are given in Chapter 7 and Chapter 8 respectively. Our specific developments for the sub-competitions and the technical challenges are described in Chapter 9. Finally, in Chapter 10, our simulation and remote control environment SimRobot, the B-Human User Shell, and some other tools are presented.

1.3 Major Changes Since 2015

The major changes made since RoboCup 2015 are described in the following sections:

3.7.8 Image Patches
   Logging of multiple “high resolution” image parts.

4.1.4 Color Segmentation / YHS2
   Full image color classification as image preprocessing.

4.3 Ball Detection
   The ball detection has been completely rewritten to handle the new black and white ball of 2016.
4.2.6 Field Features
A combination of several basic field elements now comprises so called „field features” that are used to improve the self localization.

5.1 Self-Localization
The implementation has been revised and the sensor resetting does not require any goal detections anymore. Instead, new pose alternatives can be generated based on field features.

5.4 Ball Search
Improved ball search with team field coverage.

6.3 Behavior of 2016
The behavior was improved to more precision (cf. Roles and Tactic 6.3.1).

8.7.1 Pointing Arm Motion
New arm motion engine to point at something with an outstretched arm.

9 Implementations for Sub-Competitions and Technical Challenges
In addition to the main indoor soccer competition, we also participated in the Outdoor Competition, the Drop-In Competition, as well as in both technical challenges, i.e. the No Wi-Fi Challenge and the Outdoor Challenge.
Chapter 2

Getting Started

The goal of this chapter is to give an overview of the code release package and to give instructions on how to enliven a NAO with our code. For the latter, several steps are necessary: downloading the source code, compiling the code using Visual Studio, Xcode, or make on Linux, setting up the NAO, copying the files to the robot, and starting the software. In addition, all calibration procedures are described here.

2.1 Download

The code release can be downloaded from GitHub at https://github.com/bhuman. Store the code release to a folder of your liking. After the download is finished, the chosen folder should contain several subdirectories which are described below.

**Build** is the target directory for generated binaries and for temporary files created during the compilation of the source code. It is initially missing and will be created by the build system.

**Config** contains configuration files used to configure the B-Human software. A brief overview of the organization of the configuration files can be found in Sect. 2.9.

**Install** contains all files needed to set up B-Human on a NAO.

**Make** Contains Makefiles, other files needed to compile the code, the Copyfiles tool, and a script to download log files from a NAO. In addition there are generate scripts that create the project files for Xcode, Visual Studio, CodeLite, and NetBeans.

**Src** contains the source code of the B-Human software including the B-Human User Shell (cf. Sect. 10.2).

**Util** contains auxiliary and third party libraries (cf. Sect. 11) as well as our simulator SimRobot (cf. Sect. 10.1).

2.2 Components and Configurations

The B-Human software is usable on Windows, Linux, and macOS. It consists of two shared libraries for NAOqi running on the real robot, an additional executable for the robot, the same
2.3. BUILDING THE CODE

software running in our simulator SimRobot (without NAOqi), as well as some libraries and tools. Therefore, the software is separated into the following components:

**bush** is a tool to deploy and manage multiple robots at the same time (cf. Sect. 10.2).

**Controller** is a static library that contains NAO-specific extensions of the simulator, the interface to the robot code framework, and it is also required for controlling and high level debugging of code that runs on a NAO.

**copyfiles** is a tool for copying compiled code to the robot. For a more detailed explanation see Sect. 2.5. In the Xcode project, this is called Deploy.

**libbhuman** is the shared library used by the B-Human executable to interact with NAOqi.

**libgamectrl** is a shared NAOqi library that communicates with the GameController. Additionally it implements the official button interface and sets the LEDs as specified in the rules. More information can be found at the end of Sect. 3.1.

**libqxt** is a static library that provides an additional widget for Qt on Windows and Linux. On macOS, the same source files are simply part of the library Controller.

**Nao** is the B-Human executable for the NAO. It depends on libbhuman and libgamectrl.

**qtpropertybrowser** is a static library that implements a property browser in Qt.

**SimRobot** is the simulator executable for running and controlling the B-Human robot code. It dynamically links against the components SimRobotCore2, SimRobotEditor, SimulatedNao, and some third-party libraries. SimRobot is compilable in Release, Develop, and Debug configurations. All these configurations contain debug code, but Release performs some optimizations and strips debug symbols (Linux and macOS). Develop produces debuggable robot code while linking against non-debuggable but faster Release libraries.

**SimRobotCore2** is a shared library that contains the simulation engine of SimRobot.

**SimRobotEditor** is a shared library that contains the editor widget of the simulator.

**SimulatedNao** is a shared library containing the B-Human code for the simulator. It depends on Controller, qtpropertybrowser and libqxt. It is statically linked against them.

All components can be built in the three configurations Release, Develop, and Debug. Release is meant for “game code” and thus enables the highest optimizations; Debug provides full debugging support and no optimization. Develop is a special case. It generates executables with some debugging support for the components Nao and SimulatedNao (see the table below for more specific information). For all other components it is identical to Release.

The different configurations for Nao and SimulatedNao can be looked up here:

### 2.3 Building the Code

#### 2.3.1 Project Generation

The scripts generate (or generate.cmd on Windows) in the Make/‹OS/IDE› directories generate the platform or IDE specific files that are needed to compile the components. The script
## 2.3. BUILDING THE CODE

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<thead>
<tr>
<th></th>
<th>without assertions (NDEBUG)</th>
<th>debug symbols (compiler flags)</th>
<th>debug libc(^1) (DEBUG, compiler flags)</th>
<th>optimizations (compiler flags)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Release</strong></td>
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<tr>
<td>Nao</td>
<td>✓</td>
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<tr>
<td>SimulatedNao</td>
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<td>x</td>
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<tr>
<td><strong>Develop</strong></td>
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<tr>
<td>Nao</td>
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<td>x</td>
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<td>✓</td>
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<td>SimulatedNao</td>
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<tr>
<td><strong>Debug</strong></td>
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<td>SimulatedNao</td>
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</tbody>
</table>

\(^1\) - on Windows - [http://msdn.microsoft.com/en-us/library/0b98s6w8(v=vs.140).aspx](http://msdn.microsoft.com/en-us/library/0b98s6w8(v=vs.140).aspx)

Collects all the source files, headers, and other resources if needed and packs them into a solution matching your system (i.e. Visual Studio projects and a solution file for Windows, a CodeLite project for Linux, and an Xcode project for macOS). It has to be called before any IDE can be opened or any build process can be started and it has to be called again whenever files are added or removed from the project. On Linux, the `generate` script is needed when working with CodeLite or NetBeans. Building the code from the command line, via the provided Makefile, works without calling `generate` on Linux.

### 2.3.2 Visual Studio on Windows

#### 2.3.2.1 Required Software

- Windows 8.1 64 bit or later
- Visual Studio 2015\(^1\) or later
- Cygwin x86 / x64 (available at [http://www.cygwin.com](http://www.cygwin.com)) with the additional packages `rsync`, `openssh`, `ccache`, and `clang`. Let the installer add an icon to the start menu (the Cygwin Terminal). Add the `\cygwin64\bin` directory to the beginning of the PATH environment variable (before the Windows system directory, since there are some commands that have the same names but work differently). Make sure to start the Cygwin Terminal at least once, since it will create a home directory.
- `alcommon` – For the extraction of the required alcommon library and compatible boost headers from the `NAOqi C++ SDK 2.1.4 Linux 32 bit (naoqi-sdk-2.1.4.13-linux32.tar.gz)` the script `Install/installAlcommon` can be used, which is delivered with the B-Human software. The required package has to be downloaded manually and handed over to the script. It is available at [https://community.ald.softbankrobotics.com](https://community.ald.softbankrobotics.com) (account required). Please note that this package is only required to compile the code for the actual NAO robot.

2.3.2.2 Compiling

Generate the Visual Studio project files using the script `Make/VS2015/generate.cmd` and open the solution `Make/VS2015/B-Human.sln` in Visual Studio. Visual Studio will then list all the components (cf. Sect. 2.2) of the software in the “Solution Explorer”. Select the desired configuration (cf. Sect. 2.2, `Develop` would be a good choice for starters) and build the desired project: `SimRobot` compiles every project used by the simulator, `Nao` compiles every project used for working with a real NAO, and `Utils/bush` compiles the B-Human User Shell (cf. Sect. 10.2). You may select `SimRobot` or `Utils/bush` as “StartUp Project”.

2.3.3 Xcode on macOS

2.3.3.1 Required Software

The following components are required:

- macOS 10.12 or later
- Xcode 8.1 or later
- alcommon – For the extraction of the required alcommon library and compatible boost headers from the NAOqi C++ SDK 2.1.4 Linux 32 bit (naoqi-sdk-2.1.4.13-linux32.tar.gz) the script `Install/installAlcommon` can be used, which is delivered with the B-Human software. The required package has to be downloaded manually and handed over to the script. It is available at https://community.ald.softbankrobotics.com (account required). Please note that this package is only required to compile the code for the actual NAO robot. Also note that `installAlcommon` expects the extension `.tar.gz`. If the NAOqi archive was partially unpacked after the download, e.g., by Safari, repack it again before executing the script.

2.3.3.2 Compiling

Generate the Xcode project by executing `Make/macOS/generate`. Open the Xcode project `Make/macOS/B-Human.xcodeproj`. A number of schemes (selectable in the toolbar) allow building SimRobot in the configurations `Debug`, `Develop`, and `Release`, as well as the code for the NAO\(^2\) in all three configurations (cf. Sect. 2.2). For both targets, `Develop` is a good choice. In addition, the B-Human User Shell `bush` can be built. It is advisable to delete all the schemes that are automatically created by Xcode, i.e. all non-shared ones.

When building for the NAO, a successful build will open a dialog to deploy the code to a robot (using the `copyfiles` script, cf. Sect. 2.5).\(^3\) If the `login` script was used before to login to a NAO, the IP address used will be provided as default. In addition, the option `-r` is provided by default, which will restart the software on the NAO after it was deployed. Both the IP address selected and the options specified are remembered for the next use of the deploy dialog. The IP address is stored in the file `Config/Scenes/Includes/connect.conf` that is also written by the `login` script and used by the RemoteRobot simulator scene. The options are stored in `Make/macOS/copyfiles-options.txt`. A special option is `-a`: If it is specified, the deploy dialog is not shown anymore in the future. Instead, the previous settings will be reused, i.e. building the code will automatically

---

\(^2\)Note that the cross compiler actually builds code for Linux, although the scheme says “My Mac”.

\(^3\)Before you can do that, you have to setup the NAO first (cf. Sect. 2.4).
deploy it without any questions asked. To get the dialog back, hold down the key Shift at the
time the dialog would normally appear.

2.3.3.3 Support for Xcode

Calling the script *Make/macOS/generate* also installs a lot of development support for Xcode:

**Data formatters.** If the respective file does not already exist, a symbolic link is created to
formatters that let Xcode’s debugger display summaries of several *Eigen* datatypes.

**Source file templates.** Xcode’s context menu entry *New File...* contains a category *B-
Human* that allows to create some B-Human-specific source files.

**Code snippets.** Many code snippets are available that allow adding standard constructs fol-
lowing B-Human’s coding style as well as some of B-Human’s macros.

**Source code formatter.** A system text service for formatting B-Human code is available to
be used from Xcode’s menu *Xcode→Services.*

2.3.4 Linux

The following has been tested and works on Ubuntu 16.04 64-bit. It should also work on other
Linux distributions (as long as they are 64-bit); however, different or additional packages may
be needed.

2.3.4.1 Required Software

The build has been tested using the software versions provided by the current Ubuntu distribu-
tion repositories. Earlier versions of, e.g., clang may work, but are untested.

Requirements (listed by common package names) for Ubuntu 16.04:

- clang
- qt5-default
- libqt5svg5-dev
- libglew-dev
- libxml2-dev
- libasound2-dev
- graphviz – Optional, for generating module graphs and the behavior graph.
- alcommon – For the extraction of the required alcommon library and compatible boost
  headers from the *NAOqi C++ SDK 2.1.4 Linux 32 bit* (*naoqi-sdk-2.1.4.13-linux32.tar.gz*)
  the script *Install/installAlcommon* can be used, which is delivered with the B-Human
  software. The required package has to be downloaded manually and handed over to the
  script. It is available at [https://community.ald.softbankrobotics.com](https://community.ald.softbankrobotics.com) (account re-
  quired). Please note that this package is only required to compile the code for the actual
  NAO robot.
2.4. SETTING UP THE NAO

On Ubuntu 16.04, you can execute the following command to install all requirements except for alcommon:

```
sudo apt-get install qt5-default libqt5svg5-dev libglew-dev libxml2-dev libasound2-dev clang graphviz
```

2.3.4.2 Compiling

To compile one of the components described in Section 2.2 (except Copyfiles), simply select Make/Linux as the current working directory and type:

```
make
```
to build the whole solution or

```
make <component> [CONFIG=<configuration>]
```
to build single components.

To clean up the whole solution, use:

```
make clean [CONFIG=<configuration>]
```

As an alternative, there is also support for the integrated development environments NetBeans and CodeLite that work similar to Visual Studio for Windows (cf. Sect. 2.3.2.2).

To use CodeLite, execute `Make/LinuxCodeLite/generate` and open the `B-Human.workspace` afterwards. Note that CodeLite 5 or later is required to open the workspace generated. Older versions might crash.

2.4 Setting Up the NAO

2.4.1 Requirements

First of all, download the atom system image, e.g. version 2.1.4 (`opennao-atom-system-image-2.1.4.13.2015-08-27.opn`), and the Flasher, e.g. version 2.1.0, for your operating system from the download area of [https://community.aid.softbankrobotics.com](https://community.aid.softbankrobotics.com) (account required). In order to flash the robot, you need a USB flash drive having at least 2 GB space and a network cable.

To use the scripts in the directory Install, the following tools are required:

```
sed, rsync.
```

Each script will check its requirements and will terminate with an error message if a required tool is not found.

The commands in this chapter are shell commands. They should be executed inside a Unix shell. On Windows, you must use the Cygwin Terminal to execute the commands. All shell commands should be executed from the Install directory.

2.4.2 Creating Robot Configuration Files for a NAO

Before you start to set up the NAO, you need to create configuration files for each robot you want to set up. To create the configuration files, run `createRobot` followed by `addRobotIds` in

---

4In the unlikely case that they are missing in a Linux distribution, execute `sudo apt-get install sed scp`. On Windows and macOS, they are already installed at this point.
the *Install* directory. The first script expects a team id, a robot id and a robot name. The team id is usually equal to your team number configured in *Config/settings.cfg*, but you can use any number between 1 and 254. The given team id is used as third part of the IPv4 address of the robot on both interfaces LAN and WLAN. All robots playing in the same team need the same team id to be able to communicate with each other. The robot id is the last part of the IP address and must be unique for each team id. The robot name identifies the robot and is used in the system to load robot specific configurations. Furthermore, it is used as the host name of the NAO operating system. The second file creates a table associating the *headId* and *bodyId* of each NAO to the name used by *createRobot*. These ids are the serial-numbers SoftBank Robotics uses for the NAO. Apart from the name this script expects either those ids, typed in manually, or the current ip-address of the NAO, in which case the ids will be loaded from the robot.

Before creating your first robot configuration, check whether the network configuration template files wireless and wired in *Install/Network* and *default* in *Install/Network/Profiles* match the requirements of your local network configuration.

Here is an example for creating a new set of configuration files for a robot named Penny in team three with IP xxx.xxx.3.25:

```
  cd Install
  ./createRobot -t 3 -r 25 Penny
  ./addRobotIds -ids ALDxxxxxxxxxxxx ALDxxxxxxxxxxxx Penny
```

If the robot is already running after the installation in Sect. 2.4.4 the script can use its ip (e.g. 169.254.54.28) to extract the ids from the robot directly:

```
  ./addRobotIds -ip 169.254.54.28 Penny
```

Help for both scripts is available using the option `-h`. Running *createRobot* creates all needed files to install the robot. This script also creates a directory with the robot’s name in *Config/Robots*. *addRobotIds* will store the table in *Config/Robots/robots.cfg*.

**Note:** When upgrading from an older B-Human code release running *createRobot* is not necessary. Nevertheless, the script *addRobotIds* has to be executed!

### 2.4.3 Managing Wireless Configurations

All wireless configurations are stored in *Install/Network/Profiles*. Additional configurations must be placed here and will be installed alongside the *default* configuration. After the setup will be completed, the NAO will always load the *default* configuration, when booting the operating system.

You can later switch between different configurations by calling the script *setprofile* on the NAO, which overwrites the *default* configuration.

```
  setprofile SPL_A
  setprofile Home
```

Another way to switch between different configurations is by using the tools *copyfiles* (cf. Sect. 2.5) or *bush* (cf. Sect. 10.2).
2.4.4 Setup

After the robot specific configuration files were created (cf. Sect. 2.4.2 and Sect. 2.4.3), plug in your USB flash drive and start the NAO flasher tool\(^5\). Select the opennao-atom-system-image-2.1.4.13.opn and your USB flash drive. Enable “Factory reset” and click on the write button.

After the USB flash drive has been flashed, plug it into the NAO that is switched off and press the chest button for about 5 seconds. Afterwards, the NAO will automatically install NAO OS and reboot. While installing the basic operating system, connect your computer to the robot using the network cable and configure your network for DHCP. Once the reboot is finished, the NAO will do its usual wake-up procedure. Now the NAO will say its current IP address by pressing the chest button.

Afterwards the script `installRobot` has to be executed in order to prepare the robot for the B-Human software. This script only expects the current IP address of the robot. For example run:

```
./installRobot 169.254.54.28
```

Follow the instructions on the screen until the robot reboots.

If you did not set the id of the robot in Sect. 2.4.2 you can now refer to it using the ip flag for the addRobotIds script.

Now you can use `copyfiles` (cf. Sect. 2.5) or `bush` (cf. Sect. 10.2) to copy compiled code and configuration files to the NAO.

2.5 Copying the Compiled Code

The script `copyfiles` is used to copy compiled code and configuration files to the NAO. Although `copyfiles` allows specifying the team number, it is usually better to configure the team number and the UDP port used for team communication permanently in the file `Config/settings.cfg`.

On Windows as well as on macOS, you can use your IDE to use `copyfiles`. In Visual Studio, you can run the script by “building” the project `copyfiles`, which can be built in all configurations. If the code is not up-to-date in the desired configuration, it will be built. After a successful build, you will be prompted to enter the parameters described below. On the Mac, a successful build for the NAO always ends with a dialog asking for `copyfiles`’ command line options. You can also execute the script at the command prompt, which is the only option for Linux users. The script is located in the folder `Make/<OS/IDE>`.

`copyfiles` requires two mandatory parameters. First, the configuration the code was compiled with (Debug, Develop, or Release)\(^6\), and second, the IP address of the robot. To adjust the desired settings, it is possible to set the following optional parameters:

---

\(^5\)On Linux and macOS you have to start the flasher with root permissions. Usually you can do this with `sudo ./flasher`

\(^6\)This parameter is automatically passed to the script when using IDE-based deployment.
Option | Description
--- | ---
-d | Removes all log files from the robot’s /home/nao/logs directory before copying files.
-h | --help | Prints the help.
-l <location> | Sets the location, replacing the value in the settings.cfg.
-m <n> <ip> | Copies to IP address <ip> and sets the player number to n. This option can be specified more than ones to deploy to multiple robots.
-mn <number> | Sets the magic number. Robots with different magic numbers will ignore each other when communicating.
-n <number> | Sets team number, replacing the value in the settings.cfg.
-nc | Never compiles, even if binaries are outdated. Default on Windows.
-o <port> | Overwrite team port (default is 10000 + team number).
-p <number> | Sets the player number, replacing the value in the settings.cfg.
-r | Restarts bhuman (and naoqi if necessary) after copying.
-s | Stops naoqi.
-t <color> | Sets the team color to blue, red, yellow, or black, replacing the value in the settings.cfg.
-v <percent> | Set NAO’s sound volume.
-w <profile> | Set wireless profile.
-wc | Compiles also under Windows if the binaries are outdated.

Possible calls could be:

```
./copyfiles Develop 134.102.204.229 -n 5 -t blue -p 3 -r
./copyfiles Release -m 1 10.0.0.1 -m 3 10.0.0.2
```

The destination directory on the robot is /home/nao/Config. Alternatively, the B-Human User Shell (cf. Sect. 10.2) can be used to copy the compiled code to several robots at once.

### 2.6 Working with the NAO

After pressing the chest button, it takes about 40 seconds until NAQqi is started. Currently, the B-Human software consists of two shared libraries (libbhuman.so and libgamectrl.so) that are loaded by NAQqi at startup, and one executable (bhuman), which is also loaded at startup.

To connect to the NAO, the subdirectories of Make contain a login script for each supported platform. The only parameter of that script is the IP address of the robot to login. It automatically uses the appropriate SSH key to login. In addition, the IP address specified is written to the file Config/Scenes/Includes/connect.con. Thus a later use of the SimRobot scene RemoTeRobot.ros2 will automatically connect to the same robot. On macOS, the IP address is also the default address for deployment in Xcode.

Additionally, the script Make/Linux/ssh-config can be used to output a valid ssh config file containing all robots currently present in the robots folder. Using this configuration file, one can connect to a robot using its name instead of the IP address.

There are several scripts to start and stop NAQqi and bhuman via SSH. Those scripts are copied to the NAO upon installing the B-Human software.

**naoqi** executes NAQqi in the foreground. Press Ctrl+C to terminate the process. Please note that the process will automatically be terminated if the SSH connection is closed.

**nao start|stop|restart** starts, stops or restarts NAQqi. In case libbhuman or libgamectrl were updated, copyfiles restarts NAQqi automatically.
2.7. STARTING SIMROBOT

`bhuman` executes the `bhuman` executable in the foreground. Press `Ctrl+C` to terminate the process. Please note that the process will automatically be terminated if the SSH connection is closed.

`bhumand start|stop|restart` starts, stops or restarts the `bhuman` executable. `Copyfiles` always stops `bhuman` before deploying. If `copyfiles` is started with option `-r`, it will restart `bhuman` after all files were copied.

`status` shows the status of NAOqi and `bhuman`.

`stop` stops running instances of NAOqi and `bhuman`.

`halt` shuts down the NAO. If NAOqi is running, this can also be done by pressing the chest button longer than three seconds.

`reboot` reboots the NAO.

2.7 Starting SimRobot

On Windows and macOS, SimRobot can either be started from the development environment or by starting a scene description file in `Config/Scenes`. In the first case, a scene description file has to be opened manually, whereas it will already be loaded in the latter case. On Linux, just run `Build/SimRobot/Linux/<configuration>/SimRobot`, either from the shell or from your favorite file browser, and load a scene description file afterwards. When a simulation is opened for the first time, only the scene graph is displayed. The simulation is already running, which can be noted from the increasing number of simulation steps shown in the status bar. A scene view showing the soccer field can be opened by double-clicking `RoboCup`. The view can be adjusted by using the context menu of the window or the toolbar. Double-clicking `Console` will open a window that shows the output of the robot code and that allows entering commands. All windows can be docked in the main window.

After starting a simulation, a script file may automatically be executed, setting up the robot(s) as desired. The name of the script file is the same as the name of the scene description file but with the extension `.con`. Together with the ability of SimRobot to store the window layout, the software can be configured to always start with a setup suitable for a certain task.

Although any object in the scene graph can be opened, only displaying certain entries in the object tree makes sense, namely the main scene `RoboCup`, the objects in the group `RoboCup/robots`, and all other views.

To connect to a real NAO, open the RemoteRobot scene `Config/Scenes/RemoteRobot.ros2`. You will be prompted to enter the NAO’s IP address. In a remote connection, the simulation scene is usually empty. Therefore, it is not necessary to open a scene view.

2.8 Calibrating the Robots

Correctly calibrated robots are very important since the software requires all parts of the NAO to be at the expected locations. Otherwise the NAO will not be able to walk stable and projections

---

7 On Windows, the first time starting such a file the `SimRobot.exe` must be manually chosen to open these files. Note that both on Windows and macOS, starting a scene description file bears the risk of executing a different version of SimRobot than the one that was just compiled.

8 The script might instead automatically connect to the IP address that was last used for login or deployment.
from image coordinates to world coordinates (and vice versa) will be wrong. In general, a lot of calculations will be unreliable. Two physical components of the NAO can be calibrated via SimRobot: the joints (cf. Sect. 2.8.2) and the cameras (cf. Sect. 2.8.3). Checking those calibrations from time to time is important, especially for the joints. New robots come with calibrated joints and are theoretically ready to play out of the box. However, over time and usage, the joints wear out. This is especially noticeable with the hip joint.

In addition to that, the B-Human software uses four color classes (cf. Sect. 4.1.4) which have to be calibrated, too (cf. Sect. 2.8.4). Changing locations or light conditions might require them to be adjusted.

### 2.8.1 Overall physical calibration

The physical calibration process can be split into three steps with the overall goal of an upright and straight standing robot, and a correctly calibrated camera. The first step is to get both feet in a planar position. This does not mean that the robot has to stand straight. It is done by lifting the robot up so that the bottom of the feet can be seen. The joint offsets of feet and legs are then changed until both feet are planar and the legs are parallel to one another. The distance between the two legs can be measured at the gray parts of the legs. They should be 10 cm apart from center to center.

The second step is the camera calibration (cf. Sect. 2.8.3). This step also measures the tilt of the body with respect to the feet. This measurement can then be used in the third step to improve the joint calibration and straighten the robot up (cf. Sect. 2.8.2). In some cases it may be necessary to repeat these steps.

### 2.8.2 Joint Calibration

The software supports two methods for calibrating the joints; either by manually adjusting offsets for each joint, or by using the `JointCalibrator` module which uses an inverse kinematic to do the same (cf. Sect. 8.3.3). The third step of the overall calibration process (cf. Sect. 2.8.1) can only be done via the `JointCalibrator`. When switching between those two methods, it is necessary to save the `JointCalibration`, redeploy the NAO and restart bhuman. Otherwise, the changes done previously will not be used.

Before changing joint offsets, the robot has to be set in a standing position with fixed joint angles. Otherwise, the balancing mechanism of the motion engine might move the legs, messing up the joint calibrations. This can be done with

```python
get representation: MotionRequest
```

and then set `motion = stand` in the returned statement.

When the calibration is finished it should be saved:

```python
save representation: JointCalibration
```

### Manually Adjusting Joint Offsets

First of all, the robot has to be switched to a stationary stand, otherwise the balancing mechanism of the motion engine might move the legs, messing up the joint calibration:

```python
mr StandOutput CalibrationStand
```
There are two ways to adjust the joint offsets. Either by requesting the `JointCalibration` representation with a `get` call:

```
get representation: JointCalibration
```

modifying the calibration returned and then setting it. Or by using a Data View (cf. Sect. 10.1.4.5)

```
v d representation: JointCalibration
```

which is more comfortable.

The `JointCalibration` also contains other information for each joint that should not be changed!

### Using the JointCalibrator

First set the `JointCalibrator` to provide the `JointCalibration` and switch to the `CalibrationStand`:

```
c all JointCalibrator
```

When a completely new calibration is desired, the `JointCalibration` can be reset:

```
d r module: JointCalibrator: reset
```

 Afterwards, the translation and rotation of the feet can be modified. Again either with

```
get module: JointCalibrator: offsets
```
or with:

```
v d module: JointCalibrator: offsets
```

The units of the translations are in millimeters and the rotations are in degrees.

### Straightening Up the NAO

The camera calibration (cf. Sect. 2.8.3) also calculates a rotation for the body rotation. These values can be passed to the `JointCalibrator` that will then set the NAO in an upright position. Call:

```
get representation: CameraCalibration
c all JointCalibrator
```

Copy the values of `bodyRotationCorrection` (representation `CameraCalibration`) into `bodyRotation` (representation `JointCalibration`). Afterwards, set `bodyRotationCorrection` (representation `CameraCalibration`) to zero. Another way to make these actions more or less automatically is possible by using the `AutomaticCameraCalibrator` with the automation flag (cf. Sect. 2.8.3).

The last step is to adjust the translation of both feet at the same time (and most times in the same direction) so they are perpendicular positioned below the torso. A plummet or line laser is very useful for that task.

When all is done save the representations by executing

```
s ave representation: JointCalibration
s ave representation: CameraCalibration
```

Then redeploy the NAO and restart bhuman.
2.8.3 Camera Calibration

For calibrating the cameras (cf. Sect. 4.1.2.1) using the module `AutomaticCameraCalibrator`, follow the steps below:

1. Connect the simulator to a robot on the field and place it on a defined spot (e.g. the penalty mark).

2. Run the SimRobot configuration file `AutomaticCameraCalibrator.con` (in the console type `call AutomaticCameraCalibrator`). This will initialize the calibration process and furthermore print commands or help to the simulator console that will be needed later on.

3. Announce the robot’s position on the field (cf. Sect. 4.1.2) using the `AutomaticCameraCalibrator` module (e.g. for setting the robot’s position to the penalty mark of a field, type `set module:AutomaticCameraCalibrator:robotPose rotation = 0; translation = {x = -3200; y = 0};` in the console).

4. To automatically generate the commands for the following joint calibration to correct the body rotation, you can set a flag via `set module:AutomaticCameraCalibrator:setJointOffsets true`. After you finished the optimization you can just enter the generated commands and thereby correct the rotation.

5. To start the point collection use the command `dr module:AutomaticCameraCalibrator:start` and wait for the output “Accumulation finished. Waiting to optimize...”. The process includes both cameras and will collect samples for the calibration and make the head motions to cover the whole field. The samples for the upper camera are drawn blue and the sample for the lower camera red. A drawing above the images signalizes if the sample amount is sufficient for optimization (green) or not (red).

6. If you are unhappy with the collection of some specific samples you are now able to delete samples by left-clicking onto the sample in the image in which it has been found. If there are some samples missing you can manually add them by `Ctrl + left-clicking` into the corresponding image.

7. Run the automatic calibration process using `dr module:AutomaticCameraCalibrator:optimize` and wait until the optimization has converged.
In addition to the code release of the last year [30], the module *AutomaticCameraCalibrator* has been fixed for the manual sample collection and deletion. The chance to delete or add a sample is like in the former manual versions of the calibrator. Also a color changed for the upper camera because the color and the size of samples in this camera could be mistaken with *BallSpots*.

![Figure 2.2: The three interesting camera calibration stages. a) is the start of the calibrator. b) is the view after the control start with gathered samples. c) is the stage after optimization.](image)

### 2.8.4 Color Calibration

Calibrating the color classes is split into two steps. First of all, the parameters of the camera driver must be updated to the environment’s needs. The command:

```python
get representation : CameraSettings
```

will return the current settings. Furthermore, the necessary `set` command will be generated. The most important parameters are:

**whiteBalance**: The white balance used. The available interval is $[2700, 6500]$.

**exposure**: The exposure used. The available interval is $[0, 1000]$. Usually, an exposure of 140 is used, which equals 14 ms. Be aware that high exposures lead to blurred images.

**gain**: The gain used. The available interval is $[0, 255]$. Usually, the gain is set to 50 - 70. Be aware that high gain values lead to noisy images.

**autoWhiteBalance**: Enable(1) / disable(0) the automatism for white balance. This parameter should always be disabled since a change in the white balance can change the color and mess up the color calibration. On the other hand, a real change in the color temperature of the environment will have the same result.

**autoExposure**: Enable (1) / disable (0) the automatism for exposure. When playing under static light conditions such as in the standard indoor tournament, this parameter should always be disabled, since the automation will often choose higher values than necessary, which will result in blurry images. However, for dynamic light conditions as were present in the Outdoor Competition at RoboCup 2016, using the automatism of the camera driver may be a necessity. In this case, its behavior can be altered using the parameters autoExposureAlgorithm and brightness.

The camera driver can do a one-time auto white balance. This feature can be triggered with the commands:
After setting up the parameters of the camera driver, the parameters of the color classes must be updated (cf. Sect. 4.1.4). To do so, one needs to open the views with the segmented upper and lower camera images and the color calibration view. See (cf. Sect. 10.1.4.1 and Sect. 10.1.4.1). After finishing the color class calibration and saving the current parameters, copyfiles/bush (cf. Sect. 2.5) can be used to deploy the current settings. Ensure the updated files cameraSettingsV5.cfg (or cameraSettingsV4.cfg if the NAO is a V4 model) and fieldColorsCalibrationV5.cfg (or fieldColorsCalibrationV4.cfg) are stored in the correct location.

2.9 Configuration Files

Since the recompilation of the code takes a lot of time in some cases and each robot needs a different configuration, the software uses a huge amount of configuration files which can be altered without causing recompilation. All the files that are used by the software\textsuperscript{9} are located below the directory Config.

Locations can be used to configure the software for different independent tasks. They can be set up by simply creating a new folder with the desired name within Config/Locations and placing configuration files in it. Those configuration files are only taken into account if the location is activated in the file Config/settings.cfg.

Besides the global configuration files, there are some files which depend on the robot’s head, body, or both. To differentiate the locations of these files, the names of the head and the body of each robot are used. They are defined in the file Config/Robots/robots.cfg that maps the serial numbers of the heads and the bodies of the robots to their actual names. In the Simulator, both names are always “Nao”.

To handle all these different configuration files, there are fall-back rules that are applied if a requested configuration file is not found. The search sequence for a configuration file is:

1. Config/Robots/<head name>/Head/<filename>

\textsuperscript{9}There are also some configuration files for the operating system of the robots that are located in the directory Install.
2.9. CONFIGURATION FILES

- Used for files that only depend on the robot’s **head**
  - e.g.: Robots/Amy/Head/cameraIntrinsics.cfg

2. **Config/Robots/**<body name>/Body/<filename>

- Used for files that only depend on the robot’s **body**
  - e.g.: Robots/Alex/Body/walkingEngine.cfg

3. **Config/Robots/**<head name>/<body name>/<filename>

- Used for files that depend on both, the robot’s head and **body**.
  - e.g.: Robots/Amy/Alex/cameraCalibration.cfg

4. **Config/Robots/**<head name>/<filename>

- If the head and body constellations for your robots are always the same, all files which belong to a robot may also be saved here.

5. **Config/Robots/Default/**<filename>

6. **Config/Locations/**<current location>/<filename>

7. **Config/Locations/Default/**<filename>

8. **Config/**<filename>

So, whether a configuration file is robot-dependent or location-dependent or should always be available to the software is just a matter of moving it between the directories specified above. This allows for a maximum of flexibility. Directories that are searched earlier might contain specialized versions of configuration files. Directories that are searched later can provide fallback versions of these configuration files that are used if no specialization exists.

Using configuration files within our software requires very little effort, because loading them is completely transparent for a developer when using parametrized modules (cf. Sect. 3.3.5).
Chapter 3

Architecture

The B-Human architecture [25] is based on the framework of the GermanTeam 2007 [24], adapted to the NAO. This chapter summarizes the major features of the architecture: binding, processes, modules and representations, communication, and debugging support.

3.1 Binding

The actuators and sensors (except the camera) of the NAO are accessed using the NAOqi SDK that is actually a stand-alone module framework that we do not use as such. Therefore, we deactivated all non-essential pre-assembled modules and implemented the very basic module libbhuman for accessing the actuators and sensors from another native platform process called bhuman that encapsulates the B-Human module framework.

Whenever the Device Communication Manager (DCM) reads a new set of sensor values, it notifies the libbhuman about this event using an atPostProcess callback function. After this notification, libbhuman writes the newly read sensor values into a shared memory block and raises a semaphore to provide a synchronization mechanism to the other process. The bhuman process waits for the semaphore, reads the sensor values that were written to the shared memory block, calls all registered modules within B-Human’s process Motion and writes the resulting actuator values back into the shared memory block right after all modules have been called. When the DCM is about to transmit desired actuator values (e.g. target joint angles) to the hardware, it calls the atPreProcess callback function. On this event libbhuman sends the desired actuator values from the shared memory block to the DCM.

It would also be possible to encapsulate the B-Human framework as a whole within a single NAOqi module, but this would lead to a solution with a lot of drawbacks. The advantages of the separated solution are:

- Both frameworks use their own address space without losing their real-time capabilities and without a noticeable reduction of performance. Thus, a malfunction of the process bhuman cannot affect NAOqi and vice versa.

- Whenever bhuman crashes, libbhuman is still able to display this malfunction using red blinking eye LEDs and to make the NAO sit down slowly. Therefore, the bhuman process uses its own watchdog that can be activated using the -w flag\(^1\) when starting the bhuman process. When this flag is set, the process forks itself at the beginning where one instance

\(^1\)The start up scripts bhuman and bhumand set this flag by default.
waits for a regular or irregular exit of the other. On an irregular exit the exit code can
be written into the shared memory block. The libbhuman monitors whether sensor values
were handled by the bhuman process using the counter of the semaphore. When this
counter exceeds a predefined value the error handling code will be initiated. When using
release code (cf. Sect. 2.2), the watchdog automatically restarts the bhuman process after
an irregular exit.

- Debugging with a tool such as the GDB is much simpler since the bhuman executable can
be started within the debugger without taking care of NAOqi.

The GameController (cf. Sect. 10.3) provides the library libgamectrl that handles the network
packets, sets the LEDs, and handles the official button interface. The library is already inte-
grated into the B-Human project. Since the libgamectrl is a NAOqi module, the libbhuman (cf.
Sect. 3.1) handles the data exchange with the library and provides the resulting game control
data packet to the main B-Human executable. The libbhuman also sets the team number, team
color, and player number whenever a new instance of the main B-Human executable is started,
so that the libgamectrl resets the game state to Initial.

3.2 Processes

Most robot control programs use concurrent processes. The number of parallel processes is best
dictated by external requirements coming from the robot itself or its operating system. The
NAO provides images from each camera at a frequency of 30 Hz and accepts new joint angles
at 100 Hz. For handling the camera images, there would actually have been two options: either
to have two processes each of which processes the images of one of the two cameras and a third
one that collects the results of the image processing and executes world modeling and behavior
control, or to have a single process that alternately processes the images of both cameras and also
performs all further steps. We use the latter approach, because each interprocess communication
might add delays to the system. Since the images of both cameras are processed, that single
process runs at 60 Hz. In addition, there is a process that runs at the motion frame rate of the
NAO, i.e. at 100 Hz. Another process performs the TCP communication with a host PC for
the purpose of debugging.

This results in the three processes Cognition, Motion, and Debug used in the B-Human system
(cf. Fig. 3.1). Cognition receives camera images from Video for Linux, as well as sensor data

![Figure 3.1: The processes used on the NAO](Image)
from the process Motion. It processes this data and sends high-level motion commands back to the process Motion. This process actually executes these commands by generating the target angles for the 25 joints of the NAO. It sends these target angles through the libbhuman to NAO’s Device Communication Manager, and it receives sensor readings such as the actual joint angles, acceleration and gyro measurements, etc. In addition, Motion reports about the motion of the robot, e.g., by providing the results of dead reckoning. The process Debug communicates with the host PC. It distributes the data received from it to the other two processes, and it collects the data provided by them and forwards it back to the host machine. It is inactive during actual games.

Processes in the sense of the architecture described can be implemented as actual operating system processes, or as threads. On the NAO and in the simulator, threads are used. In contrast, in B-Human’s past team in the Humanoid League, framework processes were mapped to actual processes of the operating system (i.e. Windows CE). For the sake of consistency, we will use the term “processes” in this document.

### 3.3 Modules and Representations

A robot control program usually consists of several modules, each performing a certain task, e.g. image processing, self-localization, or walking. Modules require a certain input and produce a certain output (so-called representations). Therefore, they have to be executed in a specific order to make the whole system work. The module framework introduced in [24] simplifies the definition of the interfaces of modules and automatically determines the sequence in which the modules must be executed. It consists of the blackboard, the module definition, and a visualization component (cf. Sect. 10.1.4.5).

#### 3.3.1 Blackboard

The blackboard [12] is the central storage for information, i.e. for the representations. Each process is associated with its own instance of the blackboard. Representations are transmitted through inter-process communication if a module in one process requires a representation that is provided by a module in another process. The blackboard itself is a map that associates names of representations to reference-counted instances of these representations. It only contains entries for the representations that at least one of the modules in associated process actually requires or provides.

#### 3.3.2 Module Definition

The definition of a module consists of three parts: the module interface, its actual implementation, and a statement that allows instantiating the module. Here an example:

```plaintext
MODULE(SimpleBallLocator
  ,
  , REQUIRES(BallPercept),
  , REQUIRES(FrameInfo),
  , PROVIDES(BallModel),
  , DEFINES_PARAMETERS(
    , (Vector2f)(5.f, 0.f) offset,
    , (float)(1.1f) scale,
    )
);
```

30
class SimpleBallLocator : public SimpleBallLocatorBase
{
    void update(BallModel& ballModel)
    {
        if(theBallPercept.wasSeen)
        {
            ballModel.position = theBallPercept.position * scale + offset;
            ballModel.wasLastSeen = theFrameInfo.frameTime;
        }
    }
}

MAKE_MODULE(SimpleBallLocator, modeling);

The module interface defines the name of the module (e.g. SimpleBallLocator), the representations that are required to perform its task, the representations provided by the module, and the parameters of the module, the values of which can either be defined in place or loaded from a file. The interface basically creates a base class for the actual module following the naming scheme <ModuleName>Base. The actual implementation of the module is a class that is derived from that base class. It has read-only access to all the required representations in the blackboard (and only to those), and it must define an update method for each representation that is provided. It also inherits all parameters. As will be described in Section 3.3.3, modules can expect that all their required representations have been updated before any of their provider methods is called. Finally, the MAKE_MODULE statement allows the module to be instantiated. It has a second parameter that defines a category that is used for a more structured visualization of the module configuration (cf. Sect. 10.1.4.5). This second parameter is also used to filter modules that can be loaded in the current framework environment, i.e. the process (cf. Sect. 3.2). In process Cognition the categories cognitionInfrastructure, communication, perception, modeling, and behaviorControl are available and in process Motion the categories motionInfrastructure, motionControl, and sensing. The list of available categories is defined in the main implementation file of the respective process (Src/Processes/Cognition.cpp and Src/Processes/Motion.cpp). While the module interface is usually part of the header file, the MAKE_MODULE statement has to be part of the implementation file.

