

# Software Survey 2026

## Team name

Invic

**Which division(s) are you applying for? If your used software differs between divisions, please fill out the survey once per division.**

Small Size (height < 110 cm, weight < 15 kg)

**Is your software fully or partially OpenSource? If so, where can it be found?**

Yes, we built our own software system based on the Bhuman framework, which can be found at <https://github.com/bhuman/BHumanCodeRelease>.

**Are you using any software developed by other teams? If so, list every component that you are reusing and the team that originally developed it.**

Yes, Bhuman: SimRobot.

**Are you using any datasets in your research? If you are using your own datasets, are they public?**

Yes, but they are not public

**Please list the scientific publications your team has made since the last application to RoboCup (or if not applicable in the last 2 years).**

It has not been published yet. We will release it once we have further results.

**Are there any other contributions you would like to share with the RoboCup community?**

**Which approach are you using to generate the robot walking motion?**

The ZMP Preview Controller, based on a 3D linear inverted pendulum model, generates walking motions. Its core module calculates the center of mass and foot trajectory in real-time to achieve dynamic equilibrium, integrating real-time feedback from IMUs and foot sensors for

posture stabilization, arm coordination, and reactive stride adjustments. The entire system is integrated into a modular motion framework, supporting multi-gait switching and smooth motion transitions.

**Which approach are you using to generate other motions of the robot (e.g. kicking, standing up)?**

All robot motions beyond basic walking are primarily generated using predefined keyframe animations, with each action encapsulated as a reusable MotionPhase object. These phases are dynamically created by the MotionGenerator according to behavioral requirements and executed within a unified motion engine. The entire framework supports automatic phase switching, sensor feedback adaptation, and exception handling, ensuring robustness while maintaining highly modular code.

**Do you have a kinematic or dynamic model of your robot? If so, how did you create it (e.g. measure physical robot, export from CAD model)?**

measure physical robot

**What approaches are you using in your robot's visual perception?**

Visual Perception Solution:

Deep learning is employed only for the most critical task (site boundary detection), while other tasks rely on efficient classical algorithms + geometric constraints.

Deeply integrated robot models (BodyContour, CameraMatrix),

fully automated calibration and exposure, eliminating manual parameter tuning.

The entire system is specifically designed for NAO's limited CPU resources, maintaining <30ms latency while delivering exceptional environmental adaptability.

**Are you planning with objects in Cartesian or image space? If you are using Cartesian space, how do you transform between the image space and cartesian space?**

Planning is performed entirely in Cartesian space (world coordinate system).

The conversion process from image space to Cartesian space is as follows:

Perform rolling shutter and motion correction on raw pixels;

Calculate the camera's pose relative to the robot using joint angles + IMU + kinematic model;

Map pixels to the robot's coordinates on the ground plane via inverse perspective projection;  
Then combine with self-localization results to transform into global field coordinates.

**Do you have some form of active vision (i.e. moving the robots camera based on information known about the world)?**

Yes.

**What approach are you using to localize your robot?**

Self-localization employs a hybrid probabilistic framework combining "multi-hypothesis UKF + lightweight particle filtering."

It relies entirely on field markers without utilizing goalposts;

Rapidly recovers from falls or disturbances via a sensor reset mechanism;

Resolves field symmetry ambiguities using game rules + temporal continuity;

Ultimately outputs high-precision, low-latency global pose estimates, enabling accurate walking, passing, and tactical coordination.

**Is your team performing team communication? Which communication protocol are you using?**

Yes, use UDP

**What approach are you using for navigation? Are you avoiding obstacles?**

Its navigation system employs a layered design:

Long-range □ Utilizes A + visibility maps for global path planning, avoiding teammates and opponents;

Short-range or sudden situations □ Employs reactive local obstacle avoidance, dynamically adjusting movement direction;

Obstacle information integrates visual detection with team communication, with avoidance logic spanning both planning and execution phases.

**How is the behavior of your robots structured? (e.g. Behaviour Trees)**

Finite State Machine

**Are you simulating your robot? If so, which simulator are you using and for what purpose do you use simulations?**

Yes, use SimRobot, Pre-test before computer lab session

**What operating system is running on your robot and which middleware are you using (for example Ubuntu 22.04 and ROS2 Galactic)?**

The NAO robot runs on a customized Linux system and utilizes a proprietary middleware framework developed by B-Human.

**Is there anything else you would like to share that did not fit any previous question?**