



Fast and Robust Edge-Based Localization in the Sony Four- Legged Robot League

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Outline of the Talk

- Motivation
- Detecting edges
- Monte-Carlo Localization
 - Sensor model
 - Details
- Experiments
- Localization in real games
- Conclusions

Localization in the Sony Four-Legged Robot League

- Advantages
 - Automatic positioning
 - Sharing perceptions
 - Full support of referee commands
- Challenges
 - Vision-based
 - Directed vision
 - Variable camera position
 - Limited computing power

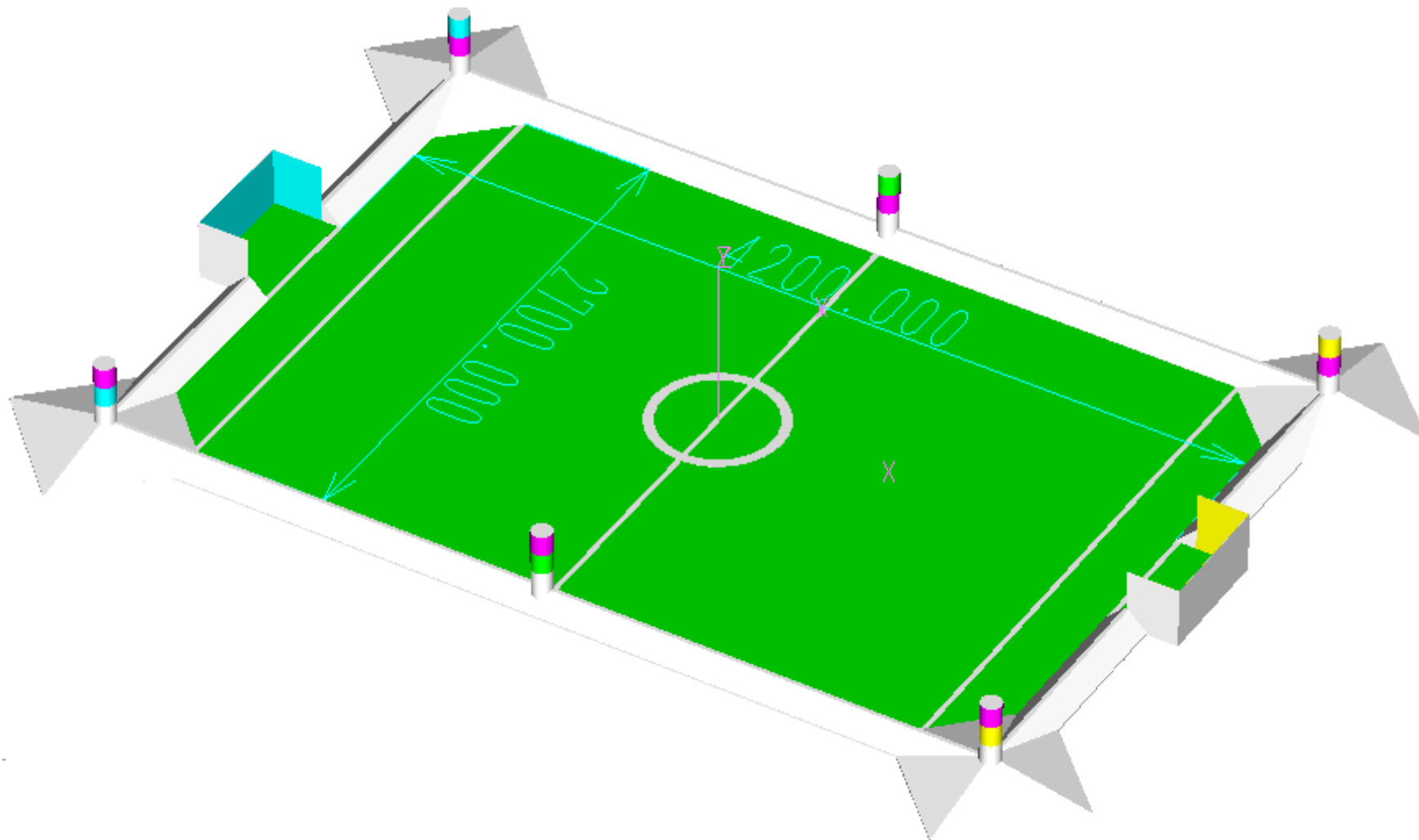


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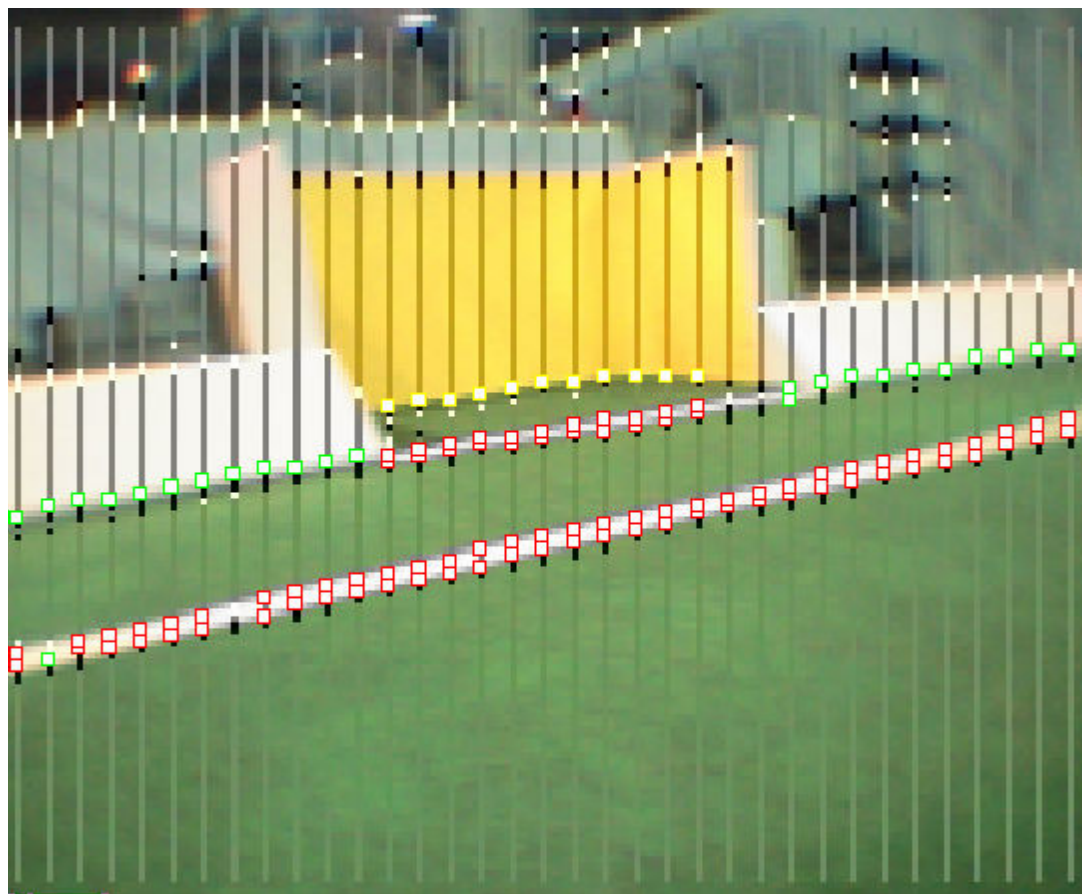


