

# A Plan Optimality Monitoring Approach to Detect Commitment Abandonment

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COIN@AAMAS, 2017

May, 2017

# Introduction

- Determining whether an agent is actually executing steps towards a goal (or has abandoned it), is important when:
  - multiple agents are trying to achieve joint goals, or
  - agents are committed for achieving goals for each other.
- **Commitment abandonment**: situation in which an agent switches from executing the actions of one plan that achieves the consequent it is committed to, to executing actions from another plan;
- We develop a **domain-independent** approach based on **planning techniques** to:
  - detect sub-optimal steps; and
  - infer whether an agent will honour a commitment

# Background: Commitments

- A **commitment**  $C(\text{DEBTOR}, \text{CREDITOR}, \text{antecedent}, \text{consequent})$  formalizes that the agent `DEBTOR` commits to agent `CREDITOR` to bring about the consequent if the antecedent holds;
- The **antecedent** and **consequent** conditions of a commitment are conjunctions or disjunctions of events and possibly other commitments;
- In this paper, we aim to **monitor** the `DEBTOR`'s behaviour (i.e., sequence of actions) to **detect** if this agent is **individually committed to carrying out a plan to achieve** the consequent for the `CREDITOR`.

# Background: Planning, Heuristics, and Landmarks

## Definition (**Planning**)

A *planning instance* is represented by a triple  $\Pi = \langle \Xi, \mathcal{I}, G \rangle$ , in which:

- $\Xi = \langle \Sigma, \mathcal{A} \rangle$  is the **domain definition**, and consists of a finite set of **facts**  $\Sigma$  and a finite set of **actions**  $\mathcal{A}$  (action costs = 1);
- $\mathcal{I}$  and  $G$  represent the **planning problem**, in which  $\mathcal{I} \subseteq \Sigma$  is the **initial state**, and  $G \subseteq \Sigma$  is the **goal state**.
- **Heuristics** are used to estimate the cost to achieve a particular goal. In this work, we use **domain-independent heuristics**;

## Definition (**Landmarks**)

Given a *planning instance*  $\Pi = \langle \Xi, \mathcal{I}, G \rangle$ , a **fact** (or **action**)  $L$  is a **landmark** in  $\Pi$  iff  $L$  must be **satisfied** (or **executed**) at some point along all valid plans that achieve  $G$  from  $\mathcal{I}$ .

## Background: Fact Partitioning

Pattison and Long (“**Domain Independent Goal Recognition**”). In STAIR, 2010) classify facts into a set of mutually exclusive fact partitions. We use such partitions to infer whether certain observations are consistent with a particular goal state, as follows:

- **Strictly Activating** is a type of fact that can never be added by any action unless defined in the initial state;
- **Unstable Activating** is a type of fact that that once deleted, cannot be re-achieved;
- **Strictly Terminal** is a type of fact that once added, cannot be deleted.

# Background: Commitment Abandonment Problem

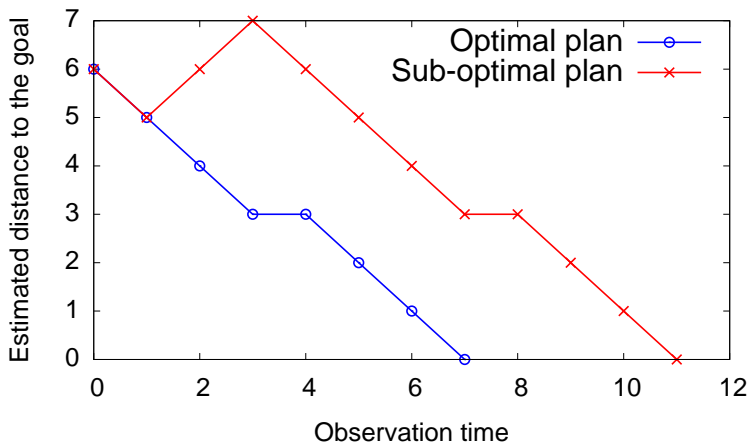
## Definition (**Commitment Abandonment Problem**)