MODULE is a macro that gets all the information about the module as parameters, i.e. they are all separated by commas. The macro ignores its second and its last parameter, because by convention, these are used for opening and closing curly brackets. These let some source code formatting tools to indent the definitions as a block. Currently, MODULE is limited to up to 80 definitions between the curly brackets. When the macro is expanded, it creates a lot of hidden functionality. Each entries that references a representation makes sure that it is created in the blackboard when the module is constructed and freed when the module is deconstructed. The information that a module has certain requirements and provides certain representations is not only used to generate a base class for that module, but is also available for sorting the providers, and can be requested by a host PC. On a host PC the information can be used to change the configuration and for visualization (cf. Sect. 10.1.4.5).

For each representation provided with PROVIDES the execution time can be determined (cf. Sect. 3.6.7) and it can be sent to a host PC or even altered by it. If the latter is not desired, PROVIDES WITHOUT MODIFY can be used instead. If a MessageID id<representation> exists, the representation can also be logged.

If a representation provided defines a parameterless method draw, that method will be called after the representation was updated. The method is intended to visualize the representation using the techniques described in Sect. 3.6.3. If the representation defines a parameterless
method `verify`, that method will be called in `Debug` and `Develop` builds after the representation was updated as well. A `verify` method should contain `ASSERT`s that check whether the contents of the representation are plausible. Both methods are only called if they are defined in the representation itself and not if they are inherited from its base class.

### 3.3.3 Configuring Providers

Since modules can provide more than a single representation, the configuration has to be performed on the level of providers. For each representation it can be selected which module provides it or that it is not provided at all. Normally, the configuration is read from the file `Config/Location/<location>/modules.cfg` during the start-up of the process, but it can also be changed interactively when the robot has a debug connection to a host PC using the command `mr` (cf. Sect. 10.1.6.3).

The configuration does not specify the sequence in which the providers are executed. This sequence is automatically determined at runtime based on the rule that all representations required by a provider must already have been provided by other providers before, i.e. those providers have to be executed earlier.

In some situations it is required that a certain representation is provided by a module before any other representation is provided by the same module, e.g., when the main task of the module is performed in the `update` method of that representation, and the other `update` methods rely on results computed in the first one. Such a case can be implemented by both requiring and providing a representation in the same module.

### 3.3.4 Pseudo-Module `default`

During the development of the robot control software, it is sometimes desirable to simply deactivate a certain provider or module. As mentioned above, it can always be decided not to provide a certain representation, i.e. all providers generating the representation are switched off. However, not providing a certain representation typically makes the set of providers inconsistent, because other providers rely on that representation, so they would have to be deactivated as well. This has a cascading effect. In many situations, it would be better to be able to deactivate a provider without any effect on the dependencies between the modules. That is what the module `default` was designed for. It is an artificial construct – so not a real module – that can provide all representations that can be provided by any module in the same process. It will never change any of the representations – so they basically remain in their initial state – but it will make sure that they exist, and thereby, all dependencies can be resolved. However, in terms of functionality, a configuration using `default` is never complete and should not be used during actual games.

### 3.3.5 Parameterizing Modules

Modules usually need some parameters to function properly. Those parameters can also be defined in the module’s interface description. Parameters behave like protected class members and can be accessed in the same way. Additionally, they can be manipulated from the console using the commands `get parameters:<ModuleName>` or `vd parameters:<ModuleName>` (cf. Sect. 10.1.6.3).

There are two different parameter initialization methods. In the hard-coded approach, the initialization values are specified as part of the C++ source file. They are defined using the
3.4. DEFINES_PARAMETERS macro. This macro is intended for parameters that may change during development but will never change again afterwards. In contrast, loadable parameters are initialized to values that are loaded from a configuration file upon module creation, i.e. the initialization values are not specified in the source file. These parameters are defined using the LOADS_PARAMETERS macro. By default, parameters are loaded from a file with the same base name as the module, but starting with a lowercase letter\(^2\) and the extension *.cfg*. For instance if a module is named SimpleBallLocator, its configuration file is simpleBallLocator.cfg. This file can be placed anywhere in the usual configuration file search path (cf. Sect. 2.9). It is also possible to assign a custom name to a module’s configuration file by passing the name as a parameter to the constructor of the module’s base class.

Only either DEFINES_PARAMETERS or LOADS_PARAMETERS can be used in a module definition. They can both only be used once. Their syntax follows the definition of generated streamable classes (cf. Sect. 3.4.4). Parameters may have any data type as long as it is streamable (cf. Sect. 3.4.3).

3.4 Serialization

In most applications, it is necessary that data can be serialized, i.e. transformed into a sequence of bytes. While this is straightforward for data structures that already consist of a single block of memory, it is a more complex task for dynamic structures, e.g. lists, trees, or graphs. Our implementation for streaming data follows the ideas introduced by the C++ iostreams library, i.e., the operators \(<<\) and \(>>\) are used to implement the process of serialization. In contrast to the iostreams library, our implementation guarantees that data is streamed in a way that it can be read back without any special handling, even when streaming into and from text files, i.e. the user of a stream does not need to know about the representation that is used for the serialized data (cf. Sect. 3.4.1).

On top of the basic streaming class hierarchy, it is also possible to derive classes from class Streamable and implement the mandatory method serialize(In*, Out*). In addition, the basic concept of streaming data was extended by a mechanism to gather information on the structure of the data while serializing it. This information is used to translate between the data in binary form and a human-readable format that reflects the hierarchy of a data structure, its variables, and their actual values.

As a third layer of serialization, two macros allow defining classes that automatically implement the method serialize(In*, Out*).

3.4.1 Streams

The foundation of B-Human’s implementation of serialization is a hierarchy of streams. As a convention, all classes that write data into a stream have a name starting with “Out”, while classes that read data from a stream start with “In”. In fact, all writing classes are derived from the class Out, and all reading classes are derivations of the class In. All classes support reading or writing basic datatypes with the exceptions of long, unsigned long, and size_t, because their binary representations have different sizes on currently used platforms (32/64 bits). They also provide the ability to read or write raw binary data.

All streaming classes derived from In and Out are composed of two components: One for read-

\(^2\)Actually, if a module name begins with more than one uppercase letter, all initial uppercase letters but the last one are transformed to lowercase, e.g. the module USControl reads the file usControl.cfg.
ing/writing the data from/to a physical medium and one for formatting the data from/to a specific format. Classes writing to physical media derive from `PhysicalOutStream`, classes for reading derive from `PhysicalInStream`. Classes for formatted writing of data derive from `StreamWriter`, classes for reading derive from `StreamReader`. The composition is done by the `OutStream` and `InStream` class templates.

A special case are the `OutMap` and the `InMap` streams. They only work together with classes that are derived from the class `Streamable`, because they use the structural information that is gathered in the `serialize` method. They are both directly derived from `Out` and `In`, respectively.

Currently, the following classes are implemented:

**PhysicalOutStream**: Abstract class

- **OutFile**: Writing into files
- **OutMemory**: Writing into memory
- **OutSize**: Determine memory size for storage
- **OutMessageQueue**: Writing into a MessageQueue

**StreamWriter**: Abstract class

- **OutBinary**: Formats data binary
- **OutText**: Formats data as text
- **OutTextRaw**: Formats data as raw text (same output as “cout”)

**Out**: Abstract class

- **OutStream<PhysicalOutStream, StreamWriter>**: Abstract template class
  - **OutBinaryFile**: Writing into binary files
  - **OutTextFile**: Writing into text files
  - **OutTextRawFile**: Writing into raw text files
  - **OutBinaryMemory**: Writing binary into memory
  - **OutTextMemory**: Writing into memory as text
  - **OutTextRawMemory**: Writing into memory as raw text
  - **OutBinarySize**: Determine memory size for binary storage
  - **OutTextSize**: Determine memory size for text storage
  - **OutTextRawSize**: Determine memory size for raw text storage
  - **OutBinaryMessage**: Writing binary into a MessageQueue
  - **OutTextMessage**: Writing into a MessageQueue as text
  - **OutTextRawMessage**: Writing into a MessageQueue as raw text

**OutMap**: Writing into a stream in configuration map format (cf. Sect. 3.4.5). This only works together with serialization (cf. Sect. 3.4.3), i.e. a streamable object has to be written. This class cannot be used directly.

- **OutMapFile**: Writing into a file in configuration map format
- **OutMapMemory**: Writing into a memory area in configuration map format
- **OutMapSize**: Determine memory size for configuration map format storage

**PhysicalInStream**: Abstract class
### 3.4. SERIALIZATION

- **InFile**: Reading from files
- **InMemory**: Reading from memory
- **InMessageQueue**: Reading from a MessageQueue

**StreamReader**: Abstract class

- **InBinary**: Binary reading
- **InText**: Reading data as text
- **InConfig**: Reading configuration file data from streams

**In**: Abstract class

**InStream<PhysicalInStream, StreamReader>**: Abstract class template

- **InBinaryFile**: Reading from binary files
- **InTextFile**: Reading from text files
- **InConfigFile**: Reading from configuration files
- **InBinaryMemory**: Reading binary data from memory
- **InTextMemory**: Reading text data from memory
- **InConfigMemory**: Reading config-file-style text data from memory
- **InBinaryMessage**: Reading binary data from a MessageQueue
- **InTextMessage**: Reading text data from a MessageQueue
- **InConfigMessage**: Reading config-file-style text data from a MessageQueue

**InMap**: Reading from a stream in configuration map format (cf. Sect. 3.4.5). This only works together with serialization (cf. Sect. 3.4.3), i.e. a streamable object has to be read. This class cannot be used directly.

- **InMapFile**: Reading from a file in configuration map format
- **InMapMemory**: Reading from a memory area in configuration map format

### 3.4.2 Streaming Data

To write data into a stream, `Tools/Streams/OutStreams.h` must be included, a stream must be constructed, and the data must be written into the stream. For example, to write data into a text file, the following code would be appropriate:

```cpp
#include "Tools/Streams/OutStreams.h"
// ...
OutTextFile stream("MyFile.txt");
stream << 1 << 3.14 << "Hello Dolly" << endl << 42;
```

The file will be written into the configuration directory, e.g. `Config/MyFile.txt` on the PC. It will look like this:

```
1 3.14 "Hello Dolly"
42
```

As spaces are used to separate entries in text files, the string “Hello Dolly” is enclosed in double quotes. The data can be read back using the following code:

```cpp
#include "Tools/Streams/InStreams.h"
// ...
InTextFile stream("MyFile.txt");
int a, d;
double b;
std::string c;
stream >> a >> b >> c >> d;
```
It is not necessary to read the symbol \texttt{endl} here, although it would also work, i.e. it would be ignored.

For writing to text streams without the separation of entries and the addition of double quotes, \texttt{OutTextRawFile} can be used instead of \texttt{OutTextFile}. It formats the data such as known from the ANSI C++ \texttt{cout} stream. The example above is formatted as following:

\begin{verbatim}
13.14 Hello Dolly
42
\end{verbatim}

To make streaming independent of the kind of the stream used, it could be encapsulated in functions. In this case, only the abstract base classes \texttt{In} and \texttt{Out} should be used to pass streams as parameters, because this generates the independence from the type of the streams:

\begin{verbatim}
#include "Tools/Streams/InOut.h"

void write(Out& stream)
{
    stream << 1 << 3.14 << " Hello Dolly" << endl << 42;
}

void read(In& stream)
{
    int a, d;
    double b;
    std::string c;
    stream >> a >> b >> c >> d;
} // ...

OutTextFile stream("MyFile.txt");
write(stream);
// ...
InTextFile stream("MyFile.txt");
read(stream);
\end{verbatim}

### 3.4.3 Streamable Classes

A class is made streamable by deriving it from the class \texttt{Streamable} and implementing the abstract method \texttt{serialize(In*, Out*)}. For data types derived from \texttt{Streamable} streaming operators are provided, meaning they may be used as any other data type with standard streaming operators implemented. To implement the \texttt{modify} functionality (cf. Sect. 3.6.6), the streaming method uses macros to acquire structural information about the data streamed. This includes the data types of the data streamed as well as that names of attributes. The process of acquiring names and types of members of data types is automated. The following macros can be used to specify the data to stream in the method \texttt{serialize}:

\texttt{STREAM_REGISTER_BEGIN} indicates the start of a streaming operation.

\texttt{STREAM_BASE(<class>)} streams the base class.

\texttt{STREAM(<attribute>[, <class>])} streams an attribute, retrieving its name in the process. The second parameter is optional. If the streamed attribute is of an enumeration type (single value, array, or vector) and that enumeration type is not defined in the current class, the second parameter specifies the name of the class in which the enumeration type is defined. The enumeration type streamed must either be defined with the \texttt{ENUM} macro (cf. Sect. 3.4.6) or a matching \texttt{getName} function must exist.
**STREAM_REGISTER_FINISH** indicates the end of the streaming operation for this data type.

These macros are intended to be used in the `serialize` method. For instance, to stream an attribute `test`, an attribute `testEnumVector` which is a vector of values of an enumeration type that is defined in this class, and an enumeration variable of a type which was defined in `SomeOtherClass`, the following code can be used:

```cpp
virtual void serialize(In* in, Out* out)
{
    STREAM_REGISTER_BEGIN;
    STREAM(test);
    STREAM(testEnumVector);
    STREAM(otherEnum, SomeOtherClass);
    STREAM_REGISTER_FINISH;
}
```

In addition to the above listed macros `STREAM_*()` there is another category of macros of the form `STREAM_*_EXT()`. In contrast to the macros described above, they are not intended to be used within the `serialize`-method, but to define the external streaming operators `operator<<(...)` and `operator>>(...)`. For this purpose, they take the actual stream to be read from or written to as an additional (generally the first) parameter. The advantage of using external streaming operators is that the class to be streamed does not need to implement a virtual method and thus can save the space needed for a virtual method table, which is especially reasonable for very small classes. Consider an example of usage as follows:

```cpp
template<typename T> Out & operator<<(Out & out, const Range<T> & range)
{
    STREAM_REGISTER_BEGIN_EXT(range);
    STREAM_EXT(out, range.min);
    STREAM_EXT(out, range.max);
    STREAM_REGISTER_FINISH;
    return out;
}
```

### 3.4.4 Generating Streamable Classes

The approach to make classes streamable described in the previous section has been proven tedious and prone to mistakes in recent years. Each new member variable has to be added in two or even three places, i.e. it must be declared, it must be streamed, and often it also must be initialized. Two macros automatically generate all this code for a streamable struct\(^3\) and optionally also initialize its member variables. The first is:

```cpp
STREAMABLE(<class>,
{ <header>,
  <comma-separated-declarations>,
})
```

The second is very similar:

```cpp
STREAMABLE_WITH_BASE(<class>, <base>, ...)
```

The parameters have the following meaning:

- **class**: The name of the struct to be declared.

\(^3\text{Meaning, the default access for members is public.}\)
**base:** Its base class. It must be streamable and its `serialize` method must not be private. The default (without `_WITH_BASE`) is the class `Streamable`.

**header:** Everything that can be part of a class body except for the attributes that should be streamable and the default constructor. Please note that this part must not contain commas that are not surrounded by parentheses, because C++ would consider it to be more than a single macro parameter otherwise. A workaround is to use the macro `COMMA` instead of an actual comma. However, the use of that macro should be avoided if possible, e.g. by defining constructors with comma-separated initializer lists outside of the `struct`'s body.

**comma-separated-declarations:** Declarations of the streamable attributes\(^4\) in four possible forms\(^5\):

\[
\begin{align*}
&\text{(type) <var>} \\
&(\text{type})(\text{init}) <var> \\
&((\text{enum-domain}) \text{type}) <var> \\
&((\text{enum-domain}) \text{type})(\text{init}) <var>
\end{align*}
\]

**type:** The type of the attribute that is declared.

**var:** The name of the attribute that is declared.

**init:** The initial value of the attribute, or in case an object is declared, the parameter(s) passed to its constructor.

**enum-domain:** If an enum is declared, the type of which is not declared in the current class, the class the enum is declared in must be specified here. The type and the optional init value are automatically prefixed by this entry with `::` in between. Please note that there is a single case that is not supported, i.e. streaming a `std::array` or `std::vector` of enums that are declared in another class, because in that case, the class name is not a prefix of the typename rather than a prefix of its type parameter. The macro would, e.g., generate `C::std::vector<E>` instead of `std::vector<C::E>`, which does not compile.

Please note that all these parts, including each declaration of a streamable attribute, are separated by commas, since they are parameters of a macro. Here is an example:

```cpp
STREAMABLE(Example,
{
  ENUM(ABC,
   {
    a,
    b,
    c,
   });

  Example()
  {
    std::memset(array, 0, sizeof(array));
  },

  (int) anInt,
  (float)(3.14f) pi,
  (int[4]) array,
  (Vector2f)(1.f, 2.f) aVector,
```\(^4\)Currently, the macros support up to 60 entries. \(^5\)The spaces are actually important when compiling on Windows.
In this example, all attributes except for \texttt{anInt} and \texttt{aLetter} would be initialized when an instance of the class is created.

The macros can even be used if the \texttt{serialize} method should do more than just streaming the member variables. If a method \texttt{onRead()} is defined within the \texttt{struct}, it will be called at the end of \texttt{serialize} if data was read allowing to implement some post-processing, e.g. to compute the values of member variables that are not streamed from the values of other member variables that were.

### 3.4.5 Configuration Maps

Configuration maps introduce the ability to handle serialized data from files in a random order. The sequence of entries in the file does not have to match the order of the attributes in the C++ data structure that is filled with them. In contrast to most streams presented in Sect. 3.4.1, configuration maps do not contain a serialization of a data structure, but rather a hierarchical representation.

Since configuration maps can be read from and be written to files, there is a special syntax for such files. A file consists of an arbitrary number of pairs of keys and values, separated by an equality sign, and completed by a semicolon. Values can be lists (encased by square brackets), complex values (encased by curly brackets) or plain values. If a plain value does not contain any whitespaces, periods, semicolons, commas, or equality signs, it can be written without quotation marks, otherwise it has to be encased in double quotes. Configuration map files have to follow this grammar:

\begin{verbatim}
map ::= record
record ::= field ';' { field ';' }
field ::= literal '=' ( literal | '{' record '}' | array )
array ::= ']' [ element { ',' element } [ ',' ] ']
    element ::= literal | '{' record '}'
    literal ::= '"' { anychar1 } '"' | { anychar2 }
\end{verbatim}

\texttt{anychar1} must escape doublequotes and the backslash with a backslash. \texttt{anychar2} cannot contain whitespace and other characters used by the grammar. However, when such a configuration map is read, each \texttt{literal} must be a valid literal for the datatype of the variable it is read into. As in C++, comments can be denoted either by // for a single line or by /* ... */ for multiple lines. Here is an example:

```cpp
// A record
defensiveGoaliePose = {
    rotation = 0;
    translation = {x = -4300; y = 0;};
};
/* An array of
 * three records
*/
kickoffTargets = [
    {x = 2100; y = 0;},
    {x = 1500; y = -1350;},
    {x = 1500; y = 1350;}
];
// Some individual values
outOfCenterCircleTolerance = 150.0;
bballCounterThreshold = 10;
```
Configuration maps can only be read or written through the streams derived from `OutMap` and `InMap`. Accordingly, they require an object of a streamable class to either parse the data in map format or to produce it. Here is an example of code that reads the file shown above:

```cpp
STREAMABLE(KickOffInfo,
{
    (Pose2f) defensiveGoaliePose,
    (std::vector<Vector2f>) kickoffTargets,
    (float) outOfCenterCircleTolerance,
    (int) ballCounterThreshold,
});

InMapFile stream("kickOffInfo.cfg");
KickOffInfo info;
if(stream.exists())
    stream >> info;
```

### 3.4.6 Enumerations

To support streaming, enumeration types should be defined using the macro `ENUM` defined in `Src/Tools/Streams/Enum.h` rather than using the C++ `enum` keyword directly. The macro’s first parameter is the name of the enumeration type. The second and the last parameter are reserved for curly brackets and are ignored. All other parameters are the elements of the defined enumeration type. It is not allowed to assign specific integer values to the elements of the enumeration type, with one exception: It is allowed to initialize an element with the symbolic value of the element that has immediately been defined before (see example below). The macro automatically defines a function `static inline const char* getName(Typename)`, which can return the string representations of all “real” enumeration elements, i.e. all elements that are not just synonyms of other elements. In addition, the function will return 0 for all values outside the range of the enumeration type.

The macro also automatically defines a constant `numOf<Typename>` which reflects the number of elements in the enumeration type. Since the name of that constant has an added “s” at the end, enumeration type names should be singular. If the enumeration type name already ends with an “s”, it might be a good idea to define a constant outside the enumeration type that can be used instead, e.g. `static const unsigned char numOfClasses = numOfClasss` for an enumeration type with the name `Class`.

The following example defines an enumeration type `Letter` with the “real” enumeration elements `a`, `b`, `c`, and `d`, a user-defined helper constant `numOfLettersBeforeC`, and an automatically defined helper constant `numOfLetters`. The numerical values of these elements are `a = 0`, `b = 1`, `c = 2`, `d = 3`, `numOfLettersBeforeC = 2`, `numOfLetters = 4`. In addition, the function `getName(Letter)` is defined that can return “a”, “b”, “c”, “d”, and 0.

```cpp
ENUM(Letter,
{
    a,
    b,
    numOfLettersBeforeC,
    c = numOfLettersBeforeC,
    d,
});
```
3.4.6.1 Iterating over Enumerations

It is often necessary to enumerate all constants defined in an enumeration type. This can easily be done using the `FOREACH_ENUM` macro:

```c
FOREACH_ENUM (Letter, letter)
{
    // do something with "letter", which is of type "Letter"
}
```

It is also possible to specify a different upper limit to only enumerate a part of the constants. The upper limit must be one of the constants defined:

```c
FOREACH_ENUM (Letter, letter, numOfLettersBeforeC)
{
    // do something with "letter", which is of type "Letter"
}
```

If the enumeration type is defined in another class, that class needs to be specified separately:

```c
FOREACH_ENUM ((MotionRequest) Motion, motion)
{
    // do something with "motion", which is of type "MotionRequest::Motion"
}
```

In that case, an optional upper limit would automatically be assumed to be defined in the other class as well, i.e. it must not be prefixed by the classes’ name.

3.4.6.2 Enumerations as Array Indices

In the B-Human code, enumerations are often used as indices for arrays, because this gives entries a name, but still allows to iterate over these entries. However, in configuration files and in the UI, it is hard to find specific entries in arrays, in particular if they have a larger number of elements. For instance, the arrays of all joint angles has 26 elements, making it hard to identify the angle of a specific joint. Therefore, a special macro (defined in `Src/Tools/Streams/EnumIndexedArray.h`) that is backed by a template class allows to define an array indexed by an enumeration type that solves this problem by streaming such an array as if it were a structure with a member variable for each of its elements named after the respective enumeration constant. Technically, such an array is derived from `std::array` and simply defines the method `serialize` (cf. Sect. 3.4.3). For example, an array of all joint angles could be defined by using the enumeration `Joints::Joint` as index:

```c
ENUM_INDEXED_ARRAY (Angle, (Joints) Joint) jointAngles;
```

`jointAngles` still behaves like an array, i.e. it is derived from `std::array<Angle, Joints::numOfJoints>`, but when, e.g., written to a file using `OutMapFile` (cf. Sect. 3.4.5), it would appear differently:

```plaintext
headYaw = 0 deg;
headPitch = 0 deg;
lShoulderPitch = 0 deg;
...
```

The class prefix of the enumeration type used as an index only needs to be specified if that type was defined in another class. However, if the elements of the array are also of an enumeration type, that type must be fully specified, i.e. including the class prefix, even if that type is defined in the same class as the array:

```c
ENUM_INDEXED_ARRAY ((MotionRequest) Motion, Letter) someArray;
```
3.5 Communication

Three kinds of communication are implemented in the B-Human framework: *inter-process communication, debug communication, and team communication.*

3.5.1 Inter-process Communication

The representations sent back and forth between the processes *Cognition* and *Motion* (so-called shared representations) are automatically calculated by the *ModuleManager* based on the representations required by modules loaded in the respective process but not provided by modules in the same process. The directions in which they are sent are also automatically determined by the *ModuleManager*.

All inter-process communication is triple-buffered. Thus, processes never block each other, because they never access the same memory blocks at the same time. In addition, a receiving process always gets the most current version of a packet sent by another process.

3.5.2 Message Queues

The *debug communication, the team communication, and logging* are all based on the same technology: *message queues*. The class *MessageQueue* allows storing and transmitting a sequence of messages. Each message has a type (defined in *Src/Tools/MessageQueue/MessageIDs.h*) and a content. Each queue has a maximum size which is defined in advance. On the robot, the amount of memory required is pre-allocated to avoid allocations during runtime. On the PC, the memory is allocated on demand, because several sets of robot processes can be instantiated at the same time, and the maximum size of the queues is rarely needed.

Since almost all data types are streamable (cf. Sect. 3.4), it is easy to store them in message queues. The class *MessageQueue* provides different write streams for different formats: messages that are stored through *out.bin* are formatted binary. The stream *out.text* formats data as text and *out.textRaw* as raw text. After all data of a message was streamed into a queue, the message must be finished with *out.finishMessage(MessageID)*, giving it a *message id*, i.e. a type.

```cpp
MessageQueue m;
m.setSize(1000); // can be omitted on PC
m.out.text << "Hello world!";
m.out.finishMessage(idText);
```

To declare a new message type, an id for the message must be added to the enumeration type *MessageID* in *Src/Tools/MessageQueue/MessageIDs.h*. The enumeration type has three sections: the first for representations that should be recorded in log files, the second for team communication, and the last for infrastructure. When changing this enumeration by adding, removing, or re-sorting message types, compatibility issues with existing log files or team mates running an older version of the software are highly probable.

Messages are read from a queue through a message handler that is passed to the queue’s method *handleAllMessages(MessageHandler*)*. Such a handler must implement the method *handleMessage(InMessage*) that is called for each message in the queue. It must be implemented in a way as the following example shows:

```cpp
class MyClass : public MessageHandler
{
    protected:
```
bool handleMessage(InMessage& message)
{
    switch(message.getMessageID())
    {
        default:     return false;
        case idText:
        {
            std::string text;
            message.text >> text;
            return true;
        }
    }
}

The handler has to return whether it handled the message or not. Messages are read from a MessageQueue via streams. Thereto, message.bin provides a binary stream, message.text a text stream, and message.config a text stream that skips comments.

### 3.5.3 Debug Communication

For debugging purposes, there is a communication infrastructure between the processes Cognition and Motion and the PC. This is accomplished by debug message queues. Each process has two of them: theDebugSender and theDebugReceiver, often also accessed through the references debugIn and debugOut. The macro OUTPUT(<id>, <format>, <sequence>) defined in Src/Tools/Debugging/Debugging.h simplifies writing data to the outgoing debug message queue. id is a valid message id, format is text, bin, or textRaw, and sequence is a streamable expression, i.e. an expression that contains streamable objects, which – if more than one – are separated by the streaming operator <<.

```cpp
OUTPUT(idText, text, "Could not load file " << filename << " from " << path);
OUTPUT(idImage, bin, Image());
```

For receiving debugging information from the PC, each process also has a message handler, i.e. it implements the method handleMessage to distribute the data received.

The process Debug manages the communication of the robot control program with the tools on the PC. For each of the other processes (Cognition and Motion), it has a sender and a receiver for their debug message queues (cf. Fig. 3.1). Messages that arrive via WLAN or Ethernet from the PC are stored in debugIn. The method Debug::handleMessage(InMessage&) distributes all messages in debugIn to the other processes. The messages received from Cognition and Motion are stored in debugOut. When a WLAN or Ethernet connection is established, they are sent to the PC via TCP/IP. To avoid communication jams, it is possible to send a QueueFillRequest to the process Debug. The command qfr to do so is explained in Section 10.1.6.3.

### 3.5.4 Team Communication

The purpose of the team communication is to send messages to the other robots in the team. These messages are always broadcasted, so all teammates can receive them. Originally, our team communication used a message queue embedded in a UDP package that could be filled from everywhere in the process Cognition using the macro TEAM_OUTPUT. However, filling the queue was mostly bundled in the module TeamDataSender later to make it more clear which
information is actually exchanged. In addition, in 2014 the `SPLStandardMessage` was introduced as the format for team communication. Therefore, the approach currently used is still based on a message queue, but some information is filtered from the queue to fill the standardized fields of the UDP packet. The remaining messages are still sent in the form of a message queue in the non-standardized part of the packet.

The non-standardized part also contains a header that allows our robots to detect that they actually received a B-Human message and can safely interpret the remaining data as a message queue. When receiving, the whole contents of a `SPLStandardMessage` are converted back to a single message queue that is then processed by the module `TeammateDataProvider`. It also implements the network time protocol (NTP) and translates time stamps contained in the messages it receives into the local time of the robot.

According to the rules, team communication packets are only broadcasted every 200 ms. The representation `TeammateData` contains a flag that states whether a team communication packet will be sent out in the current frame or not. The `TeammateDataProvider` also provides the representation `UnfilteredTeammateData` that is the input for a module that judges the reliability of the information provided by teammates during the Drop-in Player competition. This module will then provide the representation `TeammateData` instead with all messages from untrustworthy teammates removed.

### 3.6 Debugging Support

Debugging mechanisms are an integral part of the B-Human framework. They are all based on the debug message queues already described in Section 3.5.3. Since the software runs fast enough and the debug mechanisms greatly support developers when working with a NAO, these mechanisms are available in all project configurations.

#### 3.6.1 Debug Requests

*Debug requests* are used to enable and disable parts of the source code. They can be seen as runtime switches for debugging.

The debug requests can be used to trigger certain debug messages to be sent as well as to switch on certain parts of algorithms. They can be sent using the SimRobot software when connected to a NAO (cf. command `dr` in Sect. 10.1.6.3). The following macros ease the use of the mechanism as well as hide the implementation details:

- `DEBUG_RESPONSE(\langle id\rangle)` executes the following statement or block if the debug request with the name `id` is enabled.
- `DEBUG_RESPONSE_ONCE(\langle id\rangle)` executes the following statement or block *once* when the debug request with the name `id` is enabled.
- `DEBUG_RESPONSE_NOT(\langle id\rangle)` executes the following statement or block if the debug request with the name `id` is *not* enabled.

These macros can be used anywhere in the source code, allowing for easy debugging. For example:

```c
DEBUG_RESPONSE("test") test();
```
This statement calls the method test() if the debug request with the identifier “test” is enabled. Debug requests are commonly used to send messages on request as the following example shows:

```c
DEBUG_RESPONSE("sayHello") OUTPUT(idText, text, "Hello");
```

This statement sends the text “Hello” if the debug request with the name "sayHello" is activated. Please note that only those debug requests are usable that are in the current path of execution. This means that only debug requests in those modules can be activated that are currently executed. To determine which debug requests are currently available, a method called polling is employed. It asks each debug response to report the name of the debug request that would activate it. This information is collected and sent to the PC (cf. command poll in Sect. 10.1.6.3).

### 3.6.2 Debug Images

Debug images are used for low level visualization of image processing debug data. They can either be displayed as background image of an image view (cf. Sect. 10.1.4.1) or in a color space view (cf. Sect. 10.1.4.1). Each debug image consists of pixels in one of the formats defined in `Src/Tools/ImageProcessing/ PixelTypes.h`, which are RGB, BGRA, YUYV, YUV, Colored and Grayscale. Except for Colored and Grayscale, pixels of these formats are all made up of four unsigned byte values, each representing one of the color channels of the format. The YUYV format is special as one word of the debug image describes two image pixels by specifying two luminance values, but only one value per U and V channel. Thus, it resembles the YUV422 format of the images supplied by the NAO’s cameras. Debug images in the Grayscale format only contain a single channel of unsigned byte values describing the luminance of the associated pixel. The Colored format consists of only one channel, too. Values in this channel are entries of the FieldColors::Color enumeration which contains identifiers for the color classes used for image processing (cf. Sect. 4.1.4).

Debug images are supposed to be declared as instances of the template class `TImage`, instantiated with one of the pixel formats named above, or the `Image` class which is otherwise only used for the camera image. The following macros are used to transfer debug images to a connected PC.

```c
SEND_DEBUG_IMAGE(<id>, <image>) sends the debug image to the PC. The identifier given to this macro is the name by which the image can be requested.

COMPLEX_IMAGE(<id>) only executes the following statement if the creation of a certain debug image is requested. For debug images that require complex instructions to paint, it can significantly improve the performance to encapsulate the drawing instructions in this macro (and maybe additionally in a separate method).
```

These macros can be used anywhere in the source code, allowing for easy creation of debug images. For example:

```c
class Test {
private:
  TImage<GrayscaledPixel> testImage;

public:

  void doSomething() {
    // [...]
    COMPLEX_IMAGE("test") draw();
  }
}
```
The example calls the `draw()` method if the "test" image was requested, which then initializes a grayscale debug image, paints it gray and sends it to the PC.

### 3.6.3 Debug Drawings

Debug drawings provide a virtual 2-D drawing paper and a number of drawing primitives, as well as mechanisms for requesting, sending, and drawing these primitives to the screen of the PC. In contrast to debug images, which are raster-based, debug drawings are vector-based, i.e., they store drawing instructions instead of a rasterized image. Each drawing has an identifier and an associated type that enables the application on the PC to render the drawing to the right kind of drawing paper. In the B-Human system, two standard drawing papers are provided, called “drawingOnImage” and “drawingOnField”. This refers to the two standard applications of debug drawings, namely drawing in the system of coordinates of an image and drawing in the system of coordinates of the field. Hence, all debug drawings of type “drawingOnImage” can be displayed in an image view (cf. Sect. 10.1.4.1) and all drawings of type “drawingOnField” can be rendered into a field view (cf. Sect. 10.1.4.1).

The creation of debug drawings is encapsulated in a number of macros in `Src/Tools/Debugging/DebugDrawings.h`. Most of the drawing macros have parameters such as pen style, fill style, or color. Available pen styles (`solidPen`, `dashedPen`, `dottedPen`, and `noPen`) and fill styles (`solidBrush` and `noBrush`) are part of the namespace `Drawings`. Colors can be specified as `ColorRGBA`. The class also contains a number of predefined colors such as `ColorRGBA::red`. A few examples for drawing macros are:

- **DECLARE_DEBUG_DRAWING**\(<id>, <type>\) declares a debug drawing with the specified `id` and `type`.\n
- **COMPLEX_DRAWING**\(<id>\) only executes the following statement or block if the creation of a certain debug drawing is requested. This can significantly improve the performance when a debug drawing is not requested, because for each drawing instruction it has to be tested whether it is currently required or not. By encapsulating them in this macro (and maybe in addition in a separate method), only a single test is required. However, the macro **DECLARE_DEBUG_DRAWING** must be placed outside of **COMPLEX_DRAWING**.

- **DEBUG_DRAWING**\(<id>, <type>\) is a combination of **DECLARE_DEBUG_DRAWING** and **COMPLEX_DRAWING**. It declares a debug drawing with the specified `id` and `type`. It also executes the following statement or block if the debug drawing is requested.

- **CIRCLE**\(<id>, <x>, <y>, <radius>, <penWidth>, <penStyle>, <penColor>, <fillStyle>, <fillColor>\) draws a circle with the specified radius, pen width, pen style, pen color, fill style, and fill color at the coordinates \((x, y)\) to the virtual drawing paper.
3.6. DEBUGGING SUPPORT

**LINE(** id, x1, y1, x2, y2, penWidth, penStyle, penColor **)**
draws a line with the pen color, width, and style from the point (x1,y1) to the point (x2,y2) to the virtual drawing paper.

**DOT(** id, x, y, penColor, fillColor **)** draws a dot with the pen color and fill color at the coordinates (x,y) to the virtual drawing paper. There also exist two macros **MID_DOT** and **LARGE_DOT** with the same parameters that draw dots of larger size.

**DRAWTTEXT(** id, x, y, fontSize, color, text **)** writes a text with a font size in a color to a virtual drawing paper. The upper left corner of the text will be at coordinates (x,y).

**TIP(** id, x, y, radius, text **)** adds a tool tip to the drawing that will pop up when the mouse cursor is closer to the coordinates (x,y) than the given radius.

**ORIGIN(** id, x, y, angle **)** changes the system of coordinates. The new origin will be at (x,y) and the system of coordinates will be rotated by angle (given in radians). All further drawing instructions, even in other debug drawings that are rendered afterwards in the same view, will be relative to the new system of coordinates, until the next origin is set. The origin itself is always absolute, i.e. a new origin is not relative to the previous one.

These macros can be used wherever statements are allowed in the source code. For example:

```cpp
DECLARE_DEBUG_DRAWING("test", "drawingOnField");
CIRCLE("test", 0, 0, 1000, 10, Drawings::solidPen, ColorRGBA::blue,
      Drawings::solidBrush, ColorRGBA(0, 0, 255, 128));
```

This example initializes a drawing called **test** of type **drawingOnField** that draws a blue circle with a solid border and a semi-transparent inner area.

### 3.6.4 3-D Debug Drawings

In addition to the aforementioned two-dimensional debug drawings, there is a second set of macros in Src/Tools/Debugging/DebugDrawings3D.h which provide the ability to create three-dimensional debug drawings.

3-D debug drawings can be declared with the macro  **DECLARE_DEBUG_DRAWING3D**( id, type ). The id can then be used to add three dimensional shapes to this drawing. type defines the coordinate system in which the drawing is displayed. It can be set to “field”, “robot”, or any named part of the robot model in the scene description. Note that drawings directly attached to hinges will be drawn relative to the base of the hinge, not relative to the moving part. Drawings of the type “field” are drawn relative to the center of the field, whereas drawings of the type “robot” are drawn relative to the origin of the robot according to SoftBank Robotics’ documentation. B-Human uses a different position in the robot as origin, i.e. the middle between the two hip joints. An object called “origin” has been added to the NAO simulation model at that position. It is often used as reference frame for 3-D debug drawings in the current code. The optional parameter allows defining code that is executed while the drawing is requested.

The parameters of macros adding shapes to a 3-D debug drawing start with the id of the drawing this shape will be added to, followed, e.g., by the coordinates defining a set of reference points (such as corners of a rectangle), and finally the drawing color. Some shapes also have other parameters such as the thickness of a line. Here are a few examples for shapes that can be used in 3-D debug drawings:
LINE3D(<id>, <fromX>, <fromY>, <fromZ>, <toX>, <toY>, <toZ>, <size>, <color>) draws a line between the given points.

QUAD3D(<id>, <corner1>, <corner2>, <corner3>, <corner4>, <color>) draws a quadrangle with its four corner points given as 3-D vectors and specified color.

SPHERE3D(<id>, <x>, <y>, <z>, <radius>, <color>) draws a sphere with specified radius and color at the coordinates (x, y, z).

COORDINATES3D(<id>, <length>, <width>) draws the axis of the coordinate system with specified length and width into positive direction.

COMPLEX_DRAWING3D(<id>) only executes the following statement or block if the creation of the debug drawing is requested (similar to COMPLEX_DRAWING(<id>) for 2-D drawings).

DEBUG_DRAWING3D(<id>, <type>) is a combination of DECLARE_DEBUG_DRAWING3D and COMPLEX_DRAWING3D. It declares a debug drawing with the specified id and type. It also executes the following statement or block if the debug drawing is requested.

The header file furthermore defines some macros to scale, rotate, and translate an entire 3-D debug drawing:

SCALE3D(<id>, <x>, <y>, <z>) scales all drawing elements by given factors for x, y, and z axis.

ROTATE3D(<id>, <x>, <y>, <z>) rotates the drawing counterclockwise around the three axes by given radians.

TRANSATE3D(<id>, <x>, <y>, <z>) translates the drawing according to the given coordinates.

An example for 3-D debug drawings (analogously to the example for regular 2-D debug drawings):

DECLARE_DEBUG_DRAWING3D("test3D", "field");
SPHERE3D("test3D", 0, 0, 250, 75, ColorRGBA::blue);

This example initializes a 3-D debug drawing called test3D which draws a blue sphere. Because the drawing is of type field and the origin of the field coordinate system is located in the center of the field, the sphere’s center will appear 250 mm above the center point.

Watch the result in the scene view of SimRobot (cf. Sect. 10.1.3) by sending a debug request (cf. Sect. 10.1.6.3) for a 3-D debug drawing with the ID of the desired drawing prefixed by “debugDrawing3d:” as its only parameter. If you wanted to see the example described above, you would type “dr debugDrawing3d:test3D” into the SimRobot console. The rendering of debug drawings can be configured for individual scene views by right-clicking on the view and selecting the desired type of visualization in the “Drawings Rendering” submenu.

3.6.5 Plots

The macro PLOT(<id>, <number>) allows plotting data over time. The plot view (cf. Sect. 10.1.4.5) will keep a history of predefined size of the values sent by the macro PLOT and
plot them in different colors. Hence, the previous development of certain values can be observed as a time series. Each plot has an identifier that is used to separate the different plots from each other. A plot view can be created with the console commands vp and vpd (cf. Sect. 10.1.6.3).

For example, the following code statement plots the measurements of the gyro for the pitch axis in degrees. It should be placed in a part of the code that is executed regularly, e.g. inside the update method of a module.

```cpp
PLOT("gyroY", theInertialSensorData.gyro.y());
```

The macro DECLARE_PLOT(<id>) allows using the PLOT(<id>, <number>) macro within a part of code that is not regularly executed as long as the DECLARE_PLOT(<id>) macro is executed regularly.

### 3.6.6 Modify

The macro MODIFY(<id>, <object>) allows reading and modifying of data on the actual robot during runtime. Every streamable data type (cf. Sect. 3.4.3) can be manipulated and read, because its inner structure is gathered while it is streamed. This allows generic manipulation of runtime data using the console commands get and set (cf. Sect. 10.1.6.3). The first parameter of MODIFY specifies the identifier that is used to refer to the object from the PC, the second parameter is the object to be manipulated itself. When an object is modified using the console command set, it will be overridden each time the MODIFY macro is executed.

```cpp
int i = 3;
MODIFY("i", i);
MotionRequest m;
MODIFY("representation: MotionRequest", m);
```

The macro PROVIDES of the module framework (cf. Sect. 3.3) includes the MODIFY macro for the representation provided. For instance, if a representation Foo is provided by PROVIDES(Foo), it is modifiable under the name representation:Foo. If a representation provided should not be modificable, e.g., because its serialization does not register all member variables, it must be provided using PROVIDES_WITHOUT_MODIFY.

