

Non-Monotonic Reasoning for Localization in RoboCup

David Billington

Vladimir Estivill-Castro

Rene Hexel

Andre Rock



Australia

Non-Monotonic Reasoning for RoboCup

Outline

- Reasoning and Localization
- Why reasoning and modelling with logic
 - The Software Engineering justification
 - The Hybrid Intelligent Agent justification
- Running reasoning on a AIBO ERS-7
- Model Development and Results
- Conclusion



Reasoning

- Deriving conclusions from facts
 - Apparently, a fundamental characteristic of intelligence
- An expected aspect of intelligent systems
- Withdrawing conclusions in the light of new evidence is a capability usually referred to as non-monotonic reasoning

Our environment

RoboCup

- A test-bed for Multi-Agent Systems
- We know our environment, so one would expect to be able to construct a knowledge base and apply reasoning



RoboCup environment is hard

- Non-deterministic
 - I can not predict the state of the environment after I perform an action
- Not accessible
 - I can not sense all elements of the environment
- Dynamic
 - Environment changes while I decide what action to take
- Teams
 - I need to negotiate, collaborate, distribute tasks and goals
- Adversaries
 - Of unknown capabilities

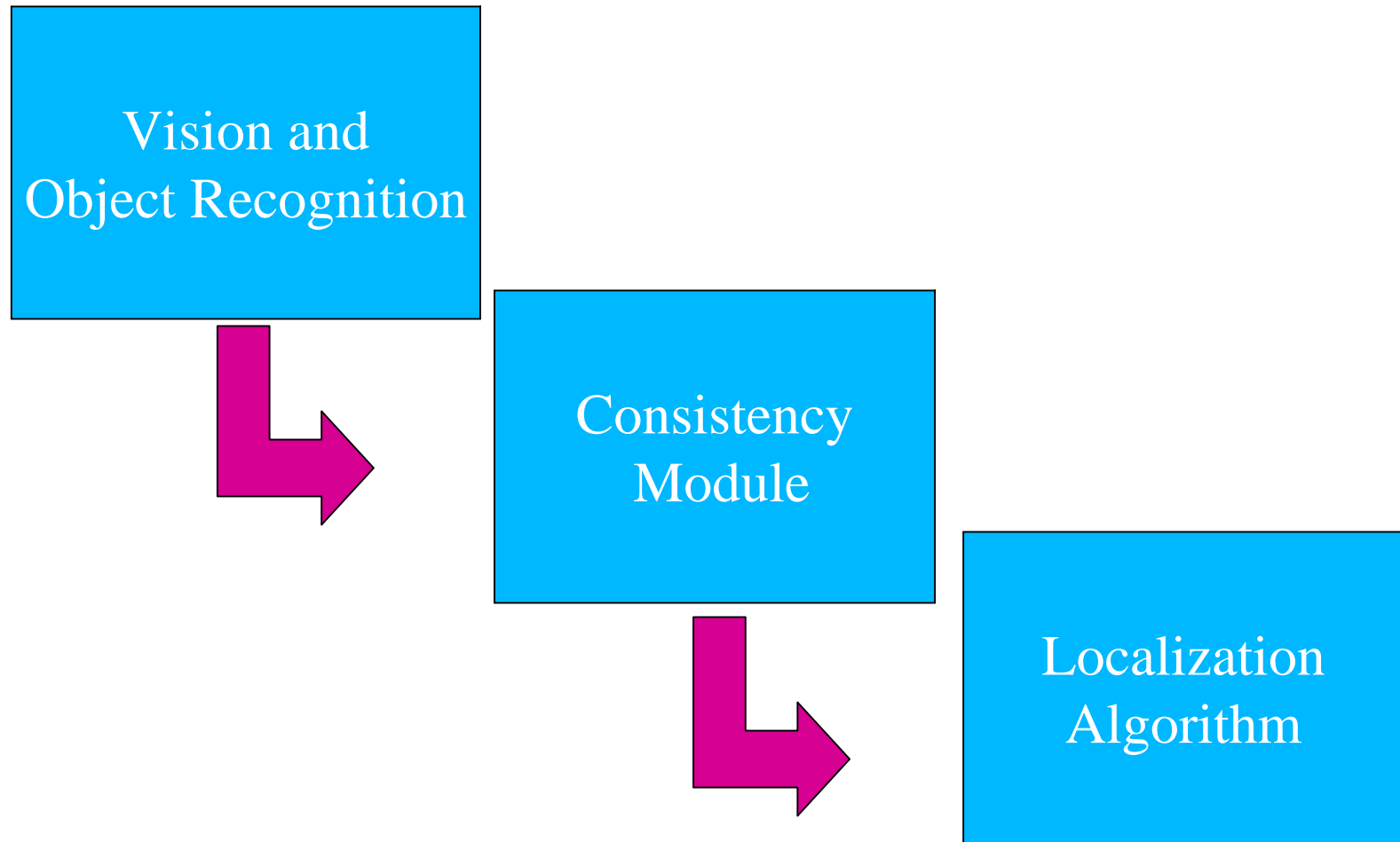
We demonstrate reasoning within the task of localization