- Domain definition (Properties and Actions)  $\Xi = \langle \Sigma, \mathcal{A} \rangle$ ;
  - Commitment  $C$ , in which  $C(\text{DEBTOR}, \text{CREDITOR}, A_t, C_t)$ ,  $\text{DEBTOR}$  is the debtor,  $\text{CREDITOR}$  is the creditor,  $A_t$  is the antecedent condition, and  $C_t$  is the consequent;
  - Initial state  $\mathcal{I}$ , *s.t.*,  $A_t \subseteq \mathcal{I}$  (when begins the monitoring process);
  - An observation sequence  $O = \langle o_1, o_2, \dots, o_n \rangle$ , representing a full observable plan execution; and
  - Threshold  $\theta$ , representing the percentage of sub-optimal actions that the  $\text{DEBTOR}$  agent can deviate to achieve the consequent state  $C_t$ .
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- The solution for a commitment abandonment problem is whether an observation sequence  $O$  has deviated more than  $\theta$  from the optimal plan to achieve the consequent  $c_t$  of commitment  $c$ .

- We use plan optimality monitoring techniques from the literature to detect sub-optimal steps (Pereira *et al.* “**Monitoring Plan Optimality using Landmarks and Domain-Independent Heuristics**”. In PAIR@AAAI, 2017.);
- This approach combines **planning techniques**, *i.e.*, landmarks and domain-independent heuristics.
  - It uses **landmarks** to obtain information about **what cannot be avoided** to achieve a goal  $G$ ; and
  - It uses **heuristics** to analyse possible **plan execution deviation**.

# Analyzing Plan Execution Deviation

- If an observation  $o_i$  results a state  $s_i$ , we consider a **deviation from a plan** to occur if  $h(s_{i-1}) < h(s_i)$ .





# Predicting Non-regressive Actions via Landmarks

- To predict which actions could be executed in the next observation, we **estimate the distance to the closest landmarks** (using  $h_{max}$ ) from the current state to the extracted landmarks  $\mathcal{L}$ , and select the following actions:
  - For every fact landmark  $l \in \mathcal{L}$  with  $h_{max}(l) = 0$ , we select actions  $a \in \mathcal{A}$  such that  $l \in pre(a)$ ; and
  - For every fact landmark  $l \in \mathcal{L}$  with  $h_{max}(l) = 1$ , we select actions  $a \in \mathcal{A}$  such that  $pre(a) \in \text{current state}$  and  $l \in eff(a)^+$ ;
- Predicted actions **may reduce the distance** to the **monitored goal** and **next landmarks**.

# Detecting Sub-Optimal Steps

- To detect sub-optimal steps (actions) in observation sequence  $O$  for a monitored goal  $G$ , we combine the techniques we developed and filter with the following condition:
  - **An observed action  $o \in O$  is considered sub-optimal if:**  
 $o \notin$  set of predicted actions AND  $(h(s_{i-1}) < h(s_i))$ .

# Commitment Abandonment Detection Approach

- We monitor the sequence of actions of a DEBTOR to infer whether it **will abandon a commitment**
  - Observed sequence should achieve the consequent from a state in which the antecedent holds
- We use a **threshold**  $\theta$ , representing the percentage of sub-optimal actions that the DEBTOR agent can deviate to achieve the consequent it is committed to, *i.e.*, a percentage of actions that CREDITOR agent agrees to deviate from the optimal.

# Determining Commitment Abandonment using Plan Optimality Monitoring and Fact Partitioning

Our approach determines that a `DEBTOR` agent has abandoned a commitment it is committed to if any one of three conditions is true:

- ① *Strictly Activating* facts that we extracted are not in the initial state;
- ② we observe the evidence of any *Unstable Activating* and/or *Strictly Terminal* facts during the execution of actions in the observations; or
- ③ the number of observed sub-optimal steps is greater than  $\theta$  defined by the `CREDITOR`.