The Field

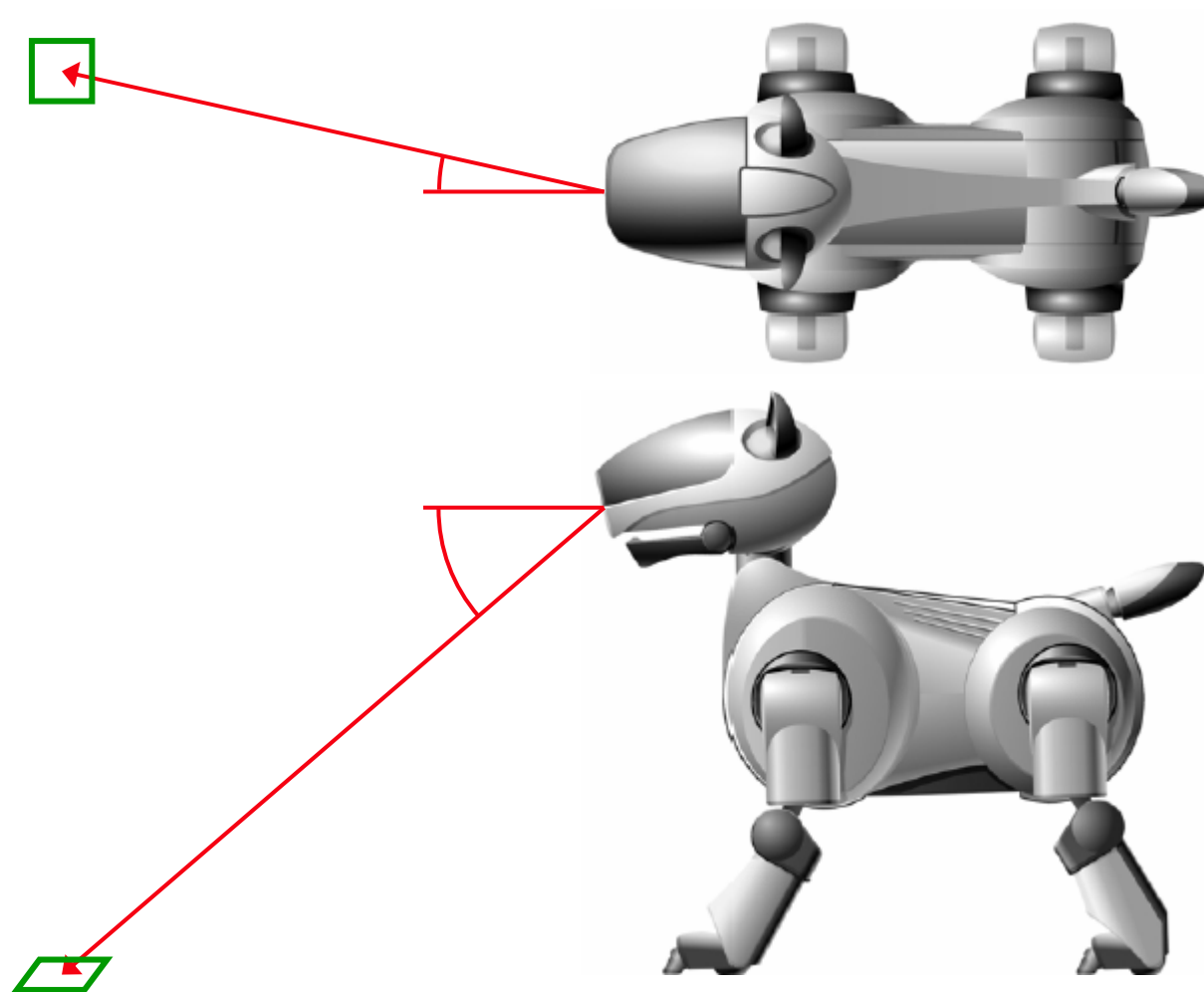


Detecting Edges

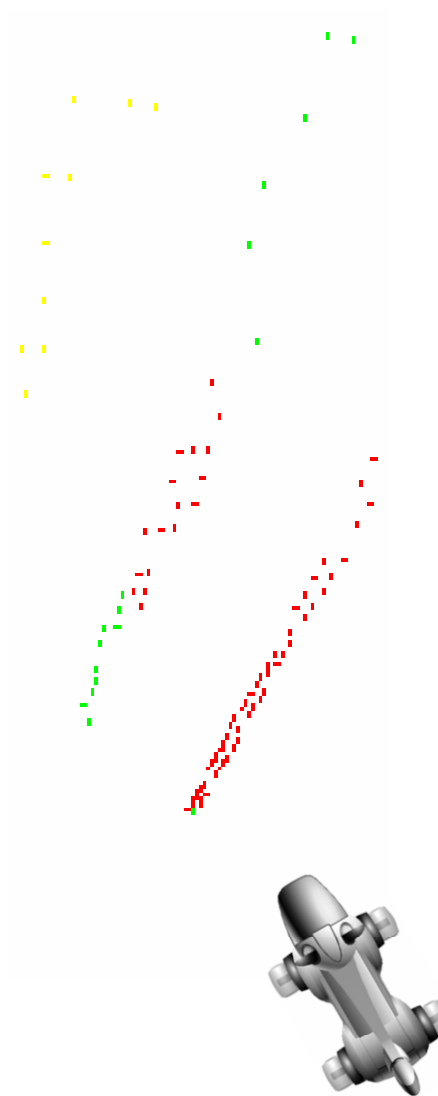
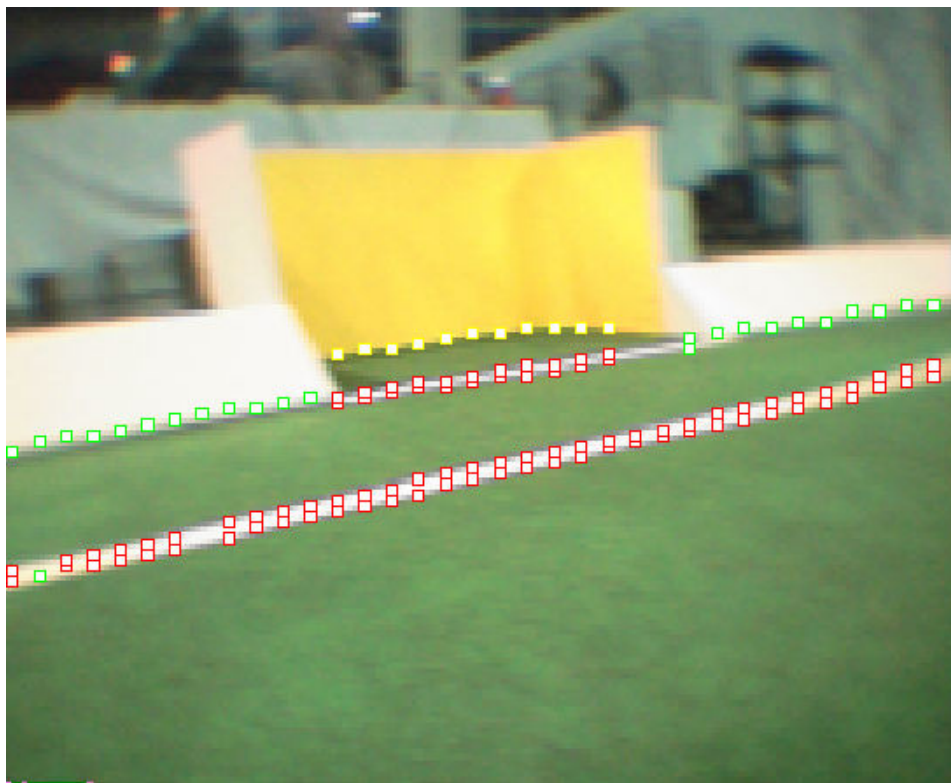
- Between field and
 - Border
 - Field lines
 - Goals
 - yellow
 - skyblue