If a single variable of an enumeration type should be modified, another macro called MODIFY_ENUM has to be used. It has an optional third parameter to which the name of the class in which the enumeration type is defined. It can be omitted if the enumeration type is defined in the current class. This is similar to the STREAM macro (cf. Sect. 3.4.3).

```cpp
class Foo
{
public :
    ENUM(Bar, {
        a, b, c,
    });

    void f()
    {
        Bar x = a;
        MODIFY_ENUM("x", x);
    }
};

class Other
3.6.7 Stopwatches

Stopwatches allow the measurement of the execution time of parts of the code. The macro 
\texttt{STOPWATCH(<id>)} (declared in \texttt{Src/Tools/Debugging/Stopwatch.h}) measures the runtime of 
the statement or block that follows. \textit{id} is a string used to identify the time measurement. To 
activate the time measurement of all stopwatches, the debug request \texttt{dr timing} has to be sent. 
The measured times can be seen in the timing view (cf. Sect. 10.1.4.5). By default, a stopwatch 
is already defined for each representation that is currently provided and for the whole processes 
\textit{Cognition} and \textit{Motion}.

An example to measure the runtime of a method called \textit{myCode}:

\begin{verbatim}
STOPWATCH("myCode") myCode();
\end{verbatim}

Often rudimentary statistics (cf. Sect. 10.1.4.5) of the execution time of a certain part of the 
code are not sufficient, but the actual progress of the runtime is needed. For this purpose 
there is another macro \texttt{STOPWATCH\_WITH\_PLOT(<id>)} that enables measuring \textit{and} plotting the 
exection time of some code. This macro is used exactly as the one without \texttt{WITH\_PLOT}, but it 
additionally creates a plot \texttt{stopwatch:myCode} (assuming the example above) (cf. Sect. 3.6.5 for 
the usage of plots).

3.7 Logging

The B-Human framework offers a sophisticated logging functionality that can be used to log 
the values of selected representations while the robot is playing. There are two different ways 
of logging data:

3.7.1 Online Logging

The online logging feature can be used to log data directly on the robot during regular games. 
It is implemented as part of the \textit{Cognition} and \textit{Motion} processes and is designed to log repre-
sentations in real-time.

Online logging starts as soon as the robot enters the \textit{ready} state and stops upon entering the 
\textit{finished} state. Due to the limited space on the NAO’s flash drive, the log files are compressed 
on the fly using Google’s snappy compression [8]. The name of the log file consists of the name of 
the process it logs and the names of the head and bodies of the robot. If connected to the 
GameController, the name of the opponent team, the half, and the player number are also added 
to the log file name. Otherwise, “Testing” is used instead. If a log file with the given name 
already exists, a number is added that is incremented for each duplicate.

To retain the real-time properties of the processes, the heavy lifting, i.e. compression and writing 
of the file, is done in separate threads without real-time scheduling, one for \textit{Cognition} logging 
and one for \textit{Motion} logging. These threads use every bit of remaining processor time that is
not used by one of the real-time parts of the system. Communication with these threads is
realized using very large ring buffers (usually around 500MB). Each element of the ring buffers
represents one second of data. If a buffer is full, the current element is discarded. Due to the
size of the buffers, writing of log files might continue for quite a while after the robot has entered
the finished state.

Due to the limited buffer size, the logging of very large representations such as the Image is
not possible. It would cause a buffer overflow within seconds rendering the resulting log file
unreadable. However, without images a log file is nearly useless, therefore loggable thumbnail
images (cf. Sect. 3.7.7) or image patches (cf. Sect. 3.7.8) are generated and used instead.

In addition to the logged representations, online log files also contain timing data, which can be
seen using the TimingView (cf. Sect. 10.1.4.5).

3.7.2 Configuring the Online Loggers

The online loggers can be configured by changing values in the files logger<Process>.cfg, which
should be located inside the configuration folder.

Following values can be changed:

logfilePath: All log files will be stored in this path on the NAO.

maxBufferSize: Maximum size of the buffer in seconds.

blockSize: How much data can be stored in a single buffer slot, i.e. in one second. Note that
maxBufferSize × blockSize is always allocated.

representations: List of all representations that should be logged.

enabled: An online logger is enabled if this is true. Note that it is not possible to enable online
loggers inside the simulator.

writePriority: Priority of the logging process. Priorities greater than zero use the real-time
scheduler, zero uses the normal scheduler, and negative values (−1 and −2) use idle pri-

orities.

depbugStatistics: If true a logger will print debugging messages to the console.

minFreeSpace: The minimum amount of disk space that should always be left on the NAO in
bytes. If the free space falls below this value the logger will cease operation.

3.7.3 Remote Logging

Online logging provides the maximum possible amount of debugging information that can be
collected without breaking the real-time capability. However in some situations one is interested
in high precision data (i.e. full resolution images) and does not care about real-time. The
remote logging feature provides this kind of log files. It utilizes the debugging connection (cf.
Sect. 3.5.3) to a remote computer and logs all requested representations on that computer. This
way the heavy lifting is outsourced to a computer with much more processing power and a bigger
hard drive. However sending large representations over the network severely impacts the NAO’s
performance resulting in loss of the real-time capability.

To reduce the network load, it is usually a good idea to limit the number of representations to
the ones that are really needed for the task at hand. Listing 3.7.3 shows the commands that
need to be entered into SimRobot to record a minimal vision log.
3.7.4 Log File Format

In general, log files consist of serialized message queues (cf. Sect. 3.5.2). Each log file consists of up to three chunks. Each chunk is prefixed by a single byte defining its type. The enum `log fileFormat` in `Src/Tools/Logging/log fileFormat.h` defines these chunk identifiers in the namespace `Logging` that have to appear in the given sequence in the log file if they appear at all:

- `logFileMessageIDs`: This chunk contains a string representation of the `MessageIDs` (cf. Sect. 3.5.2) stored in this log file. It is used to convert the `MessageIDs` from the log file to the ones defined in the version of `SimRobot` (cf. Sect. 10.1) that is replaying the log file. Thereby, log files still remain usable after the enumeration `MessageID` was changed.

- `logFileStreamSpecification`: This chunk contains the specification of all datatypes used in the log file. It is used to convert that data logged to the specifications that are defined in the version of `SimRobot` that is replaying the log file. If the specification changed, messages will appear in `SimRobot`’s console about the representations that are converted. Please note that a conversion is only possible for representations the specification of which is fully registered. This is not the case for representations that use `read` and `write` methods to serialize their data, e.g. `Image`, `JPEGImage`, `Thumbnail`, and `ImagePatches`. Therefore,
such representations cannot be converted and will likely crash SimRobot when trying to replay log files containing them after they were changed.

**logFileUncompressed**: Uncompressed log data is stored in a single MessageQueue. Its serialized form starts with an eight byte header containing two values. These are the size used by the log followed by the number of messages in the queue. Those values can also be set to -1 indicating that the size and number of messages is unknown. In this case the amount will be counted when reading the log file. The header is followed by several frames. A frame consists of multiple log messages enclosed by a idProcessBegin and idProcessFinished message. Every log message consists of its id, its size and its payload (cf. Fig. 3.2). This kind of chunk is created by remote logging.

**logFileCompressed**: This chunk contains a sequence of compressed MessageQueues. Each queue contains a single second worth of log data and is compressed using Google’s snappy [8] compression (cf. Fig. 3.3). Each compressed MessageQueue is prefixed by its compressed size. This kind of chunk is created by online logging.

The first two chunks are optional. Each log file must contain one of the latter two chunks, which is also the last chunk in the file.

### 3.7.5 Replaying Log Files

Log files are replayed using the simulator. Special modules, the CognitionLogDataProvider and the MotionLogDataProvider automatically provide all representations that are present in the log file. All other representations are provided by their usual providers, which are simulated. This way log file data can be used to test and evolve existing modules without access to an actual robot. However, it is currently only possible to replay a single log file for a single instance of the B-Human code. For online logging, this means that i.e. either a Cognition log or a Motion log can be replayed, but not both at the same time.

For SimRobot to be able to replay log files, they have to be placed in Config/Logs. Afterwards the SimRobot scene ReplayRobot.con can be used to load the log file. In addition to loading the log file this scene also provides several keyboard shortcuts for navigation inside the log file. However, the most convenient way to control log file playback is the log player view (cf. Sect. 10.1.4.5). If you want to replay several log files at the same time, simply create a file Config/Scenes/Includes/replay.con and add several sl statements (cf. Sect. 10.1.6.1) to it. The data of each log file will be fed into a separate instance of the B-Human code.

### 3.7.6 Annotations

To further enhance the usage of log files, we added the possibility for our modules to annotate individual frames of a log file with important information. This is, for example, information about a change of game state, the execution of a kick, or other information that may help us to debug our code. Thereby, when replaying the log file, we may consult a list of those annotations to see whether specific events actually did happen during the game. In addition, if an annotation

![Figure 3.3: A compressed log file consists of several regular log files and their compressed sizes.](image-url)

53
was recorded, we are able to directly jump to the corresponding frame of the log file to review the emergence and effects of the event without having to search through the whole log file. This feature is accessed via the \texttt{ANNOTATION-Macro}. An example is given below:

\begin{verbatim}
#include "Tools/Debugging/Annotation.h"
...
ANNOTATION("GroundContactDetector", "Lost GroundContact");
\end{verbatim}

It is advised to be careful to not send an annotation in each frame because this will clutter the log file. When using annotations inside of a behavior option the output option \texttt{Annotation} should be used to make sure annotations are not sent multiple times.

The annotations recorded can be displayed while replaying the log file in the annotations view (cf. Sect. 10.1.4.5).

### 3.7.7 Thumbnail Images

Logging raw YUV422 images as the cameras capture them seems impossible with the current hardware. One second would consume about 45000 KB. This amount of data could not be stored fast enough on the harddrive/USB flash drive and would fill it up within a few minutes. Thus images have to be compressed. This compression must happen within a very small time frame, since the robot has to do a lot of other tasks in order to play football. Despite the compression, the resulting images must still contain enough information to be useful.

For compressing the images in a way that fulfills the criteria mentioned above, two separate methods have been implemented. One of these generates grayscale images by taking the luminance channel (Y) of the camera image and averaging blocks of adjacent pixels. The other method compresses the YUV camera image by first eliminating every second row of pixels and then averaging every channel of adjacent blocks of pixels. Afterwards the size of each pixel is reduced to two bytes by eliminating one of the two Y channels in each pixel and using only six bits for the remaining luminance channel and five bits for each of the two chroma channels. In both of the two possible methods, averaging adjacent pixels is done using SSE instructions to speed up the computation.

The total compression rate as well as the usefulness of the resulting images depends on the \texttt{downscales} parameter of the \texttt{ThumbnailProvider}, which determines the size of the blocks of pixels being averaged as $2^{\text{downscales}} \times 2^{\text{downscales}}$ pixels. For a \texttt{downscales} parameter of 2, the resulting grayscale thumbnail images are 32 times smaller and the resulting colored images are 128 times smaller than the original camera image. Although a lot of detail is lost in the thumbnail images, it is still possible to see and assess the situations the robot was in (cf. Fig. 3.4).

### 3.7.8 Image Patches

While logging of thumbnail images allows analysis of the robot’s behavior during the game, their compression restricts any use for later image processing. Therefore, image patches were implemented which enable logging of parts of the image. An image patch is defined by an offset into the current camera image as well as its width and height. When logging image patches, only the parts of the image belonging to a patch are written into the log file. This makes it possible to log parts of every image that are important for certain perception modules. An example of this would be the perception of the ball: as described in Sect. 4.3, only areas around ball spots obtained from analyzing vertical color-classified scanlines are considered for further steps of the ball perception. Thus, if functionality of the ball perceptor is to be evaluated and
Figure 3.4: (a) Image from the upper camera of the NAO and (b) the corresponding colored thumbnail with a downscale factor of 2 as well as the corresponding grayscale thumbnails with downscale factors (c) 2 and (d) 3. (e) shows the image patches containing only areas of the image in which a potential ball candidate was seen.
adjusted based on log files, it is sufficient to log image patches containing areas around the ball spots. An example of this can be seen in figure 3.4e.

In order to assure that the perception components work normally with images reconstructed from logged image patches, all areas not covered by an image patch are filled with a dark green color resembling the field color.
Chapter 4

Perception

The perception modules run in the context of the process Cognition. They detect features in the image that was just taken by the camera. The modules can be separated into four categories. The modules of the perception infrastructure provide representations that deal with the perspective of the image taken, provide the image in different formats, and provide representations that limit the area interesting for further image processing steps. Based on these representations, modules detect features useful for self-localization, the ball, and obstacles. All information provided by the perception modules is relative to the robot’s position.

An overview of the perception modules and representations is shown in Fig. 4.1.

4.1 Perception Infrastructure

4.1.1 Using Both Cameras

The NAO robot is equipped with two video cameras that are mounted in the head of the robot. The first camera is installed in the middle forehead and the second one approx. 4cm below. The lower camera is tilted by 39.7° with respect to the upper camera and both cameras have a vertical opening angle of 47.64°. Because of that, the overlapping parts of the images are too small for stereo vision. It is also impossible to get images from both cameras at the exact same time, as they are not synchronized on a hardware level. This is why we analyze only one picture at a time and do not stitch the images together. To be able to analyze the pictures from both the upper and lower camera in real-time without loosing any images, the Cognition process runs at 60Hz.

Since the NAO is currently not able to provide images from both cameras at their maximum resolution, we use a smaller resolution for the lower camera. During normal play the lower camera sees only a very small portion of the field, which is directly in front of the robot’s feet. Therefore, objects in the lower image are close to the robot and rather big. We take advantage of this fact and run the lower camera with half the resolution of the upper camera, thereby saving a lot of computation time.

Both cameras deliver their images in the YUV422 format. The upper camera provides 640 \times 480 pixels while the lower camera only provides 320 \times 240 pixels. As the perception of features in the images relies either on color classes (e.g. for region building) or the luminance values of the image pixels (e.g. for computing edges in the image), the YUV422 images are converted to the “extracted and color-classified” EClmage. The EClmage consists of two images: the grayscaled image obtained from the Y channel of the camera image and a so-called “colored” image mapping.
Figure 4.1: Perception module graph. Perception modules are depicted as yellow rectangles. All representations they provide are shown as blue ellipses. The representations they require are shown as ellipses colored in either yellow if they are provided by other Cognition modules and orange if they are received from the process Motion.
each image pixel to a color class.

Cognition modules processing an image need to know from which camera it comes. For this reason, we implemented the representation CameraInfo, which contains this information as well as the resolution of the current image.

### 4.1.2 Definition of Coordinate Systems

The global coordinate system (cf. Fig. 4.2) is described by its origin lying at the center of the field, the $x$-axis pointing toward the opponent goal, the $y$-axis pointing to the left, and the $z$-axis pointing upward. Rotations are specified counter-clockwise with the $x$-axis pointing toward $0^\circ$, and the $y$-axis pointing toward $90^\circ$.

In the robot-relative system of coordinates (cf. Fig. 4.3), the axes are defined as follows: the $x$-axis points forward, the $y$-axis points to the left, and the $z$-axis points upward.

#### 4.1.2.1 Camera Matrix and Camera Calibration

The CameraMatrix is a representation containing the transformation matrix of the active camera of the NAO (cf. Sect. 4.1.1) that is provided by the CameraMatrixProvider. It is used for projecting objects onto the field as well as for the creation of the ImageCoordinateSystem (cf. Sect. 4.1.2.2). It is computed based on the TorsoMatrix that represents the orientation and position of a specific point within the robot’s torso relative to the ground (cf. Sect. 7.4). Using the RobotDimensions and the current joint angles, the transformation of the camera matrix relative to the torso matrix is computed as the RobotCameraMatrix. The latter is used to compute the BodyContour (cf. Sect. 4.1.3). In addition to the fixed parameters from the RobotDimensions, some robot-specific parameters from the CameraCalibration are integrated, which are necessary, because the camera cannot be mounted perfectly plain and the torso is not always perfectly vertical. A small variation in the camera’s orientation can lead to significant errors when projecting farther objects onto the field.

The process of manually calibrating the robot-specific correction parameters for a camera is a very time-consuming task, since the parameter space is quite large (8 resp. 11 parameters for calibrating the lower resp. both cameras). It is not always obvious, which parameters have to be adapted, if a camera is miscalibrated. In particular during competitions, the robots’ cameras require recalibration very often, e.g. after a robot returned from repair.

In order to overcome this problem, an automatic CameraCalibrator module was introduced. It collects points on the field lines fully autonomously. The points on the lines required are provided by the LineSpotProvider (see Sect. 4.2.1). They are collected for both cameras, after the head moved to predefined angles. Although this calibrator significantly reduces the time needed for a calibration, it has a drawback in terms of precision. The line detection has problems in detecting lines that are further away, in particular when the color calibration is not very good. This can lead to inaccurate values for the estimated tilts of the cameras.

Notable features of the automatic CameraCalibrator are:

- The user can mark arbitrary points on field lines if the automatic detection does not produce enough points. This is particularly useful during competitions because it is possible to calibrate the camera if parts of the field lines are covered (e.g. by robots or other team members).
- Since both cameras are used, the calibration module is able to calibrate the parameters of
the lower as well as the upper camera. Therefore, the user simply has to mark additional reference points in the image of the upper camera.

- In order to optimize the parameters, the Gauss-Newton algorithm is used\(^1\) instead of hill climbing. Since this algorithm is designed specific for non-linear least squares problems like this, the time to converge is drastically reduced to an average of 5–10 iterations. This has the additional advantage that the probability to converge is increased.

- During the calibration procedure, the robot stands on a defined spot on the field. Since the user is typically unable to place the robot exactly on that spot and a small variance of the robot pose from its desired pose results in a large systematical error, additional correction parameters for the \(\text{RobotPose} \) are introduced and optimized simultaneously.

- The error function takes the distance of a point to the next line in image coordinates instead of field coordinates into account. This is a more accurate error approximation because the parameters and the error are in angular space.

- A manual deletion of samples is possible by left-clicking into the image and on the point the sample has been taken. Likewise, a manual insertion of samples is now possible by \(CTRL + \) left-clicking into the image at the point you want the sample to be.

- Command generation for correcting the body rotation. In case you don’t want the \(\text{BodyRotationCorrection} \) stored in the \(\text{CameraCalibration} \), you can manually call the \text{JointCalibrator} and transfer the values or you can use the command \(\text{set module:AutomaticCameraCalibrator:setJointOffsets true} \) before running the optimization. After the optimization, a bunch of commands will be generated and you can enter them in order of appearance to transfer the values into the \text{JointCalibration}.

With these features, the module typically produces a parameter set that requires only little manual adjustments, if any. The calibration procedure is described in Sect. 2.8.3.

\(^1\)Actually, the Jacobian used in each iteration is approximated numerically
4.1.2.2 Image Coordinate System

Based on the camera transformation matrix, another coordinate system is provided which applies to the camera image. The ImageCoordinateSystem is provided by the module CoordinateSystem-Provider. The origin of the $y$-coordinate lies on the horizon within the image (even if it is not visible in the image). The $x$-axis points right along the horizon whereby the $y$-axis points downwards orthogonal to the horizon (cf. Fig. 4.4). For more information see also [29].

Using the stored camera transformation matrix of the previous cycle in which the same camera took an image enables the CoordinateSystemProvider to determine the rotation speed of the camera and thereby interpolate its orientation when recording each image row. As a result, the representation ImageCoordinateSystem provides a mechanism to compensate for different recording times of images and joint angles as well as for image distortion caused by the rolling shutter. For a detailed description of this method, applied to the Sony AIBO, see [23].

4.1.3 Body Contour

If the robot sees parts of its body, it might confuse white areas with field lines or other robots. However, by using forward kinematics, the robot can actually know where its body is visible in the camera image and exclude these areas from image processing. This is achieved by modeling the boundaries of body parts that are potentially visible in 3-D (cf. Fig. 4.5 left) and projecting them back to the camera image (cf. Fig. 4.5 right). The part of the projection that intersects with the camera image or above is provided in the representation BodyContour. It is used by image processing modules as lower clipping boundary. The projection relies on the representation ImageCoordinateSystem, i.e., the linear interpolation of the joint angles to match the time when the image was taken.

4.1.4 Color Classification

Identifying the color classes of pixels in the image is done by the ECImageProvider when computing the ECImage. In order to be able to clearly distinguish different colors and easily define color classes while still being able to compute the ECImage for every camera image in real time, the YHS2 color space is used.
YHS2 Color Space

The YHS2 color space is defined by applying the idea behind the HSV color space, i.e. defining the chroma components as a vector in the RGB color wheel, to the YUV color space. In YHS2, the hue component H describes the angle of the vector of the U and V components of the color in the YUV color space, while the saturation component S describes the length of that vector divided by the luminance of the corresponding pixel. The luminance component Y is just the same as it is in YUV. By dividing the saturation by the luminance, the resulting saturation value describes the actual saturation of the color more accurately, making it more useful for separating black and white from actual colors. This is because in YUV, the chroma components are somewhat dependent on the luminance (cf. Fig. 4.6).

Classification Method

Classifying a pixel’s color is done by first applying a threshold to the saturation channel. If it is below the given threshold, the pixel is considered to describe a non-color, i.e. black or white. In this case, whether the color is black or white is determined by applying another threshold to the luminance channel. However, if the saturation of the given pixel is above the saturation threshold, the pixel is of a certain color, if its hue value lies within the hue range defined for that color.
4.1. PERCEPTION INFRASTRUCTURE

This approach was used throughout the RoboCup 2016 and provided very good results. It is however possible to use an alternative classification method by setting the `simpleClassification` parameter of the `ECImageProvider` to `false`. In this case, colors are defined not only by a hue range, but by one range of values for each of H, S and Y. Additionally, white and black are not separated by a single threshold for the Y channel, but there is a minimum Y threshold for white and a maximum Y threshold for black to enable non-classified pixels with low saturation.

In order to classify the whole camera image in real time, both the color conversion to YHS2 and the color classification are done using SSE instructions.

Fig. 4.7 shows representations of an image from the upper camera in the YHS2 color space and a classification based on it for the colors white, black, green and “none” (displayed gray).

4.1.5 Segmentation and Region-Building

The `ColorScanlineRegionizer` scans along vertical and horizontal scanlines whose distances from each other are small enough to detect the field boundary, field lines, and the ball. These scanlines are segmented into regions, i.e. line segments of a similar color based on the colored part of the `ECImage`. In vertical direction, the sampling frequency is not constant. Instead, it is given by the `ScanGrid`, the steps of which depend on the vertical orientation of the camera and basically correspond to a distance of a bit less than the expected width of a horizontal field line at that position in the image. Whenever the color classes of two successive scan points on a vertical line differ, the region in between is scanned to find the actual position of the edge between the two
4.1. PERCEPTION INFRASTRUCTURE

Figure 4.8: (a) Visualization of the \textit{ScanGrid}: the blue dots are checked for building regions of colors. (b) and (c) show the high resolution vertical and the horizontal scanline regions, respectively.

neighboring regions. The best position is the one above which enough pixels are found that are classified as color classes different from the one of the previous region. Fig. 4.8 visualizes the subsampling.

In order to save computational time, the \textit{ScanGrid} starts at the horizon as given by the \textit{ImageCoordinateSystem}. Additionally, the vertical scanlines are divided into “low resolution” and “high resolution” scanlines. At first, only the low resolution scanlines are computed and used for determining the field boundary. Afterwards, the high resolution scanlines are computed for further processing, starting at the field boundary.

4.1.6 Detecting The Field Boundary

The rules state that if fields are further away from each other than 3 m, a barrier between them can be omitted. This means that robots can see goals on other fields that look exactly like the goals on their own field. In addition, these goals can even be closer than a goal on their field. Therefore, it is very important to know where the own field ends and to ignore everything outside. For this purpose, our current approach searches for vertical scanlines, starting from the bottom of the image going upwards. Simply using the first non-green pixel for the boundary is not an option, since pixels on field lines and all other objects on the field would comply with such a criterion. In addition, separating two or more fields would not be possible and noise could lead to false positives.

Our approach builds a score for each scan line while the pixels are scanned. For each green pixel, a reward is added to the score. For each non-green pixel a penalty is subtracted from it. The pixel where the score is the highest is then selected as boundary spot for the corresponding scan line. The rewards and penalties are modified depending on the distance of the pixel from the robot when projected to the field. Field lines tend to have more green pixels above them in the image. As a result, they are easily skipped, but the gaps between the fields tend to be small.
in comparison to the next field when they are seen from a greater distance. Thus, for non-green pixels with a distance greater than 3.5 meters, a higher penalty is used. This also helps with noise at the border of the own field. Figure 4.9 shows that most of the spots are placed on the actual boundary using this method.

Since the detection of the field lines (cf. Sect. 4.2.1) should only take place inside the area of the field, the \texttt{ColorScanlineRegionsVertical} are clipped by the \texttt{FieldBoundary} resulting in a representation called \texttt{ColorScanlineRegionsVerticalClipped} that is used for further processing. Since clipping away the area outside the field also means to remove the regions that are potentially cluttered the most, ignoring them in further processing also benefits the computation time.

Since other robots in the image also produce false positives below the actual boundary, the general idea is to calculate the upper convex hull from all spots. However, such a convex hull is prone to outliers in upward direction. Therefore, an approach similar to the one described by Thomas Reinhardt [22, Sect. 3.5.4] is used. Several hulls are calculated successively from the spots by removing the point of the hull with the highest vertical distances to its direct neighbors. These hulls are then evaluated by comparing them to the spots. The best hull is the one with the most spots in near proximity. This one is selected as boundary. An example can be seen in Fig. 4.9.

The lower camera does not see the actual field boundary most of the time. So calculating the boundary in every image would lead to oscillating representations. Thus, the field boundary is calculated on the upper image. Nevertheless, a convex hull is calculated on the lower image too, since the boundary can start there. The resulting hull is stored for one frame in field coordinates so that the movement of the NAO can be compensated in the upper image. It is then used to provide starting points for the calculations. The resulting hull is used as field boundary over two images.
4.2 Localization Features

To provide input for the self-localization (cf. 5.1), there exist multiple approaches to extract information from the visual sensors. Every (visual) knowledge relevant for position estimation that our robot is aware of is based on the field line detection. Although the penalty mark and the center circle are basically lines as well, they are – at least partly – detected separately to ensure more robustness for each single detection. Together, the representations of all three basic detections make up the first layer in the process chain of the visual perception. Each layer combines previously known information to provide more specific features. Information of every layer can be used by the self-localization. The main layers after the basic feature detection are the intersections perception and the following field feature perception.

4.2.1 Detecting Lines

The perception of field lines by the LinePerceptor relies mostly on the scanline regions. In order to find horizontal lines in the image, adjacent white vertical regions that are not within a perceived obstacle (cf. Sect. 4.5) are combined to line segments. Correspondingly, vertical line segments are constructed from white horizontal regions. These line segments and the center points of the regions they are made up of, called line spots, are then projected onto the field. Using linear regression of the line spots, the line segments are then merged together and extended to larger line percepts. During this step, line segments are only merged together if at least a given ratio of the resulting line consists of white pixels in the image. Fig. 4.10 shows the process of finding lines in the camera image.

4.2.2 Detecting the Center Circle

Besides providing the LinesPercept containing perceived field lines, the LinePerceptor also detects the center circle in an image if it is present. In order to do so, when combining line spots to line segments, their field coordinates are also fitted to circle candidates. After the LinesPercept was computed, spots on the circle candidates are then projected back into the image and adjusted so they lie in the middle of white regions in the image. These adjusted spots are then again projected onto the field and it is once again tried to fit a circle through them, excluding outliers.

If searching the center circle using this approach did not yield any results, another method of finding the center circle is applied. For this method, for all previously detected lines whose line spots describe an arc, the center points of the arcs – described by three adjacent line spots – are accumulated into clusters. If one of these clusters contains a sufficient number of center points, the average point of these center points is considered to be the center of the center circle.

If a potential center circle was found by any of these two methods, it is accepted as a valid center circle only if – after projecting spots on the circle back into the image – at least a certain ratio of the corresponding pixels is white (cf. Fig. 4.10c).

4.2.3 Line Coincidence Detection

To find points where two lines coincide, we are calculating the intersections of the previously perceived lines (cf. 4.2.1). This approach is probably much faster and more accurate\(^2\) than any separate (visual) perception such as e.g. corner search, but it means that detecting a line

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\(^2\) Assuming a pretty accurate line detection.
intersection without a line (percept) is impossible. In particular this disadvantage applies to line corners of a penalty area that extend into the image.

The IntersectionProvider fills the IntersectionPercept based on the LinesPercept. Each known line will be compared to all other known lines, excluding those that are known to be part of the center circle. Because of the way the field is set up (cf. the current rules [4]), an intersection is only possible, if the inclination of two compared lines in field coordinates is roughly $\frac{\pi}{2}$. If that is the case, the point of intersection of the two lines is calculated. In order to be valid, this point must lie in between both detected line segments. Next, we also want to specify the type of each intersection. The types of intersections are congruent with their names: L, T, and X. This means that intersections lying clearly in-between both lines are of type X while those that are clearly inside one line but roughly at one end of the other are considered to be of type T. If the point of intersection lies roughly at the ends of both lines, it is assumed to belong to an L-type intersection.

In case that parts of both lines are seen, but the point of intersection does not lie on either line – e.g. because an obstacle is standing in front of the intersection point – the lines will be
virtually stretched to some extend so that the point of intersection lies roughly at their ends. The possible distance by which a line can be extended depends on the length of its recognized part.

The *IntersectionPercept* stores the type of each intersection as well as its position, its base lines, and its rotation.

### 4.2.4 Preprocessed Lines and Intersections

Some features of the *LinesPercept* and *IntersectionPercept*, such as the spots from which the lines were fitted and image coordinates, are not needed for further processing. On the other hand, now that the positions of all lines and intersections are known, additional information can be gained by making assumptions based on the known layout of the field lines. Therefore, as a next step, the *FieldLinesProvider* provides adapted representations of the *LinesPercept* and *IntersectionPercept* for further use.

The representation *FieldLines* contains the processed line percepts (cf. Fig. 4.10d). For this, every line that is assumed to be seen on the center circle or inside the goal frame or net (cf. 4.2.6) was sorted out. Lines that extend into the goal frame are clipped to its boundaries. In addition to that, every line that is as long as a ball would appear at the corresponding position in the image is removed[^3]. Apart from removing invalid percepts, lines in *FieldLines* also have additional properties. A line is marked as *mid*, if it goes through the center circle, which means that it is probably the center line. The *long* attribute is given to any line that is longer than a specific on-field distance. A long line means the detected line probably belongs to the outer lines, center line or penalty front lines. For this reason, it is advisable to check on a distance that is longer than the penalty area sidelines or a part line inside the center circle, but as short as possible to receive the additional information as often as possible.

The representation *FieldLineIntersections* contains the processed line intersection percepts. It consists only of intersections that do not apply to any previously removed lines. If one of the lines belonging to an intersection was cut, the intersection type is corrected accordingly. Additionally, an intersection can now have a type of *mid* or *big*. If both lines of the intersection are of type *mid*, the intersection is also considered to be *mid*. Analogous to that, an intersection consisting of two *long* lines has the attribute *big*. After all this processing, L-type intersections are filtered out, if the angle between their two lines is not about ninety degrees.

### 4.2.5 Penalty Mark Perception

The current module that provides the penalty mark, the *PenaltyMarkPeceptor3000*, was developed with an emphasis on performance instead on detection rate, while not allowing too many false positives. The main idea is to find a white spot that is surrounded by (only) green and has no black inside.

Therefore, the low resolution vertical scan lines in *ColorScanlineRegionsVerticalClipped* are searched for white spots that have approximately the same height as a penalty mark would have at this position in the image. These spots are tested against a number of criteria, with the first spot passing all tests successfully being considered to be the actual penalty mark. The tests are checking on duplicates, position inside the ball, width, black inside and green around. To ensure that every area in the image in which spots were found will be checked only once, no spot will be further evaluated if the distance to any previously checked spot is less than the

[^3]: This is a competition quick hack to avoid further calculations on false-positive lines that are fitted inside the ball.
calculated in-image-width of the currently checked penalty mark. The check on area intersection with the ball prediction is theoretically unnecessary but during competition it allowed us to easily compensate for non-perfect color conditions thanks to a good ball detection. The width will be checked by comparing the calculated in-image-width of the penalty mark with a simply measured width by scanning to the left and right from the spot. In addition to the check, this will be used to relocate the in-image x-position of the spot. The scan lines for the finishing black and green checks are generated by the estimated in-image penalty mark size. The green scan should ensure that this white plot does not belong to anything else. This check is supposed to be successful only for real penalty marks and balls, with the last being sorted out by the black scan.

### 4.2.6 Field Features

The self-localization has always used field lines, their crossings, and the center circle as measurements. Since 2015, these features are complemented by the perception of the penalty marks. All of these field elements are distributed over the whole field and can be detected very reliably, providing a constant input of measurements in most situations.

Built upon this, the perception of a new category of measurements, the so-called *Field Features*, has been implemented. They are created by combining multiple basic field elements in a way that a robot pose (in global field coordinates) can be derived directly. The handling of the field symmetry, which leads to actually two poses, is described Sect. 5.1.2.

Overall, this approach provides a much higher number of reliable pose estimates than the previous goal-based approach, as the field lines on which it is based can be seen from many perspectives and have a more robust context (straight white lines surrounded by green carpet) than the noisy unknown background of goal posts.

Every *FieldFeature* is a *Pose2f*, which is the pose of this in relative field coordinates. In addition, it has the following attributes and methods:

- **isValid** is an attribute that describes, if the pose is valid or rather if the feature was detected in this frame.
- **markedPoints**, **markedLines**, and **markedIntersections** are attributes that associate several percepts with explicitly defined global field elements such as the center line. Although these lists are filled by the individual feature providers, the functionality is not described in detail in this report, as these elements are not in use right now.
- **isLikeEnoughACorrectPerception(...)** calculates if more lines are unmarked inside a given space around this feature than allowed.
- **getGlobalFeaturePosition()** calculates the two global robot poses that the robot can have according to this feature (if it is valid).
- **clear()** clears the lists **markedPoints**, **markedLines**, and **markedIntersections**. The intended use is inside a provider at the start of an update method.
- **draw()** draws the result of **getGlobalFeaturePosition()** into a world state drawing context.

Normally, all feature detections are using the preprocessed field elements (cf. 4.2.4). We are currently computing the following field features:
Penalty Area The PenaltyAreaPerceptor, which provides PenaltyArea, tries to find a penalty area by the help of the penalty mark and a line or two intersections. If a penalty mark is known, it searches a line that matches the corresponding penalty area front line. The penalty area can also be found by combining two small (not big) T-intersections or one small T-intersection and one small L-intersection.

Outer Corner The OuterCorner refers to the areas where ground lines meet side lines, which means that the outer corner is on the right or left side. To find such a corner, the OuterCornerPerceptor combines a big L-intersection with an element of the penalty area.

Mid Circle The MidCircleProvider is combining a detected center circle with a detected mid line to provide the MidCircle.

Mid Corner A MidCorner, which is provided by the MidCornerPerceptor, is simply derived from a big T-intersection.

Goal Frame Besides the obvious reason to have another field feature to get more information to put into the self-locator, the GoalFrame is also used to sort out false-positive lines that were detected inside the goal net. Thus, the GoalFrameDetector has to calculate directly based on the line percepts. In general, the module searches for points around the goal and validates the rotation with a special use of the FieldBoundary.

4.3 Detecting the Black and White Ball

The introduction of the black and white ball is the major new challenge in the Standard Platform League in 2016. Until the RoboCup 2015, the ball was orange and rather easy to detect. In particular, it was the only orange object on the field. The new ball is mainly white with a regular pattern of black patches, just as a miniature version of a regular soccer ball. The main problem is that the field lines, the goals, and the NAO robots are also white. The latter even have several round plastic parts and they also contain grey parts. Since the ball is often in the vicinity of the NAOs during a game, it is quite challenging to avoid a large number of false positives.

We use a multi-step approach for the detection of the ball. First, the vertical scan lines our vision system is mainly based on are searched for ball candidates. Then, a contour detector fits ball contours around the candidates’ locations. Afterwards, fitted ball candidates are filtered using some general heuristics. Finally, the surface pattern inside each remaining candidate is checked.

4.3.1 Searching for Ball Candidates

Our vision system scans the image vertically using scan lines of different density based on the size that objects, in particular the ball, would have at a certain position inside the image. To determine ball candidates, the BallSpotProvider searches these scan lines for sufficiently large gaps in the green that also have a sufficiently large horizontal extension and contain enough white (cf. Fig. 4.11a). Only the first and second resolutions of the scanlines are used. From a specific distance, the ball is that small inside the image that the calculation can not provide good results. For this reason, ball spots more far away than this distance will be generated on white blobs that paled of green. In addition, all initial spots are ignored if they are nearby a
4.3. DETECTING THE BLACK AND WHITE BALL

(a) Vertical scan lines and a detected ball candidate (the cross). Parts of the robot’s body are ignored (bottom left).

(b) Contrast-normalized Sobel image. The colors indicate the directions of the gradients.

(c) Visualization of the search space for the ball contour. The actual search is only performed around the ball candidate, but in single pixel steps in both dimensions.

(d) The contour with the highest response (green) and the sample grid to check the ball pattern (pixels classified as black are shown in red, white pixels in blue).

Figure 4.11: The main steps of the ball detection

previous one⁴ or are clearly inside a detected robot. In this context, “clearly inside a robot” means all parts except for feet and arms, as these elements cannot be detected accurately.

4.3.2 Fitting Ball Contours

As the position of a ball candidate is not necessarily in the center of an actual ball, the area around such a position is searched for the contour of the ball as it would appear in this part of the image given the intrinsic parameters of the camera and its pose relative to the field plane. The approach is very similar to the detection of objects in 3-D space using a stereo camera system as described by Müller et al. [19], but we only use a single image instead. For each ball candidate, the CNSBallRegionsProvider computes a rectangle that surrounds the area a ball could occupy in that part of the image. The CNSImageProvider then first joins neighboring areas to

⁴Despite this step is unnecessary for the following used ball perceptor, it saves a bit calculation time inside the ball spot provider itself, but even more inside different ball percept provider approaches.
bigger rectangles and computes a contrast-normalized Sobel image for each area (cf. Fig. 4.11b). Technically, only a single Sobel image is filled, in which only the areas surrounding the candidate positions are updated. The CNSBallPerceptor then performs the actual ball detection. It searches this contrast image in each of the areas provided by the CNSBallRegionsProvider for the best match with the expected ball contour (cf. Fig. 4.11c). The best match is then refined by adapting its hypothetical 3-D coordinates (cf. Fig. 4.11d).

4.3.3 Filtering Ball Candidates

The fitting process results in a measure, the response, for how well the image matches with the contour expected at the candidate’s location. If this value is below a threshold, the ball candidate is dropped. The threshold is dynamically determined from the amount of green that surrounds the ball candidate. On the one hand, the less green is around the candidate, the higher the response has to be to reduce the amount of false positives inside robots. However, if a ball candidate is completely surrounded by green pixels and the response was high enough to exclude the possibility of being a penalty mark, the ball candidate is accepted right away, skipping the final step described below that might be failing if the ball is rolling quickly. All candidates that fit well enough are processed in descending order of their response. As a result, the candidate with the highest response that also passes all other checks will be accepted. These other checks include that the ball radius found must be similar to the radius that would be expected at that position inside the image.

4.3.4 Checking the Surface Pattern

For checking the black and white surface pattern, a fixed set of 3-D points on the surface of the ball candidate are projected into the image (cf. Fig. 4.11d). For each of these pixels, the brightness of the image at its location is determined. Since the ball usually shows a strong gradient in the image from its bright top to a much darker bottom half, the pixels are artificially brightened depending on their position inside the ball. Then, Otsu’s method [21] is used to determine the optimal threshold between the black and the white parts of the ball for the pixels sampled. If the average brightnesses of both classes are sufficiently different, all pixels sampled are classified as being either black or white. Then, this pattern is looked up in a pre-computed table to determine whether it is a valid combination for the official ball. The table was computed from a 2-D texture of the ball surface considering all possible rotations of the ball around all three axes and some variations close to the transitions between the black and the white parts of the ball.

The approach allows our robots to detect the ball in distances of up to five meters with only a few false positive detections. It works better under good lighting conditions. In a rather dark environment, as on Field A in the indoor competition of RoboCup 2016, balls with lower responses had to be accepted in order to detect the ball at all. This resulted in more false positive detections, in particular in the feet of other robots, because they are also largely surrounded by green.

4.4 Ball Perception Statistics

Due to the change of the official ball of the Standard Platform League, the team members of B-Human developed a number of different solutions for this task. To determine which perception approach is the best, it was considered useful to raise some statistics like the perception range...
4.5. DETECTING OTHER ROBOTS AND OBSTACLES

The module BallPerceptorEvaluator provides those statistics in the representation BallPerceptorEvaluation. This representation contains the following attributes:

- **seenPercentage**: The rate of percepts over all images in percent. Please note that both cameras are taken into account for this attribute. Thus, an overall rate of 0.5 is perfectly fine if the percept is computed by just one camera.

- **guessedFalsePositives**: The guessed rate of false positives over all percepts. This value is calculated by thresholds for percept jumps in BallPerceptorEvaluator.

- **averageDifferenceUpper**: The average of differences between percept positions on the field in the upper camera.

- **averageDifferenceLower**: The average of differences between percept positions on the field in the lower camera.

- **maximumPerceptDistance**: The longest distance between robot and percept.

- **minimumPerceptDistance**: The shortest distance between robot and percept.

- **averagePerceptDistance**: The average distance between robot and percept.

The module BallPerceptorEvaluator requires the BallPercept to get the relevant data for the statistics, and the CameraInfo to check which camera received the current percept. The parameters are the following.

- **jumpAngleThreshold**: A threshold that describes the maximum angular difference between the last two percepts. Values above the threshold can be an indication for a false positive.

- **jumpOnFieldThreshold**: The threshold that describes the maximum positional difference between the last two percepts. Values above the threshold can be an indication for a false positive.

The BallPerceptorEvaluator computes the statistics by using the status and the positionOnField (relative to the robot) of the BallPercept. The module uses a lot of variables and buffers to avoid complications inside the representation BallPerceptorEvaluation. The special attribute guessedFalsePositives should not be taken too seriously. A percept is seen as a false positive if the angular and positional difference to the last percept are above the thresholds that are declared as parameters. The averaged values are calculated by using a ring buffer with a maximum of 100 elements. To avoid numerical problems, the counting variables are checked at every update cycle.

4.5 Detecting Other Robots and Obstacles

The PlayersPerceptor recognizes standing and fallen robots up to seven meters away in less than two milliseconds of computation time.

Initially, the approach searches for robot parts within the field boundary (cf. Fig. 4.12a). Starting at the boundary, it scans down in vertical lines, looking only at a subset of the pixels. The distance between these pixels is constantly growing. The space and its growing rate are calculated from the distance of the corresponding points on the field. If there are a couple of non-green
Figure 4.12: (a) Searching for a robot below the boundary of the field (depicted in orange) at a subset of the pixels (depicted as yellow dots). (b) Jersey colors are recognized on robots.

pixels scanned in a vertical line, then the lowest of these pixels gets marked as a potential robot base point. These are possible spots at the feet and hands of a robot, and they can be merged to a bounding box. After calculating the surrounding box, the method searches for a jersey within it and tries to determine the jersey color (cf. Fig. 4.12b).

For jersey association, the module uses the team color information from the GameController. If one of the teams uses black jerseys (in most cases our own team), we scan up to four times with the first scan checking the shins of the NAO, to determine whether it is a referee or not. Otherwise, we scan up to three times. All other scans are performed below the shoulders, at the height of the chest, and at the hip, when no jersey was detected in each previous scan.

During each scan, the color values of scanned pixels are analyzed by using the following conditions:

1. Is the intensity high and the saturation low enough to be white?
   Then a counter for white will be increased.

2. Do the hue and intensity values resemble that of a prototypical green color?
   Then the pixel will be ignored.

3. Is one of the jerseys black?
   
   - Yes
     - Is the intensity too high for black but low enough for a dark color and the saturation not too high for gray?
       Then it could be a black jersey.
   
   - No
     - Is the intensity too high for black?
       Then it will be checked which team color is within the ranges or next to the color of the pixel, if both are within the ranges.

Depending on which color dominates, the obstacle will be marked as an opponent, teammate, or obstacle. The HSI color values for all team colors have to be predefined prototypically.
Chapter 5

Modeling

To compute an estimate of the world state – including the robot’s position, the ball’s position and velocity, and the presence of obstacles – given the noisy and incomplete data provided by the perception layer, a set of modeling modules is necessary. Since these modules have to update their output for each new image processed, they also run in the process Cognition. Together with their dependencies, they are depicted in Fig. 6.1.

5.1 Self-Localization

A robust and precise self-localization has always been an important requirement for successfully participating in the Standard Platform League. B-Human has always based its self-localization solutions on probabilistic approaches [31] as this paradigm has been proven to provide robust and precise results in a variety of robot state estimation tasks. However, due to the high amount of noise and the problems arising from the field’s symmetry, the main state estimation module is complemented by additional modules that generate new robot pose alternatives and check, if a robot’s current orientation is point-symmetrically flipped.

5.1.1 Probabilistic State Estimation

As in previous years, the pose state estimation is handled by multiple hypotheses that are each modeled as an Unscented Kalman Filter [13]. The hypotheses management and hypotheses resetting (cf. Sect. 5.1.2) is realized as a particle filter [5]. Both approaches are straightforward textbook implementations [31], except for some adaptions to handle certain RoboCup-specific game states, such as the positioning after returning from a penalty. In addition, we only use a very low number of particles (currently 12), as multimodalities do not occur often in RoboCup games.

The current self-localization implementation is the SelfLocator that provides the RobotPose. Furthermore, for debugging purposes, the module also provides the loggable representation SelfLocalizationHypotheses which contains the states of all currently internally maintained pose hypotheses.

Overall, this combination of probabilistic methods enables a robust, precise and efficient self-localization on the RoboCup field. However, eventually, the result always depends on the quality and quantity of the incoming perceptions.
5.1.1.1 Perceptions Used for Self-Localization

In the past, B-Human used goals as a dominant feature for self-localization. When the field was smaller and the goal posts were painted yellow, they were easy to perceive from most positions and provided precise and valuable measurements for the pose estimation process. In particular the sensor resetting part (cf. Sect. 5.1.2), i.e. the creation of alternative pose estimates in case of a delocalization, was almost completely based on the goal posts perceived. In 2015, we still relied on this approach, using a detector for the white goals [27]. However, as it turned out that this detector required too much computation time and did not work reliably in some environments (requiring lots of calibration efforts), we decided to perform self-localization without goals but
by using the new complex field features (cf. Sect. 4.2.6) instead.

In addition, the self-localization still uses basic features such as field lines, their crossings, and the center circle as measurements. Since 2015, these features are complemented by the perception of the penalty marks (cf. Sect. 4.2.5). All these field elements are distributed over the whole field and can be detected very reliably, providing a constant input of measurements in most situations.

Field features are also used as measurements but not as a perception of relative landmarks. Instead, an artificial measurement of a global pose is generated, reducing the translational error in both dimensions as well as the rotational error at once. Furthermore, in contrast to the basic field elements that are not unique, no data association (cf. Sect. 5.1.1.2) is required. The handling of the field symmetry, which leads to actually two poses, is similar to the one that is done for the sensor resetting as described in Sect. 5.1.2.

5.1.1.2 Resampling based on Particle Validity

One important part of any Monte Carlo localization approach is the resampling (cf. [5]), i.e. to determine which particles become copied how often to the new particle set representing the current state’s probability distribution. For this purpose, each particle has a weighting that describes how “good” it is; the higher the weighting, the higher the likelihood that the particle will be copied to the new set. There is no general-purpose approach to compute these weights, each application requires its own way to determine the quality of a particle instead.

In our implementation, each particle’s validity is directly used as its weighting (combined with a base weighting, as described below). The validity is our measure for a particle’s compatibility to the recent perceptions and is computed during the process of data association. To use a perception in the UKF’s measurement step, it needs to be associated to a field element in global field coordinates first. For instance, given a perceived line and a particle’s current pose estimate, the field line that is most plausible regarding length, angle, and distance is determined. In a second step, it is checked whether the difference between model and perception is small enough to consider the perception for the measurement step. For all kinds of perceptions, different thresholds exist, depending on the likelihood of false positives and the assumed precision of the perception modules. After the association process has been carried out for all perceptions, the current validity $v_c$ can be computed by setting the number of successfully associated perceptions in relation to the total number of perception in this frame. For this computation, we consider different weights for different kinds of perceptions, i.e. a field feature – which is a very reliable perception – that cannot be associated has a higher impact on the validity than a field line that has not been matched.

To avoid strong oscillations of the validity, which might cause instabilities within the sample set and hence a robot pose that reacts too quickly to false measurements, a particle’s validity $v_p$ is filtered over $n$ frames (currently, $n$ is configured to 60):

$$v_p = \frac{v_{p,\text{old}} \times (n - 1) + v_c}{n} \quad (5.1)$$

Furthermore, the addition of a base weighting $w_b$ in the computation of a particle’s weighting $w_p$ contributes to a more stable resampling process:

$$w_p = w_b + (1 - w_b) \times v_p \quad (5.2)$$
5.1.2 Sensor Resetting based on Field Features

When using a particle filter on a computationally limited platform, only few particles, which cover the state space very sparsely, can be used. Therefore, to recover from a delocalization, it is a common approach to perform sensor resetting, i.e. to insert new particles based on recent measurements [18]. The new field features provide exactly this information – by combining multiple basic field elements in a way that a robot pose (in global field coordinates) can be derived directly – and thus are used by us for creating new particles. These particles are provided as AlternativeRobotPoseHypothesis by the AlternativeRobotPoseProvider.

Overall, this approach provides a much higher number of reliable pose estimates than the previous goal-based approach, as the field lines on which it is based can be seen from many perspectives and have a more robust context (straight white lines surrounded by green carpet) than the noisy unknown background of goal posts.

As false positives can be among the field features, e.g. caused by robot parts overlapping parts of lines and thereby inducing a wrong constellation of elements, an additional filtering step is necessary. All robot poses that can be derived from recently observed field features are clustered and only the largest cluster, which also needs to contain a minimum number of elements, is considered as a candidate for a new sample. This candidate is only inserted into the sample set in case it significantly differs from the current robot pose estimation.

To resolve the field’s symmetry when handling the field features, we use the constraints given by the rules (e.g. all robots are in their own half when the game state switches to Playing or when they return from a penalty) as well as the assumption that the alternative that is more compatible to the previous robot pose is more likely than the other one. An example is depicted in Fig. 5.2. This assumption can be made, as no teleportation happens in real games. Instead, most localization errors result from situations in which robots lose track of their position and accumulate translational and rotational errors.

These changes in localization – along with a growing number of robots that have a z-axis gyroscope – enabled us to reduce the number of leaving the field penalties from 15 in seven games during RoboCup 2015 to 0 in five games during the RoboCup European Open and 2 in eleven games during RoboCup 2016. In comparison, the average number of leaving the field calls at RoboCup 2016 was 35.45 times per team (5.8 per team per game), which either meant that the robots were delocalized or they were chasing a false ball they detected outside the field.

5.1.3 Handling the Field’s Symmetry

As aforementioned, the self-localization already includes some mechanisms that keep a robot playing in the right direction. However, it still happens that sometimes a single robot loses track of its playing direction. To detect such situations, a separate module – the ExpSideConfidenceProvider – compares a robot’s ball observations with those of the teammates. If the current constellation indicates that the robot has probably been flipped, the representation SideConfidence is set accordingly and the subsequently executed SelfLocator can mirror all particles.

In the past, the comparison of the ball observations has been made between the own observation and special team ball, similar to the one described in Sect. 5.2.3. As such a team ball always describes an interpolated position and might be strongly influenced by single robots, we have replaced this implementation by a list of mutual agreements. Whenever a teammate communicates a new ball, it is checked, whether both robots can agree about the ball position or if they disagree, i.e. agree about the flipped alternative. If none of both alternatives appears to be

\[\text{Actually, both were the result of human errors, as we confirmed from analyzing video footage and log files.}\]
Whenever a new alternative pose is inserted into the sample set, it has to be decided, if it can be used directly or if it needs to be flipped. This decision depends on the current robot pose. This figure visualizes the possible decisions for one example robot pose (located on the right side of the own half). The stars denote example alternative poses. A red line means: 

If a new pose is computed here, it has to be flipped before its insertion into the sample set.

Consequently, black lines mean that the pose can be used directly. As one can see, the current formula prefers the direction towards the opponent goal over the current position.

likely, the status is considered to be unknown. The ExpSideConfidenceProvider keeps an internal list of all teammates and the result of the last comparisons.

To agree about a ball position, both observations need to have occurred within a certain amount of time (currently one second) and at a similar place (currently up to one meter distance). As not all robots can see the ball frequently, each agreement / disagreement will be remembered for some time (currently eight seconds). However, if the (own or teammate) ball is located at certain positions, it will not be used to compute any agreement / disagreement: as the field is point-symmetric, the area around the field’s center is useless for any computations; as there might be false positives on penalty marks or on lines that are close to robots or close to other lines, all balls close to any of these are discarded. Furthermore, as the robots have to agree within a certain time slice, a fast rolling ball could be perceived at very different positions and has to be ignored, too.

Finally, to make a decision whether a robot is flipped or not, the list of mutual agreements is checked:

- The localization is considered as being correct, if the robot agrees with more teammates
about the ball position than it disagrees with.

- The robot is considered as flipped, if it disagrees with multiple teammates but does not agree with anyone.

## 5.2 Ball Tracking

To keep track of the current ball state, all robots maintain two ball models: a local one that estimates the ball position and velocity based on the own observations and a global one that fuses the own observations with observations communicated by the teammates. Furthermore, to predict upcoming positions of rolling balls, a model for the current carpet’s friction can be learned.

### 5.2.1 Local Ball Model

Given the ball perceptions described in Sect. 4.3, the ExpBallLocator uses Kalman filters to derive the current ball position and velocity, represented as the BallModel. Since ball motion on a RoboCup soccer field has its own peculiarities, our implementation extends the usual Kalman filter approach in several ways described below.

First of all, the problem of multimodal probability distributions, which is naturally handled by the particle filter, deserves some attention when using a Kalman filter. Instead of only one, we use twelve multivariate Gaussian probability distributions to represent the belief concerning the ball. Each of these distributions is used independently for the prediction step and the correction step of the filter. Effectively, there are twelve Kalman filters running in every frame. Only one of these distributions is used to generate the actual ball model. That distribution is chosen depending on how well the current measurement, i.e. the position the ball is currently seen at, fits and how small the variance of that distribution is. That way we get a very accurate estimation of the ball motion while being able to quickly react on displacements of the ball, for example when the ball is moved by the referee after being kicked off the field.

To further improve the accuracy of the estimation, the twelve distributions are equally divided into two sets, one for rolling balls and one for balls that do not move. Both sets are maintained at the same time and get the same measurements for the correction steps. In each frame, the

![Figure 5.3: Ball model and measurement covariances for (a) short and (b) medium distances. The orange circles show the ball models computed from the probability distribution. The larger ellipses show the assumed covariances of the measurements.](image-url)
worst distribution of each set gets reset to effectively throw one filter away and replace it with a newly initialized one.

The robot influences the motion of the ball either by kicking or just standing in the way of a rolling ball. To incorporate these influences into the ball model, the mean value of the best probability distribution from the last frame gets clipped against the robot’s feet. In such a case, the probability distribution is reset, so that the position and velocity of the ball get overwritten with new values depending on the motion of the foot the ball is clipped against. Since the vector of position and velocity is the mean value of a probability distribution, a new covariance matrix is calculated as well.

Speaking of covariance matrices, the covariance matrix determining the process noise for the prediction step is fixed over the whole process. Contrary to that, the covariance for the correction step is derived from the actual measurement; it depends on the distance between robot and ball (cf. Fig. 5.3).

The major parts of this module remained unchanged for many years. However, the introduction of the new black and white ball required the addition of a few more checks. In previous years, the number of false positive ball perceptions has been zero in most games. Hence, the ball tracking was implemented as being as reactive as possible, i.e. every perception was considered. Although the new ball perception is quite robust in general (cf. Sect. 4.3), several false positives per game cannot be avoided due to the similarity between the ball’s shape and surface and some robot parts. Therefore, there must be multiple ball perceptions within a certain area and within a maximum time frame before a perception is considered for the tracking process. This slightly reduces the module’s reactivity but is still fast enough to allow the execution of ball blocking moves in a timely manner. Furthermore, a common problem is the detection of balls inside robots that are located at the image’s border and thus are not perceived by our software. A part of these perceptions, i.e. those resulting from our teammates, is excluded by checking against the communicated teammate positions.

### 5.2.2 Friction and Prediction

For a precise Kalman filter prediction step as well as for an estimation of a rolling ball’s end position, which is required by some behaviors, a model of the friction between ball and ground is essential. For this model, we assume a linear model for ball deceleration

\[
s = v \times t + \frac{1}{2} \times a \times t^2
\]

with \(s\) being the distance rolled by the ball, \(t\) the time, and \(a\) the friction coefficient. The coefficient can be configured for each individual field and has to be set in the file `fieldDimensions.cfg` (cf. Sect. 2.9).

The coefficient can be guessed by rolling a ball and comparing its end position with the predicted end position. To make this process more convenient, the code contains the module `FrictionLearner`, which is deactivated during games, in order to automatically determine the floor-dependent friction parameter. When the module is active, a robot with remote connection can be placed on the field and the ball has to be rolled through its field of view with medium speed. Preferably, the start and end position of the ball should both be outside the robot’s field of view and the field should be even. The module buffers all ball perceptions and after the ball is outside the field of view or has come to a stop, a least squares optimization determines the coefficient \(a\) that provides the best explanation for the ball’s movement. The result is printed to the SimRobot console window afterwards. As this process is subject to noise and outliers
happen quite often, it should be carried out multiple times and the results have to be checked and compared carefully by the user.

All computations for ball state prediction that involve friction have been encapsulated in the class BallPhysics to be used by different modules.

Some modules, such as the ball perception (cf. Sect. 4.3), might require the current ball position estimate before the BallModel has been updated in current execution frame. Hence, the BallPredictor provides a BallPrediction by using the previous frame’s ball estimate and applying odometry and motion updates.

5.2.3 Team Ball Model

The team ball model is calculated locally by each robot, but takes the ball models of all teammates into account. This means that the robot first collects the last valid ball model of each teammate, which is in general the last received one, except for the case that the teammate is not able to play, for instance because it is penalized or fallen down. In this case, the last valid ball model is used. The only situation in which a teammate’s ball model is not used at all is if the ball was seen outside the field, which is considered as a false perception. After the collection of the ball models, they are combined in a weighted sum calculation to get the team ball model. There are four factors that are considered in the calculation of the weighted sum:

- The approximated validity of the self-localization: the higher the validity, the higher the weight.
- The time since the ball was last seen: the higher the time, the less the weight.
- The time since the ball should have been seen, i.e. the time since the ball was not seen although it should have appeared in the robot’s camera image: the higher the time, the less the weight.
- The approximated deviation of the ball based on the bearing: the higher the deviation, the less the weight.

Based on these factors, a common ball model, containing an approximated position and velocity, is calculated and provided as the representation TeamBallModel.

Among other things, the team ball model is currently used to make individual robots hesitate to start searching for the ball if they currently do not see it but their teammates agree about its position.

5.3 Obstacle Modeling

Similar to the previously described ball tracking, all robots maintain two different models for the obstacles in their environment, one based solely on own observations and own incorporating information sent by teammates.

5.3.1 Local Obstacle Model

To compute the ObstacleModel, the ObstacleModelProvider creates an Extended Kalman Filter for each obstacle. All obstacles are held in a list. The position of an obstacle is relative to the
position of the robot on the field plane. Sources for measurements are visually recognized goal posts, robots and sensed contacts with arms and feet. Arm and feet contacts are interpreted as an obstacle somewhere near the shoulder or foot. In the prediction step of the Extended Kalman Filter, the odometry offset since the last update is applied to all obstacle positions. The update step tries to find the best match for a given measurement, by using Euclidean distance and updates that Kalman sample. If there is no suitable sample to match, a new sample is created. Due to noise in images and motion, not every measurement might create an obstacle if no best match was found. To prevent false positive obstacles in the model, there is a minimum number of measurements in a fixed duration required to generate an obstacle. If an obstacle was not seen for some time, it is removed. This might also happen if an obstacle cannot be perceived although its estimated position is in the current field of view.

5.3.2 Global Obstacle Model

As in some situations, a local model as base for team cooperation is not sufficient, an additional global obstacle model – the TeamPlayersModel – is computed by the TeamPlayersLocator, which makes use of communicated percepts.

5.3.2.1 Positions of Teammates

For the coordination of the own team, for instance for the role selection or for the execution of certain tactics, it is important to know the current positions of all teammates. Computing these positions does not require special calculations such as filtering or clustering, because each robot sends its already filtered position to the teammates via team communication. This means that each robot is able to get an accurate assumption of all positions of its teammates just by listening to the team communication. Besides to the coordination of the team, the positions of the teammates are of particular importance for distinguishing whether perceived obstacles belong to own or opponent players.

5.3.2.2 Positions of Opponent Players

In contrast to the positions of the teammates, it is more difficult to estimate the positions of the opponent players. The opponent robots need to be recognized by vision to compute their positions. Compared to only using the local models, which are solely based on local measurements and may differ for different teammates, it is more difficult to get a consistent model among the complete team based on all local models. For a “global” model of opponent robot positions, which is intended here, it is necessary to merge the models of all teammates to get accurate positions. The merging consists of two steps, namely clustering of the positions of all local models and reducing each cluster to a single position.

For the clustering of the individual positions an $\varepsilon$-neighborhood is used, which means that all positions that have a smaller distance to each other than a given value are put into a cluster. It is important for the clustering that only positions are used that are located near a teammate.

After the clustering is finished, the positions inside each cluster that represents a single opponent robot have to be merged to a single position. For this reduction, the covariances of the measurements are merged using the measurement update step of a Kalman filter. The result of the whole merging procedure is a set of positions that represent the estimated positions of the opponent robots. In addition, the four goal posts are also regarded as opponent robots because they are obstacles that are not teammates.
5.4 Field Coverage

In recent years, the reaction of our robots to losing knowledge about the ball position was to just turn on the spot and wait for the ball to show up in the field of view again. This was sufficient in times of near perfect ball perception even across the whole field. However, the introduction of the new ball in 2016 brought up new problems. It is no longer possible to detect the ball at very far ranges. In addition, there exist several configurations that obstruct ball detection from certain perspectives, such as a ball between the legs of a robot. As a consequence, there are more periods of time during which no robot knows the current ball position. To overcome this problem, we reintegrated our cooperative ball search (as presented in [15]), which takes into account which parts of the field are actually visible for which robot and accordingly computes search areas for each individual robot. Due to the team-wide ball model (cf. 5.2.3), a robot only needs to actively search for the ball if all team members lost knowledge of the ball position. Consequently, if a robot is searching, it can be sure that its team members do the same. Knowing which parts of the field are visible to the team members, searching the ball can happen dynamically and much more effective.

5.4.1 Local Field Coverage

To keep track of which parts of the field are visible to a robot, the field is divided into a very coarse grid of cells, each cell being a square that has a size of $0.25m^2$. To determine which of the cells are currently visible, the current field of view is projected onto the field. Then, all cells whose center lies within the field of view are candidates for being marked as visible. There may be other robots obstructing the view to certain parts of the field. Thus, depending on the point of view of the viewing robot, another robot may create a “shadow” on the field. No cell whose center lies within such a shadow is marked as visible. An example local field coverage is depicted in Fig. 5.4.

Having determined the set of visible cells, each of the cells gets a time stamp. These time stamps are later used to build the global field coverage model. They are also used to determine the least recently seen cell, which can be used to generate the head motion to scan the field while searching for the ball.

A special situation arises when the ball goes out. If this happens, the time stamps of the cells are reset to values depending on where the ball is being put back onto the field. That way, the least recently seen cell of the grid – the cell which the robot has the most outdated information about – is now the cell in which the ball most likely is. This cell is determined by the last intersection of the trajectory of the ball with a field line before the GameController sent the information that the ball is out. In addition, all other cells that are near the field border receive later time stamps so that the side lines become targets for the ball search in case of a referee error or an error in the estimation of the ball motion.

Another special case is the ball’s movement as represented in the BallModel. If the ball touches a cell, its time stamp is set back to zero. This way, if the ball is lost while moving, the robots will search for it along its last known path first.

5.4.2 Global Field Coverage

To make use of the field coverage grids of the other robots, each robot has to communicate its grid to its teammates. So, in addition to its own local field coverage grid, each robot maintains a global field coverage grid, which is incrementally updated in every team communication cycle.
This global grid looks roughly the same for all teammates, so that calculations based on the grid come to sufficiently similar results for all teammates.

Since we want to change the actual time stamp of each cell in case the ball goes out, we actually need two time stamps in team communication. One to determine the time when the cell was last seen as described above and one to keep track of the time when this value was changed. The former will be called coverage value from here on. The merging of the grids is then done simply by looking at the time stamp of each cell in the existing global grid and a received local grid. If the received cell has a more recent time stamp it is carried over to the global grid.

Given this year’s field dimensions, 4 byte time stamps, and a cell’s edge length of $\frac{1}{2}m$, there are $2 \times 4 \text{ bytes} \times \frac{6m \times 9m}{(\frac{1}{2}m)^2} = 1728 \text{ bytes}$ which have to be sent in every team communication cycle in addition to all other communication our robots do during gameplay. Since the resulting bandwidth requirement would be beyond the bandwidth cap set by this year’s rules, the grid is not sent as a whole but in intervals of 18 cells each. In addition, the time stamps are compressed into 2 byte values and a base time stamp for each interval. This way only $18 \times 2 \times 2 + 4 \text{ bytes} = 76 \text{ bytes}$ are required each cycle.

To calculate each robot’s patrol target, the cells of the global grid are sorted descending by their coverage value. For each cell in this list, the distance to each robot is calculated. If the distance of the cell to the current robot is less than the distance to one of its teammates, the patrol target is found.
5.5 Whistle Recognition

The implementation for detecting a referee whistle is more of less the same as in the last two years. The approach is based on the correlation between the robot’s current audio input and a previously recorded reference whistle. As different whistles might be used during a competition and as a single whistle might produce quite different sounds, our 2016 implementation allows to compare the audio input with multiple reference whistles. The whistle recognition is implemented in the MultiWhistleRecognizer module, providing the Whistle representation. In addition to this detection, we also implemented a mechanism for a team-wide majority vote.

5.5.1 Correlating Whistle Signals

For the whistle detection, we use a cross-correlation. To calculate the correlation between a given whistle signal and a actual recorded signal in a fast way, we transform the actual signal into the frequency domain, multiply it with the conjugate complex stored reference whistle signal and transform it back into the time domain, where we detect the compliance with the reference signal.

First, we need a reference signal $s_{\text{ref}}$. The reference signal is also recorded by the robot, where we take a window length of $N$ samples. We use the sampling frequency $f_s$ of the robot’s audio hardware without downsampling. The signal must be extended with zeros by length $N$, because the result of the correlation is twice the window length. In addition, correlating signals by Fourier transforming them results in a cyclic correlation, which is avoided by the zero padding. This signal is transformed to the frequency domain by a Fast Fourier Transformation (we use the highly efficient FFTW3 implementation by [6]).

$$\mathcal{F}(s_{\text{ref}}[n]) = S_{\text{ref}}[k].$$  

(5.4)

The underline denotes complex values. This signal is stored as a reference and has not to be recalculated in every step.

For the actual recorded signal $s_{\text{act}}$, we use the same procedure by adding zeros to get the doubled length and transform it into the frequency domain to get $S_{\text{act}}$.

We make use of the fact that a correlation can be represented by a convolution using one signal in reorder. The advantage of a convolution is that it can be calculated easily in the frequency domain by just multiplying the signals. As we do not want to store the signal in reverse order, we can transform the reversion into the frequency domain, too. Thus, we get

$$s_{\text{ref}}[n] * s_{\text{act}}[-n] = S_{\text{ref}}[k] \cdot S_{\text{act}}^*[k],$$  

(5.5)

where $S_{\text{act}}^*$ denotes the conjugate complex of the signal.

As it does not matter which signal is conjugated, we store the reference signal as conjugate. Multiplying both signals gives us the result in the frequency domain and the inverse Fourier Transform is the correlation result $s_{\text{corr}}$ in the time domain.

The best result can be achieved if the reference signal is equal to the actual signal or the negative actual signal. Using this value, which is the autocorrelation value, and dividing the correlation signal $s_{\text{corr}}$ by this factor, we always get a percent value of the best possible result. An example is depicted in Fig. 5.5.

If the percentage is above a defined threshold, the whistle is detected. As step by step formulation, we have the following implementation in discrete form:
5.5. WHISTLE RECOGNITION

Figure 5.5: Correlation between the current signal and a pre-recorded reference whistle signal over 600 frames (equal 10 seconds). During this time, a whistle has been blown twice. The threshold for whistle detection during this experiment has been set to 0.15. The differently colored lines depict the different results from the two used audio channels.

\[
\begin{align*}
    s'_{\text{act}}[n] &= (s_{\text{act}}[n] \ 0_N) \\
    S_{\text{act}}[k] &= \mathcal{F}(s'_{\text{act}}[n]) \\
    s_{\text{corr}}[n] &= \mathcal{F}^{-1}(S_{\text{act}}[k] \cdot S^*_{\text{ref}}[k]) \\
    \varphi_{\text{cc}} &= \max(|s_{\text{corr}}[n]|) \\
    \varphi_{\text{acmax}} \in [0, 1] \\
    w &= \begin{cases} 
        \text{true} & \text{for } \varphi_{\text{cc}} \geq \alpha \\
        \text{false} & \text{otherwise} 
    \end{cases}
\end{align*}
\]

\( \varphi \) denotes the correlation value for the cross correlation and the auto correlation of the reference signal respectively. The parameters \( n \) and \( k \) are discrete time and frequency respectively.

The correlation procedure is repeated for every stored reference whistle. The computed Whistle representation contains the information about the signal that has the best correlation. The module’s required computing time scales linearly with the number of reference whistles. During RoboCup 2016, we used the reference signals of four different whistles. However, all computations are only executed during the game’s SET state.

5.5.2 Majority Vote

To decide whether or not a whistle has been blown, we don’t rely on the detection of each individual robot alone. Instead, we take all detections on the different audio channels on each robot into account and start a majority vote on whether or not the team should start playing. To do this, each robot sends the time stamp and its confidence of the last whistle detection to the teammates via team communication. If the robot never detected a whistle, this time stamp is zero. The confidence is calculated as follows:

- 100: If the robot detected the signal on both audio channels
- 66: If the robot detected the signal on one audio channel while the other one is deactivated due to being damaged
- 33: If the robot detected the signal only on one audio channel although both microphones are supposed to be working fine
- 0: If the robot did not detect a signal
- If both audio channels are deactivated (i.e. the robot is deaf)

All detections that happened after the beginning of the last SET state are now taken and sorted by their time stamps. Each detection is now grouped with all other detections that happened within 500 milliseconds of each other. The total confidence of each group is then divided by the number of non-deaf robots. If the average confidence of one group is above 50, the team starts playing.
Chapter 6

Behavior Control

The part of the B-Human system that performs the action selection is called Behavior Control. It also runs in the context of the process Cognition. The behavior is modeled using the C-based Agent Behavior Specification Language (CABSL). The main module – BehaviorControl2015 [sic!] – provides many representations that either control the actions of the robot or are sent to teammates. It is accompanied by several additional modules such as the VGPathPlanner (cf. Sect. 6.5), the KickPoseProvider (cf. Sect. 6.6), the CameraControlEngine (cf. Sect. 6.7), and the LEDHandler (cf. Sect. 6.8) as well as the RoleProvider (cf. Sect. 6.3.1).

This chapter begins with a short overview of the behavior specification language CABSL and how it is used in a simple way. Afterwards, it is shown how to set up a new behavior. Both issues are clarified by examples. The major part of this chapter is a detailed explanation of the full soccer behavior used by B-Human at RoboCup 2016.

6.1 CABSL

CABSL is a language that has been designed to describe an agent’s behavior as a hierarchy of state machines and is a derivative of the State Machine Behavior Engine (cf. [26, Chap. 5]) that we used a few years ago. CABSL solely consists of C++ preprocessor macros and, therefore, can be compiled with a normal C++ compiler.

For using it, it is important to understand its general structure. In CABSL, the following base elements are used: options, states, transitions, actions. A behavior consists of a set of options that are arranged in an option graph. There is a single starting option from which all other options are called; this is the root of the option graph. Each option is a finite state machine that describes a specific part of the behavior such as a skill or a head motion of the robot, or it combines such basic features. Each option starts with its initial state. Inside a state, an action can be executed which may call another option as well as execute any C++ code, e.g., modifying the representations provided by the Behavior Control. Further, there is a transition part inside each state, where a decision about a transition to another state (within the option) can be made. Like actions, transitions are capable of executing C++ code. In addition to options (cf. Sect. 6.1.2), CABSL supports the use of so-called Libraries (cf. Sect. 6.1.2), which may be used inside actions and transitions. Libraries are collections of functions, variables, and constants, which help keeping the code clean and readable without the need of additional modules and representations, while providing a maximum of freedom if extensive calculations have to be made.

This structure is clarified in the following subsections with several examples.
Figure 6.1: Behavior control module graph. Behavior control modules are depicted as yellow rectangles. All representations they provide are shown as blue ellipses. The representations they require are shown as ellipses colored in either yellow if they are provided by other Cognition modules and orange if they are received from the process Motion.
6.1.1 Options