- ▀ Dynamically selecting proper inputs for localization
 - The classical example in RoboCup for the Aibo league is that
 - A frame where both goals are visible indicates something wrong with the object recognition task

Possible solutions

- Introduce sanity checks
 - Filter out the frame if both goals are visible
- Pass it out to localization and expect the sophistication of the algorithm (capacity to handle error in sensor input) to handle these cases
 - Kalman Filter
 - Markov Localization
 - Monte-Carlo localization

Our approach



Our approach

Consistency
Module

Non-monotonic logic that combines facts known
about the environment with what is reported
as visible in this frame

Why non-monotonic logic

- To reason about the inconsistent information provided by the sensors (vision)
- Without reasoning, all localization methods must determine
 - $Prob(\text{visible scene} \mid \text{position})$

Problem with localization methods

■ Illustration

■ $Prob(\text{front goal visible} \ \& \ \text{back goal visible} \mid \text{position}) = 0$

- Not the best answer, or defines a large set of special cases
- It is hard to express it as function of

$Prob(\text{front goal visible} \mid \text{position})$

and

$Prob(\text{front goal visible} \mid \text{position})$

Plausible logic

- Only non-monotonic logic with an efficient non-looping algorithm
- Can prove using factual information and also plausible information
- 3 types of rules
 - $A \rightarrow l$ (factual information)
 - $Human(x) \rightarrow Mammal(x)$ [All humans are mammals]
 - $B \Rightarrow f$ (plausible situations)
 - $Bird(x) \Rightarrow Fly(x)$ [Birds usually fly]
 - $A \not\Rightarrow \neg l$
 - $\{sick(x), Bird(x)\} \not\Rightarrow Fly(x)$ [Sick birds may not fly]

Plausible logic (cont)

- Rules are in an acyclic hierarchy
 - $R_i > R_j$
 - Rule i is more informative than rule j .
- Conclusion with one rule may be defeated by the more informative rule

Why reasoning and modelling with logic

■ The Software Engineering justification

- All the “intelligence” (logic) about what makes sense in an image (or sequence of images) is properly encapsulated in a human understandable logic
 - Not a series of “if ..then ...else” statements of C++ in the code
 - Can test completeness and correctness
 - Can be updated easily

■ The Hybrid Intelligent Agent justification

- A higher level description that allows reasoning

14

Illustration

- Naturally to develop rules systems where the new rules redefine exception to the previous ones
 - 3 laws
 1. A robot may not harm a human
 2. A robot must obey a human unless it contradict law 1
 3. A robot must protect itself unless contradicts rule 1 or 2
 - Ripple down rules
 - Rules are defined and new rules are subsequently added to revise the cases not covered by the more general rules
 - A tree that is a hierarchy of rules
 - No formal reasoning

Modelling with standard logic

- A first model

$$\{See(x)\} \cup \{\neg See\{y\} : y \in Landmarks-\{x\}\} \rightarrow Cs(x)$$

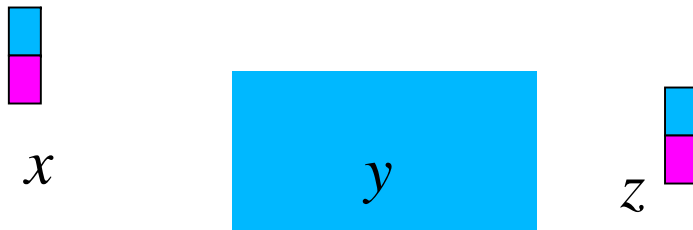
- If I only see one object, then it is consistent

Modelling with standard logic

- A second model

$$C(x,y)=\{See(x),See(y)\} \cup \{\neg See\{z\} : z \in Landmarks-\{x,y\}\}$$

1. $\{SeeLtoR(x,y), FactLtoR(x,y,z)\} \cup C(x,y) \rightarrow Cs(x,y)$
2. $\{SeeLtoR(y,x), FactLtoR(x,y,z)\} \cup C(x,y) \rightarrow CsI(x,y)$
3. $\{CsI(x,y), Post(x), Goal(y)\} \rightarrow Cs(x)$
4. $\{CsI(x,y), Post(x), Post(y), BigSmall(x,y)\} \rightarrow Cs(x)$