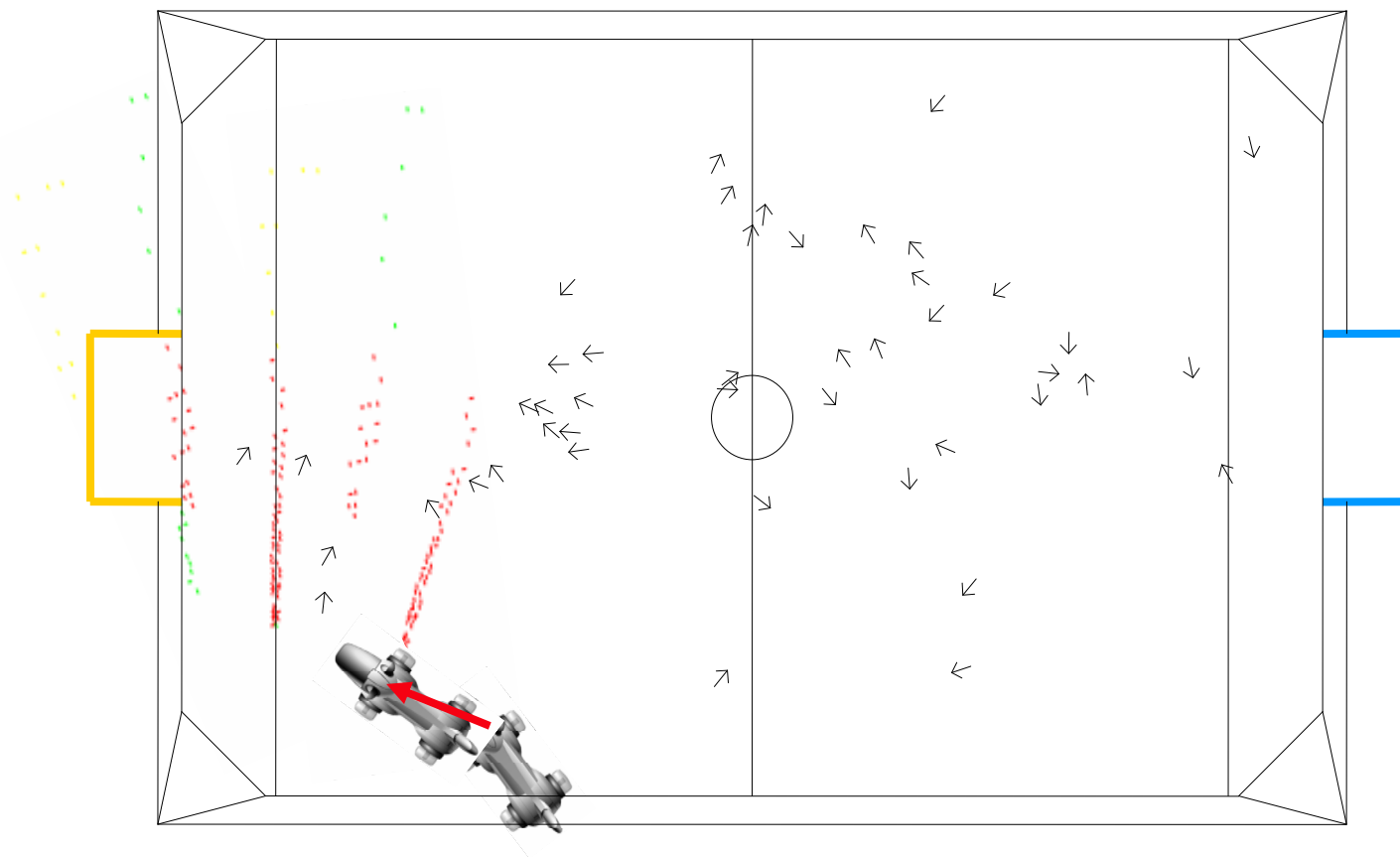
Projection on the Field



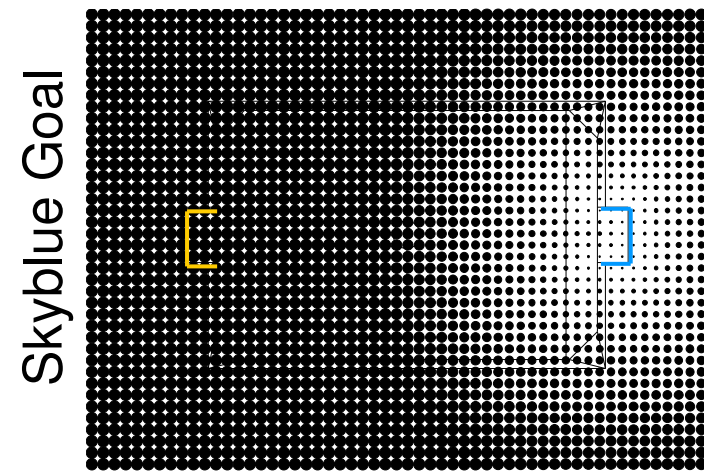
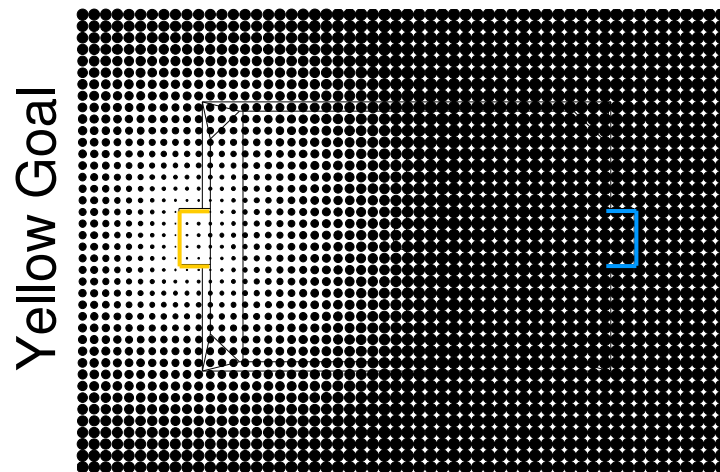