```plaintext
option(exampleOption)
{
    initial_state(firstState)
    {
        transition
        {
            if(booleanExpression)
                goto secondState;
            else if(libExample.boolFunction())
                goto thirdState;
        }
        action
        {
            providedRepresentation.value = requiredRepresentation.value * 3;
        }
    }
}

state(secondState)
{
    action
    {
        SecondOption();
    }
}

state(thirdState)
{
    transition
    {
        if(booleanExpression)
            goto firstState;
    }
    action
    {
        providedRepresentation.value = RequiredRepresentation::someEnumValue;
        ThirdOption();
    }
}
```

Special elements within an option are common transitions as well as target states and aborted states.

Common transitions consist of conditions that are checked all the time, independent from the current state. They are defined at the beginning of an option. Transitions within states are only “else-branches” of the common transition, because they are only evaluated if no common transition is satisfied.

Target states and aborted states behave like normal states, except that a calling option may check whether the called option currently executes a target state or an aborted state. This can come in handy if a calling option should wait for the called option to finish before transitioning to another state. This can be done by using the special symbols `action_done` and `action_aborted`.

Note that if two or more options are called in the same action block, it is only possible to check whether the option called last reached a special state.
In addition, options can have parameters that can be used like normal function parameters. The specification of parameters uses the same syntax as the streamable data type definition in our
Figure 6.2: The behavior view shows the actual parameters, e.g. of the option SetHeadPanTilt (left: source, right: view). The display of values equaling default parameters (specified in a second pair of parentheses) is suppressed. On the right of the view, the number of seconds an option or state is active is shown.

system. Thereby, the actual parameters of each option can be recorded in log files and they can also be shown in the behavior dialog (cf. Fig. 6.2). Please note that the source code shown in Fig. 6.2 is still understood by a C++ compiler and thereby by the editors of C++ IDEs.

### 6.1.2 Libraries

A library is a normal C++ class, a single object that is instantiated as part of the behavior control and that is accessible by all options. In contrast to options, libraries can have variables that keep their values beyond a single execution cycle. Libraries must be derived from the class LibraryBase and an instance variable with their type must be added to the file Libraries.h. In addition, if you want your library to be useable by other libraries, it has to be added to the files LibraryBase.h and LibraryBase.cpp. Please refer to the example behavior contained in this code release on how to do this.

Here is an example for a header file that declares a library:

```cpp
class LibExample : public LibraryBase
{
public:
    LibExample();
    void preProcess() override;
    void postProcess() override;
    bool boolFunction(); // Sample method
};
```

Here is the matching implementation file:

```cpp
#include "../LibraryBase.h"

namespace Behavior2015
{
    #include "LibExample.h"
    #include "AnotherLib.h"
}```
6.2 Setting Up a New Behavior

The code release comes with a basic behavior but you might want to set up your own one with a new name. To do so, you simply need to copy the existing behavior in `Src/Modules/BehaviorControl/BehaviorControl2015` into a new folder `Src/Modules/BehaviorControl/<NewBehavior>` and subsequently replace all references to the old name within the cloned folder. This basically means to rename the module `BehaviorControl2015` to your desired name (this includes renaming the configuration file `Config/Locations/Default/behaviorControl2015.cfg`) and to update the namespace definitions of the source files within `<NewBehavior>/*.cpp` and `<NewBehavior>/*.h`. When renaming the namespace and the module, you may not choose the same name for both.

6.3 Behavior Used at RoboCup 2016

The behavior is split in two parts, “BodyControl” and “HeadControl”. The “BodyControl” part is the main part of the behavior and is used to handle all game situations by making decisions and calling the respective options. It is also responsible for setting the `HeadControlMode` which is then used by the “HeadControl” to move the head as described in Sect. 6.3.7. Both parts are called in the behavior’s main option named `Soccer`, which also handles the initial stand up of the robot if its chest button is pressed as well as the sit down motion, if the chest button is pressed thrice.

The “BodyControl” part starts with an option named `HandlePenaltyState` which makes sure that the robot stands still in case of a penalty and listens to chest button presses to penalize and unpenalize the robot manually. If the robot is not penalized, the option `HandleGameState` is called, which in turn will make sure that the robot stands still, if one of the gamestates `initial`, `set`, or `finished` is active. In the “Ready” state, it will call the option `ReadyState` which lets the robots walk to their kickoff positions.

When in “Playing” state, an option named `PlayingState` will be called which handles different game situations. First, it handles kickoff situations which are described in detail in its own
section (cf. Sect. 6.3.6). In addition it checks whether a special situation, where all robots have to behave the same, has to be handled before invoking the role selection (cf. Sect. 6.3.1) where each robot gets chosen for one of the different roles (cf. Sect. 6.3.1).

The special situations are:

**StopBall:** If the ball was kicked in the direction of the robot, it tries to catch it by getting in its way and execute a special action to stop it.

**TakePass:** If the striker decides to pass to a robot, that robot immediately turns in the direction of the ball and tries to take the pass.

**SearchForBall:** If the whole team has not seen the ball for a while, the field players start turning around to find it. While turning, the robot aligns its head to the positions provided by the FieldCoverageProvider (cf. Sect. 5.4). If, after a whole turn, no ball was observed, each robot patrols to a position provided by the GlobalFieldCoverageProvider (cf. Sect. 5.4). This also includes a special handling for balls that went out.

### 6.3.1 Roles and Tactic

The intention of this year’s major behavior changes was to improve the behavior of last year by more exact calculations instead of good approximations. The general behavior ideas of last year were to play defensively and to let the robots move less. To accomplish this, we played with the following lineup:

1. **One Keeper:** The goalkeeper robot.
2. **Two Defenders:** Robots that are positioned defensively to help the Keeper by defending their own goal.
3. **One Striker:** The ball-playing robot.
4. **One Supporter:** A robot that is positioned offensively to help the Striker.

In a lineup with less than five robots, at least one robot should be the Defender. The ideas behind the lineup at the overall behavior are to prevent opponent teams from dribbling into our goal and to save battery power and avoid joint heat. Saving battery power is important for us since we had issues to play multiple long games on one tournament day with the batteries available to us. In the same scenario, it is important to avoid joint heat and the consequential joint stiffness loss to perform motions on the highest possible level, when they are really needed.

This year, we have been using the same lineup as in 2015 but the behavior differs in Keeper, Defender (slightly), KickPoseProvider (cf. 6.6) and ball search (cf. 5.4).

The role selection is done by the RoleProviderRoyale. The general procedure is that every robot is calculating a role selection suggestion for itself and each connected teammate, but uses the role selection of the captain robot. The captain robot is determined by the lowest (not penalized) connected teammate number. The reason behind this procedure is to have always a suggestion available but inhibit the chance that the robots play with different role selections at the same time.

The keeper role is always assigned to the robot with the number one and will not be changed at any point. The Striker role will be assigned to the robot which can reach the ball fastest. The calculation will favor the current striker to prevent oscillations. All other available robots
will now distribute to the two Defender roles and one Supporter role. To decide which robot becomes a Supporter and which a Defender, the global x-position will be used with a favor for the previous role. If there are more robots than the goalkeeper and the striker, at least one defender will assigned.

6.3.2 Striker

The main task of the striker is to go to the ball and to kick it into the opponent goal. To achieve this simple-sounding target, several situations need to be taken into account:

**GoToBallAndKick:** Whenever no special situation is active (such as kick-off, duel, search for the ball . . . ), the striker walks towards the ball and tries to score. This is the main state of the striker and after any other state, which was activated, has reached its target, it returns to GoToBallAndKick. To reach a good position near the ball, the striker walks to the KickPose (cf. Sect. 6.6). Depending on the distance to the target, the coordinates are then either passed to the PathPlanner (cf. Sect. 6.5) to find the optimal way to a target far away or to the LibWalk when it comes to close range positioning and obstacle avoidance. Each strategy then provides new coordinates, which are passed to the WalkingEngine (cf. Sect. 8.3).

After reaching the pose, the striker either executes the kick that was selected by the KickPoseProvider (cf. Sect. 6.6) or chooses another state in case of an opponent robot blocking the way to the goal.

**Duel:** Whenever there is a close obstacle between the ball and the goal, the robot will choose an appropriate action to avoid that obstacle. Depending on the position of the robot, the ball, and the obstacle, there are several actions to choose from (cf. Fig. 6.3):

- The robot is near the left or right sideline.
  - If the robot is oriented to the opponent goal and the straight way is not blocked by an obstacle, it will perform a fast forward kick.
  - If the direct way to the goal is blocked or the robot is oriented away from the opponent goal, the robot will perform a sidewards kick to the middle of the field.

- The robot is in the middle of the field.
  - If the obstacle is blocking the direct path to the goal, the robot will perform a sidewards kick. The direction of the kick depends on the closest teammate and other obstacles. In an optimal situation, a supporter should stand besides the striker to receive the ball.
  - If the obstacle leaves room for a direct kick towards the opponent goal, the robot will perform a fast forward kick.

The kick leg is always chosen by the smallest distance from the leg to the ball. If the robot has not seen the ball for more than three seconds while in duel mode, it will move backwards for a short time to reduce the probability of getting punished for pushing in case the ball is no longer close to the robot. Duel is deactivated in the vicinity of the opponent’s penalty box, because we do not want the robot to kick the ball away from the goal if there is an obstacle (i.e. the keeper or a goalpost) in front of it.

**DribbleDuel:** This is a special kind of duel behavior. It is chosen if there is an opponent between the ball and the goal and our robot stands in front of the ball facing the side of
6.3. BEHAVIOR USED AT ROBOCUP 2016

Figure 6.3: A set of possible tackling situations. The black lines mark the desired kick direction.

the field. In that case, the robot just takes its arms back to avoid contact with the opponent robot and starts to run to the ball to quickly dribble it away from the opponent’s feet.

**DribbleBeforeGoal:** If the ball is lying near the opponent ground line, so that it is unlikely to score a goal, the striker aligns behind the ball and slowly moves in the direction of the opponent’s penalty mark. If the robot drifts too far away from its target, it starts again to align. If it is again possible to score directly, the state *GoToBallAndKick* is executed.

**WalkNextToKeeper:** If the ball is inside the own penalty area, the robot will try to block the way to the ball so that opponents cannot reach it. If the keeper is ready to kick the ball out of the penalty area, the robot will move out of the way.

**KickingOut:** The striker will kick the ball out of the field, if the position at which it should be put in again is strategically better than the current ball position.

6.3.3 Supporter

Our supporter knows two modes: Directly supporting the striker or waiting near the opponent goal. Directly supporting means following the striker to catch the ball in case the striker loses the ball against an opponent. The aim of waiting near the opponent goal is to catch the ball in case the opponent defense successfully blocked a long range shot at their goal.

6.3.4 Defender

A Defender has five modes, which differ only slightly. These are: back-left, back-middle, back-right, forward-left and forward-right – the areas where the robot can position itself. Whereas the role Defender is taken upon consultation with the robot’s team members, the decision of the taken mode is made by the Defender alone. It makes its decision on the basis of the received pose of the second Defender (or on the absence of such data). If there are two Defenders, they
must be either left or right and never on the same side. In addition, one of them must hold a back-mode but the other one may be forward. Otherwise, if there is just one defending robot, it must be a back one.