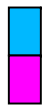
The world

Modelling with standard logic

- A second model

$$C(x,y) = \{See(x), See(y)\} \cup \{\neg See\{z\} : z \in Landmarks - \{x,y\}\}$$

1. $\{SeeLtoR(x,y), FactLtoR(x,y,z)\} \cup C(x,y) \rightarrow Cs(x,y)$
2. $\{SeeLtoR(y,x), FactLtoR(x,y,z)\} \cup C(x,y) \rightarrow CsI(x,y)$
3. $\{CsI(x,y), Post(x), Goal(y)\} \rightarrow Cs(x)$
4. $\{CsI(x,y), Post(x), Post(y), BigSmall(x,y)\} \rightarrow Cs(x)$



x

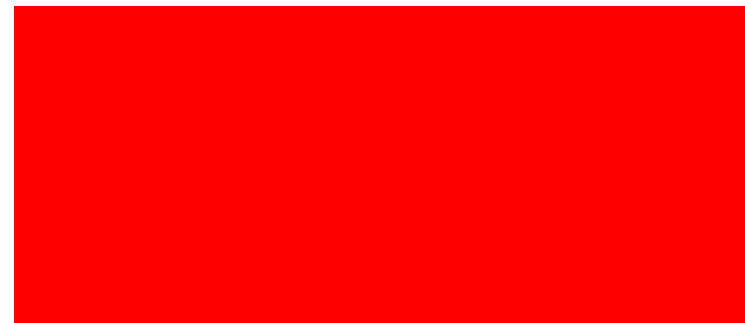


y



z

The world



Vision

Modelling with standard logic

- A second model

$$C(x,y) = \{See(x), See(y)\} \cup \{\neg See\{z\} : z \in Landmarks - \{x,y\}\}$$

1. $\{SeeLtoR(x,y), \textcolor{green}{FactLtoR}(x,y,z)\} \cup C(x,y) \rightarrow Cs(x,y)$
2. $\{SeeLtoR(y,x), \textcolor{green}{FactLtoR}(x,y,z)\} \cup C(x,y) \rightarrow CsI(x,y)$
3. $\{CsI(x,y), \textcolor{red}{Post}(x), \textcolor{red}{Goal}(y)\} \rightarrow Cs(x)$
4. $\{CsI(x,y), \textcolor{red}{Post}(x), \textcolor{red}{Post}(y), \textcolor{red}{BigSmall}(x,y)\} \rightarrow Cs(x)$



x

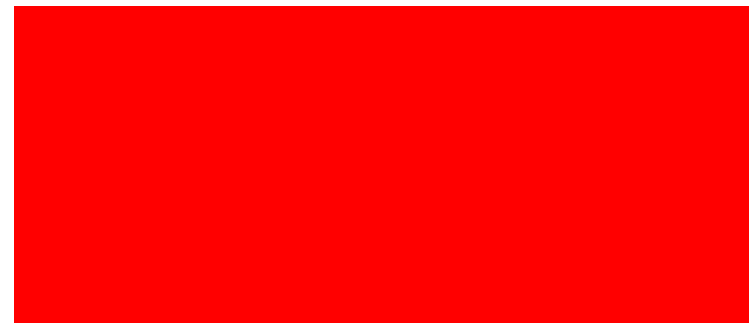


y



z

The world



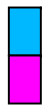
Vision

Modelling with standard logic

- A second model

$$C(x,y) = \{See(x), See(y)\} \cup \{\neg See\{z\} : z \in Landmarks - \{x,y\}\}$$

1. $\{SeeLtoR(x,y), \textcolor{green}{FactLtoR}(x,y,z)\} \cup C(x,y) \rightarrow Cs(x,y)$
2. $\{SeeLtoR(y,x), \textcolor{green}{FactLtoR}(x,y,z)\} \cup C(x,y) \rightarrow CsI(x,y)$
3. $\{CsI(x,y), \textcolor{red}{Post}(x), \textcolor{red}{Goal}(y)\} \rightarrow Cs(x)$
4. $\{CsI(x,y), \textcolor{red}{Post}(x), \textcolor{red}{Post}(y), \textcolor{red}{BigSmall}(x,y)\} \rightarrow Cs(x)$



x



y



z

The world



Vision

Problems with standard logic

- ▶ Rapidly we have the same situation
 - Many different cases coded essentially independently
 - Seeing exactly 3 objects need 26 rules
 - Seeing exactly 4 objects needs 120 rules
 - Proves most C++ is incomplete
 - (and perhaps inconsistent)
 - Survives because of the frame rate
 - Concerns on correctness/reliability of intelligent systems