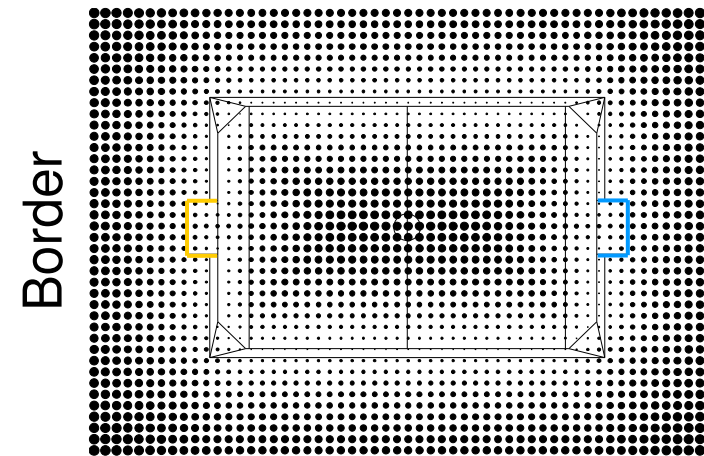
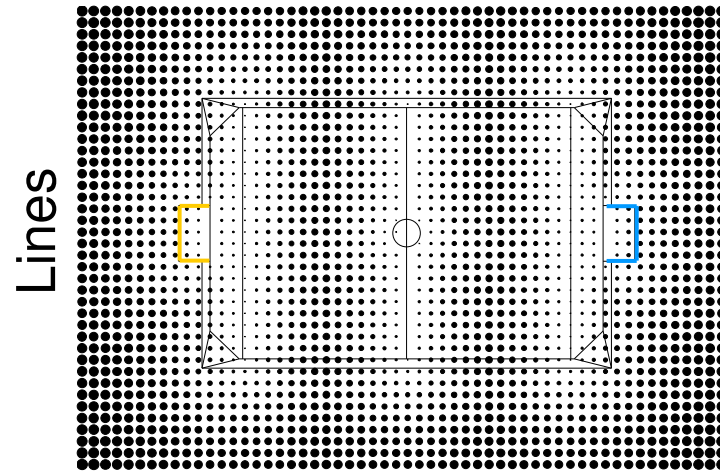
Projection on the Field