In each frame, the Defender recomputes

1. if it is left or right (or in special case middle),
2. if it is forward,
3. which position it should take, and
4. if it should move.

The Left / Right Decision

If the Defender is left, it means that it will calculate a position on the left side of the virtual line between the calculated goalkeeper position and the ball. A middle defending position means that the Defender stands on this line. This position is just a special case if neither a Keeper is inside the penalty area nor a team member performs a second Defender. In all other cases, you do not want to interfere with the direct goalkeeper sight of the ball.

- Being a single Defender (with a present Keeper), a robot will always decide for the same side like last frame as long as the ball is not too far on the other side.
- In the case of two Defenders, each robot will compare its (signed) distance to the goalkeeper-ball line and the distance of the last known position of the other Defender to that line. If the robot can clearly decide if it is more left of the line, it takes the decision. If this is not the case, it tries to decide by looking up which of the both poses is clearly more left on the field (given the global coordinate system as described in Sect. 4.1.2). If the robot still does not know which side it belongs to, it uses the decision of the last frame.

The Back / Forward Decision

A back position is defined as a position directly outside the penalty area, a forward position could be farther away. In the case of a single defending robot, the position will be back as mentioned above. In the other case, the robot decides on the basis of its actual position, if it is distinctly forward or back. It does the same for the position of the second Defender. The robot will hold its decision, if it differs from the position about the second Defender or if it is back and the ball is not far away. If both cases are not applicable and the robots are clearly looking in different directions, the robot that is looking in the direction of the opponent goal will move forward. If none of the cases were applicable and the Defender robot is on the same side as the ball, it is moving forward. If the ball side is not clear, both Defenders stay back.

The Position Selection

All selected positions for the Defender are located on a semicircle, which is centered around the center of the own goalposts. The radius is decided by the previously described back / forward mode. If it is back, the semicircle goes barely around the penalty area, but the maximum semicircle for a forward Defender gives the own goal a wide berth. As long as the ball is not inside the own penalty area, the forward radius is always smaller than the distance of the
6.3. BEHAVIOR USED AT ROBOCUP 2016

semicircle origin to the ball. The **Defender** will take a position on the selected semicircle with that he can cover the space between the goalpost on his side and the goalie blocking coverage area. This means that a forward **Defender** is standing nearer to the goalie-ball-line, which results in a restriction to the goalkeeper’s sight of the ball. At that range our goalie will not see the new ball as good as the old one. Thus, this positioning is still acceptable. To avoid that the distance of the back-standing robots to the penalty area gets too large, the semicircles may be cut along the lines of the penalty area. There is also a special handling when the ball is entering the penalty area or is lying close to the own ground line.

**The Decision to Move**

Beside the general idea, this year, methods corresponding to this topic were not implemented. Instead, our robots always wanted to move to the calculated position. The underlying path finding modules inhibits this for very short distances.

It is important that the **Defender** is doing its job, but as the last instance, between opponent scoring or a successful defense, there is still the goalkeeper. To give him time for a timely reaction, it must see the ball early enough. To maximize this time, the **Defenders** are also performing arm movements to make themselves as thin as possible, if the goalkeeper-ball line comes close to the robot.

6.3.5 Keeper

The keeper’s main task is to defend the own goal. Hence, the keeper mainly stays inside the own penalty area. There are only two situations in which the robot leaves the own goal. Either the keeper is taken out of the field or it walks towards the ball to kick it away. When the robot was taken from the field and put back on the field, the keeper walks directly towards its own goal again. Certainly, the time to reach the own goal should take as little time as possible. To achieve this, the robot walks straight forward to the own goal as long as it is still far away and starts walking omni-directionally when it comes closer to the target to reach the desired orientation. When walking omni-directionally, the keeper avoids obstacles in the same way as the other players do. The second situation, in which the robot leaves the own goal, is given if the ball is close to the goal and has stopped moving. In this case, the robot walks towards the ball in order to kick it away. The keeper uses the same kicking method as the striker with the difference that obstacle avoidance is deactivated. Since a goalie is not penalized for pushing in its own penalty box, this solution is acceptable.

In any other situation, the robot will not leave its position within the own penalty area and executes different actions according to different situations. Normally, the keeper walks to positions computed like described below. So while guarding the goal, the keeper will switch its head control depending on how long the ball was not seen and in which half of the field the ball was seen last. For example, while the ball is in the opponent half, the keeper looks for important landmarks to improve its localization. But once the ball passes the middle line, the keeper invariably tracks the ball.

If the ball is rolling towards the goal, the goalie has to select whether it should spread its legs or dive sidewards to catch the ball. The decision is made by considering the velocity of the ball as well as its distance. The estimates are used to calculate both the remaining time until the ball will intersect the lateral axis of the goalie and the position of intersection. In case of a close intersection, the goalie changes to a wide defensive posture to increase its range (for approximately four seconds, conforming to the rules of 2016). The diving is initiated when the
Figure 6.4: The blue figure indicates the pose of a goalkeeper that is rotated towards the ball (depicted as a small orange circle). The area between ball and goal is shown in red. The areas that are covered with the keeper actions standing / blocking / diving are colored in green / violet / orange.

Ball intersects the goalkeeper’s lateral axis in a position farther away from the farthest possible point of the defensive posture. The decision whether to execute such a motion is consolidated over multiple frames (about a third of a second in total) to compensate for falsely positive ball estimates and to avoid an unnecessary dive.

Sometimes, after a rough situation in the penalty area, the keeper’s rotation is wrong by 180°, letting the robot turn in the wrong direction and stare into the goal net. This situation is resolved by the self-locator by checking for a close FieldBoundary and getting no percepts of major field elements for a certain time. In this case, the goalie localization is turned back by 180 degrees, letting the behavior move to the right pose again.

In case the keeper has not seen the ball for several seconds (the actual value is currently set to 7s) it will, under some further circumstances, initiate a ball searching routine. This is only done if its absolute position on the field is considered to be within the own penalty area and the ball position communicated by the teammates is considered invalid. The first condition grants that the keeper does not start searching for the ball while it is walking back to its goal after returning from a penalty. If the ball position communicated by the teammates is considered valid, the keeper may be able to find the ball by turning towards the communicated position. If it is not valid, it will refrain from its normal positioning as described below and take a position in the middle of the goal. Once it reaches this position, it will continually turn its body left and right to cover the greatest possible range of view until either of its teammates or the goalie himself has found the ball.

The main idea of the goalie positioning is to inhibit the opponent robot to score a goal, i.e. to inhibit the ball to move at a straight line into the goal. This year, we are directly calculating which parts of the field will be covered with a specific robot pose. The perfect position would be if the robot is able to cover the whole goal by just standing. Of course, in nearly every situation this is not recommendable because of the distance to the goal the robot must hold to meet this condition. The second best situation would be if our robot is able to cover the whole goal with
our blocking motion (sitting down and spreading the legs) since this motion needs nearly no
time to perform. This is why we are trying to maximize the goal coverage with this motion at
all time. This intention is restricted by the secondary idea of the goalie positioning: the goalie
should also be able to perform his dive motion at any moment. It is possible to perform a dive
motion if the goalie is inside the penalty area and if it will not dive into the goal post/frame to
prevent damages. In addition, a restriction is added that the robot is not touching the penalty
area lines with the feet to keep the lines detectable for the vision system. This should help
the robot to maintain an accurate self-localization which is essential to find the best spot for
covering the goal.

With this goalie positioning, we are able to cover (nearly) the whole goal at all time and to
perform the dive motion before the ball crosses the keeper. The coverage at an example situation
is depicted in Fig. 6.4;

6.3.6 Kickoff

The kickoff options control the positioning behavior of all robots during the whole READY and
SET game states as well as the action selection during the beginning of PLAYING.

6.3.6.1 Positioning

For the two main kickoff situations (defensive and offensive), distinct sets of kickoff poses are
specified in a configuration file. Each pose set is ordered by the importance of the respective
poses, i.e. if robots are missing, the least important positions are not taken. We do not rely on
a fixed pose assignment (except for the goalkeeper, of course) based on robot numbers as such
an approach has two significant drawbacks. Firstly, an important position might be left empty,
if one robot is missing. Secondly, some robots might be forced to walk farther than necessary
and thus risk to come too late.

As our self-localization provides a very precise pose estimate and all robots receive the poses of
their teammates via team communication, a robust dynamic pose assignment is possible. Our
approach is realized by two sequential assignment rules:

1. The robot that is closest to the center position will walk to that position. If our team has
   kickoff, this position is equal to the center of the field. If the opponent team has kickoff,
   this position is between the center of the field and our own goal, right at the border of the
   center circle.

2. For the remaining field players, the configuration that has the smallest sum of squared
distances (between robots and positions) is determined.

The first rule ensures that the central position will always be occupied, if there is at least one
field player left. This position is definitely the most important one as it is the one closest to the
ball. Furthermore, assigning this position to the closest robot strongly increases the likelihood
that this robot actually reaches it in time. A robot that has a long way to walk during the
READY state might be blocked by other robots or the referees. These delays could lead to a
manual placement position which is too far away from the ball position.

We prefer a correct orientation to a correct position. Thus, to avoid having robots facing their
own goal, the positioning procedure is stopped a few seconds before the end of the READY
state to give a robot some time to take a proper orientation.
6.3.6.2 Actions after the Kickoff

When the game state changes from SET to PLAYING, not all robots immediately switch to their default playing behavior as some rules have to be considered.

If a team does not have kickoff, its robots are not allowed to enter the center circle until the ball has been touched or 10 seconds have elapsed in the PLAYING state. Therefore, our robots switch to a special option that calls the normal soccer behavior but stops the robots before entering the center circle without permission. The robots actively check, if the ball has already been moved in order to enter the center circle as early as possible.

If a team has kickoff, its robots are not allowed to score before the ball has left the center circle. Therefore, the normal playing behavior, which would probably try to make a direct shot into the goal, is postponed and a kickoff action is carried out. Our kickoff taker prefers to make a medium-length shot into the opponent half. The direction of the shot depends on the perceived obstacles around the center circle. If there seem to be no obstacles at all, the kickoff taker dribbles the ball to a position outside the center circle.

Both kickoff variants are terminated, i.e. the behavior is switched to normal playing, in two cases: if the ball has left the center circle or if a certain timeout (currently 15 seconds) has been reached.

6.3.7 Head Control

The head control sets the angles of the two head joints and thereby the direction in which the robot looks. In our behavior, the head control is an independent option (HeadControl) running parallel to the game state handling and is called from the root option Soccer. This ensures that only one head control command is handled per cycle. However, the head control command is given by the game state handling option by setting the symbol theHeadControlMode. As long as the symbol does not change, the selected head control command is executed continuously.

The problem of designing and choosing head control modes is that the information provided by our vision system is required from many software modules with different tasks. For instance, the BallPerceptor, and the LinePerceptor provide input for the modeling modules (cf. Sect. 5) that provide localization information, and the ball position. However, providing images from all relevant areas on the field is often mutually exclusive. For instance, when the ball is located in a different direction from the robot than the goal, it cannot look at both objects at the same time. In addition to only being able to gather some information at a time, speed constraints come into play, too. The solution to move the head around very fast to look at important areas more often proves impractical, since not only the images become blurred above a certain motion speed, but also because a high motion speed has a negative influence on the robot’s walk stability due to the forces resulting from fast rotational movements of the head. With these known limitations, we had to design many head control modes for a variety of needs. We are able to use three different ways of setting the position of the head. We can specify the absolute angels of the head joints (option SetHeadPanTilt), a position on the field (option SetHeadTargetOnGround), or a position relative to the robot (option SetHeadTarget). The head control modes used in our behavior are:

**off**: In this mode, the robot turns off its head controlling joints.

**lookForward**: In this mode, the pan and tilt angles are set to static values, so that the robot looks forward. This mode is used before the game, after the game, and when the robot is penalized.
lookLeftAndRight: In this mode, the robot moves its head left and right in a continuous motion. This mode is used for a first orientation after the robot has been penalized and gets back to the game. It is also used when the robot or the whole team searches for the ball.

lookAtGlobalBall: As the name implies, this head control moves the head to the global ball estimate, which is primarily used by the keeper, whenever the ball is covered by field players.

lookAtGlobalBallMirrored: This mode is basically the same as the one above, but it moves the head toward the position of the global ball mirrored at the center point, assuming the robot might be at the mirrored position from where it thinks to be.

lookAtBall: When using this head control mode, the robot looks at its ball estimate. However, if the ball is not seen for some time, the head will be moved toward the global ball estimate. In case the ball is still not seen after some additional time, the robot just changes to lookActive mode. The mode lookAtBall is used, whenever a robot kicks the ball or is preparing to kick it as well as when the robot is in a duel. Nonetheless, this mode is important for the keeper, when the ball moves toward the goal and the keeper must prepare to catch it. Additionally, this mode is used by the robot roles used for penalty shoot-out (cf. Sect. 6.4).

lookAtBallMirrored: This is basically the same as the mode above, but the robot looks at the estimated ball position mirrored at the center point. This mode is used for checking, if the ball is mirrored from where the robot thinks it is.

lookAtOwnBall: This mode is nearly the same as lookAtBall, but – as the name implies – without taking the global ball estimate into account.

lookAtOwnBallMirrored: This mode is analog to lookAtBallMirrored. It is basically the same as lookAtOwnBall, but it takes the mirrored position into account. This mode is used for checking, if the ball is mirrored from where the robot thinks it is.

lookActive: In this mode, the robot looks to the most interesting point on the field, which is calculated by libLook. During the game, the importance or unimportance of different points on the field – for example field features, field lines, and the ball position – changes, depending on the current situation. Therefore, our solution is based on having a set of points of interest which consists of static points on the field as well as of dynamic points like the ball estimate or the global ball estimate. All points that are reachable by the head are assigned a value representing the robot’s interest in that point, depending on the robot’s role and the time that has passed since the point has been seen. It also handles synchronized head control, i.e. when the striker looks away from the ball, all other robots that are currently using this head control mode will focus the ball. Whether the striker looks away from the ball is checked in the three basic head control options SetHeadPanTilt, SetHeadTargetOnGround, and SetHeadTarget. The synchronized head control shall ensure that at least one team member sees the ball. Additionally, while using this head control mode, every robot that is currently not looking at the ball will turn its head as fast as possible toward the ball, when the global ball estimate moves toward the robot’s current position. This head control mode is the default one and is always used when none of the other ones fits better.

lookActiveWithBall: This head control mode is basically the same as lookActive, but it limits the reachability of points of interest, as it requires the head control to always keep
6.4 Penalty Shoot-out Behavior

The penalty shoot-out behavior of 2016 did not change mainly to previous years: Since the keeper can simply track the ball and jump to catch it, the main idea for the striker was to kick the ball as close to the goal post as possible. In addition, the ball shall be kicked with as much force as possible without any lack of precision. On the other side, the keeper shall also catch balls that are kicked near the goal post. The keeper’s jump on the goal line would leave some space between the robot and the goal post. To remove this possibility for a ball passing the keeper and to cover the whole space between the goal posts, the keeper’s position must be closer to the ball. Thus, the keeper shall move toward the ball before the striker kicks it.

The RoboCup 2016 penalty shoot-out behavior is embedded in the behavior described above. It is implemented by introducing two additional roles: PenaltyStriker, and PenaltyKeeper (cf. Sect. 6.3.1). Depending on the secondary game state and the current kick-off team, the RoleProviderRoyale assigns one of these two roles to the robot. Game state handling and head control is processed the same way as during an ordinary game.

In detail, the two roles work as follows:

**PenaltyKeeper:** The penalty keeper starts moving toward a position in the middle of the penalty area, using the time the striker needs for getting to the ball. As described above, this is done to cover the whole goal with a keeper jump. How much time the keeper has to take the required position is defined in Behavior2015Parameters. When the position is reached or the defined time has passed, the keeper has some additional time to check its rotation toward the ball. After that, the penalty keeper sits down as preparation to catch the ball. When the ball moves towards the keeper, depending on where the ball is assumed to pass the robot, the keeper either jumps left, jumps right, or uses the genuflect movement, i.e. a sit down movement where the robot spreads its legs. Regarding the head control, the penalty keeper always keeps the ball in sight of view, using lookActiveWithBall while taking position and lookAtBall after that (cf. Sect. 6.3.7).

**PenaltyStriker:** In the beginning, the robot randomly decides, if it wants to kick at the right edge of the goal or at the left edge. The robot moves towards the ball, already rotating in the direction, it wants to kick to. Reaching a specific distance from the ball, the striker stops two times for a short time, so it will not accidentally run against the ball. Then the robot makes the last few steps towards the ball, getting in position for the kick. In the end, our standard kick is executed. There exists a fallback behavior, in case there is just little time left for the shot. This behavior lets the robot just move towards the ball and kick it, using the KickPoseProvider (cf. Sect. 6.6). For the head control, the penalty striker simply uses lookAtBall, because it needs to focus at the ball all the time. This is acceptable, because due to its position on the field and its rotation toward the goal, the penalty area is also always in sight. Similar to the penalty keeper, the striker has a localization behavior for searching the ball, which is activated when the ball is not seen for some time.
6.5 Path Planner

In some situations, robots have to walk longer distances to reach their next target location, e.g., when walking to their kickoff positions or when walking to a distant ball. In these cases, a purely reactive control can be disadvantageous, because it usually would not consider obstacles that are further away, which might result in getting stuck. Therefore, our robots use a path planner in these situations since 2011. Until 2014, it was based on the Rapidly-Exploring Random Tree approach [17] with re-planning in each Cognition cycle. Although the planner worked quite well, it had two major problems: On the one hand, the randomness sometimes resulted in suboptimal paths and in oscillations\(^1\). On the other hand, it seemed that the RRT approach is not really necessary for solving a 2-D planning problem, as the planner actually did. Thus, it was slower than it needed to be.

6.5.1 Approach

Therefore, a new planner was developed for RoboCup 2015. It is a visibility-graph-based 2-D A* planner (cf. Fig. 6.5). It represents obstacles as circles on which they can be surrounded and the path between them as tangential straight lines. As a result, a path is always an alternating sequence of straight lines and circle segments. There are four connecting tangents between each pair of non-overlapping obstacle circles, only two between circles that overlap, and none if one circle contains the other. With up to nine other robots on the field, four goal posts, and the ball, the number of edges in the visibility graph can be quite high. Thus, the creation of the entire graph could be a very time-consuming task. Therefore, the planner creates the graph while planning, i.e. it only creates the outgoing edges from nodes that were already reached by the A* planning algorithm. Thereby, the A* heuristic (which is the Euclidean distance) not only speeds up the search, but it also reduces the number of nodes that are expanded. When a node is expanded, the tangents to all other visible nodes that have not been visited before are computed. Visible means that no closer obstacle circles intersect with the tangent, which would prevent traveling directly from one circle to another. To compute the visibility efficiently, a sweep line approach is used. As a result, the planning process never took longer than 1 ms per Cognition cycle in typical games.

6.5.2 Avoiding Oscillations

Re-planning in each Cognition cycle bears the risk of oscillations, i.e. repeatedly changing the decision, for instance, to avoid the closest obstacle on either the left or the right side. The planner introduces some stability into the planning process by adding an extra distance to all outgoing edges of the start node based on how far the robot had to turn to walk in the direction of that edge and whether the first obstacle is passed on the same side again (no extra penalty) or not (extra penalty). Note that this does not violate the requirement of the A* algorithm that the heuristic is not allowed to overestimate the remaining distance, because the heuristic is never used for the outgoing edges of the start node.

6.5.3 Overlapping Obstacle Circles

The planning process is a little bit more complex than it appears at first sight: As obstacles can overlap, ingoing and outgoing edges of the same circle are not necessarily connected, because

\(^1\)Although the planner tried to keep each new plan close to the previous one.
Figure 6.5: Visualization of the planning process. The robot’s position is shown as a small rectangle right of the big orange square. The target position is depicted as small circle left of the center circle. Obstacles are shown as red and yellow squares, some of which are ignored. Obstacle circles are depicted in yellow. Yellow lines indicate expanded edges. The shortest path is depicted in bright green.

the robot cannot walk on their connecting circle segment if this is also inside another obstacle region. Therefore, the planner manages a set of walkable (non-overlapping) segments for each circle, which reduces the number of outgoing edges that are expanded when a circle is reached from a certain ingoing edge. However, this also breaks the association between the obstacle circles and the nodes of the search graph, because since some outgoing edges are unreachable from a certain ingoing one, the same circle can be reached again later through another ingoing edge that now opens up the connection to other outgoing edges. To solve this problem, circles are cloned for each yet unreached segment, which makes the circle segments the actual nodes in the search graph. However, as the graph is created during the search process, this cloning also only happens on demand.

6.5.4 Forbidden Areas

There are two other extensions in the planning process. Another source for unreachable segments on obstacle circles is a virtual border around the field. In theory, the shortest path to a location could be to surround another robot outside of the carpet. The virtual border makes sure that no paths are planned that are closer to the edge of the carpet than it is safe. On demand, the planner can also activate lines surrounding the own penalty area to avoid entering it. The lines prevent passing obstacles on the inner side of the penalty area. In addition, edges of the
visibility graph are not allowed to intersect with these lines. To give the planner still a chance to find a shortest path around the penalty area, four obstacle circles are placed on its corners in this mode. A similar approach is also used to prevent the robot from walking through the goal nets.

6.5.5 Avoiding Impossible Plans

In practice, it is possible that the robot should reach a position that the planner assumes to be unreachable. On the one hand, the start position or the target position could be inside obstacle circles. In these cases, the obstacle circles are “pushed away” from these locations in the direction they have to be moved the least to not overlap with the start/target position anymore before the planning is started\(^2\). On the other hand, due to localization errors, the start and target location could be on different sides of lines that should not be passed. In these cases, the closest line is “pushed away”. For instance, if the robot is inside its penalty area although it should not be, this would move the closest border of the penalty area far enough inward so that the robot’s start position appears to be outside for the planning process so that a plan can be found.

6.6 Kick Pose Provider

To calculate the optimal position for a kick toward the opponent’s goal, a module called \textit{KickPoseProvider} is used. Most of the time, at least some parts of the goal are covered by opponent robots. Therefore, we need to calculate the part of the goal that has the largest opening angle as seen from the ball (cf. Fig. 6.6). The center of this part will be the target for our kick.

Now we need to choose which kick to use. For every possible kick, the module requires information about how long it takes to perform the kick and what offset the robot has to have to the ball, including a rotation offset. The rotation offset defines an angle by which the target pose for performing the kick must be rotated to have the ball move in the intended direction after the kick. Based on this information, the module calculates a \textit{KickPose} for each kick, i.e. the point the robot has to stand at and the direction it has to face to execute the kick. Afterwards, each pose is evaluated to find the best of the possible poses.

The evaluation of the different poses is based on several criteria. The most important one is how long it will take the robot to execute the specified kick. This is broken down into how long it takes for the robot to reach the pose and how long it takes to perform the actual kick. Other things that are taken into account are whether the kick is strong enough for the ball to reach the opponent goal and the time since the ball was last seen. Some other constant properties of the kick influence the evaluation via the execution time of the kick. If, for instance, a kick is rather unstable and should only be used if the robot is already standing at an almost perfect position for executing the kick, the probability of the kick being chosen can be reduced by increasing its stored execution time.

6.7 Camera Control Engine

The \textit{CameraControlEngine} takes the \textit{HeadMotionRequest} provided by the \textit{BehaviorControl2015} and modifies it to provide the \textit{HeadAngleRequest}, which is used to set the actual head angles.

\(^2\text{Note that when executing the plan, these situations are handled differently to avoid bumping into other robots.}\)
Figure 6.6: Visualization of all openings between opponent robots (red squares) as seen from the ball. The red opening was chosen to be the kick target.

The main function of this module is to make sure that the requested angles are valid. In addition, it provides the possibility to either use the so called *PanTiltMode*, where the user can set the desired head angles manually, or the target mode. In target mode, the *CameraControlEngine* takes targets on the field (provided by the user) and calculates the required head angles by using inverse kinematics (cf. Sect. 8.3.3). It must also be ensured that the cameras cover as much of the field as possible. Therefore, the *CameraControlEngine* calculates which camera suits best the provided target and sets the angles accordingly. Also, because of this, most of the time the provided target will not be in the middle of either camera image.

### 6.8 LED Handler

The LEDs of the robot are used to show information about the internal state of the robot, which is useful when it comes to debugging the code.
6.8. LED HANDLER

Right Eye

<table>
<thead>
<tr>
<th>Color</th>
<th>Role</th>
<th>Additional information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Blue</td>
<td>All</td>
<td>No ground contact</td>
</tr>
<tr>
<td>Magenta</td>
<td>All</td>
<td>Robot is connected to external power source</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(Only if no Head LEDs are available.)</td>
</tr>
<tr>
<td>Blue</td>
<td>Keeper</td>
<td></td>
</tr>
<tr>
<td>White</td>
<td>Defender</td>
<td></td>
</tr>
<tr>
<td>Green</td>
<td>Supporter</td>
<td></td>
</tr>
<tr>
<td>Red</td>
<td>Striker</td>
<td></td>
</tr>
</tbody>
</table>

Left Eye

<table>
<thead>
<tr>
<th>Color</th>
<th>Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Yellow</td>
<td>No ground contact</td>
</tr>
<tr>
<td>White</td>
<td>Ball was seen</td>
</tr>
<tr>
<td>Blue</td>
<td>Field Feature was seen</td>
</tr>
<tr>
<td>Red</td>
<td>Ball and Field Feature were seen</td>
</tr>
</tbody>
</table>

Torso (Chest Button)

<table>
<thead>
<tr>
<th>Color</th>
<th>State</th>
</tr>
</thead>
<tbody>
<tr>
<td>Off</td>
<td>Initial, finished</td>
</tr>
<tr>
<td>Blue</td>
<td>Ready</td>
</tr>
<tr>
<td>Green</td>
<td>Playing</td>
</tr>
<tr>
<td>Yellow</td>
<td>Set</td>
</tr>
<tr>
<td>Red</td>
<td>Penalized</td>
</tr>
</tbody>
</table>

Feet

- The left foot shows the team color. For instance, if the team is currently the red team, the color of the LED is red. If the team color is black, this LED will be switched off.

- The right foot shows whether the team has kick-off or not. If the team has kick-off, the color of the LED is white, otherwise it is switched off. In case of a penalty shootout, the color of the LED is green, when the robot is the penalty taker, and yellow if it is the goal keeper.

Ears

- The right ear shows the battery level of the robot. For each 10% of battery loss, an LED is switched off.

- The left ear shows the number of players connected through the wireless. For each connected player, two LEDs are switched on (upper left, lower left, lower right, upper right). Without game controller access, two LEDs are blinking, the rest are off.
Head (if available)

If the robot is charging, the LEDs will be turned on and off one after another. Due to the circular arrangement of the LEDs, this looks like a rotating beam of light. If these LEDs are not available and the robot is charging, the right eye will be colored magenta.
Chapter 7

Proprioception (Sensing)

The NAO has an inertial measurement unit (IMU) with three acceleration sensors (for the $x$-, $y$-, and $z$-direction) and two or three gyroscopes (for the rotation around the $x$-, $y$- and in V5 NAOs also the $z$-axes). By means of these measurements, the IMU board calculates rough approximations of the robot’s torso angles relative to the ground, which are provided together with the other sensor data. Furthermore, the NAO has a sensor for the battery level, eight force sensing resistors on the feet, two ultrasound sensors with different measurement modes, and sensors for the load, temperature, and angle of each joint. We currently do not use the data from the ultrasound sensors.

In the process Motion, the NaoProvider receives all sensor readings from the NaoQi module libbhuman (cf. Sect. 3.1), adds a configured calibration bias for each sensor, and provides them as FsrSensorData, InertialSensorData, JointSensorData, KeyStates, and SystemSensorData. The

![Figure 7.1: Sensing module graph. Sensing modules are depicted as yellow rectangles. All representations they provide are shown as blue ellipses. The representations they require are shown as ellipses colored in either yellow if they are provided by other Motion modules and orange if they are received from the process Cognition.](image)

111
JointAngles encapsulates the JointSensorData for the whole system, which is used to set the WristJoints angles in case they are not supported by the NAO version.

The RobotModel (cf. Sect. 7.2) provides the positions of the robot’s limbs and the GroundContactState (cf. Sect. 7.1), which is the information whether the robot’s feet are touching the ground. Based on the JointAngles, the InertialData, the RobotModel and the currently executed motion, it is possible to calculate the TorsoMatrix (cf. Sect. 7.4), which describes a transformation from the ground to an origin point within the torso of the NAO. The representation InertiaData is also considered to detect whether a robot has fallen down by the module FallDownStateDetector, which provides the FallDownState (cf. Sect. 7.5). Figure 7.1 shows all Sensing modules and the provided representations.

7.1 Ground Contact Recognition

Since it may happen during official soccer matches that a robot is manually placed or it gets lifted because of a penalty, it is useful for several reasons (e.g. localization and behavior) to know whether it is standing or walking on the ground or not. It also comes in handy when the robot stops moving automatically after it got lifted, since it is much easier to place a standing robot on the field instead of a moving one. The GroundContactState, which is actually a simple Boolean value indicating whether there is at least one foot on the ground, should not be confused with the FallDownState that indicates whether the robot is in a horizontal position.

Due to the inaccuracy and failure of many force sensing resistors in the past, these are currently not the basis for the ground contact detection. The current GroundContactDetector measures the noise in the accelerometers and gyroscopes and calculates the average values over 60 frames. If all values exceed a preset threshold, the GroundContactState becomes false. The assumption for this method is that movements that cause high values in the accelerometers and gyroscopes tend to be the ones that cause losing ground contact like lifting the robot, or the robot falling down.

The force sensing resistors of newer NAO robots seem to be much more reliable and might be used in the near future.

7.2 Robot Model Generation

The RobotModel is a simplified representation of the robot. It provides the positions and rotations of the robot’s limbs relative to its torso as well as the position of the robot’s center of mass (CoM). All limbs are represented by homogeneous transformation matrices (Pose3D) whereby each limb maps to a joint. By considering the measured joint angles of the representation JointAngles, the calculation of each limb is ensured by the consecutive computations of the kinematic chains. Similar to the inverse kinematic (cf. Sect.8.3.3), the implementation is customized for the NAO, i.e., the kinematic chains are not described by a general purpose convention such as Denavit-Hartenberg parameters to save computation time.

The CoM is computed by equation (7.1) with \( n \) = number of limbs, \( \vec{r}_i \) = position of the center
of mass of the \(i\)-th limb relative to the torso, and \(m_i\) = the mass of the \(i\)-th limb.

\[
\vec{r}_{com} = \frac{\sum_{i=1}^{n} \vec{r}_i m_i}{\sum_{i=1}^{n} m_i} \tag{7.1}
\]

For each limb, \(\vec{r}_i\) is calculated by considering the representation RobotDimensions and the position of its CoM (relative to the limb origin). The limb CoM positions and masses are provided in the representation MassCalibration. They can be configured in the file massCalibration.cfg. The values used were taken from the NAO documentation by SoftBank Robotics.

### 7.3 Inertia Sensor Data Filtering

The InertialDataFilter module determines the orientation of the robot’s torso relative to the ground. Therefore, the calibrated IMU sensor readings (InertialSensorData) and the measured stance of the robot (RobotModel) are processed using an Unscented Kalman filter (UKF) \[13\].

A three-dimensional rotation matrix, which represents the orientation of the robot’s torso, is used as the estimated state in the Kalman filtering process. In each cycle, the rotation of the torso is predicted by adding an additional rotation to the estimated state. The additional rotation is computed using the readings from the gyroscope sensor. After that, two different cases are handled. In the case that it is obvious that the feet of the robot are not evenly resting on the ground (when the robots fell down or is currently falling), the acceleration sensors are used to update the estimated rotation of torso, in order to avoid accumulating an error caused by slightly miscalibrated or noisy sensors. In the other case, an orientation of the robot’s torso computed from the RobotModel, assuming that at least one foot of the robot rests evenly on the ground, is used to update the estimate. The resulting orientation is provided in the representation InertialData.

### 7.4 Torso Matrix

The TorsoMatrix describes the three-dimensional transformation from the projection of the middle of both feet on the ground up to the center of hip within the robot torso. Additionally, the TorsoMatrix contains the alteration of the position of the center of hip including the odometry. Hence, the TorsoMatrix is used by the WalkingEngine for estimating the odometry offset. The CameraMatrix within the Cognition process is computed based on the TorsoMatrix.

In order to calculate the TorsoMatrix, the vector of each foot from ground to the torso (\(f_l\) and \(f_r\)) is calculated by rotating the vector from the torso to each foot (\(t_l\) and \(t_r\)). This can be calculated by the kinematic chains, according to the estimated rotation (cf. Sect. 7.3). The estimated rotation is represented as rotation matrix \(R\).

\[
f_l = -R \cdot t_l \tag{7.2}
\]
\[
f_r = -R \cdot t_r \tag{7.3}
\]

The next step is to calculate the span \(s\) between both feet (from left to right) by using \(f_l\) and \(f_r\):
Now, it is possible to calculate the translation part of the torso matrix \( p_{tm} \) by using the longer leg. The rotation part is already known since it is equal to \( R \).

\[
p_{tm} = \begin{cases} 
 \frac{s}{2} + f_l & \text{if } (f_l)_z > (f_r)_z \\
 -\frac{s}{2} + f_r & \text{otherwise}
\end{cases}
\] (7.5)

The change of the position of the center of hip is determined by using the inverted torso matrix of the previous frame and concatenating the odometry offset. The odometry offset is calculated by using the change of the span \( s \) between both feet and the change of the ground foot’s rotation as well as the new torso matrix.

### 7.5 Detecting a Fall

Although we try our best at keeping our robots upright, it occasionally happens that one of them falls over. In such a case, it is helpful to decrease the stiffness in the joints and bring the head into a safe position to protect the robot’s hardware from unnecessary damage. The task of detecting such a situation is realized by the \texttt{FallDownStateDetector}, which provides the \texttt{FallDownState}. One major requirement for the detection system is the certainty that the robot is going to fall and there is no way to avoid it. If the system makes a false positive, the robot might get into trouble and cause an unnecessary collapse by itself. For that reason, fall detection is split up into two phases. In the first one, the robot detects if it is staggering. In a situation like this, we bring the head into a safe position, which can be done without risking its stability. If the robot exceeds another given threshold in its tilt angle, the system enters the second phase which will power down the joints to a low hardness. It has been shown that this is a better approach than completely switching them off, because deactivated joints tend to gain too much momentum before hitting the floor. To determine the robot’s tilt angle, the \texttt{FallDownStateDetector} utilizes the \texttt{InertialData}. Still, there are some exceptions, when the \texttt{FallDownState} is undefined and thus the safety procedures mentioned above are not triggered, in particular, if the keeper is diving, as these motions switch off the joints on their own.

To start an appropriate get-up motion after a fall, the \texttt{FallDownStateDetector} determines, whether the robot is lying on its front, its back, its left side, or its right side. The latter two cases appear to be highly unlikely, but are not impossible. Note that we distinguish cases, in which the robot falls sideward first, then continuing to fall on its back or front. We use this information to correct the robot’s odometry data, in order to speed up regaining knowledge of its rotation on the field after standing up again.

### 7.6 Arm Contact Recognition

Robots should detect whether they touch obstacles with their arms in order to improve close-range obstacle avoidance. When getting caught in an obstacle with an arm, there is a good chance that the robot gets turned around, causing its odometry data to get erroneous. This, in turn, affects the self-localization (cf. Sect. 5.1).

In order to improve the precision as well as the reactivity, the \texttt{ArmContactModelProvider} is executed within the \texttt{Motion} process. This enables the module to query the required joint angles 100 times per second. The difference of the intended and actual position of the shoulder joint
per frame is calculated and also buffered over several frames. This allows to calculate an average error. Each time this error exceeds a certain threshold, an arm contact is reported. Using this method, small errors caused by arm motions in combination with low stiffness can be smoothed. Hence, we are able to increase the detection accuracy. Due to the joint slackness, the arm may never reach certain positions. This might lead to prolonged erroneous measurements. Therefore, a malfunction threshold has been introduced. Whenever an arm contact continues for longer than this threshold, all further arm contacts will be ignored until no contact is measured.

The current implementation provides several features that are used to gather information while playing. For instance, we are using the error value to determine in which direction an arm is being pushed. Thereby, the average error is converted into compass directions relative to the robot. Additionally, the ArmContactModelProvider keeps track of the time and duration of the current arm contact. This information may be used to improve the behavior.

In order to detect arm contacts, the first step of our implementation is to calculate the difference between the measured and commanded shoulder joint angles. Since we noticed that there is a small delay between receiving new measured joint positions and commanding them, we do not compare the commanded and actual position of one shoulder joint from one frame. Instead we are using the commanded position from $n$ frames earlier as well as the newest measured position.

In our approach, the joint position of one shoulder consists of two components: $x$ for pitch and $y$ for roll. Given the desired position $p$ and the current position $c$, the divergence $d$ is simply calculated as:

$$
\begin{align*}
  d.x &= c.x - p.x \\
  d.y &= c.y - p.y
\end{align*}
$$

In order to overcome errors caused by fast arm movements, we added a bonus factor $f$ that decreases the average error if the arm currently moves fast. The aim is to decrease the precision, i.e. to increase the detection threshold for fast movements in order to prevent false positives. The influence of the factor $f$ can be modified with the parameter speedBasedErrorReduction and is calculated as:

$$
f = \max\left(0, 1 - \frac{|\text{handSpeed}|}{\text{speedBasedErrorReduction}}\right)
$$

So for each arm, the divergence value $d_a$ actually being used is:

$$
d_a = d \cdot f
$$

As mentioned above, the push direction is determined from the calculated error of an arm shoulder joint. This error has two signed components $x$ and $y$ denoting the joint’s pitch and roll divergences. One component is only taken into account if its absolute value is greater than its contact threshold.

Table 7.1 shows how the signed components are converted into compass directions for the right arm. The compound directions NW, NE, SE, SW are constructed by simply combining the above rules if both components are greater than their thresholds, e.g. $x < 0 \land y < 0$ results into direction SE. The push direction of each arm is used to add an obstacle to the robot’s obstacle model, causing the robot to perform an evasion movement when it hits an obstacle with one of its arms.

\footnote{Currently, $n = 5$ delivers accurate results.}
Table 7.1: Converting signed error values into compass directions for the right shoulder and with $-y$ for the left

Arm contacts are also used to move the arm out of the way to avoid further interference with the obstacle (cf. Sect. 8.7). Please note that while arm motions from the ArmMotionEngine are active, no arm contacts are detected for that arm.
Chapter 8

Motion Control

The B-Human motion control system generates all kinds of motions needed to play soccer with a robot. They are split into the different types of motions walk, stand, kick, getUp and specialAction. These motions are generated by the corresponding whole body motion engines. Additionally there are engines for the head and specific arm motions. This split was done in order for the walking engine to be able to use the upcoming arm and head movements for its balancing algorithms. The walking motion and a corresponding stand are dynamically generated by the WalkingEngine (cf. Sect. 8.3). Some kicking motions are generated by the KickEngine [20] that models kicks by using Bézier splines and inverse kinematics. Getting up from a position lying on the ground is a special task that includes basic balancing and is done with the GetUpEngine (cf. Sect. 8.5). All other whole body motions are provided by the module SpecialActions in the form of sequences of static joint angle patterns (cf. Sect. 8.4). The head motion is generated by the HeadMotionEngine (cf. Sect. 8.6). The modules responsible for specific arm motions are the ArmKeyFrameEngine and PointAtEngine (cf. Sect. 8.7). All motion engines generate joint angles. Depending on the selected motions (cf. Sect. 8.1) these angles are combined into the JointRequest at the end of the process Motion (cf. Sect. 8.2). Figure 8.1 shows all motion control modules and the representations they provide.

8.1 Motion Selection

According to the representations MotionRequest and ArmMotionRequest, the MotionSelector calculates which motions to execute and how to interpolate between them, while switching from one to another. This information is then provided in the representations LegMotionSelection and ArmMotionSelection.

The MotionSelector determines which arm motion engine (ArmKeyFrameEngine and PointAtEngine) and whole-body motion engine (WalkingEngine, KickEngine, SpecialActions, or GetUpEngine) is to be executed and calculates interpolation ratios in order to switch between them. In doing so this module takes into account which motions and motion engines are demanded, which motions and motion engines are currently being executed, and whether those motions can be interrupted. Almost all motions can be interrupted but not in all states. The interruptibility of a motion is handled by the corresponding motion engine in order to ensure the motion is not interrupted in an unstable state.
8.2 Motion Combination

Motion combination is the task of merging the generated joint angles of all active motions into one, according to the ratios provided by the MotionSelector. This merging is done by a linear interpolation. Since joint angles for the arms have to be calculated before the walk motion can be generated, there are three modules performing this task:

The ArmMotionCombinator merges the joint angles for the arms, by taking into account the outputs of all modules producing such angles. Its output is the ArmJointRequest containing those angles and the ArmMotionInfo describing the executed motions per arm.

The LegMotionCombinator is responsible for merging the joint angles for the legs and outputs them in form of the LegJointRequest.

The MotionCombinator is the module executed last in the motion cycle. It combines the ArmJointRequest and LegJointRequest into the final JointRequest containing all joint angles for the NAO to be executed. In addition, it handles emergency situations, where the robot is falling by choosing appropriate angles and stiffnesses. Finally it fills the representations MotionInfo and OdometryData that contain data such as the current position in the walk cycle, whether the motion is stable, and the odometry position.
8.3 Walking

The walk and stand motions are generated by the WalkingEngine. This engine is split into two modules, called WalkingEngine and WalkingEngineLegProvider. The module WalkingEngine is the main component. It generates an abstract walk state for the legs in the form of the representation WalkingEngineState as well as joint angles for the arms as WalkArmRequest and StandArmRequest. Together with the abstract walk state and the ArmJointRequest generated by the ArmMotionCombinator, the WalkingEngineLegProvider then produces joint angles for the legs. These angles are held by the representations WalkLegRequest and StandLegRequest. The distinction between the two sets of joint angles (Stand*Request and Walk*Request) exists, because the MotionCombinators (cf. Sect. 8.2) expect two individual joint angle sets for standing and walking. However, both of the sets provided are identical to allow the WalkingEngine to handle the transition between standing and walking on its own.