Implementing Plausible Logic on SONY Aibo

- Besides Plausible Logic we develop a Logic Programming Language - DPL
 - Create definitions, macros, determine what to prove
- A HASKELL implementation of the inference algorithm of plausible logic
 - A program in DPL that proves off-line
 - Finds the equivalent logic expression to Cs(Front goal) in terms of **World** predicates and **Test** predicates
- A simulator for validation of-line and gluing code
- A Template method in the consistency module on the Aibo

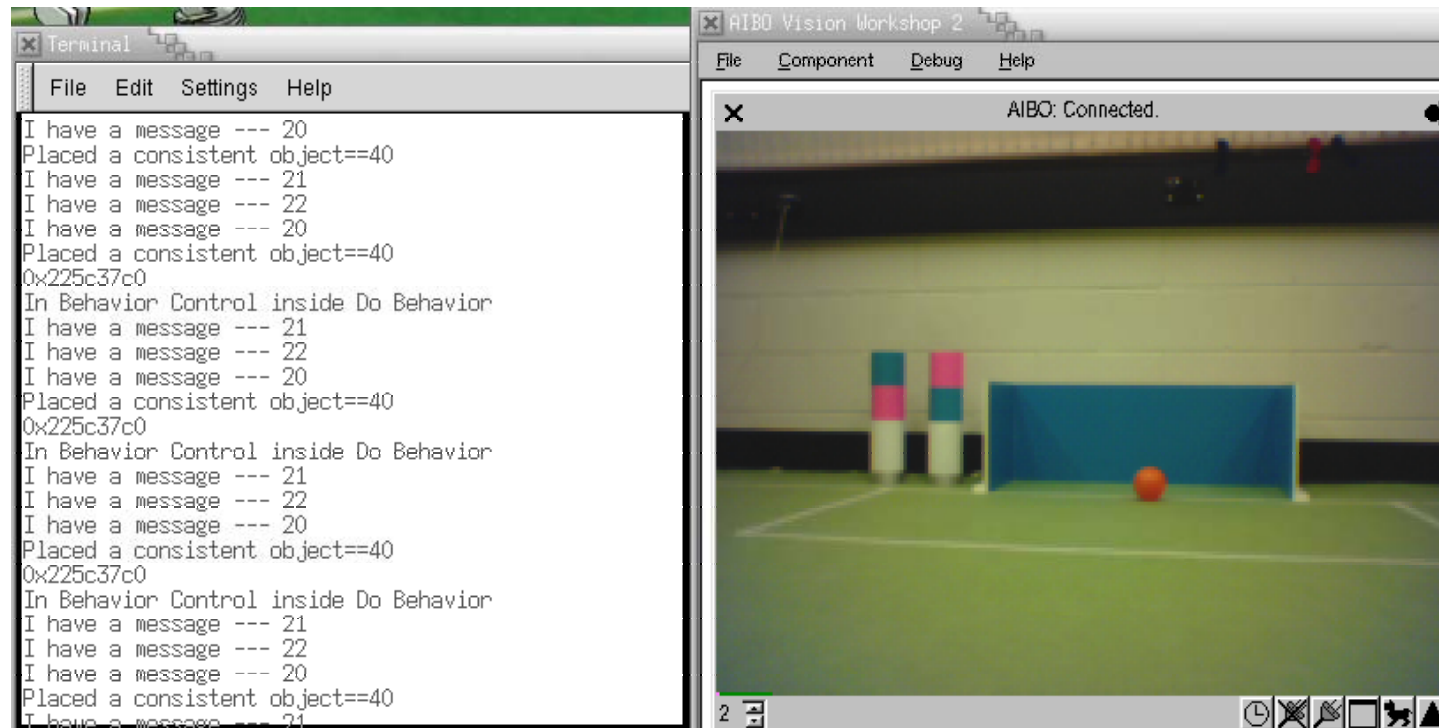
Model 1

- ▀ $R1: \Rightarrow \sim Cs(x) .$
- ▀ $R2: See(x) \Rightarrow Cs(x) .$
- ▀ $R2 > R1 .$
- ▀ Validates the system and implementation process