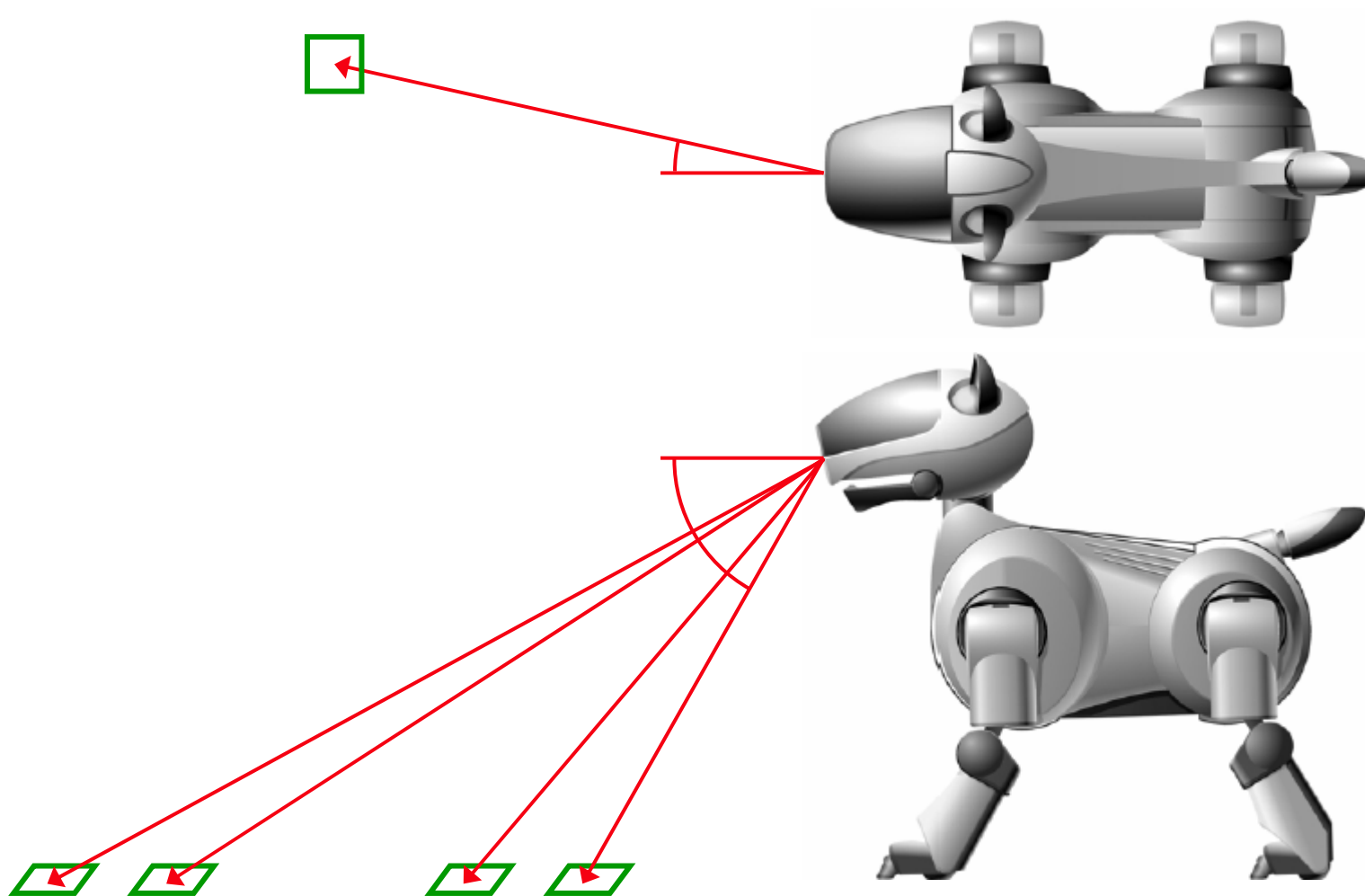
Approach



Assigning Observations to Field Model



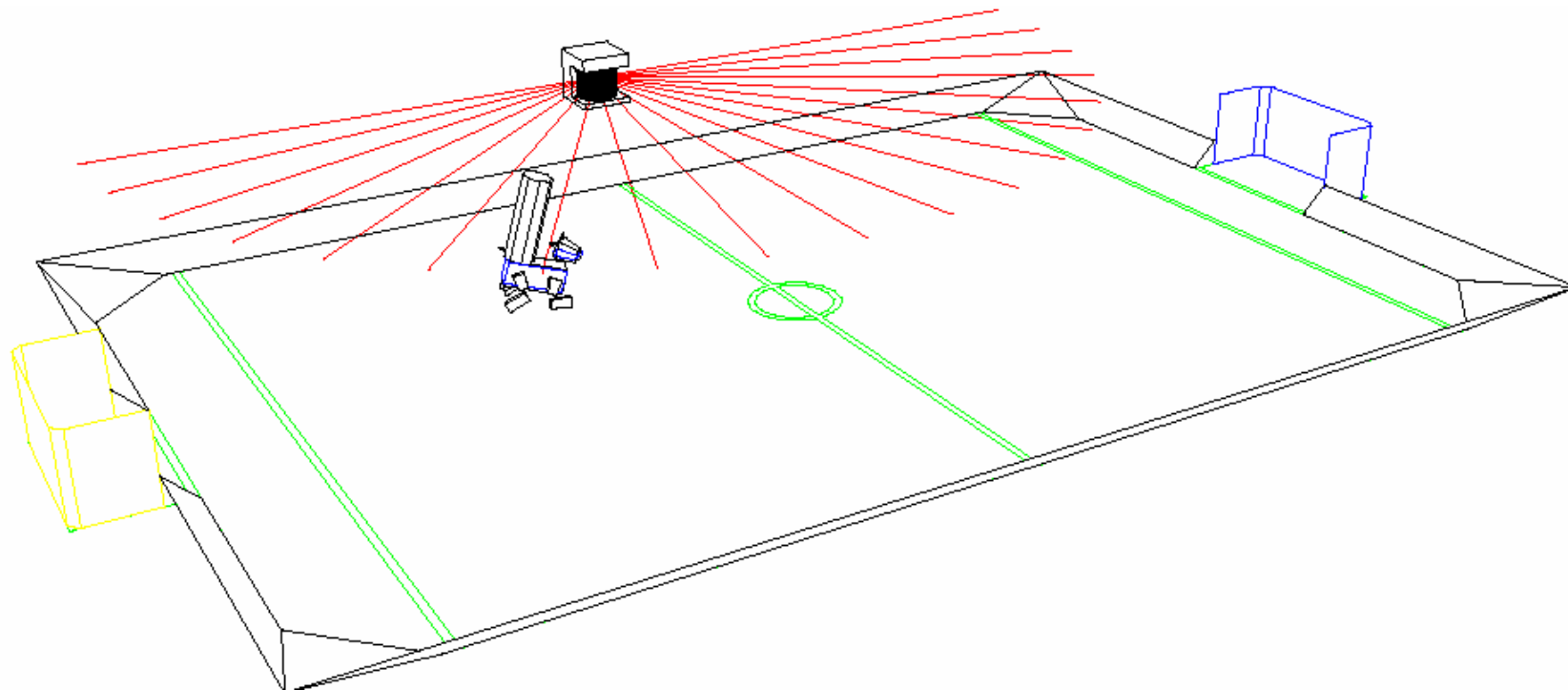
Sensor Model



Details

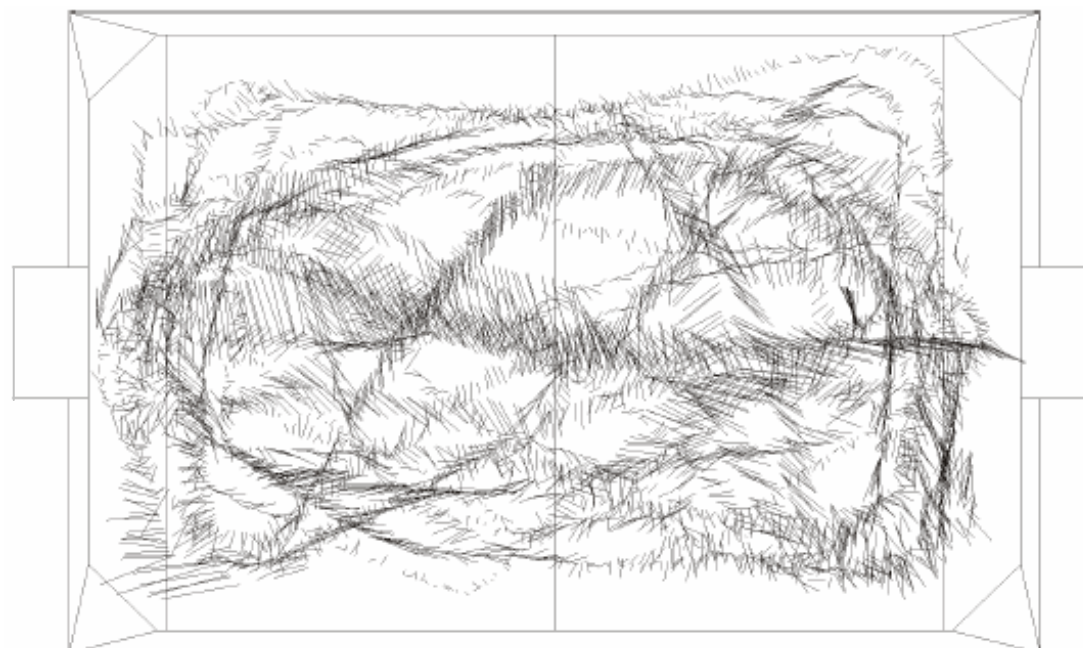
- Probability of samples
 - Probability is adapted slowly
 - Separate probabilities for different edge types
 - Samples are randomly moved, weighted by their probabilities
- Sensor resetting
 - Draw samples based on the ratio of their probability and the average probability
 - Replace them by candidate postures that can be derived from observations
- Calculating candidates in advance
 - A large number of random postures is generated
 - Their distance to the edge they are pointing to is determined
 - The postures are indexed by their distance and edge type