The walking motions are generated based on the computationally inexpensive model of an inverted pendulum. Using this model, a trajectory for the robot’s center of mass can be planned where the supporting foot does not tip over one of the foot’s edges. The planned trajectory can then be transformed into a series of joint angles where the center of mass follows the planned trajectory. To compensate inaccuracies of the pendulum model and to react on external forces, the resulting motion of the center of mass is observed based on data of the TorsoMatrix and the RobotModel. If the observed motion does not correspond to the planned center of mass trajectory, the planned trajectory is slightly adjusted to ensure that further steps can be performed without causing the robot to fall over. This approach was released and described in further detail in [10] and [9].

Changes

- Compared to the releases of the previous years the walking engine underwent a heavy refactoring, mainly by moving the calculations of the linear inverted pendulum model onto separate classes called LIP and LIP3D.

- The acceleration of the walking Engine is now correctly influenced by the parameter speed-MaxChange.

- The in-walk kicks were replaced by a new engine (cf. 8.3.2).

8.3.1 Load Balancing

RoboCup games can take quite a while, in particular the play-offs, in which the clock is stopped when not in playing state. During such a game, the power consumption is very high, which has mainly two negative effects: On the one hand, weak rechargeable batteries can reach a critical
limit. On the other hand, the joints are getting very hot\(^1\), which results in weaker and less precise joint movements, and ultimately in the robots falling down more often.

For the past two years we have addressed the high power consumption problem with a special behavior (cf. 6.3.1) that is more optimized to avoid the robots have to continuous walk around the field. Instead, they try to stand still as often as (strategically) possible. But even when standing, the heat of the joints can quickly go up if the weight is not evenly distributed over both legs.

Therefore, while standing, our robots are continuously detecting the cumulative electrical currents for each leg over a short amount of time and then lean the torso over towards the side the leg which had the smaller power consumption so far. Thereby, the robots are able to distribute the load more equally and to keep the temperature from increasing while the robot is standing. The calculations are done in the StandBodyRotationProvider which is providing the StandBodyRotation.

As can be seen in Fig. 8.2, on a robot playing during the final of the RoboCup German Open 2015, the increase in joint temperature was always slowed down when the robot was standing.

\[\text{Figure 8.2: The temperature in } ^\circ\text{C of the hottest joint of B-Human’s robot “Kripke” in the second half of the final of the RoboCup German Open 2015. The robot also played the whole first half before. The time is shown as the number of vision frames until the end of the half, i.e. in } \frac{1}{60}\text{ secs. The vertical blue lines mark the begin and end of phases in which the robot was standing.}\]

8.3.2 In-Walk Kicks

Besides walking and standing, the WalkingEngine has also minor kicking capabilities. The tasks of walking and kicking are often treated separately, both solved by different approaches. In the presence of opponent robots, such a composition might waste precious time as certain transition phases between walking and kicking are necessary to ensure stability. A common sequence is to walk, stand, kick, stand, and walk again. Since direct transitions between walking and kicking modules are likely to let the robot stumble or fall over, the WalkingEngine is able to carry out simple kicks while walking.

At the moment, there are three kick types: forward, sidewaysInner, and turnOut. Those kicks are described individually by the config files located in Config/WalkKicks. Every kick is defined by its duration, step sizes before and during the kick, and key frames for the 3D position and rotation of the kicking foot. These key frames are later interpolated by cubic splines to

\[^1\text{The joint temperature is an estimation done by NAO’s operating system NAOqi and not an actual measurement.}\]
describe the actual trajectory of the foot, by overlaying the original trajectory of the swinging foot and thereby describing the actual kicking motion. In doing so the overlaying trajectory starts and stops at 0 in all dimensions both in position and velocity. Thus, the kick retains the start and end positions as well as the speeds of a normal step. The instability resulting from the higher momentum of the kick is compensated by the walk during the steps following the kick.

8.3.3 Inverse Kinematic

The inverse kinematics for the ankles of the NAO are a central component of the module WalkingEngine. In general, they are a handy tool for generating motions, but solving the inverse kinematics problem analytically for the NAO is not straightforward, because of two special circumstances:

- The axes of the hip yaw joints are rotated by 45 degrees.
- These joints are also mechanically connected among both legs, i.e., they are driven by a single servo motor.

The target of the feet is given as homogeneous transformation matrices, i.e., matrices containing the rotation and the translation of the foot in the coordinate system of the torso. In order to explain our solution we use the following convention: A transformation matrix that transforms a point \( p_A \) given in coordinates of coordinate system \( A \) to the same point \( p_B \) in coordinate system \( B \) is named \( A_{2B} \), so that \( p_B = A_{2B} \cdot p_A \). Hence the transformation matrix \( Foot_{2Torso} \) is given as input, which describes the foot position relative to the torso. The coordinate frames used are depicted in Fig. 8.3.

The position is given relative to the torso, i.e., more specifically relative to the center point between the intersection points of the axes of the hip joints. So first of all the position relative to the hip is needed\(^2\). It is a simple translation along the \( y \)-axis\(^3\)

\[
Foot_{2Hip} = Trans_y \left( \frac{l_{dist}}{2} \right) \cdot Foot_{2Torso} \tag{8.1}
\]

with \( l_{dist} \) = distance between legs. Now the first problem is solved by describing the position in a coordinate system rotated by 45 degrees, so that the axes of the hip joints can be seen as orthogonal. This is achieved by a rotation around the \( x \)-axis of the hip by 45 degrees or \( \frac{\pi}{4} \) radians.

\[
Foot_{2HipOrthogonal} = Rot_x \left( \frac{\pi}{4} \right) \cdot Foot_{2Hip} \tag{8.2}
\]

Because of the nature of the kinematic chain, this transformation is inverted. Then the translational part of the transformation is solely determined by the last three joints, by which means they can be computed directly.

\[
Hip_{Orthogonal2Foot} = Foot_{2HipOrthogonal}^{-1} \tag{8.3}
\]

The limbs of the leg and the knee form a triangle, in which an edge equals the length of the translation vector of \( Hip_{Orthogonal2Foot} \) (\( l_{trans} \)). Because all three edges of this triangle are known (the other two edges, the lengths of the limbs, are fix properties of the NAO) the angles

\(^2\)The computation is described for one leg and can be applied to the other leg as well.

\(^3\)The elementary homogeneous transformation matrices for rotation and translation are noted as \( Rot_{<axis>}(<\text{angle}) \) resp. \( Trans_{<axis>}(<\text{translation}) \).
of the triangle can be computed using the law of cosines (8.4). Knowing that the angle enclosed
by the limbs corresponds to the knee joint, that joint angle is computed by equation (8.5).

\[ c^2 = a^2 + b^2 - 2ab \cos \gamma \]  

\[ \gamma = \arccos \left( \frac{l_{\text{upperLeg}}^2 + l_{\text{lowerLeg}}^2 - l_{\text{trans}}^2}{2l_{\text{upperLeg}}l_{\text{lowerLeg}}} \right) \]  

(8.5)

Because \( \gamma \) represents an interior angle and the knee joint is being stretched in the zero-position,
the resulting angle is computed by

\[ \delta_{\text{knee}} = \pi - \gamma \]  

(8.6)

Additionally, the angle opposite to the upper leg has to be computed, because it corresponds to
the foot pitch joint:
8.3. WALKING

$$\delta_{footPitch1} = \arccos \frac{l_{lowerLeg}^2 + l_{trans}^2 - l_{upperLeg}^2}{2 \cdot l_{lowerLeg} \cdot l_{trans}}$$  \hspace{1cm} (8.7)

Now the foot pitch and roll joints combined with the triangle form a kind of pan-tilt-unit. Their joints can be computed from the translation vector using atan2.\(^4\)

$$\delta_{footPitch2} = \text{atan}2(x, \sqrt{y^2 + z^2})$$  \hspace{1cm} (8.8)

$$\delta_{footRoll} = \text{atan}2(y, z)$$  \hspace{1cm} (8.9)

where \(x, y, z\) are the components of the translation of Foot2HipOrthogonal. As the foot pitch angle is composed by two parts it is computed as the sum of its parts.

$$\delta_{footPitch} = \delta_{footPitch1} + \delta_{footPitch2}$$  \hspace{1cm} (8.10)

After the last three joints of the kinematic chain (viewed from the torso) are determined, the remaining three joints that form the hip can be computed. The joint angles can be extracted from the rotation matrix of the hip that can be computed by multiplications of transformation matrices. For this purpose another coordinate frame Thigh is introduced that is located at the end of the upper leg, viewed from the foot. The rotation matrix for extracting the joint angles is contained in HipOrthogonal2Thigh that can be computed by

$$\text{HipOrthogonal2Thigh} = \text{Thigh2Foot}^{-1} \cdot \text{HipOrthogonal2Foot}$$  \hspace{1cm} (8.11)

where Thigh2Foot can be computed by following the kinematic chain from foot to thigh.

$$\text{Thigh2Foot} = \text{Rot}_x(\delta_{footRoll}) \cdot \text{Rot}_y(\delta_{footPitch}) \cdot \text{Trans}_z(l_{lowerLeg}) \cdot \text{Rot}_y(\delta_{knee}) \cdot \text{Trans}_z(l_{upperLeg})$$  \hspace{1cm} (8.12)

To understand the computation of those joint angles, the rotation matrix produced by the known order of hip joints (yaw (\(z\)), roll (\(x\)), pitch (\(y\))) is constructed (the matrix is noted abbreviated, e.g. \(c_x\) means \(\cos \delta_x\)).

$$\text{Rot}_{\text{Hip}} = \text{Rot}_z(\delta_z) \cdot \text{Rot}_x(\delta_x) \cdot \text{Rot}_y(\delta_y) = \begin{pmatrix} c_y c_z - s_x s_y s_z & -c_x s_y & c_z s_y + c_y s_z s_x \\
-c_z s_y + c_x s_z & c_x c_z & c_y c_z s_y + s_z s_y s_x \\
-c_x s_y & c_z s_y & c_x c_y \end{pmatrix}$$  \hspace{1cm} (8.13)

The angle \(\delta_x\) can obviously be computed by \(\arcsin r_{21}\).\(^5\) The extraction of \(\delta_y\) and \(\delta_z\) is more complicated, they must be computed using two entries of the matrix, which can be easily seen by some transformation:

$$\begin{pmatrix} r_{01} \\ r_{11} \end{pmatrix} = \begin{pmatrix} \cos \delta_x \cdot \sin \delta_z & \sin \delta_x \\ \cos \delta_x \cdot \cos \delta_z \end{pmatrix} = \begin{pmatrix} \sin \delta_z \\ \cos \delta_z \end{pmatrix} = \tan \delta_z$$  \hspace{1cm} (8.14)

Now \(\delta_z\) and, using the same approach, \(\delta_y\) can be computed by

\(^4\)atan2(\(y, x\)) is defined as in the C standard library, returning the angle between the \(x\)-axis and the point \((x, y)\).

\(^5\)The first index, zero based, denotes the row, the second index denotes the column of the rotation matrix.
\[ \delta z = \delta_{\text{hipYaw}} = \text{atan2}(-r_{01}, r_{11}) \quad (8.15) \]

\[ \delta y = \delta_{\text{hipPitch}} = \text{atan2}(-r_{20}, r_{22}) \quad (8.16) \]

At last the rotation by 45 degrees (cf. eq. 8.2) has to be compensated in joint space.

\[ \delta_{\text{hipRoll}} = \delta_x - \frac{\pi}{4} \quad (8.17) \]

Now all joints are computed. This computation is done for both legs, assuming that there is an independent hip yaw joint for each leg.

The computation described above can lead to different resulting values for the hip yaw joints. From these two joint values a single resulting value is determined, in which the interface allows setting the ratio. This is necessary, because if the values differ, only one leg can realize the desired target. Normally, the support leg is supposed to reach the target position exactly. By applying this fixed hip joint angle the leg joints are computed again. In order to face the six parameters with the same number of degrees of freedom, a virtual foot yaw joint is introduced, which holds the positioning error provoked by the fixed hip joint angle. The decision to introduce a foot yaw joint was mainly taken because an error in this (virtual) joint has a low impact on the stability of the robot, whereas other joints (e.g. foot pitch or roll) have a huge impact on stability. The computation is almost the same as described above, except it is the other way around. The need to invert the calculation is caused by the fixed hip joint angle and the additional virtual foot joint, because the imagined pan-tilt-unit is now fixed at the hip and the universal joint is represented by the foot.

This approach can be realized without any numerical solution, which has the advantage of a constant and low computation time and a mathematically exact solution instead of an approximation.

### 8.4 Special Actions

Special actions are hardcoded motions that are provided by the module `SpecialActions`. By executing a special action, different target joint values are sent consecutively, allowing the robot to perform actions such as a goalie dive or the high stand posture. Those motions are defined in `.mof` files that are located in the folder `Config/mof`. A `.mof` file starts with the unique name of the special action, followed by the label `start`. The following lines represent sets of joint angles, separated by a whitespace. The order of the joints is as follows: head (pan, tilt), left arm (shoulder pitch/roll, elbow yaw/roll, wrist, hand), right arm (shoulder pitch/roll, elbow yaw/roll, wrist, hand), left leg (hip yaw-pitch/roll/pitch, knee pitch, ankle pitch/roll), and right leg (hip yaw-pitch/roll/pitch, knee pitch, ankle pitch/roll). A `*` does not change the angle of the joint (keeping, e.g., the joint angles set by the head motion engine), a `–` deactivates the joint. Each line ends with two more values. The first decides whether the target angles will be set immediately (the value is 0); forcing the robot to move its joints as fast as possible, or whether the angles will be reached by interpolating between the current and target angles (the value is 1). The time this interpolation takes is read from the last value in the line in milliseconds. If the values are not interpolated, the robot will set and hold the values for that amount of time instead.

6Ignored
It is also possible to change the stiffness of the joints during the execution of a special action, which can be useful, e.g., to achieve a stronger kick while not using the maximum stiffness as default. This is done by a line starting with the keyword *stiffness*, followed by a value between 0 and 100 for each joint (in the same order as for specifying actual joint angles). In the file *Config/stiffnessSettings.cfg* default values are specified. If only the stiffness of certain joints should be changed, the others can be set to ‘*’. This will cause those joints to use the default stiffness. At the end of one stiffness line the time has to be specified that it will take to reach the new stiffness values. This interpolation time runs in parallel to the interpolation time between target joint angles. In addition, the stiffness values defined will not be reached if another stiffness command is executed before the interpolation time has elapsed.

_Transitions_ are conditional statements. If the currently selected special action is equal to the first parameter, the special action given in the second parameter will be executed next, starting at the position of the label specified as last parameter. Note that the currently selected special action may differ from the currently executed one, because the execution costs time. Transitions allow defining constraints such as _to switch from A to B, C has to be executed first_. There is a wildcard condition _allMotions_ that is true for all currently selected special actions. Furthermore, there is a special action called _extern_ that allows leaving the module _SpecialActions_, e.g., to continue with walking. _extern.mof_ is also the entry point to the special action module. Therefore, all special actions must have an entry in that file to be executable. A special action is executed line by line, until a transition is found the condition of which is fulfilled. Hence, the last line of each _mof_ file contains an unconditional transition to _extern_.

An example of a special action:

```
motion_id = stand
label start
stiffness 20 20 60 60 60 60 60 60 60 60 90 90 90 90 90 90 90 90 90 50
* * 0 -50 -2 -40 0 0 0 -50 -2 -40 0 0 -6 -1 -43 92 -48 0 -6 -1 -43 92 -48 -1 1 100
transition allMotions extern start
```

To receive proper odometry data for special actions, they have to be manually set in the file *Config/specialActions.cfg*. It can be specified whether the robot moves at all during the execution of the special action, and if yes, how it has moved after completing the special action, or whether it moves continuously in a certain direction while executing the special action. It can also be specified whether the motion is stable, i.e., whether the camera position can be calculated correctly during the execution of the special action. Several modules in the process _Cognition_ will ignore new data while an unstable motion is executed to protect the world model from being impaired by unrealistic measurements.

### 8.5 Get Up Motion

Besides the complex motion engines (e.g., walking and kicking) there is also a smaller, task-specific engine called the _GetUpEngine_. As the name implies, this engine specializes in getting a lying robot back on its feet.

While using a key-frame-based approach, we included so called critical parts in order to verify certain body poses, e.g. if the robot’s torso is near an upright position. Whenever one of the critical parts is unsuccessfully passed, the motion engine stops the current execution and initiates a recovery movement in order to try again from the start. After two attempts, a hardware defect is assumed causing the engine to set the stiffness of all joints to zero. Thus, the robot stays lying down in order to prevent any more damage. Furthermore, a cry for help is initiated causing the robot to play the sound “help me”.

There is only one way to terminate the execution of this motion engine: get the robot into an upright position. This can either be done by a successful standing up movement or (after two unsuccessful attempts) manually by holding the robot upright as a referee would do following the fallen robot rule.

As we all know robot soccer is often very rough and even robots that are almost finished with a stand-up movement could be in distress. Thus, we included the possibility to enable a PID-controlled gyroscope based balancing (similar to the one in KickEngine by Müller et al. [20]) starting at a certain predefined part of the movement until it is finished, e.g. if the robot is standing upright again.

It is desirable that a motion engine is easy to use. For that reason, the last feature is that the engine is able to distinguish on its own whether the robot has fallen on its front, back, or not at all. Thereby, the torso rotation of the representation \texttt{InertiaData} is used to select the appropriate motion, e.g., standing up from lying on the front side or standing up from lying on the back side. In case that the robot has not fallen at all, the engine terminates immediately, i.e. flags itself as “leaving is possible” and initiates a stand. Furthermore, after each unsuccessful stand up attempt, the engine reviews the torso rotation in case an interfering robot has changed it.

Thus, if a fall is detected, one only needs to activate the engine by setting the \texttt{motion} to \texttt{getUp} in the representation \texttt{MotionRequest} and wait for the termination (i.e. the “leaving is possible” flag).

8.5.1 Modify Get Up Motions

After a couple of years, we finally needed to adjust our stand-up motions, which did not feel very comfortable. For that reason, we tried to improve that process by using special actions (cf. Sect. 8.4) during creation and convert them afterwards into \texttt{getUpEngine.cfg}. In doing so we first included \texttt{getUpEngineDummy.mof} as a dummy file that is later used to edit and execute the stand-up motion to change.

To load the data from an existing motion into the dummy file we use \texttt{set module:GetUpEngine:generateMofOf nameOfMotion; in simulator. This looks up the motion name in file getUpEngine.cfg and converts the data of that specific stand-up motion from the config-file into the file getUpEngineDummy.mof. Afterwards it can be written back to the file getUpEngine.cfg with \texttt{set module:GetUpEngine:generateMotionFromDummyMof nameOfMotion;}. Once you have generated the file getUpEngineDummy.mof you can use it as normal special action. Of course this includes the simulator command \texttt{mof} that compiles and copies special actions from your hard drive to the robot without restart.

As all conversions are done on file level, it is advisable to always convert in offline mode. Otherwise, in online mode, you have to look up the files on the robot’s memory drive instead of your own harddrive.

Since GetUpEngine motions are a bit different from normal special actions, we included some special comments to mark critical phases (“@critical), the phase the balance mechanism should be activated (“@balanceStartLine), and how the odometry changes after the whole motion is done (“@odometryOffset x y rotation). These special comments are used if the motion is converted back into the file getUpEngine.cfg and implemented by the GetUpEngine. If the motion is executed by the module SpecialActions, the special comments are ignored. Thus, balancing and upright position verification are not enabled.

At last the GetUpEngine is now able to change the joint stiffness during execution. This can be marked during the creation process as special action with the \texttt{stiffness} command. As special
actions and the GetUpEngine are functioning different from each other, there is a compatibility issue regarding the stiffness. In the GetUpEngine the stiffness is applied at the start of each interpolation process in a single joints key-frame. In the module SpecialActions the stiffness is interpolated in parallel to the joints key-frame interpolations. Thus, the stiffness interpolation time can differ from the joints interpolation time and stretch over more than a single joints key-frame. Consequently, comparable behaviors of both engines are only reached if you set the interpolation time of stiffness key-frames to 0.

The following example shows how to use the special comments if a motion is created as special action:

```plaintext
motion_id = getUpEngineDummy
label start
   stiffness 20 20 60 60 60 60 60 60 60 60 60 60 60 60 60 60 60 60 60 60 60 60
         0 0 99 74 -5 -80 -90 0 99 -74 5 80 90 0 0 0 0 0 60 0 0 0 0 0 60 0 1 500
   stiffness 30 30 90 90 90 90 90 90 90 90 90 90 90 90 90 90 90 90 90 90
         0 0 99 74 -5 -80 -90 0 99 -74 5 80 90 0 0 0 0 0 60 0 0 0 0 0 60 0 1 500

   "@critical (upright check at start of interpolation to next line)
         0 0 99 74 -5 -80 -90 0 99 -74 5 80 90 0 0 0 0 0 60 0 0 0 0 0 60 0 1 500
   "@balanceStartLine (turn on balance at start of interpolation to next line)
         0 0 99 74 -5 -80 -90 0 99 -74 5 80 90 0 0 0 0 0 60 0 0 0 0 0 60 0 1 500

label repeat
         0 0 99 74 -5 -80 -90 0 99 -74 5 80 90 0 0 0 0 0 60 0 0 0 0 0 60 0 1 500

   "@odometryOffset 230.00 -40.00 -45.00

transition getUpEngineDummy getUpEngineDummy repeat
transition allMotions extern start
```

## 8.6 Head Motions

Besides the motion of the body (arms and legs), the head motion is handled separately. This task is encapsulated within the separate module HeadMotionEngine for two reasons. The first reason is that the modules WalkingEngine and KickEngine manipulate the center of mass. Therefore these modules need to consider the mass distribution of the head before its execution in order to compensate for head movements.

The other reason is that the module smoothens the head movement by limiting the speed as well as the maximum angular acceleration of the head. In encapsulating this, the other motion modules do not have to concern themselves with this.

The HeadMotionEngine takes the representation HeadAngleRequest generated by the CameraControlEngine (cf. Sect. 4.1.1) and produces the representation HeadJointRequest.

## 8.7 Arm Motions

In addition to the arm motions that are calculated combined with the leg motions (e.g. during a special action or a kick) we also use two motion engines that provide only output for the arm joints: The PointAtEngine to point and the ArmKeyFrameEngine to follow key frames. Both motions will be selected only during a walking or standing (leg) motion.
8.7.1 Pointing at Something

The functionality to point at something with an outstretched arm was firstly used in this year’s No Wi-Fi Challenge (cf. Sect. 9.4) and Drop-In Competition’s gestures (cf. Sect. 9.1.2).

A geometric inverse kinematic is used to find the target shoulder joint angles. All other joints are set to zero\(^7\). The result is shown in Fig. 8.4. To ensure a smooth motion a simple cubic easing out function over the target joint angles is used. No output will be generated if the requested point is (roughly) too close to the own body or to the shoulder joints or if the requested point is located clearly behind the robot or in the workspace of the robot’s other arm to prevent a self-collision.

A point-at motion can be triggered by filling out the attribute `pointToPointAt` of the `ArmMotionRequest`. This is done by the behavior output options `PointAtWithArm` or `PointAt`. The former lets you select the pointing arm while the latter will choose it for you. In any case you have to specify the point you want to point at as three-dimensional coordinates relative to the robot’s origin.

8.7.2 Key Frame Motion

Although we are mainly using this feature to move the arms behind the robot’s back to prevent (further) contact with other robots, it generally supports any desired arm key frame motion. A motion is defined by a set of states, which consist of all target angles of an arm (similar to a special action, cf. Sect. 8.4). Upon executing an arm motion, the motion engine interpolates intermediate angles between two states to provide a smooth motion. When reaching the final state, the arm remains there until the engine gets a new request. The `ArmKeyFrameEngine` performs all the calculations and interpolations for the desired arm motion.

\(^7\)Plus/minus two degrees for the elbow roll joints because of the restricted robot’s kinematics.
Arm motions are configured in the file `armKeyFrameEngine.cfg`. It defines the sets of targets as well as some other configuration entries:

**actionDelay:** Delay (in ms) after which an arm motion can be triggered again by an arm contact.

**targetTime:** Time (in Motion frames) after which an arm should be moved back to its default position (applies only for motions triggered by arm contact).

**allMotions:** Array of different arm motions. The entry with the id `useDefault` is a native motion, which should neither be modified nor should it get another index within the array.

Despite the available motions, you may add further ones to the file following the same structure under a new `id`. The new `id` has to be appended to the end of the enumeration `ArmKeyFrameId` within the representation `ArmKeyFrameRequest`.

```
// [..] config entries left out
allMotions = [

  // [..] native motions left out

  { id = sampleArmMotion;
    states = [
      { angles=[-2.05, 0.2, 0.174, -0.2, -90 deg, 0 deg]; stiffness = [90, 60, 80, 90, 90, 90]; steps = 80; },
      { angles=[-2.11, -0.33, 0.4, -0.32, -90 deg, 1]; stiffness = [90, 60, 80, 90, 90, 90]; steps = 80; }
    ];
  }
];
```

Listing 8.1: Example of an arm motion definition

It is important to know that all motions will be only declared for the left arm. Inside the `ArmKeyFrameEngine`, the motion will then be correctly mirrored for the right arm. The angles respectively stiffness values of each state correspond in their order of occurrence to: ShoulderPitch, ShoulderRoll, ElbowYaw, ElbowRoll, WristYaw, and Hand. For each stiffness entry, the special value $-1$ may be used, which indicates the use of the default stiffness value for that joint as specified in the file `stiffnessSettings.cfg`. The attribute `step` specifies how many Motion frames should be used to interpolate intermediate angles between two entries in the array `states`.

Arm motions can be triggered by filling out the representation `ArmKeyFrameRequest`. This is done by the behavior output options `KeyFrameArms`, `KeyFrameLeftArm`, or `KeyFrameRightArm`. The first will start the motion selected for both arms. The representation mentioned has two attributes for each arm describing the motion to execute:

**motion:** The id of the motion to execute. `useDefault` means to use the default arm motions produced by the `WalkingEngine`, `reverse` means that you request the reverse motion of the last executed (key frame) motion.

**fast:** If set to `true`, interpolation will be turned off for this motion.
Chapter 9

Sub-Competitions and Technical Challenges

In addition to the main indoor soccer competition, B-Human participated in all other competitions that were held in 2016, i.e. the Outdoor Competition, the Drop-In Competition, as well as in both technical challenges, i.e. the No Wi-Fi Challenge and the Outdoor Challenge. For these competitions, most modules that have been described in the previous chapters can be used, too. However, each competition requires certain adaptions and/or the implementation of a few new software components. This chapter describes these specific developments.

9.1 Drop-In Competition

In the Drop-in Player Competition, robots from different teams have to play together. They have a common goal, namely to win their matches, but also compete with each other, because being seen by the human judges as both active and cooperative counts for the overall score. The main challenge of the Drop-in Player Competition is that all robots of a team run different software. Therefore, it is not quite clear for each robot what to expect from its teammates. Although the league defines a standard message that has to be exchanged among the team members, it is not guaranteed that all robots fill in all the fields with correct information. There is always the risk that a player gets a negative scoring for behaving strangely, just because it was fed wrong information by its teammates.

Since its introduction, we have always participated successfully in the Drop-In Player Competition. For these games, our major adaptions concern the robot behavior as well as the interpretation of the information sent by our teammates.

9.1.1 Reliability of Teammates

As some information given by other players might be imprecise or wrong, the reliabilities of all teammates are estimated during a game. In previous years, these reliabilities were used by the behavior to decide whether we try to carry out some actions all by ourselves or to support other robots and let them, for instance, play the ball. In 2016, we changed our implementation such that all messages from teammates that are currently not considered as reliable become discarded completely. Thereby, missing checks of the reliability – of which we discovered quite a few – cannot introduce unreliable information into our system anymore.

The reliability estimation is done by continuously analyzing the consistency, completeness, and
9.1. DROP-IN COMPETITION

plausibility of the standard messages received. This analysis consists of a number of different
tests. Currently, only the content of the most important elements is considered: the ball position
and the robot pose.

In a first step, it is checked, whether the robot pose coordinates have been set to valid numbers
at all. This is done by buffering the values and checking, whether the value ranges exceed a
certain threshold. This approach allows to exclude teammates that do not set this field (i.e.
leaving the default value) as well as teammates that – for whatever reason – use meters instead
of millimeters as length unit. In addition, to exclude teammates that use degrees instead of
radians, the value for the robot’s rotation has to be within the interval \([-\pi, \pi]\). Although
forbidden by the current rules, during the RoboCup 2016 Drop-In Competition, we experienced
a case in which a teammate sent a message with invalid floating point data such as Infinity
or Not a Number. Thus, we implemented several additional checks for all floating point fields. If
all these numerical and plausibility checks have been passed, the communicated robot position
as well as the ball position must be on the field.

Finally, it is checked, whether communicated ball perceptions are compatible to our own per-
ceptions, i.e. if the difference (in absolute field coordinates) between both positions is below
are given threshold. Only if this check is successful multiple times, the other robot has reached
reliability status and is treated by the behavior similar to a teammate in a normal game. This
state lasts for at least several seconds. If both robots see the ball at the same place many times,
the state length might even be extended to a few minutes.

Regardless of the teammates’ reliability, the team-wide localization symmetry resetting (cf.
Sect. 5.1.3) is disabled in all Drop-In games. As these games are not as tough as many normal
games, it is less likely that a robot loses its orientation. In addition, we only use NAO V5 robots
in Drop-In games, as their z-axis gyroscopes prevent most orientation problems.

9.1.2 Behavior

Major parts of the robot behavior are identical to our behavior for the normal competition.
However, the role selection is limited to striker and supporting behaviors. As we believe that
our robot can better contribute to the overall team performance when being a field player, the
goalkeeper role is avoided. For the role selection, we actually would accept suggestions by our
teammates, if a majority of the trusted teammates agrees on a role for our robot.

As passes are rewarded with positive scores, our drop-in player tries to pass to teammates
that report an adequate target position. For normal games, these passes are not activated.
In addition, the supporting behaviors actively take promising positions and wait for passes by
our teammates. When playing normal games, the decision whether or not to play the ball
is completely based on the communication with the teammates. However, in drop-in games
there might be ball-playing teammates that are not trustworthy enough to be considered in
our decision making. This is why we additionally consider perceptions of robots: if the vision
system reports a teammate close to the ball, we do not try to play the ball and thereby avoid
any negative scores for bad team play.

One new element that we have introduced for the 2016 competition are gestures. In general, the
audience and the judges cannot know, if some action is intended by the robot or just happened
accidentally. Therefore, we implemented two gestures: waving and pointing. Whenever we play
a pass to a teammate or decided to let a teammate play the ball, the robot is pointing towards
that robot to express its intention in a comprehensible manner. If the robot positions to receive
a pass from a teammate, it raises its arm and waves to indicate its waiting status. For the arm
motions, the approaches described in Sect. 8.7 have been used.
9.1.3 Analysis and Results

We won the 2016 Drop-In Player Competition and thus the Most Valuable Player award with the maximum possible score of 300 points [3]. In contrast to the 2015 competition – in which we showed a reasonable performance but lost many games – we won all six games. Many thanks to our teammates who contributed to this success by scoring goals and helping to defend our goal!

Two thirds of the overall score are based on the average judge rating. We received the full 200 judge points as our average judge score per game (18.1 points) was the highest of all teams. We do not know the detailed scores by any individual judge but after having seen the games and given the scoring rules defined in [4], one can assume the following: In most games, our robot had a lot of ball contacts and scored multiple goals. Hence, many scores might involve an exceptional game participation, which is awarded with 10 points. In addition, an average positioning (worth 5 points) combined with a few individual positive team play actions (+1 – +4 points) might result in the score that we achieved. Personal communication with judges revealed that our team play gestures have been noticed and appreciated and thus probably contributed to our success.

9.2 Outdoor Competition

We successfully applied for the participation in the Outdoor Competition and thus played our first matches on artificial lawn and with natural lighting (cf. Fig. 9.1) instead of playing the first preliminary round of the normal indoor competition. Except for some adaptions to our vision system and a lot of tuning of our robots’ walking parameters, we used the same code as for the indoor competition.

9.2.1 Vision Adptions

Most of this year’s vision system changes, especially using the YHS color space for color classification (cf. Sect. 4.1.4), were made in order to be able to play under the natural lighting conditions of the outdoor competition. Because of these adjustments, we were able to calibrate our robots for the outdoor competition the same way as during the normal indoor tournament, achieving similar results except for circumstances with very sharp shadow edges or very bright light. However, even in these cases we made our vision still work well enough by using the automatic exposure mechanism of the camera driver. It turned out that for the automatic exposure, the adaptive weighted auto exposure mode for lowlights combined with a low brightness setting provided good results for both color classification and ball detection, which relies on clear edges, throughout the outdoor competition.

9.2.2 Analysis and Results

Overall, our team was able to play under outdoor conditions in a reasonable manner. In five matches, we scored 18 goals and received only one goal. Unfortunately, the goal against us was scored in the final, which we lost 0:1 against the Nao Devils.

The lighting conditions only caused minor perception problems. Hence, localization and ball tracking worked almost as good as in the indoor competition. However, making our robots walk robustly on the artificial lawn turned out to be a major problem. Several robots that played decently on the indoor carpet were not able to walk at all, others tumbled a lot. As future
9.3 OUTDOOR CHALLENGE

Figure 9.1: Outdoor Competition final between B-Human and Nao Devils.

competitions will be played on a similar, we will definitely need a new walk as the current one cannot be considered as competitive to some others.

9.3 Outdoor Challenge

As we also participated in the Outdoor Competition, no additional preparations regarding vision and walking have been necessary for this challenge. We only adapted the robots’ behaviors to increase their chances of receiving the maximum score for an attempt, i.e. shot at the goal by the striker and execution of a proper blocking motion by the goalkeeper. Hence, the goalkeeper was programmed to keep standing and looking for the ball until the ball is eventually rolling towards the goal. The striker starts each attempt with a localization sequence. After it has seen enough features to be sure about the current pose on the field, it performs short kicks towards the goal. It will try to score not before the ball has reached a position that allows a shot that is likely to enter the goal.

Due to an improper robot configuration in the first round (making the striker unable to localize) and bad luck in the second round (the shot towards the goal stopped before it reached the goal), our robots performed worse than expected. Most test runs scored more points. Overall, the score was still good enough to reach a shared second place in this challenge.
9.4 No Wi-Fi Challenge

The standard communication channel for the robots is Wi-Fi. The task for this challenge was to implement a channel that is not depending on that infrastructure. During the challenge, two robots are placed on the opposite sides of the field, facing each other. In the first round, one robot has to transmit a location on the field to the other robot which has to point at it with his arm (cf. Sect. 8.7.1). In the second round, as much as possible arbitrary binary data has to be transmitted.

9.4.1 Evaluated Communication Approaches

As the rules for this challenge do not require the use of any specific communication approach, we evaluated different ways of transmitting data, such as sound, infrared light, and the recognition of robot’s eye LEDs. Infrared transmitters and receivers are part of the NAO robots. They allow the transmission of decent amounts of data, but unfortunately not over the distances required in this challenge. One alternative approach was to change the transmitting robot’s eye colors in a fast manner and to perceive these changes with the camera of the receiving robot. This method seems to work in principle but was not finished early enough for thorough testing and tuning.

Eventually, we used sound communication in this challenge. We developed two different approaches in parallel: one sophisticated approach that had the potential of transmitting a high number of bytes within a short amount of time as well as one basic backup approach that was very slow but very robust.

The sophisticated approach was based on BPSK (binary phase-shift keying) modulation. It works by detecting the phase of a signal and deriving the transmitted symbols out of it. Until the RoboCup, the modulation, demodulation, basic filtering, and a package abstraction layer for data transparency were implemented but further filtering was necessary to correctly demodulate noisy signals. Unfortunately, due to the complex nature of digital signal processing, these filtering techniques did not work correctly when the challenge took place. It is impossible to estimate the maximum bitrate of this unsuccessful approach over a noisy channel without adequate on-site testing, but in our software testing environment, which added noise around the carrier frequency, it worked with over 100 bit/s.

The remainder of this section describes the approach that has been actually used in the challenge.

9.4.2 Transmitting and Detecting Sounds

The basic idea of our sound transmission is quite straightforward:

1. We defined a set of sounds, each of them represents a certain value.
2. Data is encoded and converted to a sequence of sounds that are played by the transmitting robot.
3. The receiving robot recognizes the sounds, converts them back to the assigned values, and decodes the data.

The sounds are normal audio files that are played just in the same way as other sounds. Each sound has a length of 300 ms and contains a generated signal of a certain frequency. We chose frequencies in the range $[1750, 4000]$ Hz as these do not occur often in surrounding sounds.
described in the following section, we use quaternary symbols and thus require four different sounds\(^1\). This small number of sounds in combination with the length of each sound (which is required for a robust detection) indicates that this approach is only able to transmit a few bits per second. However, our major intention was to reliably score some points and thus we did not make any major efforts in tuning the bandwidth.

For the recognition of the different sounds, we use a Fast Fourier Transformation. In contrast to the whistle recognition described in Sect. 5.5.1, we do not perform a correlation with a previously recorded signal but just perform the transformation of the audio signal to the frequency domain by the FFT. As we know the frequencies that are used by the transmitting robot, their detection is now just a simple lookup, whether the amplitudes of these frequencies are higher than a preconfigured threshold. The sound recognition is implemented in the module `NoWirelessSoundRecognizer`.

### 9.4.3 Base 4 Hamming Encoding

One of our main goals in this challenge was to make sure that all data received and used was error free. Thus, we encoded all data using a linear error-correcting code called Hamming(8, 4)-code by Richard Hamming [11]. For each 4 bit of data this code adds 4 additional so-called parity bits. Using this additional redundant information one is able to detect errors in up to 3 of these 8 bits and to correct one error when decoding.

Since transmitting single bits was too slow, we decided to use a quaternary instead of a binary format for our data. As described by Leonid Bystrykh [2], it is possible to encode quaternary symbols in the same way as bits. Using this technique, we are able to transmit 8 bits of data using 8 quaternary symbols while preserving the high level of error detection and correction described above.

The decoding is implemented in the `HammingSoundToDataConverter` and the encoding as well as the playing of the aforementioned sound files is implemented in the `HammingSoundTransmitter`.

### 9.4.4 Analysis and Results

When we decided to use the robust backup solution, we already expected that we would not be able to win this challenge as the method’s maximum possible data transmission rate is quite low. During the challenge at RoboCup 2016, we successfully transmitted data from the close positions in the first part as well as a few bytes in the second part. However, similar to the Outdoor Challenge, our final tests (a few minutes before the challenge was held) on the same field were more successful as we were also able to reliably transmit data from the far positions, too.

In the end, we reached the third place in this challenge, with a score close to the Nao Devils but far away from the HULKs [3], who demonstrated the by far best solution.

---

\(^1\)Due to implementation details, we actually used four pairs of sounds and thus eight different sounds in total.
Chapter 10

Tools

The following chapter describes B-Human’s simulator, SimRobot, as well as the B-Human User Shell (bush), which is used during games to deploy the code and the settings to several NAO robots at once. As the official GameController and the tools accompanying it were also developed by B-Human, they are also described at the end of this chapter.

10.1 SimRobot

The B-Human software package uses the physical robotics simulator SimRobot [16, 14] as front end for software development. The simulator is not only used for working with simulated robots, but it also functions as graphical user interface for replaying log files and connecting to actual robots via Ethernet or WiFi.

10.1.1 Architecture

Three dynamic libraries are created when SimRobot is built. These are SimulatedNao, SimRobotCore2 and SimRobotEditor (cf. Fig. 10.1). SimRobotCore2 is an enhancement of the previous simulation core. It is the most important part of the SimRobot application, because it models the robots and the environment, simulates sensor readings, and executes commands given by the controller or the user. The core is platform-independent and it is connected to a user interface and a controller via a well-defined interface.

The library SimulatedNao is in fact the controller that consists of the two projects SimulatedNao and Controller. SimulatedNao creates the code for the simulated robot. This code is linked together with the code created by the Controller project to the SimulatedNao library. In the scene files, this library is referenced by the controller attribute within the Scene element.

10.1.2 B-Human Toolbar

The B-Human toolbar is part of the general SimRobot toolbar which can be found at the top of the application window (see Fig. 10.2).

\footnote{The actual names of the libraries have platform-dependent prefixes and suffixes, i.e. .dll, .dylib, and lib .so.}
10.1 SIMROBOT

Figure 10.1: This figure shows the most important libraries of SimRobot (excluding foreign libraries). The rectangles represent the projects, which create the appropriate library. Dynamic libraries are represented by ellipses and the static one by a diamond. Note: The static library will be linked together with the SimulatedNao code. The result is the library SimulatedNao.

- Lets the robot stand up.
- Lets the robot sit down.
- Allows moving the robot’s head by hand.

10.1.3 Scene View

The scene view (cf. Fig. 10.3 right) appears if the scene is opened from the scene graph, e.g., by double-clicking on the entry RoboCup. The view can be rotated around two axes, zoomed, and it supports several mouse operations:

- Left-clicking an object allows dragging it to another position. Robots and the ball can be moved in that way.

Figure 10.2: This figure shows the three buttons from the BHToolBar.
• Left-clicking while pressing the Shift key allows rotating objects around their centers. The axes of the rotating objects can be selected first by pressing x, y or z key, before pressing and holding the Shift key.

• Select an active robot by double-clicking it. Robots are active if they are defined in the compound robots in the scene description file (cf. Sect. 10.1.5). Robot console commands are sent to the selected robot only (see also the command robot).

10.1.4 Information Views

In the simulator, information views are used to display debugging output such as debug drawings. Such output is generated by the robot control program, and it is sent to the simulator via message queues (Sect. 3.5). The views are interactively created using the console window, or they are defined in a script file. Since SimRobot is able to simulate more than a single robot, the views are instantiated separately for each robot. There are fifteen kinds of information views, which are structured here into the five categories cognition, behavior control, sensing, motion control, and general debugging support. All information views can be selected from the scene graph (cf. Fig. 10.3 left).

