

# Corpus-Based Incremental Intention Recognition via Bayesian Network Model Construction

Han The Anh (h.anh@fct.unl.pt)

Luís Moniz Pereira (lmp@di.fct.unl.pt)



ICAPS "Goal, Activity and Plan Recognition" workshop  
Freiburg, Germany, 12 June, 2011

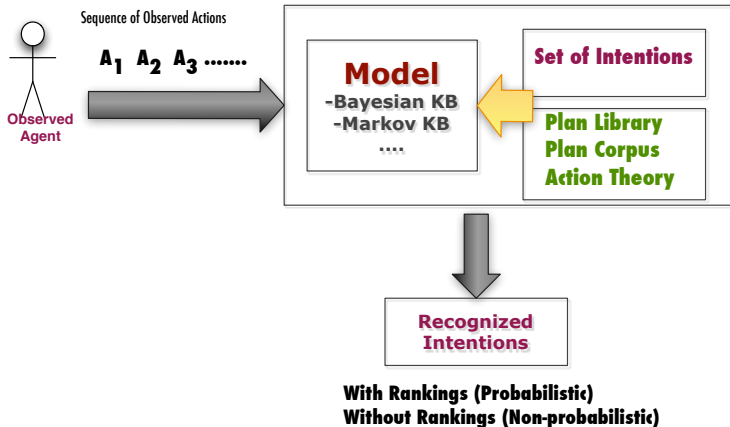
# Outline

- 1 Introduction
  - Intention Recognition
- 2 Models & Algorithms
  - Bayesian Network Construction
  - Operators for Handling Bayesian Networks for Intention Recognition (IRBNs)
  - Relations Amongst Intentions
- 3 Results & Conclusions
  - Evaluation Metrics
  - Linux Plan Corpus
  - Prisoner's Dilemma Corpus

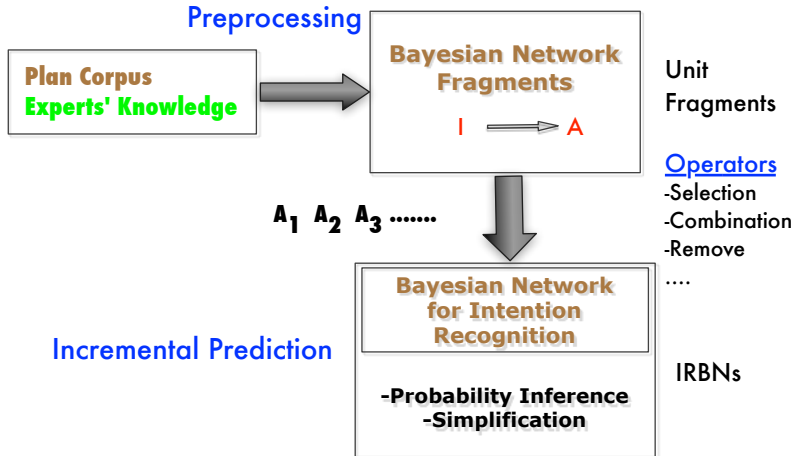
# Why Intention Recognition?

- **Intention recognition (IR)**: inferring intentions of a single agent (individual) or a group of agents (collective) based on their observed actions.
- **Why IR is necessary?** Acting on environment, agent may have to deal with other agents
  - Ease interactions, improve cooperation and coordination, especially when communication is limited.
  - Defend from potential hostile behaviors, plan in advance to take advantage.

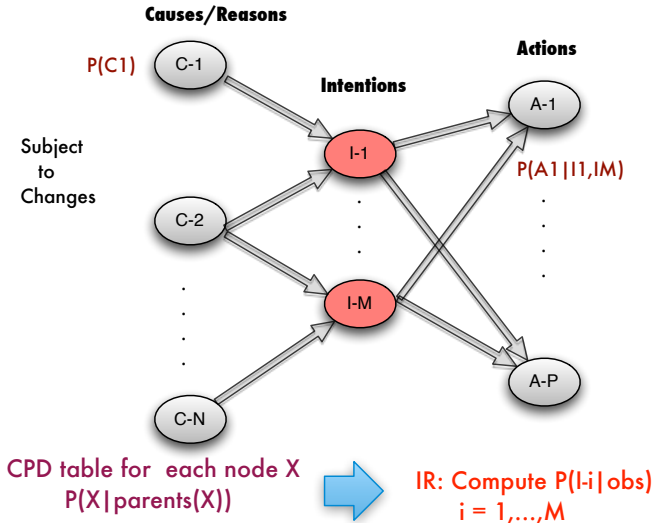
# Intention recognition systems: Typical Components