Model 2

- $R1: \Rightarrow \sim Cs(x).$
- $R2: See(x) \Rightarrow Cs(x).$
- $R2 > R1.$
- $R3: \{See(x), See(y), Opp(x, y) \Rightarrow \sim Cs(x).$
- $R3 > R2$
- $R4: \{See(x), See(y), SeeLtoR(y, x), LR(x, y)\} \Rightarrow \sim Cs(x)$
- $R4: \{See(x), See(y), SeeLtoT(y, x), LR(x, y) \Rightarrow \sim cs(y)$
- $R4 > R2$

Illustration

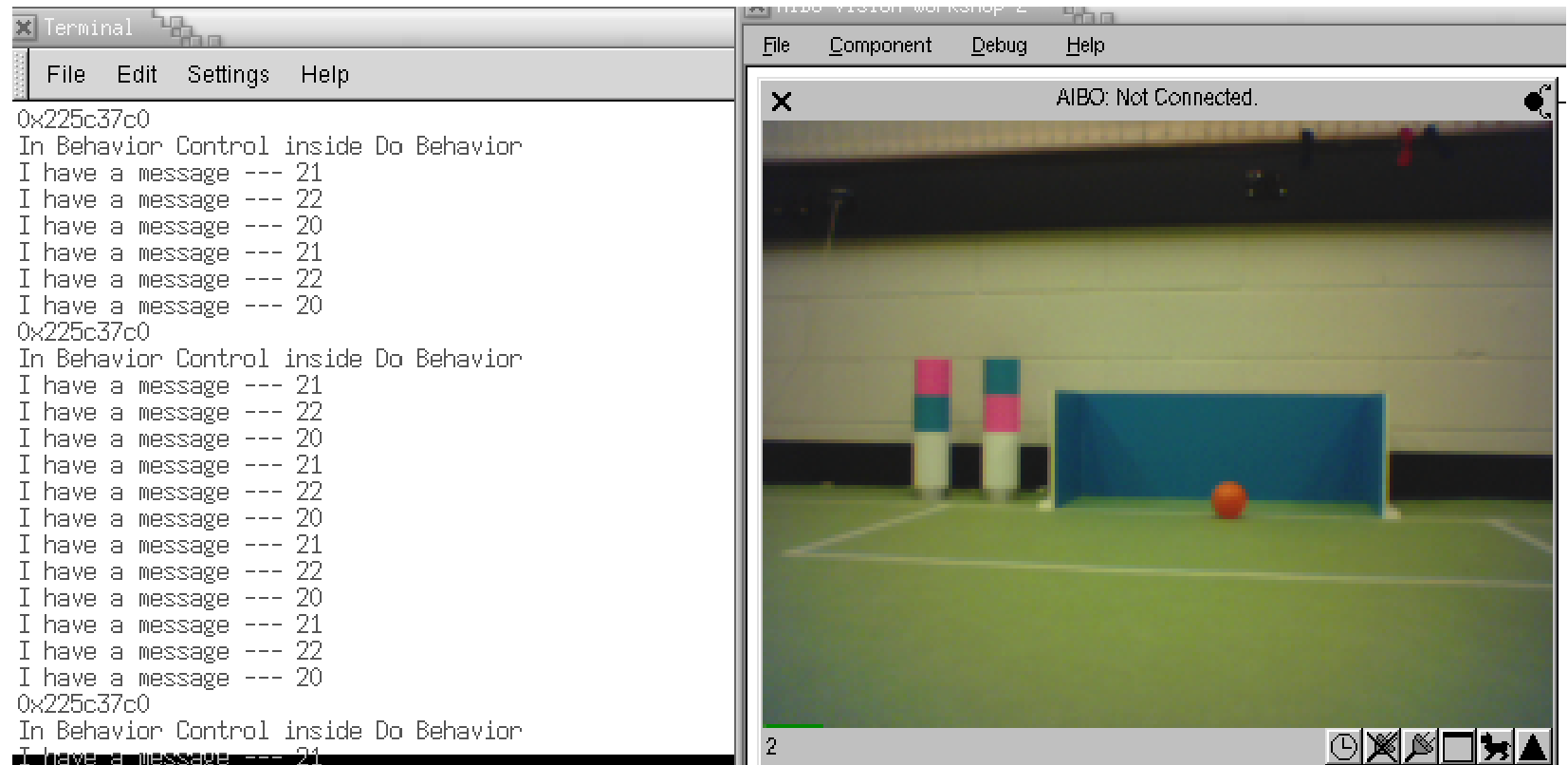


The left post is correct but the right post and goal are inverted

Model 3

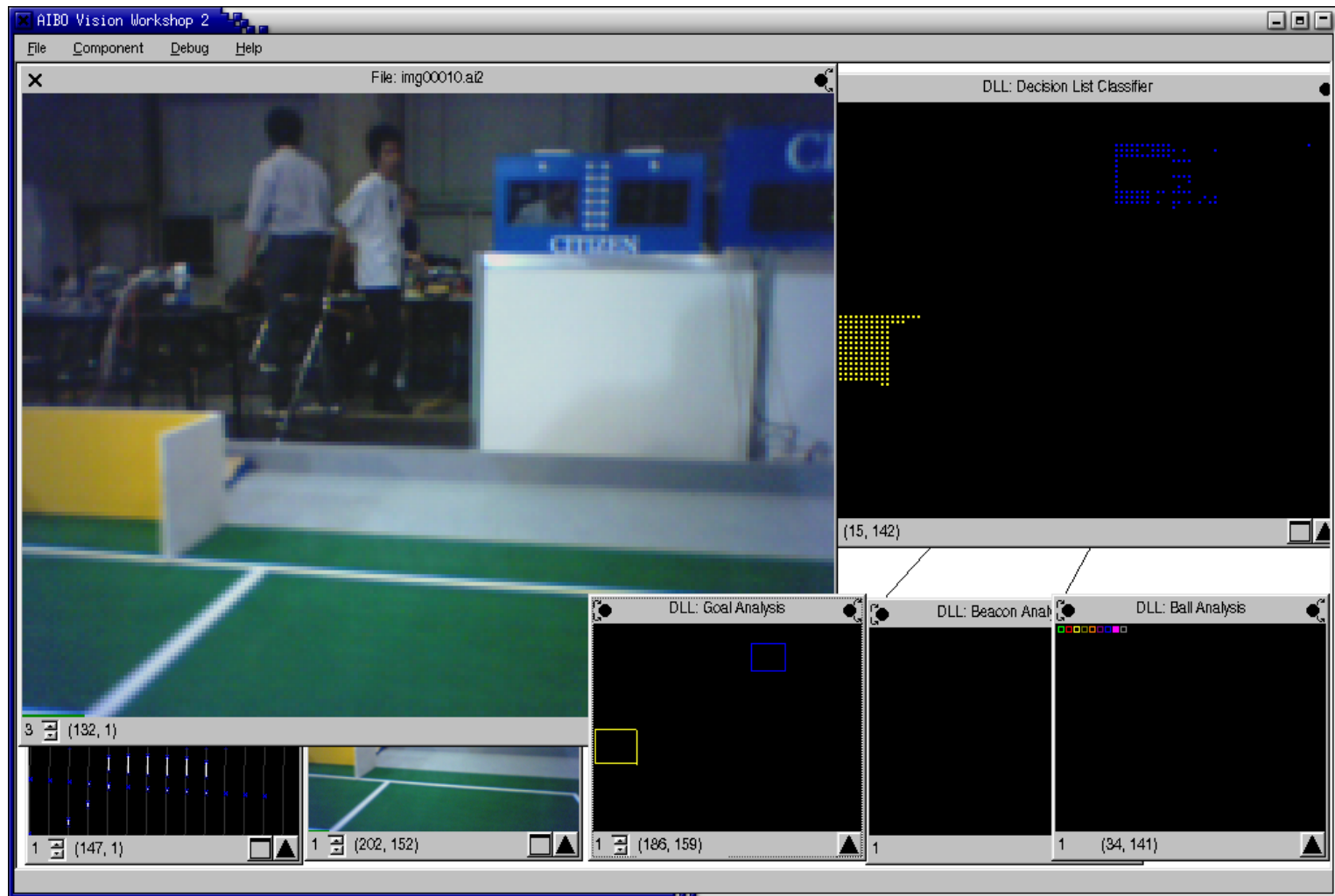
- $R1: \Rightarrow \sim Cs(x).$
- $R2: See(x) \Rightarrow Cs(x).$
- $R2 > R1.$
- $R3: \{See(x), See(y), Opp(x, y) \Rightarrow \sim Cs(x).$
- $R3 > R2$
- $R4: \{See(x), See(y), SeeLtoR(y, x), LR(x, y)\} \Rightarrow \sim Cs(x)$