Experimental Setup



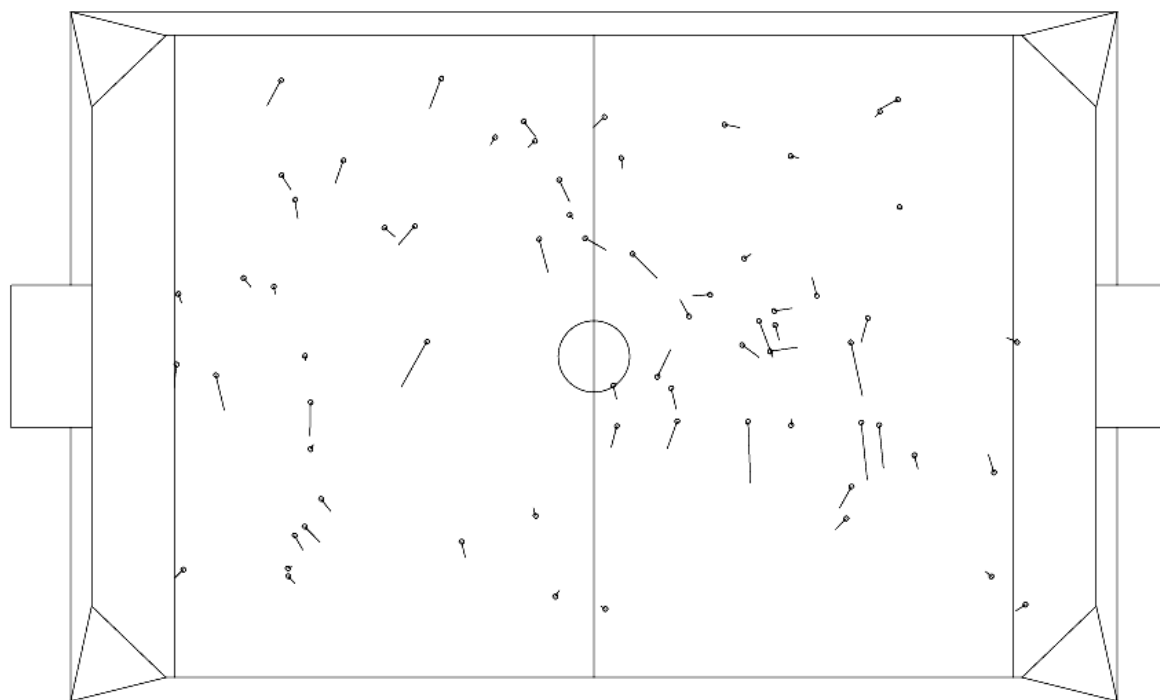
Experiment 1

- Robot continuously moving (by joystick)
- Approx. 5300 measurements
- Average error < 10.5 cm
(field size is 420 x 270 cm²)

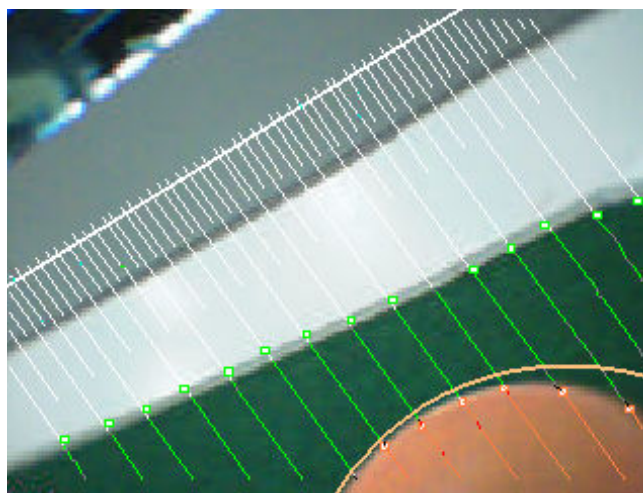
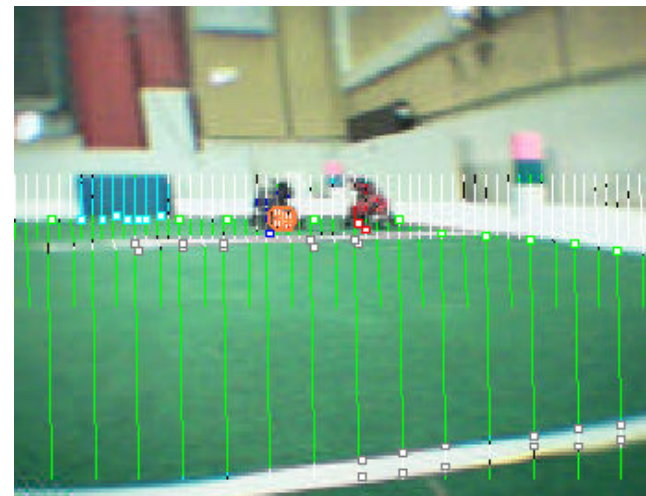
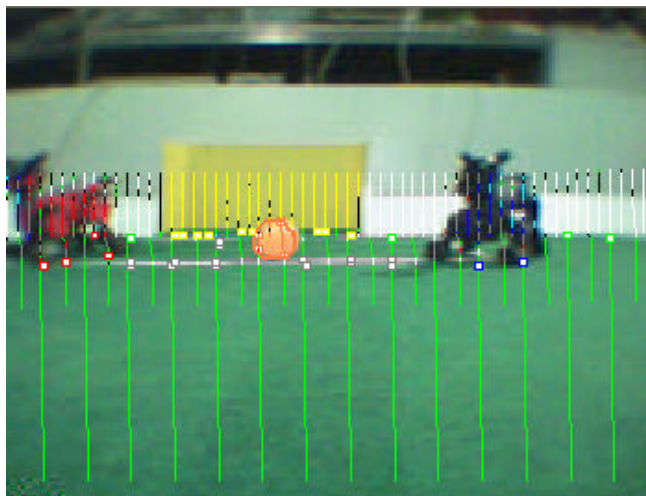