10.1.4.1 Cognition

Image Views

An image view (cf. left of Fig. 10.4) displays debug information in the coordinate system of the camera image. It is defined by an debug image, an optional flag for the image, an optional
switch for JPEG compression, an optional switch for segmentation and a name using the console command \texttt{vi}. More formally its syntax is defined as:

\begin{verbatim}
vi <debug image> [upper] [jpeg] [segmented] [name]
\end{verbatim}

The \textit{debug image} can be the regular camera image and any other image that can be sent using the \texttt{SEND\_DEBUG\_IMAGE} macro. The following \textit{debug image} are currently defined:

\begin{description}
\item[ColoredImage:] A color segmented image. Only contains the relevant colors.
\item[GrayscaledImage:] The Y channel of the image.
\item[SaturatedImage:] An image only containing the calculated saturation visualized as a grayscale image.
\item[HuedImage:] An image only containing the calculated hue visualized as a grayscale image.
\item[cnsImage:] A contrast normalized Sobel image which is a Sobel image containing information about the direction. The angles are visualized with different colors (cf. Fig. 4.11b).
\item[image:] The image provided by the camera.
\item[netThumbnail:] A downscaled compressed version of the image.
\item[BinaryDebugImage:] A binarized version of grayscale image. Provided by the module \texttt{AutomaticHeadPitchCalibrator}.
\item[imagePatches:] Image cutouts smaller than the full image, containing all channels (e.g. could be used to only use the information around specific points of interest).
\item[corrected:] Similar to \textit{image}, but without the rolling shutter effects.
\item[horizonAligned:] Similar to \textit{image}, but aligned to the horizon.
\item[none:] Displays an empty background.
\end{description}

The default is to show data based on the images taken by the lower camera. With \texttt{upper}, the upper camera is selected instead. The switch \texttt{jpeg} will cause the NAO to compress the images before sending them to SimRobot. This might be useful if SimRobot is connected with the NAO.
via WiFi. The switch segmented will cause the image view to classify each pixel and draw its color class instead of the pixel’s value itself.

The console command vid adds the debug drawings. For instance, the image view with the flag segmented is defined as:

```
vi image segmented
vid segmentedLower representation:LinesPercept:image
vid segmentedLower representation:BallPercept:image
vid segmentedLower representation:PlayersPercept:image
```

You can deactivate debug drawings by appending off to a vid command.

**Color Calibration View**

The colorCalibration view provides direct access to the parameter settings for all existing color classes. The parameters of a color class can be changed by moving the sliders. Color classes are defined in the YHS color space as described in Sect. 4.1.4.

The following buttons are added to the toolbar when the colorCalibration view is focused.

- ![Icon with two squares](image)
  - saves the local color configuration
- ![Icon with left and right arrows](image)
  - undoes the latest slider change
- ![Icon with orange square](image)
  - max black/white saturation
- ![Icon with black square](image)
  - Y threshold for black
- ![Icon with green square](image)
  - YHS for your own team’s color
- ![Icon with purple square](image)
  - YHS for the opponent team’s color
- ![Icon with four squares](image)
  - toggles the expand color mode
- ![Icon with three squares](image)
  - redoes a reverted slider change
- ![Icon with white square](image)
  - Y threshold for white
- ![Icon with green square](image)
  - YHS for green

Furthermore each button will show the current settings of the corresponding color class in the colorCalibration view. Green and both jersey colors use range selectors for the YHS color space. Black and White are defined through Y thresholds and a value for the maximum white/black saturation. Figure 10.5 shows an example calibration for green.
Color Space Views

Color space views visualize image information in 3-D (cf. Fig. 10.6). They can be rotated by clicking into them with the left mouse button and dragging the mouse afterwards. There are three kinds of color space views:

**Image Color Channel:** This view displays an image while using a certain color channel as height information (cf. Fig. 10.6 left).

**Image Color Space:** This view displays the distribution of all pixels of an image in a certain color space ($HIS$, $RGB$, or $YCbCr$). It can be displayed by selecting the entry *all* for a certain color space in the scene graph (cf. Fig. 10.6 right).

The two kinds of views have to be added manually for the camera image or any debug image. For instance, to add a set of views for the camera image under the name *raw*, the following command has to be executed:

```v3 image raw```

Field Views

A field view (cf. right of Fig. 10.4) displays information in the system of coordinates of the soccer field. The command to create and manipulate it is defined similar to the one for the image views. For instance, the view *worldState* is defined as:

```# field views
vf worldState
vfd worldState fieldLines
vfd worldState goalFrame
vfd worldState fieldPolygons
vfd worldState representation:RobotPose
# ...

# views relative to robot
vfd worldState origin:RobotPose
vfd worldState representation:BallModel:endPosition
# ...

# back to global coordinates
vfd worldState origin:Reset```
Figure 10.7: Option graph view

Please note that some drawings are relative to the robot rather than relative to the field. Therefore, special drawings exist (starting with origin: by convention) that change the system of coordinates for all drawings added afterwards, until the system of coordinates is changed again.

The field can be zoomed in or out by using the mouse wheel, touch gestures, or the page up/down buttons. It can also be dragged around with the left mouse button or by touch gestures. Double clicking the view resets it to its initial position and scale.

10.1.4.2 Behavior Control

Option Graph View

The option graph view (cf. Fig. 10.7) can be found under Docs in the scene graph. It is a static view that only displays a graph with all the options of a CABS behavior. It has the same name as the behavior the option graph of which it displays, i.e. currently BehaviorControl2015. The colors of the options visualize to which part of the behavior each option belongs:

- **Skills** are shown in yellow.
- **GameControl** options are shown in red.
- **HeadControl** options are shown in brown.
- **Output** options are shown in blue.
- **Tools** are shown in gray.
- **Roles** are shown in green.
- **Everything else** is shown in white.

The graph can be zoomed in or out by using the mouse wheel, touch gestures, or the page up/down buttons. It can also be dragged around with the left mouse button or by touch gestures.
10.1. SIMROBOT B-Human 2016

Figure 10.8: Behavior view, sensor data view and joint data view

Activation Graph View

The activation graph view (cf. left of Fig. 10.8) shows all the options and states that are currently active in the behavior.

10.1.4.3 Sensing

Sensor Data View

The sensor data view displays all the sensor data taken by the robot, e.g. accelerations, gyro measurements, pressure readings, and sonar readings (cf. middle view in Fig. 10.8). To display this information, the following debug requests must be sent:

- `dr representation: InertialSensorData`
- `dr representation: SystemSensorData`
- `dr representation: FsrSensorData`
- `dr representation: KeyStates`

Joint Data View

Similar to sensor data view the joint data view displays all the joint data taken by the robot, e.g. requested and measured joint angles, temperatures, and loads (cf. right view in Fig. 10.8). To display this information, the following debug requests must be sent:

- `dr representation: JointRequest`
- `dr representation: JointSensorData`

10.1.4.4 Motion Control

Kick View

The basic idea of the kick view shown in Figure 10.9 is to visualize and edit basic configurations of motions for the KickEngine described in [20]. In doing so the central element of this view is a 3-D robot model. Regardless of whether the controller is connected to a simulated or a real robot, this model represents the current robot posture.

Since the KickEngine organizes motions as a set of Bézier curves, the movement of the limbs can easily be visualized. Thereby the sets of curves of each limb are represented by combined cubic
Bézier curves. Those curves are attached to the 3-D robot model with the robot center as origin. They are painted into the three-dimensional space. Each curve is defined by equation 10.1:

\[ c(t) = \sum_{i=0}^{n} B_{i,n}(t)P_i \]  

(10.1)

To represent as many aspects as possible, the kick view has several sub views:

**3-D View:** In this view each combined curve of each limb is directly attached to the robot model and therefore painted into the 3-dimensional space. Since it is useful to observe the motion curves from different angles, the view angle can be rotated by clicking with the *left mouse button* into the free space and dragging it in one direction. In order to inspect more or less details of a motion, the view can also be zoomed in or out by using the *mouse wheel* or the *page up/down* buttons.

A motion configuration is not only visualized by this view, it is also editable. Thereby the user can click on one of the visualized control points (cf. Fig. 10.9 at *E*) and drag it to the desired position. In order to visualize the current dragging plane, a light green area (cf. Fig. 10.9 at *B*) is displayed during the dragging process. This area displays the mapping between the screen and the model coordinates and can be adjusted by using the *right mouse button* and choosing the desired axis configuration.

Another feature of this view is the ability to display unreachable parts of motion curves. Since a motion curve defines the movement of a single limb, an unreachable part is a set of points that cannot be traversed by the limb due to mechanic limitations. The unreachable parts will be clipped automatically and marked with orange color (cf. Fig. 10.9 at *C*).
1-D/2-D View: In some cases a movement only happens in the 2-dimensional or 1-dimensional space (for example: Raising a leg is a movement along the z-axis only). For that reason more detailed sub views are required. Those views can be displayed as an overlay to the 3-D view by using the context menu, which opens by clicking with the right mouse button and choosing Display 1D Views or Display 2D Views. This only works within the left area, where the control buttons are (cf. Fig. 10.9 left of D). By clicking with the right mouse button within the 3-D view, the context menu for choosing the drag plane appears. The second opportunity to display a sub view is by clicking at the BikeEdit entry in the menu. So the same menu, which appears as context menu, will be displayed.

Because clarity is important, only a single curve of a single phase of a single limb can be displayed at the same time. If a curve should be displayed in the detailed views, it has to be activated. This can be done by clicking on one of the attached control points.

The 2-D view (cf. Fig. 10.10) is divided into three sub views. Each of these sub views represents only two dimensions of the activated curve. The curve displayed in the sub views is defined by equation 10.1 with $P_i = \left(\frac{c_{x_i}}{c_{y_i}}\right)$, $P_i = \left(\frac{c_{y_i}}{c_{z_i}}\right)$ or $P_i = \left(\frac{c_{z_i}}{c_{x_i}}\right)$.

The 1-D sub views (cf. Fig. 10.10) are basically structured as the 2-D sub views. The difference is that each single sub view displays the relation between one dimension of the activated curve and the time $t$. That means that in equation 10.1 $P_i$ is defined as: $P_i = c_{x_i}$, $P_i = c_{y_i}$, or $P_i = c_{z_i}$.
Figure 10.11: The module view shows a part of the modules in the process Motion.

As in the 3-D view, the user has the possibility to edit a displayed curve directly in any sub view by drag and drop.

**Editor Sub View:** The purpose of this view is to constitute the connection between the real structure of the configuration files and the graphical interface. For that reason, this view is responsible for all file operations (for example open, close, and save). It represents loaded data in a tabbed view, where each phase is represented in a tab and the common parameters in another one.

By means of this view the user is able to change certain values directly without using drag and drop. Values directly changed will trigger a repainting of the 3-D view, and therefore, changes will be visualized immediately. This view also allows phases to be reordered by drag and drop, to add new phases, or to delete phases.

To save or load a motion the kick view has to be the active view, so some buttons to do this appearing at the tool bar (cf. Fig. 10.9 at $G$).

### 10.1.4.5 General Debugging Support

**Module Views**

Since all the information about the current module configuration can be requested from the robot control program, it is possible to generate a visual representation automatically. The graphs, such as the one that is shown in Figure 10.11, are generated by the program *dot* from the *Graphviz* package [7]. Modules are displayed as yellow rectangles and representations are shown as blue ellipses. Representations that are received from another process are displayed in orange and have a dashed border. If they are missing completely due to a wrong module configuration, both label and border are displayed in red. The modules of each process can either be displayed as a whole, or separated into the categories that were specified as the second parameter of the macro `MAKE_MODULE` when they were defined. There is a module view for the process Cognition and its categories `infrastructure, communication, perception, modeling, and behaviorControl`, and one for the process Motion and its categories `infrastructure, sensing, and motionControl`. 
10.1. SIMROBOT

Figure 10.12: Plot view and timing view

The module graph can be zoomed in or out by using the mouse wheel, touch gestures, or the page up/down buttons. It can also be dragged around with the left mouse button or by touch gestures.

Plot Views

Plot views allow plotting data sent from the robot control program through the macro PLOT (cf. Fig. 10.12 left). They keep a history of the values sent of a defined size. Several plots can be displayed in the same plot view in different colors. A plot view is defined by giving it a name using the console command vp and by adding plots to the view using the command vpd (cf. Sect. 10.1.6.3).

For instance, the view on the left side of Figure 10.12 was defined as:

```
vp orientationX 200 -1 1
vpd orientationX module:InertialDataProvider:internalOrientation:x
```

Timing View

The timing view displays statistics about all currently active stopwatches in a process (cf. Fig. 10.12 right). It shows the minimum, maximum, and average runtime of each stopwatch in milliseconds as well as the average frequency of the process. All statistics sum up the last 100 invocations of the stopwatch. Timing data is transferred to the PC using debug requests. By default the timing data is not sent to the PC. Execute the console command dr timing to activate the sending of timing data. Please note that time measurements are limited to full milliseconds, so the maximum and minimum execution durations will always be given in this precision. However, the average can be more precise.

Data View

SimRobot offers two console commands (get & set) to view or edit anything that the robot exposes using the MODIFY macro. While those commands are enough to occasionally change some variables, they can become quite annoying during heavy debugging sessions.

For this reason, we introduced a new dynamic data view. It displays modifiable content using a property browser (cf. Fig. 10.13 left). Property browsers are well suited for displaying hierarchical data and should be well known from various editors such as Microsoft Visual Studio or Eclipse.
Figure 10.13: The data view can be used to remotely modify data on the robot.

A new data view is constructed using the command \texttt{vd} in SimRobot. For example \texttt{vd representation:ArmContactModel} will create a new view displaying the \textit{ArmContactModel}. Data views can be found in the data category of the scene graph.

The data view automatically updates itself three times per second. Higher update rates are possible, but result in a much higher CPU usage.

To modify data, just click on the desired field and start editing. The view will stop updating itself as soon as you start editing a field. The editing process is finished either by pressing enter or by deselecting the field. By default, modifications will be sent to the robot immediately. This feature is called the auto-set mode. It can be turned off using the context menu (cf. Fig. 10.13 right). If the auto-set mode is disabled, data can be transmitted to the robot using the \texttt{send} item from the context menu.

Once the modifications are finished, the view will resume updating itself. However you may not notice this since modification freezes the data on the robot side. To reset the data, use the \textit{unchanged} item from the context menu. As a result, the data will be unfrozen on the robot side and you should see the data changing again.

\textbf{Log Player View}

The log player (cf. Fig. 10.1.4.5 left) allows to control the replay of log files using the following buttons:

- ▶️ start playback  ❳ go to the previous frame with an image
- ■ stops playback  < go to the previous frame
- ✿ repeat the current frame  > go to the next frame
- ⚪ run in a loop  ⚪ go to the next frame with an image

In addition, the current frame can be directly selected using a slider.

\textbf{Annotation View}

The annotation view displays all annotations (cf. Sect. 3.7.6) contained in a log file (cf. Fig. 10.14 right). Double clicking an annotation will cause the log file to jump to the given frame number.
10.1. SIMROBOT B-Human 2016

Figure 10.14: The log player view and the annotation view

It is also possible to view annotations of Motion and Cognition modules when using the simulated NAO or a direct debug connection to a real NAO. This has to be activated by using the following debug request:

```
dr annotation
```

10.1.5 Scene Description Files

The language of scene description files is an extended version of RoSiML [16]. To use this new version and the new SimRobotCore2, the scene file has to end with .ros2, such as `BH2016.ros2`. In the following, the most important elements, which are necessary to add robots, dummies, and balls, are shortly described (based upon `BH2016.ros2`). For a more detailed documentation see Appendix A.

```
<Include href=".
```

First of all the descriptions of the NAO, the ball and the field are included. The names of include files end with .rsi2.

```
<Compound name="robots">
```

This compound contains all active robots, i.e. robots for which processes will be created. So, all robots in this compound will move on their own. However, each of them will require a lot of computation time. In the tag `Body`, the attribute `ref` specifies which NAO model should be used and `name` sets the robot name in the scene graph of the simulation. Legal robot names are “robot1” … “robot10”, where the first five robots are assumed to play in the blue team (with player numbers 1…5) while the other five play in the red team (again with player numbers 1…5). The standard color of the NAO’s jersey is set to blue. To set it to red, use `<Set name="NaoColor" value="red">` within the tag `Body`.

```
<Compound name="extras">
```

This compound contains passive robots, i.e. robots that just stand around, but that are not controlled by a program. Passive robots can be activated by moving their definition to the compound `robots`, but the referenced model has to be changed from “NaoDummy” to “Nao”.

149
**Predefined Scenes**

A lot of scene description files can be found in `Config/Scenes`. There are two types of scene description files: the ones required to simulate one or more robots, and the ones that are sufficient to connect to a real robot.

Simulating multiple robots is expensive. To overcome this and enable decent framerates, some scenes come in special variants. In scenes ending with `PerceptOracle` percepts are provided by the simulator. However, models, for example the `BallModel`, still have to be calculated. The suffix `Fast` further implies that even models are provided. Moreover, scenes ending in `WithPuppet` contain an additional robot which is used to display joint angles.

**BH2016 (PerceptOracle|Fast):** A single robot with five dummies.

**Game2016 (PerceptOracle|Fast):** A game five against five.

**KickViewScene (RemoteWithPuppet):** For working on kicks (cf. Sect. 10.1.4.4).

**ReplayRobot (WithPuppet):** Used to replay log files (cf. Sect. 3.7.5).

**RemoteRobot (WithPuppet):** Connects to a remote robot and displays images and data.

### 10.1.6 Console Commands

Console commands can either be directly typed into the console window or they can be executed from a script file. There are three different kinds of commands. The first kind will typically be used in a script file that is executed when the simulation is started. The second kind are *global commands* that change the state of the whole simulation. The third type is *robot commands* that affect currently *selected robots* only (see command `robot` to find out how to select robots).

#### 10.1.6.1 Initialization Commands

**sc** `<name> <a.b.c.d>`

Starts a remote connection to a real robot. The first parameter defines the *name* that will be used for the robot. The second parameter specifies the IP address of the robot. The command will add a new robot to the list of available robots using name, and it adds a set of views to the scene graph. When the simulation is reset or the simulator is exited, the connection will be terminated.

**sl** `<name> <file>`

Replays a log file. The command will instantiate a complete set of processes and views. The processes will be fed with the contents of the log file. The first parameter of the command defines the *name* of the virtual robot. The name can be used in the `robot` command (see below), and all views of this particular virtual robot will be identified by this name in the tree view. The second parameter specifies the name and path of the log file. If no path is given, `Config/Logs` is used as a default. Otherwise, the full path is used. `.log` is the default extension of log files. It will be automatically added if no extension is given.

When replaying a log file, the replay can be controlled by the *Log Player* (cf. Sect. 10.1.4.5) or the command `log` (see below). It is even possible to load a different log file during the replay.
loc <location>
Set the location of the scene. This command is special, because it only has an effect if run directly from the script that is executed when a scene is loaded. It is always executed first even if it is not the first command in the script, because the location must be set before any robot code is executed, as the location influences the search path of the configuration files loaded.

10.1.6.2 Global Commands

ar off | on
Enable or disable the automatic referee.

call <file>
Executes a script file. A script file contains commands as specified here, one command per line. The default location for scripts is the directory from which the simulation scene was started, their default extension is .con.

cls
Clears the console window.

dt off | on
Switches simulation dragging to real-time on or off. Normally, the simulation tries not to run faster than real-time. Thereby, it also reduces the general computational load of the computer. However, in some situations it is desirable to execute the simulation as fast as possible. By default, this option is activated.

echo <text>
Prints text into the console window. The command is useful in script files to print commands that can later be activated manually by pressing the Enter key in the printed line.

gc initial | ready | set | playing | finished | kickOffBlue | kickOffRed | outByBlue | outByRed | gameDropIn | gamePlayoff | gameRoundRobin
The command sets the current state of the GameController.

help [<pattern>], ? [<pattern>]
Displays a help text. If a pattern is specified, only those lines are printed that contain the pattern.

robot ? | all | <name> {<name>}
Selects a robot or a group of robots. The console commands described in the next section are only sent to selected robots. By default, only the robot that was created or connected last is selected. Using the robot command, this selection can be changed. Type robot ? to display a list of the names of available robots. A single simulated robot can also be selected by double-clicking it in the scene view. To select all robots, type robot all.

st off | on
Switches the simulation of time on or off. Without the simulation of time, all calls to SystemCall::getCurrentSystemTime() will return the real time of the PC. However if the simulator runs slower than real-time, the simulated robots will receive less sensor readings than the real ones. If the simulation of time is switched on, each step of the simulator will advance the time by 10 ms. Thus, the simulator simulates real-time, but it is running slower. By default this option is switched off.
# <text>
Comment. Useful in script files.

## 10.1.6.3 Robot Commands

**ac both | upper | lower**

Accept Camera: Change the process that provides drawings in the field and 3-D views.

**bc <red%> <green%> <blue%>**

Defines the background color for 3-D views. The color is specified in percentages of red, green, and blue intensities. All parameters are optional. Missing parameters will be interpreted as 0%.

**cameraCalibrator <view> (on | off)**

This command activates or deactivates the CameraCalibrator module for the given view. By default you may want to use the “raw” view. For a detailed description of the Camera- Calibrator see Sect. 4.1.2.1.

**ci off | on [<fps>]**

Switches the calculation of images on or off. With the optional parameter fps, a customized image frame rate can be set. The default value is 60. The simulation of the robot’s camera image costs a lot of time, especially if several robots are simulated. In some development situations, it is better to switch off all low level processing of the robots and to work with ground truth world states, i.e., world states that are directly delivered by the simulator. In such cases there is no need to waste processing power by calculating camera images. Therefore, it can be switched off. However, by default this option is switched on. Note that this command only has an effect on simulated robots.

**dr ? [<pattern>] | off | <key> ( off | on )**

Sends a debug request. B-Human uses debug requests to switch debug responses (cf. Sect. 3.6.1) on or off at runtime. Type dr ? to get a list of all available debug requests. The resulting list can be shortened by specifying a search pattern after the question mark. Debug responses can be activated or deactivated. They are deactivated by default. Specifying just off as only parameter returns to this default state. Several other commands also send debug requests, e.g., to activate the transmission of debug drawings.

**get ? [<pattern>] | <key> [%]**

Shows debug data or shows its specification. This command allows displaying any information that is provided in the robot code via the MODIFY macro. If one of the strings that are used as first parameter of the MODIFY macro is used as parameter of this command (the modify key), the related data will be requested from the robot code and displayed. The output of the command is a valid set command (see below) that can be changed to modify data on the robot. A question mark directly after the command (with an optional filter pattern) will list all the modify keys that are available. A question mark after a modify key will display the type of the associated data structure rather than the data itself.

**jc hide | show | motion ( 1 | 2 ) <command> | ( press | release ) <button>**

Sets a joystick command. If the first parameter is press or release, the number following is interpreted as the number of a joystick button. Legal numbers are between 1 and 40. Any text after this first parameter is part of the second parameter. The <command> parameter can contain any legal script command that will be executed in every frame.
while the corresponding button is pressed. The prefixes press or release restrict the execution to the corresponding event. The commands associated with the 26 first buttons can also be executed by pressing Ctrl+Shift+A...Ctrl+Shift+Z on the keyboard. If the first parameter is motion, an analog joystick command is defined. There are two slots for such commands, number 1 and 2, e.g., to independently control the robot’s walking direction and its head. The remaining text defines a command that is executed whenever the readings of the analog joystick change. Within this command, $1...8$ can be used as placeholders for up to eight joystick axes. The scaling of the values of these axes is defined by the command js (see below). If the first parameter is show, any command executed will also be printed in the console window. hide will switch this feature off again, and hide is also the default.

\textbf{jm} \texttt{<axis>} \texttt{<button>} \texttt{<button>}
Maps two buttons on an axis. Pressing the first button emulates pushing the axis to its positive maximum speed. Pressing the second button results in the negative maximum speed. The command is useful when more axes are required to control a robot than the joystick used actually has.

\textbf{js} \texttt{<axis>} \texttt{<speed>} \texttt{<threshold>} \texttt{[<center>]}  
Set axis maximum speed and ignore threshold for command \texttt{jc motion <num>}. \texttt{axis} is the number of the joystick axis to configure (1...8). \texttt{speed} defines the maximum value for that axis, i.e., the resulting range of values will be $[-\text{speed}...\text{speed}]$. The \texttt{threshold} defines a joystick measuring range around zero, in which the joystick will still be recognized as centered, i.e., the output value will be 0. The \texttt{threshold} can be set between 0 and 1. An optional parameter allows for shifting the center itself, e.g., to compensate for the bad calibration of a joystick.

\textbf{kick}  
Adds the \textit{KickEngine} view to the \textit{SceneGraph}.

\textbf{log} \texttt{? mr [list]} | \texttt{start} | \texttt{stop} | \texttt{pause} | \texttt{forward [image]} | \texttt{backward [image]} | \texttt{fastForward} | \texttt{fastBackward} | \texttt{repeat} | \texttt{goto <number>} | \texttt{time <minutes> <seconds>} | \texttt{clear} | ( \texttt{keep} | \texttt{remove} ) \texttt{<message>} \texttt{[<message>]} | ( \texttt{load} | \texttt{save} | \texttt{saveImages [raw]} ) \texttt{<file>} | \texttt{saveTiming <file>} | \texttt{cycle} | \texttt{once} | \texttt{full} | \texttt{jpeg}  
The command supports both recording and replaying log files. The latter is only possible if the current set of robot processes was created using the initialization command sl (cf. Sect. 10.1.6.1). The different parameters have the following meaning:

? Prints statistics on the messages contained in the current log file.

\textbf{mr} \texttt{[list]}
Sets the provider of all representations from the log file to \texttt{CognitionLogDataProvider} or \texttt{MotionLogDataProvider}. If \texttt{list} is specified the module request commands will be printed to the console instead.

\textbf{start} | \texttt{stop}
If replaying a log file, starts and stops the replay. Otherwise, the commands will start and stop the recording.

\textbf{pause} | \texttt{forward [image]} | \texttt{backward [image]} | \texttt{repeat} | \texttt{goto <number>} | \texttt{time <minutes> <seconds>}
The commands are only accepted while replaying a log file. \texttt{pause} stops the replay without rewinding to the beginning, \texttt{forward} and \texttt{backward} advance a single step in the respective direction. With the optional parameter \texttt{image}, it is possible to step...
from image to image. *repeat* just resends the current message. *goto* allows jumping to a certain position in the log file. If the log file contains *GameInfos*, the command *time* allows to jump to the first frame with a certain remaining game time.

**fastForward | fastBackward**
Jump 100 steps forward or backward.

**clear | ( keep | remove ) <message>**
clear removes all messages from the log file, while keep and remove only delete a selected subset based on the set of message ids specified.

**( load | save | saveImages [raw]) <file>**
These commands *load* and *save* the log file stored in memory. If the filename contains no path, *Config/Logs* is used as default. Otherwise, the full path is used. *log* is the default extension of log files. It will be automatically added if no extension is given. The option *saveImages* saves only the images from the log file stored in memory to the disk. The default directory is *Config/Files*. They will be stored in the format defined by the extension of the filename specified. If the extension is omitted, *.bmp* is used. The files saved contain either RGB or YCbCr images. The latter is the case if the option *raw* is specified.

**saveTiming <file>**
Creates a comma separated list containing the data of all stopwatches for each frame. *<file>* can either be an absolute or a relative path. In the latter case, it is relative to the directory *Config*. If no extension is specified, *.csv* is used.

**cycle | once**
The two commands decide whether the log file is only replayed once or continuously repeated.

**full | jpeg**
These two commands decide whether uncompressed images received from the robot will also be written to the log file as full images, or JPEG-compressed. When the robot is connected by cable, sending uncompressed images is usually a lot faster than compressing them on the robot. By executing *log jpeg* they can still be saved in JPEG format, saving a log memory space during recording as well as disk space later. Note that running image processing routines on JPEG images does not always give realistic results, because JPEG is not a lossless compression method, and it is optimized for human viewers, not for machine vision.

**mof** Recompiles all special actions and if successful, the result is sent to the robot.

Sends a module request. This command allows selecting the module that provides a certain representation. If a representation should not be provided anymore, it can be switched off. Deactivating the provision of a representation is usually only possible if no other module requires that representation. Otherwise, an error message is printed and the robot is still using its previous module configuration. Sometimes, it is desirable to be able to deactivate the provision of a representation without the requirement to deactivate the provision of all other representations that depend on it. In that case, the provider of the representation can be set to *default*. Thus no module updates the representation and it simply keeps its previous state.

A question mark after the command lists all representations. A question mark after a representation lists all modules that provide this representation. The parameter *modules*
lists all modules with their requirements and provisions. All three listings can be filtered by an optional pattern. save saves the current module configuration to the file modules.cfg which it was originally loaded from. Note that this usually has not the desired effect, because the module configuration has already been changed by the start script to be compatible with the simulator. Therefore, it will not work anymore on a real robot. The only configuration in which the command makes sense is when communicating with a remote robot.

msg off | on | disable | enable | log <file>
Switches the output of text messages on or off, or redirects them to a text file. All processes can send text messages via their message queues to the console window. As this can disturb entering text into the console window, printing can be switched off. However, by default text messages are printed. In addition, text messages can be stored in a log file, even if their output is switched off. The file name has to be specified after msg log. If the file already exists, it will be replaced. If no path is given, Config/Logs is used as default. Otherwise, the full path is used. .txt is the default extension of text log files. It will be automatically added if no extension is given.

mv <x> <y> <z> [<rotx> <roty> <rotz>]
Moves the selected simulated robot to the given metric position. x, y, and z have to be specified in mm, the rotations have to be specified in degrees. Note that the origin of the NAO is about 330 mm above the ground, so z should be 330.

mvb <x> <y> <z>
Moves the ball to the given metric position. x, y, and z have to be specified in mm. Note that the origin of the ball is about 32.5 mm above the ground.

poll
Polls for all available debug requests and debug drawings. Debug requests and debug drawings are dynamically defined in the robot control program. Before console commands that use them can be executed, the simulator must first determine which identifiers exist in the code that currently runs. Although the acquiring of this information is usually done automatically, e.g., after the module configuration was changed, there are some situations in which a manual execution of the command poll is required. For instance if debug responses or debug drawings are defined inside another debug response, executing poll is necessary to recognize the new identifiers after the outer debug response has been activated.

pr none | illegalBallContact | playerPushing | illegalMotionInSet | inactivePlayer | illegalDefender | leavingTheField | kickOffGoal | requestForPickup | manual
Penalizes a simulated robot with the given penalty or unpenalizes it, when used with none. When penalized, the simulated robot will be moved to the sideline, looking away from the field. When unpenalized, it will be turned, facing the field again, and moved to the sideline that is further away from the ball.

qfr queue | replace | reject | collect <seconds> | save <seconds>
Sends a queue fill request. This request defines the mode how the message queue from the debug process to the PC is handled.

replace
Replace is the default mode. If the mode is set to replace, only the newest message of each type is preserved in the queue (with a few exceptions). On the one hand, the
queue cannot overflow, on the other hand, messages are lost, e.g. it is not possible to
receive 60 images per second from the robot.

**queue**
Queue will insert all messages received by the debug process from other processes into
the queue, and send it as soon as possible to the PC. If more messages are received
than can be sent to the PC, the queue will overflow and some messages will be lost.

**reject**
Reject will not enter any messages into the queue to the PC. Therefore, the PC will
not receive any messages.

**collect <seconds>**
This mode collects messages for the specified number of seconds. After that period
of time, the collected messages will be sent to the PC. Since the TCP stack requires
a certain amount of execution time, it may impede the real-time behavior of the
robot control program. Using this command, no TCP packages are sent during the
recording period, guaranteeing real-time behavior. However, since the message queue
of the process Debug has a limited size, it cannot store an arbitrary number of
messages. Hence the bigger the messages, the shorter they can be collected. After
the collected messages were sent, no further messages will be sent to the PC until
another queue fill request is sent.

**save <seconds>**
This mode collects messages for the specified number of seconds, and it will afterwards
store them on the memory stick as a log file under `/home/nao/Config/logfile.log`. No
messages will be sent to the PC until another queue fill request is sent.

**si reset | ( upper | lower ) [number] [<file>]**
Saves the raw image of a robot. The image will be saved as bitmap file. If no path is
specified, `Config/raw_image.bmp` will be used as default option. If `number` is specified, a
number is appended to the filename that is increased each time the command is executed.
The option `reset` resets the counter.

**set ? | [<pattern>] | <key> ( ? | unchanged | <data>)**
Changes debug data or shows its specification. This command allows changing any infor-
mation that is provided in the robot code via the `MODIFY` macro. If one of the strings that
are used as first parameter of the `MODIFY` macro is used as parameter of this command
(the `modify key`), the related data in the robot code will be replaced by the data structure
specified as second parameter. It is best to first create a valid `set` command using the `get`
command (see above). Afterwards that command can be changed before it is executed. If
the second parameter is the key word `unchanged`, the related `MODIFY` statement in the code
does not overwrite the data anymore, i.e., it is deactivated again. A question mark directly
after the command (with an optional filter pattern) will list all the modify keys that are
available. A question mark after a modify key will display the type of the associated data
structure rather than the data itself.

**save ? | [<pattern>] | <key> [<path>]**
Save debug data to a configuration file. The keys supported can be queried using the
question mark. An additional pattern filters the output. If no path is specified, the name
of the configuration file is looked up from a table, and its first occurrence in the search
path is overwritten. Otherwise, the path is used. If it is relative, it is appended to the
directory `Config`. 

156
v3 ? [<pattern>] | <image> [jpeg] [<name>]
 Adds a set of 3-D color space views for a certain image (cf. Sect. 10.1.4.1). The image can either be the camera image (simply specify image) or a debug image. It will be JPEG compressed if the option jpeg is specified. The last parameter is the name that will be given to the set of views. If the name is not given, it will be the same as the name of the image. A question mark followed by an optional filter pattern will list all available images.

vf <name>
 Adds a field view (cf. Sect. 10.1.4.1). A field view is the means for displaying debug drawings in field coordinates. The parameter defines the name of the view.

vfd ? [<pattern>] | ( <name> | all ) ( ? [<pattern>] | <drawing> ( on | off ) )
 (De)activates a debug drawing in a field view. The first parameter is the name of a field view that has been created using the command vf (see above) or all to add the drawing to all field views. The second parameter is the name of a drawing that is defined in the robot control program. Such a drawing is activated when the third parameter is on or is missing. It is deactivated when the third parameter is off. The drawings will be drawn in the sequence they are added, from back to front. Adding a drawing a second time will move it to the front. A question mark directly after the command will list all field views that are available. A question after a valid field view will list all available field drawings. Both question marks have an optional filter pattern that reduces the number of answers.

vi ? [<pattern>] | <image> [jpeg] [segmented] [upperCam] [<name>] [gain <value>]
 Adds an image view (cf. Sect. 10.1.4.1). An image view is the means for displaying debug drawings in image coordinates. The image can either be the camera image (simply specify image), a debug image, or no image at all (none). It will be JPEG-compressed if the option jpeg is specified. If segmented is given, the image will be segmented using the current color table. The default is to show data based on the images taken by the lower camera. With upperCam, the upper camera is selected instead. The next parameter is the name that will be given to the set of views. If the name is not given, it will be the same as the name of the image plus the word Segmented if it should be segmented. With the last parameter the image gain can be adjusted, if no gain is specified the default value will be 1.0. A question mark followed by an optional filter pattern will list all available images.

vid ? [<pattern>] | ( <name> | all ) ( ? [<pattern>] | <drawing> ( on | off ) )
 (De)activates a debug drawing in an image view. The first parameter is the name of an image view that has been created using the command vi (see above) or all to add the drawing to all image views. The second parameter is the name of a drawing that is defined in the robot control program. Such a drawing is activated when the third parameter is on or is missing. It is deactivated when the third parameter is off. The drawings will be drawn in the sequence they are added, from back to front. Adding a drawing a second time will move it to the front. A question mark directly after the command will list all image views that are available. A question mark after a valid image view will list all available image drawings. Both question marks have an optional filter pattern that reduces the number of answers.

vp <name> <numOfValues> <minValue> <maxValue> [<yUnit> <xUnit> <xScale>]
 Adds a plot view (cf. Sect. 10.1.4.5). A plot view is the means for plotting data that was defined by the macro PLOT in the robot control program. The first parameter defines the name of the view. The second parameter is the number of entries in the plot, i.e. the size of the x axis. The plot view stores the last numOfValues data points sent for each plot
and displays them. \texttt{minValue} and \texttt{maxValue} define the range of the \textit{y} axis. The optional parameters serve the capability to improve the appearance of the plots by adding labels to both axes and by scaling the time-axis. The label drawing can be activated by using the context menu of the plot view.

\texttt{vpd} \ ? \ [\texttt{<pattern>}] \ | \ <name> \ ( \ ? \ [\texttt{<pattern>}] \ | \ <drawing> \ ( \ ? \ [\texttt{<pattern>}] \ | \ <color> \ | \ off ))
Plots data in a certain color in a plot view. The first parameter is the name of a plot view that has been created using the command \texttt{vp} (see above). The second parameter is the name of plot data that is defined in the robot control program. The third parameter defines the color for the plot. \texttt{black}, \texttt{red}, \texttt{green}, \texttt{blue}, \texttt{yellow}, \texttt{cyan}, \texttt{magenta}, \texttt{orange}, \texttt{violet}, \texttt{gray}, and six-digit hexadecimal RGB color codes are supported. The plot is deactivated when the third parameter is \texttt{off}. The plots will be drawn in the sequence they were added, from back to front. Adding a plot a second time will move it to the front. A question mark directly after the command will list all plot views that are available. A question after a valid plot view will list all available plot data. Both question marks have an optional filter pattern that reduces the number of answers.

\texttt{vd} \ <\texttt{debug data}> \ (\texttt{on} \ | \ \texttt{off})
Show debug data in a window or disable the updates of the data. Data views can be found in the data category of the scene graph. Data views provide the same functionality as the \texttt{get} and \texttt{set} commands (see above). However they are much more comfortable to use.

### 10.1.6.4 Input Selection Dialog

In scripting files, it is sometimes necessary to let the user choose between some values. Therefore we implemented a small input dialog. After the user has selected a value it will then be passed as input parameter to the scripting command. To use the input dialog you just have to type the following expression:

\texttt{$\{\texttt{<label>},\texttt{<value1>},\texttt{<value2>},...\}$}

Inside the brackets there has to be a list of comma-separated values. The first value will be interpreted as a label for the input box. All following values will then be selectable via the dropdown list. The following example shows how we used it to make the IP address selectable. The user’s choice will be passed to the command \texttt{sc} that establishes a remote connection to the robot with the selected IP address. Figure 10.15 shows the corresponding dialog.

\texttt{sc Remote \$\{IP \ address:,10.0.1.1,192.168.1.1\}$}
10.2. B-HUMAN USER SHELL

10.1.7 Recording a Remote Log File

To record a log file, the robot shall at least send images, camera info, joint data, sensor data, key states, odometry data, the camera matrix, and the image coordinate system. The following script connects to a robot and configures it to do so. In addition, it prints several useful commands into the console window, so they can be executed by simply setting the cursor in the corresponding line and pressing the Enter key. As these lines will be printed before the messages coming from the robot, one has to scroll to the beginning of the console window to use them. Note that both the IP address in the second line and the filename behind the line log save have to be changed.

```
# connect to a robot
sc Remote 10.1.0.101

# request everything that should be recorded
dr representation:JPEGImage
dr representation:CameraInfo
dr representation:JointAngles
dr representation:InertialSensorData
dr representation:KeyStates
dr representation:OdometryData
dr representation:CameraMatrix
dr representation:ImageCoordinateSystem

# print some useful commands
echo log start
echo log stop
echo log save <filename>
echo log clear
```

10.2 B-Human User Shell

The B-Human User Shell (bush) accelerates and simplifies the deployment of code and the configuration of the robots. It is especially useful when controlling several robots at the same time, e.g., during the preparation for a soccer match.

10.2.1 Configuration

Since the bush can be used to communicate with the robots without much help from the user, it needs some information about the robots. Therefore, each robot has a configuration file Config/Robots/<RobotName>/network.cfg, which defines the name of the robot and how it can be reached by the bush.\(^2\) Additionally you have to define one (or more) teams, which are arranged in tabs. The data of the teams is used to define the other properties, which are required to deploy code in the correct configuration to the robots. The default configuration file of the teams is Config/teams.cfg which can be altered within the bush or with a text editor. Each team can have the configuration variables shown in Table 10.1.

10.2.2 Commands

The bush supports two types of commands. There are local commands (cf. Tab. 10.2) and commands that interact with selected robot(s) (cf. Tab. 10.3). Robots can be selected by checking their checkbox or with the keys F1 to F10.

\(^2\)The configuration file is created by the script createRobot described in Sect. 2.4.2.
10.2.3 Deploying Code to the Robots

For the simultaneous deployment of several robots the command `deploy` should be used. It accepts a single optional parameter that designates the build configuration of the code to be deployed to the selected robots. If the parameter is omitted the default build configuration of the currently selected team is used. It can be changed with the drop-down menu at the top of the `bush` user interface.

Before the `deploy` command copies code to the robots, it checks whether the binaries are up-to-date. If needed, they are recompiled by the `compile` command, which can also be called independently from the `deploy` command. Depending on the platform, the `compile` command uses `make`, `xcodebuild`, or `MSBuild` to compile the binaries required.

After all the files required by the NAO are copied, the `deploy` command generates a new `settings.cfg` according to the configuration tracked by the `bush` for each of the selected robots. After updating the file `settings.cfg`, the `bhuman` software has to be restarted for changes to take effect. This can easily be done with the command `restart`. If it is called without any parameter, it restarts only the `bhuman` software but it can also be used to restart NAOqi and `bhuman`, and the entire operating system of the robot if you call it with one of the parameters `naoqi`, `full`, or `robot`. To inspect the configuration files copied to the robots, you can use the command `show`, which knows most of the files located on the robots and can help you finding the desired files with tab completion.