- $R4: \{See(x), See(y), SeeLtoT(y, x), LR(x, y) \Rightarrow \sim cs(y)$
- $R4 > R2$
- $R5: \{See(x), See(y), See(z), SeeLtoR(y, z), SeeLtoR(z, x), Adj(x, y, z)\} \Rightarrow Cs(y)$
- $R5: \{See(x), See(y), See(z), SeeLtoR(y, z), SeeLtoR(z, x), Adj(x, y, z)\} \Rightarrow Cs(x)$
- $R5: \{See(x), See(y), See(z), SeeLtoR(z, x), SeeLtoR(x, y), Adj(x, y, z) \Rightarrow Cs(x)$
- $R5: \{See(x), See(y), See(z), SeeLtoR(z, x), SeeLtoR(x, y), Adj(x, y, z) \Rightarrow Cs(y)$
- $R5 > R4$
- $R6: \{See(x), See(y), See(z), SeeLtoR(x, z), SeeLtoR(z, y), LR(x, y), LR(y, z), Opp(x, z)\} \Rightarrow Cs(x)$
- $R6: \{See(x), See(y), See(z), SeeLtoR(x, z), SeeLtoR(z, y), LR(x, y), LR(y, z), Opp(x, z)\} \Rightarrow Cs(y)$
- $R6 > R3$
- $R6 > R4$