# Experiments and Evaluation (1 of 2)

- We evaluate our approach over 8 planning domains, most of which are inspired by real-world scenarios;
  - **Precision:** percentage of the abandoned commitments inferred that were actually abandoned (quality);
  - **Recall:** percentage of actually abandoned commitments inferred by the approach (quantity);
  - **F1-score:** harmonic mean between *Precision* and *Recall*.
- We use 6 domain-independent heuristics:
  - $h_{adjsum}$ ,  $h_{adjsum2}$ ,  $h_{adjsum2M}$ ,  $h_{combo}$ ,  $h_{ff}$ , and  $h_{sum}$ ;
- We manually generated the dataset from medium and large planning problems, generating plans that either abandoned (ultimately went to a different goal) or did not abandon their corresponding goals/consequent, varying the number of sub-optimal actions.

# Experiments and Evaluation (2 of 2)

Domain	O	T	Precision	Recall	F1-score
			$\theta$ (0% / 5% / 10%)	$\theta$ (0% / 5% / 10%)	$\theta$ (0% / 5% / 10%)
DRIVER-LOG (30) <i>h<sub>adjsum2M</sub></i>	20.0	0.83	1.0 / 1.0 / 1.0	1.0 / 1.0 / 1.0	1.0 / 1.0 / 1.0
DEPOTS (30) <i>h<sub>adjsum2</sub></i>	18.6	1.79	1.0 / 1.0 / 1.0	1.0 / 1.0 / 0.8	1.0 / 1.0 / 0.88
EASY-IPC-GRID (30) <i>h<sub>ff</sub></i>	17.3	0.95	1.0 / 1.0 / 1.0	1.0 / 1.0 / 1.0	1.0 / 1.0 / 1.0
FERRY (30) <i>h<sub>adjsum2</sub></i>	13.5	0.38	1.0 / 1.0 / 1.0	1.0 / 0.8 / 0.8	1.0 / 0.88 / 0.88
LOGISTICS (30) <i>h<sub>adjsum2</sub></i>	21.0	0.56	1.0 / 1.0 / 1.0	1.0 / 1.0 / 1.0	1.0 / 1.0 / 1.0
SATELLITE (30) <i>h<sub>adjsum2M</sub></i>	23.5	5.4	0.8 / 1.0 / 1.0	0.8 / 0.6 / 0.6	0.8 / 0.75 / 0.75
SOKOBAN (30) <i>h<sub>combo</sub></i>	22.8	5.2	0.83 / 1.0 / 1.0	1.0 / 0.6 / 0.6	0.91 / 0.75 / 0.75
ZENO-TRAVEL (30) <i>h<sub>adjsum2</sub></i>	10.0	1.1	1.0 / 1.0 / 1.0	0.8 / 0.8 / 0.8	0.88 / 0.88 / 0.88

- |O| is the average number of observed actions in a plan execution;
- T is the average monitoring time (in seconds); and
- $\theta$  is threshold value varying at 0%, 5%, and 10%.

- Geib and Goldman. “**Recognizing Plan/Goal Abandonment**”. In IJCAI, 2003;
- Kafali *et al.* “**GOSU: Computing GOal Support with Commitments in Multiagent Systems**”. In ECAI, 2014; and
- Kafali and Yolum. “**PISAGOR: A Proactive Software Agent for Monitoring Interactions**”. In Knowledge and Information Systems, 2016.

- **Contribution:**

- We formalized the commitment abandonment problem using planning;
- Our approach is domain-independent and require minimal domain knowledge; and
- We show that our approach has high accuracy (very good results) in almost all domains (apart from `SATELLITE`).

- **Limitations:**

- We only deal with full observability;
- Our approach assumes a centralized monitor;

- **Future Work:**

- Detect commitment abandonment using multiple monitors; and
- Deal with partial observability (*i.e.*, missing observations).



Thank you!  
Questions?