## System Description

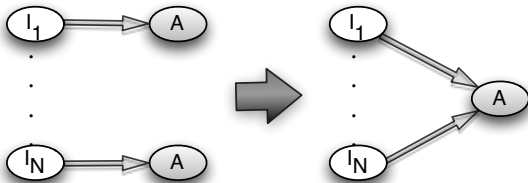


# Bayesian Network for Intention Recognition (IRBN)



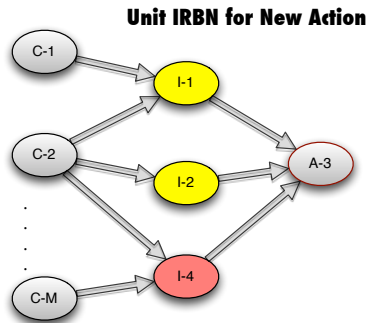
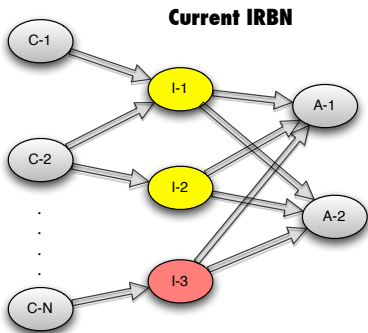
Unit Fragments  $\rightarrow$  Unit IRBN

Figure: Combine context-dependently selected unit fragments of a new action A.



# Integrate into the current model ...

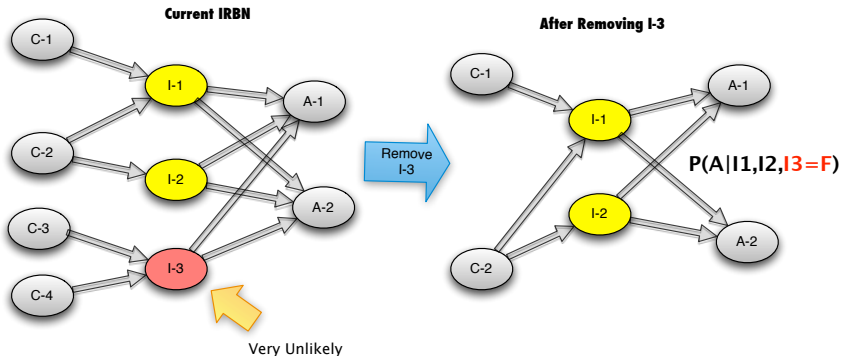
When a new action  $A_3$  arises, integrate it into the current model





# Remove Irrelevant Intentions

If some intentions are found out to be irrelevant, e.g. when their probability is very small, they are removed from the model.



# Incremental Intention Recognition Algorithm

**Repeat** until one intention remains or time limit is reached.

- 1 If **new actions** are observed: **combine** unit IRBNs for them with the current IRBN.
- 2 **Compute conditional probability** of each intention given current observed actions. **Remove** irrelevant intentions.

## Relations Amongst Intentions: multiple intentions

- In case agent **pursues multiple intentions simultaneously**, intentions may support or exclude each other.
- Two **mutually exclusive intentions** cannot be parents of an action:  $P(I_1 = T, I_2 = T) = 0 \rightarrow P(A|I_1, I_2, \dots)$  is **undefined**.
- Need to combine mutually exclusive intentions in a single node.

# Single Intention Case

- In case agent pursues a single intention (Linux Plan corpus), all intentions are mutually exclusive.
- They are combined in a single node.
- $A_1, \dots, A_m$ : current observed actions. Then,

$$P(I = I_j | A_1, \dots, A_m) = \frac{P(I_j) \prod_{i=1}^m P(A_i | I_j)}{\sum_{j=1}^n P(I_j) \prod_{i=1}^m P(A_i | I_j)}$$

- The recognizer has **linear complexity** on the number of intentions being modeled.

# Evaluation Metrics

- $Seq = a_1, \dots, a_n$ : sequence of actions achieving intention  $I$ .
- **N-best prediction**:  $correct(A) = 1$  if  $I$  is one of  $N$  most likely intentions, and 0 otherwise.

$$precision(Seq) = \left( \sum_{i=1}^n correct(a_i) \right) / z$$

$$recall(Seq) = \left( \sum_{i=1}^n correct(a_i) \right) / Z$$

$$convergence(Seq) = (z - t + 1) / z$$

$z$ : #predictions;  $Z$ : #prediction opportunities;  $t$ : smallest number from that on correctly predicts.

# Linux Plan Corpus

- One of rare, often used, plan corpora available for evaluating intention/plan recognizers.
- **goals** = tasks in Linux (e.g., find a file, copy some files);  
**actions** = Linux commands (e.g., find, cp, cd, ls).
- 19 goals; 43 actions (commands); 547 sessions.  
Collected from Linux users.