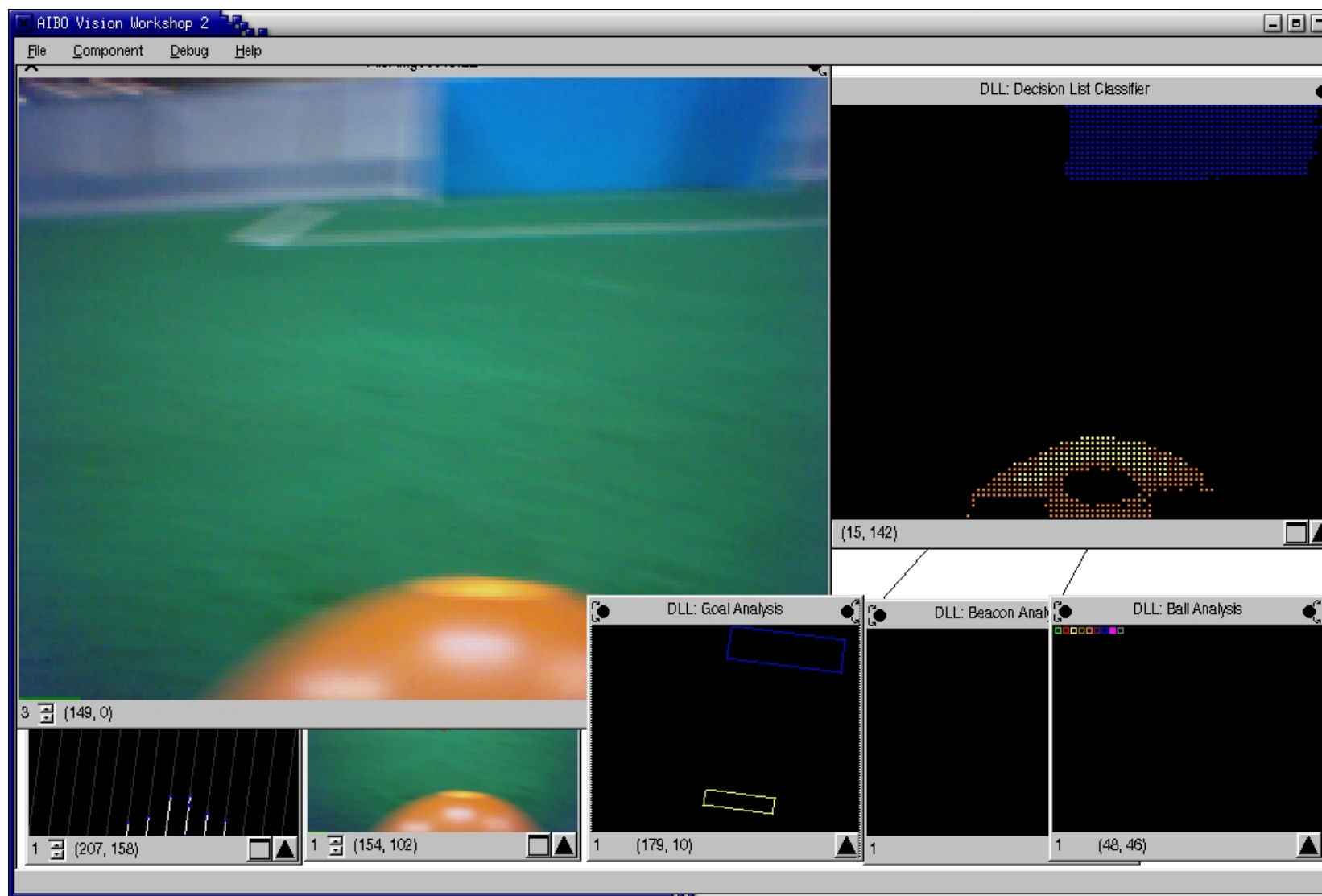
Illustration

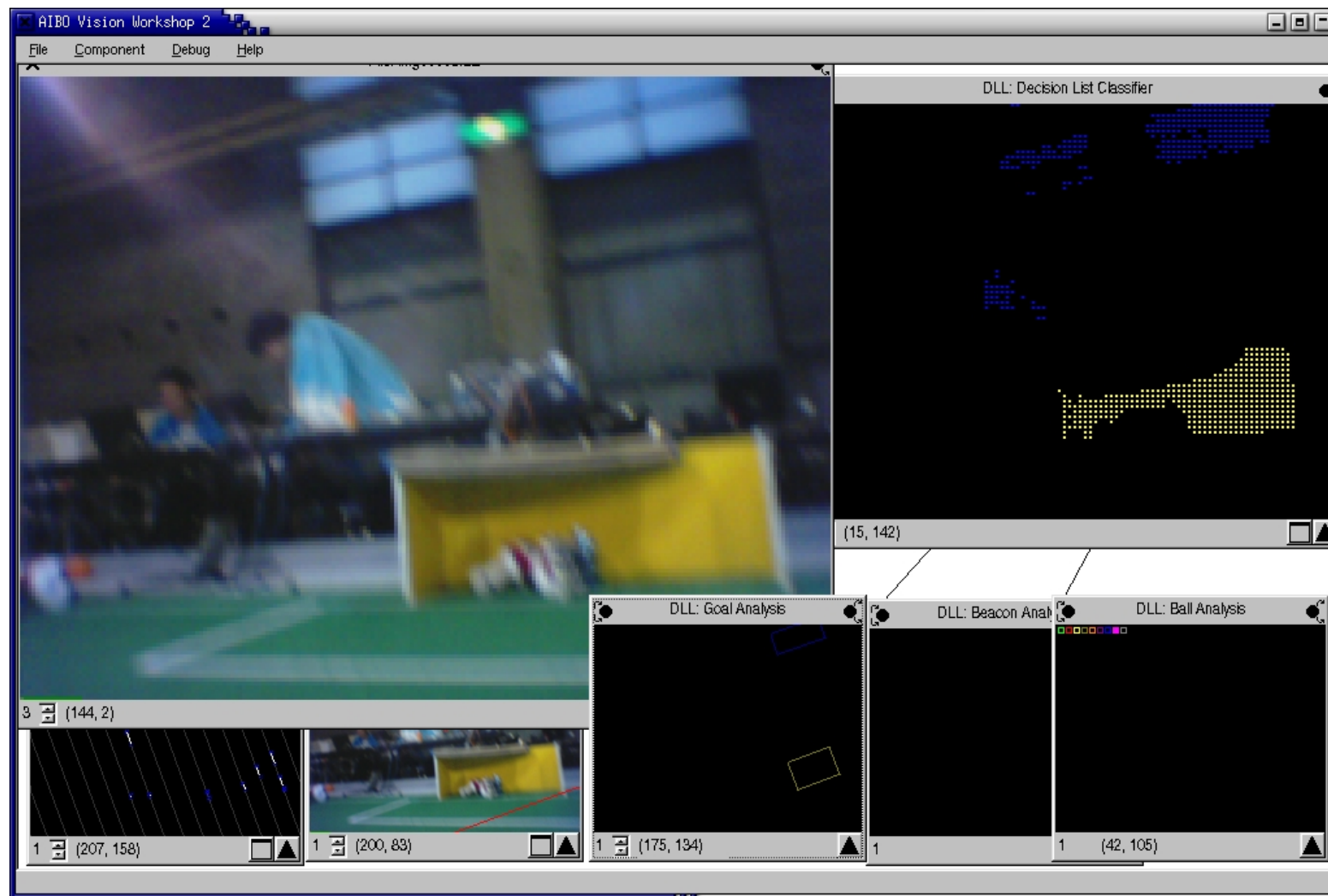


The left post and goal appear in the correct order,
but the right post appears left most

The module in action







Discussion

- CPU times were very positive
 - Model 1 : 44 microseconds
 - Model 2 : 60 microseconds
 - Model 3: 110 microseconds
 - On ERS-7 SONY Aibo

Conclusion

- The initial progress on logic and reasoning within AI has largely been discarded from mobile robotics in favour of reactive architectures
- We demonstrate the use of non-monotonic reasoning in the challenging application of RoboCup
- Plausible logic is the only non-monotonic logic with an algorithm that detects loops

Researcher
griffith
university
percebo

THANK
YOU