## Plan Recognition for Behavior Estimation in a Robotic Soccer Player

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# RoboCup



Figure: RoboCup

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Figure: RoboCup

Plan recognition is the task of recognizing goals and plans based on often incomplete observations executed by agents and properties of agent behavior in an environment [Sukthankar et al., 2014]

#### Some applications of plan recognition

- traffic monitoring [Pynadath and Wellman, 2013]
- ▶ dialog system [Carberry, 1990]
- crime detection and prevention
- military [Agmon et al., 2008]

- ➤ The plan recognition itself have your formulation in discrete world.
- But in robotic we deal with continuous states (angle, position, velocity)
  - ► The main method to deal with continuous states is discretization.
  - However, a wrong discretization lead a loss of information [Nash and Koenig, 2013]

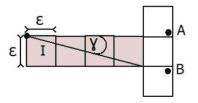


Figure: Example of Discretized Plan Recognition [Kaminka et al., 2018]

#### The Conclusion

- ▶ The planner could define the  $\epsilon$  discretization ad-hoc.
  - Mirroring

- ► In real robot applications a path is a soft trajectories between states points.
- In a robotic competition, is easy to consider that any team has you own path generation with your own desirable dynamic.
  - Propose: Identify the path generation function.

The cubicle function can generate any trajectory for this problem.

$$T_{x} = A_{3}^{x} t^{3} + A_{2}^{x} t^{2} + A_{1}^{x} t + A_{0}^{x}$$
 (1)

$$T_y = A_3^y t^3 + A_2^y t^2 + A_1^y t + A_0^y$$
 (2)

$$T_{\theta} = A_3^{\theta} t^3 + A_2^{\theta} t^2 + A_1^{\theta} t + A_0^{\theta} \tag{3}$$

where t is the real time and  $A_i^{\eta}$ ,  $i \in \{0, 1, 2, 3\}$ ,  $\eta \in \{x, y, \theta\}$  are the unknown coefficients.

Domain Formulation
A domain theory is a tuple where

$$W = \langle T, V, A, cost_i \rangle \tag{4}$$

- ▶ T is a finite set of fluents ex:  $T = \{x(r), y(r), \theta(r)\}.$
- V is the set of values.
- A is a set of actions.
- cost is a measure between transformations.
- fluents literal ex:  $s = \{x(r) = 10.14, y(r) = 7.13, \theta(r) = 45^{\circ}\}.$

Problem Formulation
A recognition problem is tuple where

$$R := \langle W, O, I, G \rangle \tag{5}$$

- W is the domain
- ▶ O is a sequence of observations
- L is the initial states.
- ► G is a set of goals

The main objective is

$$\alpha_R = \operatorname*{argmax}_{\alpha \in W} P(\alpha|O) \tag{6}$$

With Bayes Rules we can compute the  $P(\alpha|O)$ 

$$P(\alpha|O) = \beta P(O|\alpha)P(\alpha)$$

$$= \beta P(O|\alpha)P(\alpha|g)P(g),$$
(7)

where P(g) is a uniform distribution probability that robot is pursuing the goal g

The main question in Bayes rule is compute the terms  $P(O|\alpha)$  and  $P(\alpha|g)$ 

- ▶  $P(O|\alpha)$  Synthetized a trajectory that passes through the observations and continues until reach g.
- $ightharpoonup P(\alpha|g)$  We can use following approximation

$$\forall g \in G, P(\alpha|g) := \frac{cost_i(\hat{\alpha}_g)}{cost_i(\alpha)}.$$
 (8)

$$\forall g \in G, P(\alpha|g) := \frac{cost_i(\hat{\alpha}_g)}{cost_i(\alpha)}.$$
 (9)

Possible Cost Function

- ► Shortest path
- Less time.
- Less battery use.
- A linear combinations of latter options.

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