## Intention Recognition Results on Linux Plan Corpus

$\tau$  – confidence level: make prediction only when probability  $\geq \tau$ .

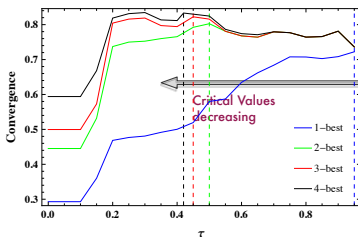
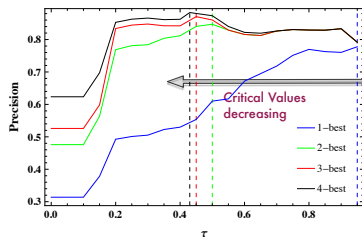


Figure: Precision and Convergence for  $\tau \in [0, 1]$ ;  $N = 1, 2, 3, 4$ .

We obtained better results than existent corpus-based intention recognizers.

<b>N-best</b>	<b>1-best</b>	<b>2-best</b>	<b>3-best</b>	<b>4-best</b>
$\tau$	0.95	0.5	0.45	0.42
<b>Precision</b>	0.786	0.847	0.870	0.883
<b>Recall</b>	0.308	0.469	0.518	0.612
<b>Converg.</b>	0.722	0.799	0.822	0.824

Table: Intention Recognition Results on the Linux Plan Corpus



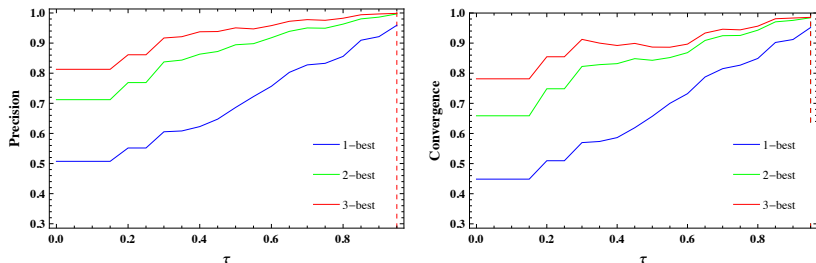
# Iterated Prisoner's Dilemma Corpus (IPD)

Prisoner's Dilemma is a symmetric two-player non-zero game defined by the payoff matrix

$$\begin{array}{cc} & C & D \\ C & (R, R) & (S, T) \\ D & (T, S) & (P, P) \end{array}$$

Intentions to be recognized are strategies in IPD: TFT, WSLS, GTFT, GRIM, FBF, etc.

## Intention Recognition Results on IPD Plan Corpus

Figure: Precision and Convergence for  $\tau \in [0, 1]$ ;  $N = 1, 2, 3$

## Summary

- Incremental IR model via incrementally constructing a Bayesian network model.
- Evaluated on Linux Plan corpus, obtaining better results than existent systems.
- Present a new benchmark for intention/plan recognition based on social dilemmas (Prisoner's Dilemma, Stag hunt,...).
- Able to handle multiple intention case by representing relations amongst intention nodes.

## Future Work

- Create plan corpora to evaluate multiple-intention recognition (suggestions of Planners for this purpose are greatly appreciated)
- Apply to explain the role of intention recognition for the evolution of cooperation (see our papers in refs).
- We have provided in another IPD benchmark corpus for testing context-dependent aspect (BMAW@UAI paper).

Thank you!

QUESTIONS?

# IPD Plan Corpus

- Intentions are famous strategies in IPD: TFT, WSLS, GTFT, GRIM, FBF. There are infinite number of strategies.
- Corpus actions:  $s_1 \dots s_M \xi$ , where  $s_i \in \{E, R, T, S, P\}$  – states of the  $M$  last interactions;  $\xi \in \{C, D\}$  – current move.  
 $\Sigma_1 = \{EC, RC, TC, SC, PC, ED, RD, TD, SD, PD\}$
- A plan session of TFT:  $[EC, RC, SD, PD, TD]$

<b>round :</b>	0	1	2	3	4	5
<b>TFT :</b>	–	C	C	D	D	D
<b>X :</b>	–	C	D	D	C	D
<b>TFT-states :</b>	E	R	S	P	T	P

## Generating Training and Testing Datasets

- 7 strategies: *AICC*, *AIID*, *TFT*, *GTFT*, *WSLS*, GRIM and FBF.
- Training corpus is generated by playing with each strategy all the possible combinations 10 times ( $r = 5, \dots, 10$ ).
- The testing dataset is generated by playing a random choice with each strategy in each round ( $r = 5, \dots, 10$ ).