10.2.4 Managing Multiple Wireless Configurations

Since the robot soccer competition generally takes place on more than just a single field and normally each field has its own WiFi access point, the robots have to deal with multiple configuration files for their wireless interfaces. When a robot is deployed the selected wireless
<table>
<thead>
<tr>
<th>Entry</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>name</td>
<td>The name of the team.</td>
</tr>
<tr>
<td>number</td>
<td>The team number.</td>
</tr>
<tr>
<td>port</td>
<td>The port, which is used for team communication messages. This entry is optional. If this value is omitted, the port is generated from the team number.</td>
</tr>
<tr>
<td>color</td>
<td>The team color in the first half. This entry is optional. It is only required if no game controller is running, which overwrites the team color.</td>
</tr>
<tr>
<td>location</td>
<td>The location, which should be used by the software (cf. Sect. 2.9). This entry is optional. It is set to <strong>Default</strong> if it is omitted.</td>
</tr>
<tr>
<td>buildConfig</td>
<td>The name of the configuration, which should be used to deploy the NAO code (cf. Sect. 2.2).</td>
</tr>
<tr>
<td>wlanConfig</td>
<td>The name of the configuration file, which should be used to configure the wireless interface of the robots.</td>
</tr>
<tr>
<td>volume</td>
<td>The audio volume, to which the robots should be set to.</td>
</tr>
<tr>
<td>deployDevice</td>
<td>The device, which should be used to connect to the robots. Either a Ethernet or a WiFi connection can be established. The default entry is auto. This chooses the best device, depending on ping times.</td>
</tr>
<tr>
<td>magicNumber</td>
<td>The magic number for the team communication. Robots with different numbers will ignore each other. The default number is -1, which will set a random number between 0 and 255.</td>
</tr>
<tr>
<td>players</td>
<td>The list of robots the team consists of. The list must have ten entries, where each entry must either be a name of a robot (with an existing file <code>Config/Robots/&lt;RobotName&gt;/network.cfg</code>), or an underscore for empty slots. The first five robots are the main players and the last five their substitutes.</td>
</tr>
</tbody>
</table>

Table 10.1: Configuration variables in the file `Config/teams.cfg`

configuration in the drop down menu will be used.

### 10.2.5 Substituting Robots

The robots known to `bush` are arranged in two rows. The entries in the upper row represent the playing robots and the entries in the lower row the robots which stand by as substitutes. To select which robots are playing and which are not, you can move them by drag&drop to the appropriate position. Since this view only supports ten robots at a time, there is another view called `RobotPool`, which contains all other robots. It can be pulled out at the right side of the `bush` window. The robots displayed there can be exchanged with robots from the main view. If a robot is deployed in the second row, NAOqi will be shut down. This is important because the robots should be operational, to be replaced as fast as possible, but are not allowed to send and receive packages.

### 10.2.6 Monitoring Robots

The `bush` displays some information about the robots’ states as you can see in Figure 10.16: wireless connection pings, wired connection pings, and remaining battery charge level. The latter is shown by the power bar. If the power bar is colored green, a power source is detected.
and if it’s red no power source is detected. Besides the color change of the power bar the robot’s head LEDs will display a rotating pattern to indicate a connected external power source.

### 10.3 GameController

A RoboCup game has a human referee. Unfortunately the robots cannot understand him or her directly. Instead the referee’s assistant relays the decisions to the robots using a software called GameController. From this year on the official GameController is the one we created. It is written in Java 1.7.
10.3.1 Architecture

The architecture (cf. Fig. 10.17) is based on a combination of the model-view-controller (MVC) and the command pattern.
The GameController communicates with the robots using a C data structure called `RoboCupGameControlData` as mentioned in the RoboCup SPL rules. It contains information about the current game and penalty state of each robot. It is broadcasted via UDP several times per seconds. Robots may answer using the `RoboCupGameControlReturnData` C data structure.

Both C data structures were translated to Java for the GameController.\(^3\) They only hold the

\(^3\)Their names leave out the prefix “RoboCup”

Figure 10.18: The sequences between some threads

Figure 10.19: Start screen of the **GameController**
The main screen of the GameController information and provide conversion methods from and to a byte stream. Unfortunately, the GameControlData does not contain all the information needed to fully describe the current state of the game. For example, it lacks information about the number of timeouts or penalties that a team has taken and the game time is only precise up to one second. Therefore, the class GameControlData is extended by a class called AdvancedData. This class holds the complete game state. From the classical MVC point of view, the AdvancedData is the model.

The view component is represented by the package GUI. All GUI functionality is controlled via the interface GCGUI. GCGUI only provides an update method with an instance of the class AdvancedData as a parameter. This update method is called frequently. Therefore, the GUI can only display the same game state as the one that is transmitted.

The controller part of the model-view-controller architecture is kind of tricky, because it has to deal with parallel inputs from the user, a ticking clock, and robots via the network. To simplify the access to the model, we only allow access to it from a single thread. Everything that modifies the model is encapsulated in action classes, which are defined by extending the abstract class GCAction. All threads (GUI, timing, network) register actions at the EventHandler. The EventHandler executes the actions on the GUI thread (cf. Fig. 10.18).

Each action knows, based on the current game state, whether it can be legally executed according to the rules. For example, switching directly from the initial to the playing state, penalizing a robot for holding the ball in the ready state or decreasing the goal count is illegal.

### 10.3.2 UI Design

After launching the GameController, a small window will appear to select the basic settings of the game (cf. Fig. 10.19). The most basic decision is the league. You can choose between SPL, SPL Drop-in and three Humanoid leagues: Kid-, Teen- and Adult-Size. You can select, which teams will play and whether it is a play-off game, as well as you can choose between a
fullscreen and a fully scalable windowed mode. For SPL teams, it can also be switched between the primary and the secondary jersey color. After pressing the start button, the main GUI will appear.

The look of the GUI (cf. Fig. 10.20) is completely symmetric and all buttons are as big as possible. In addition, keyboard shortcuts are provided for most buttons. Thus making it possible to operate the GameController in a more efficient way. Buttons are only enabled if the corresponding actions are legal. This should decrease the chance that the operator of the GameController presses a wrong button. Since mistakes such as penalizing the wrong robot can still occur, the GUI provides an undo functionality that allows to revert actions. This clearly distinguishes normal actions that must follow the rules from corrective actions that are only legal because they heal a mistake that was made before.

All undoable actions are displayed in a timeline at the bottom of the GameController. By double clicking on one of the actions in the timeline, the state will be reverted to the state right before that action has been executed. However the game time will only be reverted if a transition between different game states is reverted as well.

When testing their robots, most teams want to be able to do arbitrary state transitions with the GameController. Therefore, it has a functionality to switch in and out of a test mode. While being in test mode, all actions are allowed at any time.

10.3.3 Game State Visualizer

The GameController comes with two additional tools within the same Java project. One of them is the GameStateVisualizer (cf. Fig. 10.21). Its purpose is to show the state of the game to the audience. This includes the teams playing, the current score, the time remaining, and some other information. The GameStateVisualizer listens through the network to the GameController and displays everything of interest.
For the purpose of testing the network or packages the robots should receive, there is a test mode which can be switched on or off by pressing \textit{F11}. In the test mode, the \textit{GameStateVisualizer} shows everything that is contained in the packages and how it should be interpreted.

### 10.3.4 Log Analyzer

The \textit{GameController} writes some status information and all decisions entered to a log file with a unique name. So after a competition, hundreds of log files exist, not only the ones of official games, but also of practice matches, test kick-offs, etc. The \textit{LogAnalyzer} allows to interactively select the log files that resulted from official games and then convert from log files to files that can be imported by a spreadsheet application. The \textit{LogAnalyzer} also cleans the data, e.g. by removing decisions that were reverted later, so they do not impede statistics made later.

Right after you launch the \textit{LogAnalyzer}, it quickly parses each log file and then analyzes some meta information to make a guess, whether and why a log file may not belong to a real game. It then lists all logs in a GUI with that guess (cf. Fig. 10.22). While dealing with hundreds of log-files, you can select the right ones within a few minutes by comparing them to the timetable of the event. Afterwards a file containing the consolidated information of all the selected logs in the form of comma separated values can be created.

### 10.4 Team Communication Monitor

With the start of the Drop-in Player Competition in 2014, a standard communication protocol, i.e. the \textit{SPLStandardMessage}, was introduced to allow robots of different teams to exchange information. While implementing the standard protocol correctly is a necessity for the Drop-in Player Competition to work, it is also an opportunity for teams to lower the amount of coding.
they have to do to develop debugging tools, because when all teams use the same communication protocol, the tools they wrote could also be shared more easily. Therefore, B-Human developed the TeamCommunicationMonitor (TCM) as a standard tool to monitor the network traffic during SPL games. This project has been partially funded by the RoboCup Federation.

The TCM visualizes all data currently sent by the robots and highlights illegal messages. Furthermore, it uses knowledge about the contents of SPLStandardMessages in order to visualize them in a 3-D view of the playing field. This makes it also suitable for teams to debug their network code. Visualized properties of robots are their position and orientation, their fallen and penalty states (the latter is received from the GameController), where they last saw the ball, and their player numbers. A screenshot of the TCM can be seen in Fig. 10.23.

In order to display all messages of robots sending packets, the TCM binds sockets to all UDP ports that may be used by a team as their team number, i.e. ports 10000 to 10099, and listens for any incoming packets. When a packet is received, it is parsed as an SPLStandardMessage, marking all fields as invalid that do not contain legal values according to the definition of the SPLStandardMessage. As the TCM only listens and does not send anything, it may run on multiple computers in the same network at once without any interferences.

The TCM identifies robots using their IP address and assumes them to be sending on the port that matches their team number. The team number sent by the robots as part of the SPLStandardMessage is marked as invalid if it does not match the port on which the messages are received. For each robot, the TCM holds an internal state containing the last received message as well as the time stamps of the most recently received messages in order to calculate the number of messages per second. A robot’s state is discarded if no message was received from the robot for at least two seconds.

Displaying the 3-D view is done with OpenGL using the JOGL library [1]. The visualization subsystem is also extensible to enable teams to write plug-ins containing drawings that visualize data from the non-standardized part of their messages. We have written a plugin to display perceptions of goals, obstacles, and whistles as well as basic status information of the robots which our team communicates via the non-standardized message part.
Besides just visualizing received messages, the TCM also stores all received messages both from robots and the GameController in log files. These can be replayed later, allowing teams and organizers to check the communication of robots after the game is over.

The TCM is developed as part of the repository of the GameController and it shares parts of its code.

After having been successfully used at the RoboCup German Open 2015 to ensure valid messages from all teams during the Drop-in games, the TCM was publicly released in early June 2015 and was installed on the referee PC on each SPL field at both RoboCup 2015 and RoboCup 2016.
Chapter 11

Acknowledgements

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In addition, we want to thank the authors of the following software that is used in our code:

**Artistic Style:** Source code formatting in the *AStyle for B-Human* text service on macOS.
(http://astyle.sourceforge.net)

**AT&T Graphviz:** For generating the graphs shown in the options view and the module view of the simulator.
(http://www.graphviz.org)

**ccache:** A fast C/C++ compiler cache.
(http://ccache.samba.org)

**clang:** A compiler front end for the C, C++, Objective-C, and Objective-C++ programming languages.
(http://clang.llvm.org)

**Eigen:** A C++ template library for linear algebra: matrices, vectors, numerical solvers, and related algorithms.
(http://eigen.tuxfamily.org)

**getModKey:** For checking whether the shift key is pressed in the Deploy target on macOS.
(http://allancraig.net/index.php?option=com_docman&Itemid=100, not available anymore)

**ld:** The GNU linker is used for cross linking on Windows and macOS.
(http://sourceware.org/binutils/docs-2.21/ld)

**libjpeg:** Used to compress and decompress images from the robot’s camera.
(http://www.ijg.org)

**libjpeg-turbo:** For the NAO we use an optimized version of the libjpeg library.
(http://libjpeg-turbo.virtualgl.org)
libqxt: For showing the sliders in the camera calibration view of the simulator.
  (https://bitbucket.org/libqxt/libqxt/wiki/Home)

libxml2: For reading simulator’s scene description files.
  (http://xmlsoft.org)

mare: Build automation tool and project file generator.
  (http://github.com/craflin/mare)

ODE: For providing physics in the simulator.
  (http://www.ode.org)

OpenGL Extension Wrangler Library: For determining, which OpenGL extensions are supported by the platform.
  (http://glew.sourceforge.net)

Qt: The GUI framework of the simulator.
  (http://www.qt.io)

qtpropertybrowser: Extends the Qt framework with a property browser.
  (https://github.com/qtproject/qt-solutions/tree/master/qtpropertybrowser)

snappy: Used for the compression of log files.
  (http://google.github.io/snappy)
Bibliography


Appendix A

The Scene Description Language

A.1 EBNF

In the next section the structure of a scene description file is explained by means of an EBNF representation of the language. In the following, you can find an explanation of the symbols used.

Symbols surrounded by ?( . . . )?

Parentheses with question marks mean, that the order of all elements between them is irrelevant. That means every permutation of elements within those brackets is allowed. For example: something =?( firstEle secondEle thirdEle )?; can also be written as something =?( secondEle firstEle thirdEle )? or as something =?( thirdEle firstEle secondEle )? and so on.

Symbols surrounded by !

!x,y,...! means, that each rule is required. In fact the exclamation marks should only underline that all elements between them are absolutely required. In normal EBNF-Notation a rule like Hinge = ...bodyClass, axisClass!... can be written as Hinge = ...bodyClass axisClass.....

+ [...]+

+ [x,y]+ means, that x and y are optional. You could also write somewhat = +[x, y, z]+ as somewhat = [x] [y] [z].

{ ... }

Elements within curly braces are repeatable optional elements. These brackets have the normal EBNF meaning.

“...”

Terminal symbols are marked with quotation marks.

A.2 Grammar

appearanceClass = Appearance | BoxAppearance | SphereAppearance | CylinderAppearance | CapsuleAppearance | ComplexAppearance;
axisClass = Axis;
bodyClass = Body;
compoundClass = Compound;
deflectionClass = Deflection;
extSensorClass = Camera | DepthImageSensor | SingleDistanceSensor
 | ApproxDistanceSensor;
frictionClass = Friction | RollingFriction;
geometryClass = Geometry | BoxGeometry | CylinderGeometry
 | CapsuleGeometry | SphereGeometry;
infrastructureClass = Simulation | Include;
intSensorClass = Accelerometer | Gyroscope | CollisionSensor;
jointClass = Hinge | Slider;
lightClass = Light;
massClass = Mass | BoxMass | InertiaMatrixMass | SphereMass;
materialClass = Material;
motorClass = ServoMotor | VelocityMotor;
primitivesClass = Normals;
rotationClass = Rotation;
sceneClass = Scene;
setClass = Set;
solverClass = Quicksolver;
texCoordsClass = TexCoords;
translationClass = Translation;
userInputClass = UserInput;
verticesClass = Vertices;

Accelerometer = "<Accelerometer></Accelerometer>" | "<Accelerometer/>";
Gyroscope = "<Gyroscope></Gyroscope>" | "<Gyroscope/>";
CollisionSensor = "<CollisionSensor>" ?( +[translationClass, rotationClass]+ {geometryClass} )? "</CollisionSensor>";

Appearance = "<Appearance>" ?( +[translationClass, rotationClass]+ {setClass | appearanceClass} )? "</Appearance>";
BoxAppearance = "<BoxAppearance>" ?( !surfaceClass! +[translationClass, rotationClass]+ {setClass | appearanceClass} )? "</BoxAppearance>";
ComplexAppearance = "<ComplexAppearance>" ?( !surfaceClass, verticesClass,
primitiveGroupClass! +[translationClass, rotationClass, normalsClass, texCoordsClass]+ {setClass | appearanceClass | primitiveGroupClass} )? "</ComplexAppearance>";
CapsuleAppearance = "<CapsuleAppearance>" ?( !surfaceClass! +[translationClass, rotationClass]+ {setClass | appearanceClass} )? "</CapsuleAppearance>";
CylinderAppearance = "<CylinderAppearance>" ?( !surfaceClass! +[translationClass, rotationClass]+ {setClass | appearanceClass} )? "</CylinderAppearance>";
SphereAppearance = "<SphereAppearance>" ?( !surfaceClass! +[translationClass, rotationClass]+ {setClass | appearanceClass} )? "</SphereAppearance>";
ApproxDistanceSensor = "<ApproxDistanceSensor>" ?( +[translationClass, rotationClass]+ )? "</ApproxDistanceSensor>";
Camera = "<Camera>" ?( +[translationClass, rotationClass]+ )? "</Camera>";
DepthImageSensor = "<DepthImageSensor>" ?( +[translationClass, rotationClass]+ )? "</DepthImageSensor>";
SingleDistanceSensor = "<SingleDistanceSensor>" ?( +[translationClass,
rotationClass]+ )? "</SingleDistanceSensor>";

UserInput = "<UserInput>"/UserInput" | "</UserInput>"/;

Mass = "<Mass>" ?( +[translationClass, rotationClass]+ {setClass | massClass} )? "</Mass>";

BoxMass = "<BoxMass>" ?( +[translationClass, rotationClass]+ {setClass | massClass} )? "</BoxMass>";

InertiaMatrixMass = "<InertiaMatrixMass>" ?( +[translationClass, rotationClass]+ {setClass | massClass} )? "</InertiaMatrixMass>";

SphereMass = "<SphereMass>" ?( +[translationClass, rotationClass]+ {setClass | massClass} )? "</SphereMass>"

Geometry = "<Geometry>" ?( +[translationClass, rotationClass, materialClass]+ {setClass | geometryClass} )? "</Geometry>";

BoxGeometry = "<BoxGeometry>" ?( +[translationClass, rotationClass, materialClass]+ {setClass | geometryClass} )? "</BoxGeometry>";

CylinderGeometry = "<CylinderGeometry>" ?( +[translationClass, rotationClass, materialClass]+ {setClass | geometryClass} )? "</CylinderGeometry>";

CapsuleGeometry = "<CapsuleGeometry>" ?( +[translationClass, rotationClass, materialClass]+ {setClass | geometryClass} )? "</CapsuleGeometry>";

SphereGeometry = "<SphereGeometry>" ?( +[translationClass, rotationClass, materialClass]+ {setClass | geometryClass} )? "</SphereGeometry>";

Axis = "<Axis>" ?( +[motorClass, deflectionClass]+ {setClass} )? "</Axis>";

Hinge = "<Hinge>" ?( !bodyClass, axisClass! +[translationClass, rotationClass]+ {setClass} )? "</Hinge>";

Slider = "<Slider>" ?( !bodyClass, axisClass! +[translationClass, rotationClass]+ {setClass} )? "</Slider>";

Body = "<Body>" ?( !massClass! +[translationClass, rotationClass]+ {setClass | jointClass | appearanceClass | geometryClass | massClass | intSensorClass | extSensorClass} )? "</Body>";

Material = "<Material>" ?( {setClass | frictionClass} )? "</Material>";

Friction = "<Friction>/Friction>" | "</Friction>"/

RollingFriction = "<RollingFriction>/RollingFriction>" | "</RollingFriction>/";

ServoMotor = "<ServoMotor>/ServoMotor>" | "</ServoMotor>";

VelocityMotor = "<VelocityMotor>/VelocityMotor>" | "</VelocityMotor>";

Simulation = "<Simulation>" !sceneClass! "</Simulation>";

Scene = "<Scene>" ?( +[solverClass]+ {setClass | bodyClass | compoundClass | lightClass} )? "</Scene>";

Compound = "<Compound>" ?( +[translationClass, rotationClass]+


A.3 Structure of a Scene Description File

A.3.1 The Beginning of a Scene File

Every scene file has to start with a \(<Simulation>\) tag. Within a \(Simulation\) block a \(Scene\) element is required, but there is one exception: files included via \(<Include href=\ldots>\) must start with \(<Simulation>\), but there is no \(Scene\) element required. A \(Scene\) element specifies which controller is loaded for this scene via the \(controller\) attribute (in our case all scenes set the \(controller\) attribute to \(SimulatedNao\), so that the library \(SimulatedNao\) is loaded by SimRobot). It is recommended to include other specifications per \(Include\) before the scene description starts (compare with \(BH2016.ros2\)), but it is not necessary.

A.3.2 The ref Attribute

An element with a name attribute can be referenced by the \(ref\)-attribute using its name, i.e. elements that are needed repeatedly in a scene need to be defined only once. For example there is only one description of a NAO in its definition file (\(NaoV4H21.rsi2\)), but NAOs with different jersey colors are needed on a field. For each NAO on the field, there is a reference to the original model. The positioning of the NAOs is done by \(Translation\) and \(Rotation\) elements. The color is set by a \(Set\) element, which is described below.

\[
\begin{align*}
\text{Body} \text{name=\"Nao\">} \\
\hspace{1cm} \text{Set} \text{name=\"NaoColor\" value=\"blue\"/>} \\
\hspace{1cm} :: \\
\hspace{1cm} </Body> \\
:: \\
\begin{align*}
\text{Body} \text{ref=\"Nao\" name=\"BlueNao\">} \\
\hspace{1cm} \text{Translation} \text{x=\"-2\" y=\"0.4\" z=\"320\,mm\"/>} \\
\hspace{1cm} </Body> \\
\text{Body} \text{ref=\"Nao\" name=\"RedNao\">} \\
\hspace{1cm} \text{Translation} \text{x=\"-1.5\" y=\"-0.9\" z=\"320\,mm\"/>} \\
\hspace{1cm} \text{Rotation} \text{z=\"180\,degree\"/>} \\
\hspace{1cm} \text{Set} \text{name=\"NaoColor\" value=\"red\"/>} \\
\hspace{1cm} </Body> \\
::
\end{align*}
\]


A.3.3 Placeholders and Set Element

A placeholder has to start with a $ followed by an arbitrary string. A placeholder is replaced by the definition specified within the corresponding Set element. The attribute name of a Set elements specifies the placeholder, which is replaced by the value specified by the attribute value of the Set element.

In the following code example, the color of NAO’s jersey is set by a Set element. Within the definition of the body Nao named RedNao, the Set element sets the placeholder color to the value red. The placeholder named NaoColor of Nao, which is defined in the general definition of a NAO, is replaced by red in all elements of the model, also in the ones that are just referenced, such as the appearances of individual body parts. So the Surface elements reference a Surface named nao-red.

```xml
<ComplexAppearance name="CHEST_COORD_MACROCOL">
  <Surface ref="nao-$NaoColor"/>
  <Vertices ref="CHESTCOORD"/>
</ComplexAppearance>

<Surface name="nao-red" diffuseColor="rgb(100%, 0%, 100%)" ambientColor="rgb(20%, 12%, 12%)"/>
```

A.4 Attributes

A.4.1 infrastructureClass

- **Include** This tag includes a file specified by href. The included file has to start with `<Simulation>`.
  
  - href

- **Simulation**
  This element does not have any attributes.

A.4.2 setClass

- **Set** This element sets a placeholder referenced by the attribute name to the value specified by the attribute value

  - name The name of a placeholder.
    
    * Use: required
    * Range: String
  
  - value The value the placeholder is set.
    
    * Use: required
    * Range: String
A.4.3 sceneClass

- **Scene** Describes a scene and specifies the controller of the simulation.
  - **name** The identifier of the scene object (must always be `RoboCup`).
    - *Use*: optional
    - *Range*: String
  - **controller** The name of the controller library (without prefix `lib`; in our case it is `SimulatedNao`).
    - *Use*: optional
    - *Range*: String
  - **color** The background color of the scene, see A.4.23.
  - **stepLength**
    - *Units*: s
    - *Default*: 0.01s
    - *Use*: optional
    - *Range*: (0, MAXFLOAT]
  - **gravity** Sets the gravity in this scene.
    - *Units*: $\frac{mm}{s^2}, \frac{m}{s^2}$
    - *Default*: $-9.80665\frac{m}{s^2}$
    - *Use*: optional
  - **CFM** Sets ODE cfm (constraint force mixing) value.
    - *Default*: -1
    - *Use*: optional
    - *Range*: [0, 1]
  - **ERP** Set ODE erp (error reducing parameter) value.
    - *Default*: -1
    - *Use*: optional
    - *Range*: [0, 1]
  - **contactSoftERP** Sets another erp value for colliding surfaces.
    - *Default*: -1
    - *Use*: optional
    - *Range*: [0, 1]
  - **contactSoftCFM** Sets another cfm value for colliding surfaces.
    - *Default*: -1
    - *Use*: optional
    - *Range*: [0, 1]

A.4.4 solverClass

- **Quicksolver**
  - **iterations**
    - *Default*: -1
A.4. ATTRIBUTES

* Use: optional
* Range: \((0, \text{MAXINTEGER}]\)

- skip
  * Default: 1
  * Use: optional
  * Range: \((0, \text{MAXINTEGER}]\)

A.4.5 bodyClass

- **Body** Specifies an object that has a mass and can move.
  - name The name of the body.
    * Use: optional
    * Range: String

A.4.6 compoundClass

- **Compound** A Compound is a non-moving object. In contrast to the *Body* element, a compound does not require a *Mass* element as child.
  - name The name of the compound.
    * Use: optional
    * Range: String

A.4.7 jointClass

- **Hinge** Defines a hinge. To define the axis of the hinge, this element requires an axis element as child element. Furthermore, a body element is required to which the hinge is connected.
  - name The name of the hinge.
    * Use: optional
    * Range: String

- **Slider** Defines a slider. Requires an axis element to specify the axis and a body element, which defines the body this slider is connected to.
  - name The name of the slider.
    * Use: optional
    * Range: String

A.4.8 massClass

- **Mass** All this mass classes define the mass of an object.
  - name The name of the mass declaration.
    * Use: optional
    * Range: String
• **BoxMass**
  
  - name The name of the boxMass declaration.
    - * Use: optional
    - * Range: String
  
  - value The mass of the box.
    - * Units: g, kg
    - * Use: required
    - * Range: [0, \textit{MAXFLOAT}]
  
  - width The width of the box.
    - * Units: mm, cm, dm, m, km
    - * Use: required
    - * Range: \([-\textit{MAXFLOAT}, \textit{MAXFLOAT}]\]
  
  - height The height of the box.
    - * Units: mm, cm, dm, m, km
    - * Use: required
    - * Range: \([-\textit{MAXFLOAT}, \textit{MAXFLOAT}]\]
  
  - depth The depth of the box.
    - * Units: mm, cm, dm, m, km
    - * Use: required
    - * Range: \([-\textit{MAXFLOAT}, \textit{MAXFLOAT}]\]

• **SphereMass**
  
  - name The name of the sphereMass declaration.
    - * Use: optional
    - * Range: String
  
  - value The mass of the sphere.
    - * Units: g, kg
    - * Use: required
    - * Range: [0, \textit{MAXFLOAT}]
  
  - radius The radius of the sphere.
    - * Units: mm, cm, dm, m, km
    - * Use: required
    - * Range: [0, \textit{MAXFLOAT}]

• **InertiaMatrixMass** The matrix of the mass moment of inertia. Note that this matrix is a symmetric matrix.
  
  - name The name of the InertiaMatrixMass declaration.
    - * Use: optional
    - * Range: String
  
  - value The total mass.
    - * Units: g, kg
    - * Use: required
A.4. ATTRIBUTES

- **Range**: \([0, \text{MAXFLOAT}]\)
  - x The center of mass in x direction.
    - **Units**: mm, cm, dm, m, km
    - **Default**: 0
    - **Use**: optional
    - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)
  - y The center of mass in y direction.
    - **Units**: mm, cm, dm, m, km
    - **Default**: 0
    - **Use**: optional
    - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)
  - z The center of mass in z direction.
    - **Units**: mm, cm, dm, m, km
    - **Default**: 0
    - **Use**: optional
    - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- ix Moment of inertia around the x-axis when the object is rotated around the x-axis.
  - **Units**: \(g \times \text{mm}^2, kg \times m^2\)
  - **Use**: required
  - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- ixy Moment of inertia around the y-axis when the object is rotated around the x-axis.
  - **Units**: \(g \times \text{mm}^2, kg \times m^2\)
  - **Default**: 0
  - **Use**: optional
  - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- ixz Moment of inertia around the z-axis when the object is rotated around the x-axis.
  - **Units**: \(g \times \text{mm}^2, kg \times m^2\)
  - **Default**: 0
  - **Use**: optional
  - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- iyy Moment of inertia around the y-axis when the object is rotated around the y-axis.
  - **Units**: \(g \times \text{mm}^2, kg \times m^2\)
  - **Use**: required
  - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- iyz Moment of inertia around the z-axis when the object is rotated around the y-axis.
  - **Units**: \(g \times \text{mm}^2, kg \times m^2\)
  - **Default**: 0
  - **Use**: optional
  - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- izz Moment of inertia around the z-axis when the object is rotated around the z-axis.
  - **Units**: \(g \times \text{mm}^2, kg \times m^2\)
  - **Use**: required
  - **Range**: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)
A.4.9 geometryClass

- **Geometry** Elements of geometryClass specify the physical structure of an object.
  - name
    * **Use:** optional
    * **Range:** String

- **BoxGeometry**
  - color A color definition, see A.4.23
  - name
    * **Use:** optional
    * **Range:** String
  - width The width of the box.
    * **Units:** mm, cm, dm, m, km
    * **Use:** required
    * **Range:** $[-\text{MAXFLOAT, MAXFLOAT}]$
  - height The height of the box.
    * **Units:** mm, cm, dm, m, km
    * **Use:** required
    * **Range:** $[-\text{MAXFLOAT, MAXFLOAT}]$
  - depth The depths of the box.
    * **Units:** mm, cm, dm, m, km
    * **Use:** required
    * **Range:** $[-\text{MAXFLOAT, MAXFLOAT}]$

- **SphereGeometry**
  - color A color definition, see A.4.23
  - name
    * **Use:** optional
    * **Range:** String
  - radius The radius of the sphere.
    * **Units:** mm, cm, dm, m, km
    * **Use:** required
    * **Range:** $[-\text{MAXFLOAT, MAXFLOAT}]$

- **CylinderGeometry**
  - color A color definition, see A.4.23
  - name
    * **Use:** optional
    * **Range:** String
  - radius The radius of the cylinder.
    * **Units:** mm, cm, dm, m, km
A.4. ATTRIBUTES

• CapsuleGeometry
  – color A color definition, see A.4.23
  – name
    * Use: optional
    * Range: String
  – radius The radius of the capsule.
    * Units: mm, cm, dm, m, km
    * Use: required
    * Range: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)
  – height The height of the capsule.
    * Units: mm, cm, dm, m, km
    * Use: required
    * Range: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

A.4.10 materialClass

• Material Specifies a material.
  – Use: required
  – Range: String
  – name The name of the material.
    * Use: optional
    * Range: String

A.4.11 frictionClass

• Friction Specifies the friction between this material and an other material.
  – material The other material the friction belongs to.
    * Use: required
    * Range: String
  – value The value of the friction.
    * Use: required
    * Range: \([0, \text{MAXFLOAT}]\)

• RollingFriction Specifies the rolling friction of an material.
  – material The other material the rolling friction belongs to.
A.4.12 appearanceClass

- **Appearance** The appearance elements specify only shapes for the surfaces, so all appearance elements require a *Surface* specification. Appearance elements do not have a physical structure. Therefore a geometry has to be defined.

  - name The name of this appearance.
    - *Use: optional
    - *Range: String

- **BoxAppearance**

  - name The name of this appearance. To specify how it should look like an element of the type surfaceClass is needed.
    - *Use: optional
    - *Range: String
  
  - width The width of the box.
    - *Units: mm, cm, dm, m, km
    - *Use: required
    - *Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$

  - height The height of the box.
    - *Units: mm, cm, dm, m, km
    - *Use: required
    - *Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$

  - depth The depth of the box.
    - *Units: mm, cm, dm, m, km
    - *Use: required
    - *Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$

- **SphereAppearance**

  - name
    - *Use: optional
    - *Range: String
  
  - radius The radius of the sphere.
    - *Units: mm, cm, dm, m, km
    - *Use: required
    - *Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$

- **CylinderAppearance**
- **name**
  - *Use:* optional
  - *Range:* String
- **height** The height of the cylinder.
  - *Units:* mm, cm, dm, m, km
  - *Use:* required
  - *Range:* $[-MAXFLOAT, MAXFLOAT]$  
- **radius** The radius of the cylinder.
  - *Units:* mm, cm, dm, m, km
  - *Use:* required
  - *Range:* $[-MAXFLOAT, MAXFLOAT]$  

- **CapsuleAppearance**
  - **name**
    - *Use:* optional
    - *Range:* String
  - **height** The height of the capsule.
    - *Units:* mm, cm, dm, m, km
    - *Use:* required
    - *Range:* $[-MAXFLOAT, MAXFLOAT]$  
  - **radius** The radius of the capsule.
    - *Units:* mm, cm, dm, m, km
    - *Use:* required
    - *Range:* $[-MAXFLOAT, MAXFLOAT]$  

- **ComplexAppearance**
  - **name**
    - *Use:* optional
    - *Range:* String

### A.4.13 translationClass

- **Translation** Specifies a translation of an object.
  - **x** Translation in x direction.
    - *Units:* mm, cm, dm, m, km
    - *Default:* 0m
    - *Use:* optional
    - *Range:* $[-MAXFLOAT, MAXFLOAT]$  
  - **y** Translation in y direction.
    - *Units:* mm, cm, dm, m, km
    - *Default:* 0m
    - *Use:* optional
A.4.14 rotationClass

- **Rotation** Specifies the rotation of an object.
  - x Rotation around the x-axis.
    * **Units**: radian, degree
    * **Default**: 0degree
    * **Use**: optional
  - y Rotation around the y-axis.
    * **Units**: radian, degree
    * **Default**: 0degree
    * **Use**: optional
  - z Rotation around the z-axis.
    * **Units**: radian, degree
    * **Default**: 0degree
    * **Use**: optional

A.4.15 axisClass

- **Axis** Specifies the axis of a joint.
  - x
    * **Default**: 0
    * **Use**: optional
    * **Range**: \([-MAXFLOAT, MAXFLOAT]\)
  - y
    * **Default**: 0
    * **Use**: optional
    * **Range**: \([-MAXFLOAT, MAXFLOAT]\)
  - z
    * **Default**: 0
    * **Use**: optional
    * **Range**: \([-MAXFLOAT, MAXFLOAT]\)
  - cfm
    * **Default**: -1
    * **Use**: optional
    * **Range**: \([0, 1]\)
A.4.16 deflectionClass

- **Deflection** Specifies the maximum and minimum deflection of a joint.
  - min The minimal deflection.
    * **Use:** required
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - max The maximal deflection.
    * **Use:** required
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - init The initial deflection.
    * **Default:** 0
    * **Use:** optional
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - stopCFM
    * **Default:** -1
    * **Use:** optional
    * **Range:** \([0, 1]\]
  - stopERP
    * **Default:** -1
    * **Use:** optional
    * **Range:** \([0, 1]\]

A.4.17 motorClass

- **ServoMotor**
  - maxVelocity The maximum velocity of this motor.
    * **Units:** radian/s, degree/s
    * **Use:** required
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - maxForce The maximum force of this motor.
    * **Units:** N
    * **Use:** required
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - p The p value of the motor’s pid interface.
    * **Use:** required
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - i The i value of the motor’s pid interface.
    * **Default:** 0
    * **Use:** optional
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - d The d value of the motor’s pid interface.
    * **Default:** 0
• **Use:** optional
• **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

- **VelocityMotor**
  - maxVelocity The maximum velocity of this motor.
    - **Units:** radian/s, degree/s
    - **Use:** required
    - **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)
  - maxForce The maximum force of this motor.
    - **Units:** N
    - **Use:** required
    - **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

### A.4.18 surfaceClass

- **Surface** Defines the appearance of a surface.
  - diffuseColor The diffuse color, see A.4.23.
  - ambientColor The ambient color of the light, see A.4.23.
  - specularColor The specular color, see A.4.23.
  - emissionColor The emitted color of the light, see A.4.23.
  - shininess The shininess value.
    - **Default:** 0.0
    - **Use:** optional
    - **Range:** \([0.0, 128.0]\)
  - diffuseTexture A texture.
    - **Use:** optional
    - **Range:** String

### A.4.19 intSensorClass

- **Gyroscope** Mounts a gyroscope on a body.
  - name The name of the gyroscope.
    - **Use:** optional
    - **Range:** String

- **Accelerometer** Mounts an accelerometer on a body.
  - name The name of the accelerometer.
    - **Use:** optional
    - **Range:** String

- **CollisionSensor** A collision sensor which uses geometries to detect collisions with other objects.
  - name The name of the collision sensor.
    - **Use:** optional
    - **Range:** String
A.4.20 extSensorClass

- **Camera** Mounts a camera on a body.
  - name Name of the camera.
    - * Use: optional
    - * Range: String
  - imageWidth The width of the camera image.
    - * Use: required
    - * Range: Integer \(>0\)
  - imageHeight The height of the camera image.
    - * Use: required
    - * Range: Integer \(>0\)
  - angleX Opening angle in x.
    - * Units: degree, radian
    - * Use: required
    - * Range: Float \(>0\)
  - angleY Opening angle in y.
    - * Units: degree, radian
    - * Use: required
    - * Range: Float \(>0\)

- **SingleDistanceSensor**
  - name The name of the sensor.
    - * Use: optional
    - * Range: String
  - min The minimum distance this sensor can measure.
    - * Default: 0
    - * Use: optional
    - * Range: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
  - max The maximum distance this sensor can measure.
    - * Default: 999999.f
    - * Use: optional
    - * Range: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]

- **ApproxDistanceSensor**
  - name The name of the sensor.
    - * Use: optional
    - * Range: String
  - min The minimum distance this sensor can measure.
    - * Default: 0
    - * Use: optional
    - * Range: \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\]
– **max** The maximum distance this sensor can measure.
  * **Default:** 999999.f
  * **Use:** optional
  * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

– **angleX** The maximum angle in x-direction the ray of the sensor can spread.
  * **Units:** degree, radian
  * **Use:** required
  * **Range:** Float > 0

– **angleY** The maximum angle in y-direction the ray of the sensor can spread.
  * **Units:** degree, radian
  * **Use:** required
  * **Range:** Float > 0

• **DepthImageSensor**
  
  – **name**
    * **Use:** optional
    * **Range:** String

  – **imageWidth** The width of the image.
    * **Use:** required
    * **Range:** Integer > 0

  – **imageHeight** The height of the image.
    * **Default:** 1
    * **Use:** optional
    * **Range:** Integer > 0

  – **angleX**
    * **Units:** degree, radian
    * **Use:** required
    * **Range:** Float > 0

  – **angleY**
    * **Units:** degree, radian
    * **Use:** required
    * **Range:** Float > 0

  – **min** The minimum distance this sensor can measure.
    * **Default:** 0
    * **Use:** optional
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

  – **max** The maximum distance this sensor can measure.
    * **Default:** 999999.f
    * **Use:** optional
    * **Range:** \([-\text{MAXFLOAT}, \text{MAXFLOAT}]\)

  – **projection** The kind of projection.
    * **Default:** perspective
    * **Use:** optional
    * **Range:** perspective, spheric
A.4.21 userInputClass

- **UserInput** Combines an actuator and a sensor. The values set for the actuator are directly returned by the sensor. Using the actuator view, this allows to feed user input to the controller.
  - name Name of the user input.
    - Use: optional
    - Range: String
  - type The kind of data for which user input is provided.
    - Default: length
    - Use: optional
    - Range: angle, angularVelocity, length, velocity, acceleration
  - min The minimum value that can be set.
    - Units: Units matching the type
    - Use: required
    - Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$
  - max The maximum value that can be set.
    - Units: Units matching the type
    - Use: required
    - Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$
  - default The value returned by the sensor if no value is set in the actuator view.
    - Units: Units matching the type
    - Default: 0
    - Use: optional
    - Range: $[-\text{MAXFLOAT}, \text{MAXFLOAT}]$

A.4.22 lightClass

- **Light** Definition of a light source.
  - diffuseColor Diffuse color definition, see A.4.23
  - ambientColor Ambient color definition, see A.4.23
  - specularColor Specular color definition, see A.4.23
  - x The x position of the light source.
    - Units: mm, cm, dm, m, km
    - Use: optional
    - Range: o
  - y The y position of the light source.
    - Units: mm, cm, dm, m, km
    - Use: optional
    - Range: o
  - z The z position of the light source.
    - Units: mm, cm, dm, m, km
∗ **Use:** optional
∗ **Range:** 0
  − constant attenuation The constant attenuation of the light.
    ∗ **Use:** optional
    ∗ **Range:** $[0.f, MAXFLOAT]$ 
  − linear attenuation The linear attenuation of the light.
    ∗ **Use:** optional
    ∗ **Range:** $[0.f, MAXFLOAT]$ 
  − quadratic attenuation The quadratic attenuation of the light.
    ∗ **Use:** optional
    ∗ **Range:** $[0.f, MAXFLOAT]$ 
  − spotCutoff
    ∗ **Units:** mm, cm, dm, m, km
    ∗ **Use:** optional
    ∗ **Range:** $[-MAXFLOAT, MAXFLOAT]$ 
  − spotDirectionX The x direction of the light spot.
    ∗ **Units:** mm, cm, dm, m, km
    ∗ **Use:** optional
    ∗ **Range:** $[-MAXFLOAT, MAXFLOAT]$ 
  − spotDirectionY The y direction of the light spot.
    ∗ **Units:** mm, cm, dm, m, km
    ∗ **Use:** optional
    ∗ **Range:** $[-MAXFLOAT, MAXFLOAT]$ 
  − spotDirectionZ The z direction of the light spot.
    ∗ **Units:** mm, cm, dm, m, km
    ∗ **Use:** optional
    ∗ **Range:** $[-MAXFLOAT, MAXFLOAT]$ 
  − spotExponent
    ∗ **Use:** optional
    ∗ **Range:** $[0.f, 128.f]$ 

### A.4.23 Color Specification

There are two ways of specifying a color for a color-attribute.

- **HTML-Style** To specify a color in html-style the first sign of the color value has to be a # followed by hexadecimal values for red, blue, green (and maybe a fourth value for the alpha-channel). These values can be one-digit or two-digits, but not mixed.
  - #rgb e.g. #f00
  - #rgba e.g. #0f0a
  - #rrggbbaa e.g. #f80011
  - #rrggbbaa e.g. #1038bc
• **CSS-Style** A css color starts with rgb (or rgba) followed by the values for red, green, blue put into brackets and separated by commas. The values for r, g, b has to be between 0 and 255 or between 0% and 100%, the a-value has to be between 0 and 1.

  - rgb(r, g, b) e.g. rgb(255, 128, 0)
  - rgba(r, g, b, a) e.g. rgba(0%, 50%, 75%, 0.75)
Appendix B

Camera Kernel Module

The default kernel module for the NAO cameras bundled by SoftBank Robotics has some disadvantages. In low light situations the camera image can turn dark. This happens frequently in duelling situations for the lower camera, when the robots stand close to one another and produce a lot of shadows. Additionally, when playing with auto exposure, the power line frequency needs to be known to the camera in order to avoid flicker from lights. Apart from those problem fixes two new features where added to the driver. A gamma correction setting; and most importantly a one time auto white balance. For a detailed description on all settings, and on how to build the module, please read the README.md of the BKernel repository on GitHub at https://github.com/bhuman/BKernel.