Experiment 2

- Robot walks to random positions (approx. 70)
- Average error in positioning < 9.5 cm
- Average error in localization < 8.5 cm



Edge-based Localization in Real Games

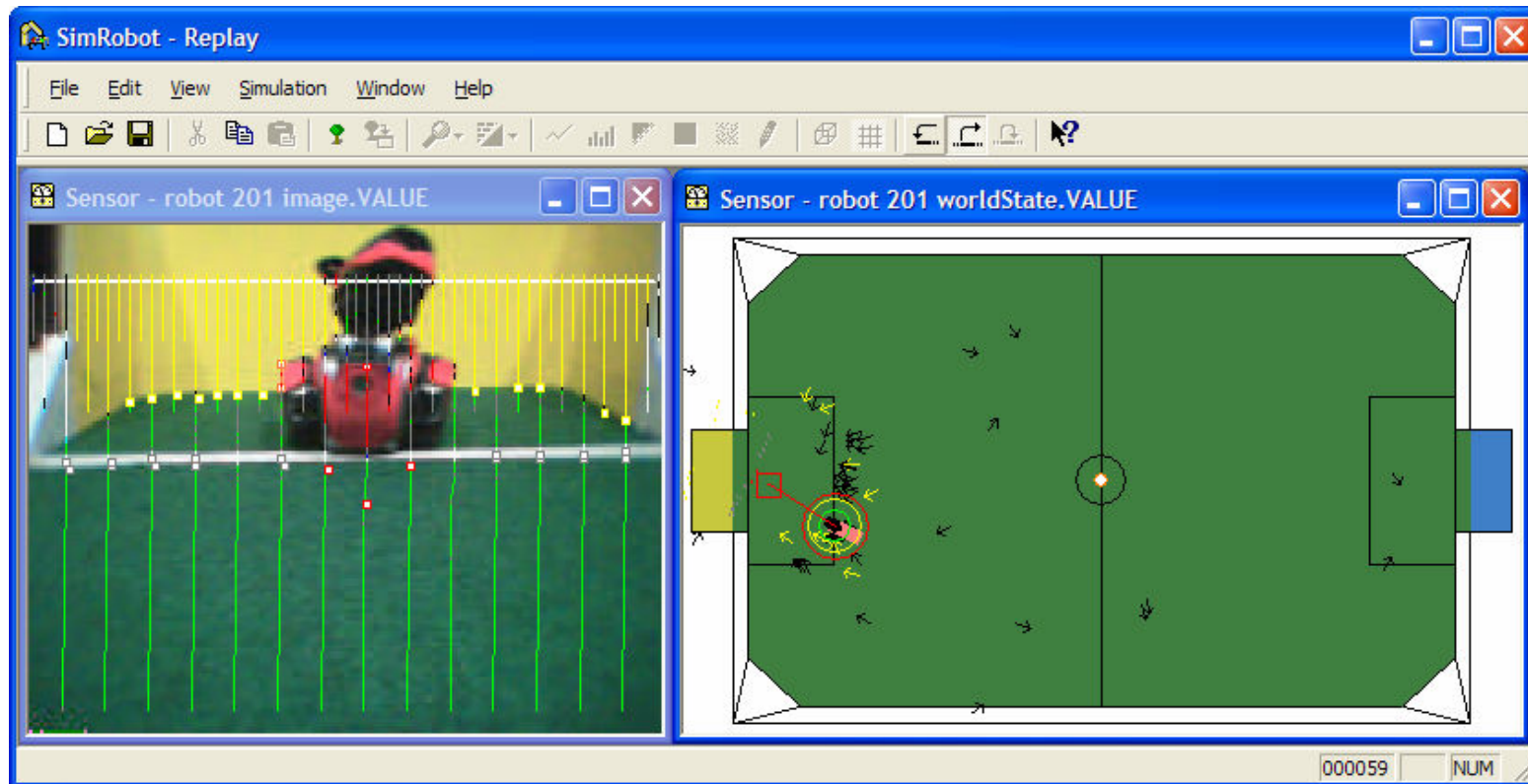




Improvements since Writing the Paper

- Candidate postures only result from goal points
- Samples are moved in direction of candidate postures, they are not replaced by them
- The speed of this motion depends on the speed of the robot (the faster the robot walks, the slower the samples adapt)
- Samples are also moved according to the assignment of measured points to model points (weighted by the distance to the measured points)

Example



Conclusions

- Fast and robust Monte-Carlo localization
- Using edges between field and border/lines/goals
- Average error < 10.5 cm
- Works in real games

- In RoboCup 2003
 - Played with combined localizer (edges + landmarks)
 - Demonstrated match (GT vs. GT) without landmarks
- In RoboCup 2004
 - Removal of landmarks?



Other Talks by Members of the GermanTeam

Vision 1, 1.1

Thursday 14:30

Matthias Jünger, Jan Hoffmann, Martin Löttsch

A Real-Time Auto-Adjusting Vision System for Robotic Soccer

Vision 1, 1.2

Thursday, 14:55

Ingo Dahm, Sebastian Deutsch, Matthias Heibel, André Osterhues

Fully Autonomous Robust Color Classification

AI 1, 1.2

Thursday 14:55

Andrea Miene, Ubbo Visser, Otthein Herzog

Recognition and prediction of motion situations based on a qualitative motion description

AI 2, 2.1

Thursday 16:25

Martin Löttsch, Joscha Bach, Hans-Dieter Burkhard, Matthias Jünger

Designing Agent Behavior with the Extensible Agent Behavior Specification Language

Vision 2, 2.3

Thursday, 17:15

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Vision 2, 2.4

Thursday, 17:40

Kai Hübner

A Symmetry Operator and its Application to the RoboCup
