

Informatics 2D – Reasoning and Agents

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Lecture 19 – Planning and Acting in the Real World II
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adapted from slides by Michael Rovatsos

Where are we?

Last time ...

- ▶ Looked at methods for real-world planning
- ▶ Sensorless planning and contingent planning
- ▶ Fully and partially observable environments

Today ...

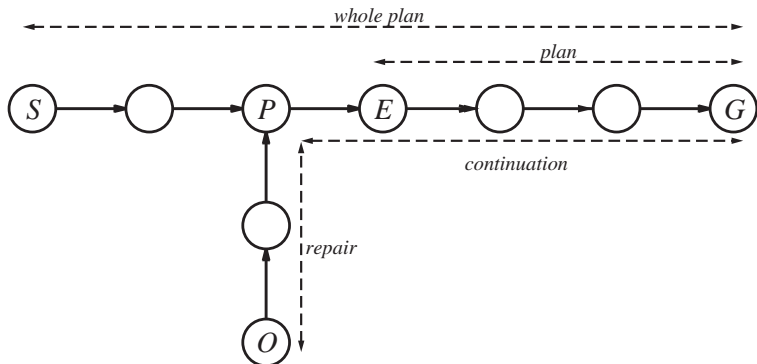
- ▶ **Planning and Acting in the Real World II**

Execution monitoring and replanning

- ▶ **Execution monitoring** = checking whether things are going according to plan (necessitated by unbounded indeterminacy in realistic environments)
 - ▶ Action monitoring = checking whether next action is feasible
 - ▶ Plan monitoring = checking whether remainder of plan is feasible
- ▶ **Replanning** = ability to find new plan when things go wrong (usually repairing the old plan)
- ▶ Taken together these methods yield powerful planning abilities

Action monitoring and replanning

- ▶ While attempting to get from S to G , a problem is encountered in E , agent discovers actual state is O and plans to get to P and execute the rest of the original plan



Plan monitoring

- ▶ Action monitoring often results in suboptimal behaviour, executes everything until actual failure
- ▶ **Plan monitoring** checks preconditions for entire remaining plan
- ▶ Can also take advantage of **serendipity** (unexpected circumstances might make remaining plan easier)
- ▶ In partially observable environments things are more complex (sensing actions have to be planned for, they can fail in turn, etc.)

Hierarchical decomposition in planning

- ▶ **Hierarchical decomposition** seems a natural idea to improve planning capabilities.
- ▶ **Key idea**: at each level of the hierarchy, activity involves only small number of steps (i.e. small computational cost)
- ▶ **Hierarchical task network** (HTN) planning: initial plan provides only high-level description, refined by **action refinements**
- ▶ Refinement process continued until plan consists only of **primitive actions**

Representing action decompositions

- ▶ Each **high level action (HLA)** has (at least) one **refinement** into a sequence of actions.
- ▶ The actions in the sequence may be HLAs or primitive.
 - ▶ So HLAs form a hierarchy!
- ▶ If they're all primitive, then that's an **implementation** of the HLA.

Example: Go to SF Airport

Refinement(Go(Home, SFO),
PRECOND:At(Car, Home)
STEPS:[Drive(Home, SFO LongTermParking)
Shuttle(SFO LongTermParking, SFO)])

Refinement(Go(Home, SFO),
PRECOND:Cash, At(Home)
STEPS:[Taxi(Home, SFO)])

Refinements can be Recursive

Refinement(Navigate([a, b], [x, y]),
PRECOND: $a = x, b = y$
STEPS: [])

Refinement(Navigate([a, b], [x, y]),
PRECOND: *Connected*([a, b], [a - 1, b])
STEPS: [*Left*, *Navigate*([a - 1, b], [x, y])])

Refinement(Navigate([a, b], [x, y]),
PRECOND: *Connected*([a, b], [a + 1, b])
STEPS: [*Right*, *Navigate*([a + 1, b], [x, y])])

High-Level Plans

- ▶ High-Level Plans (HLP) are a sequence of HLAs.
- ▶ An implementation of a High Level Plan is the concatenation of the an implementation of each of its HLAs.
- ▶ An HLP achieves the goal from an initial state if **at least one** of its implementations does this.
- ▶ **Not all implementations of an HLP have to reach the goal state!**
- ▶ The agent gets to decide which implementation of which HLAs to execute.

Searching for Primitive Solutions

- ▶ The **HLA plan library** is a **hierarchy**:
 - ▶ (Ordered) Daughters to an HLA are the sequences of actions provided by one of its refinements;
 - ▶ Because a given HLA can have more than one refinement, there can be more than one node for a given HLA in the hierarchy.
- ▶ This hierarchy is essentially a **search space of action sequences** that conform to knowledge about how high-level actions can be broken down.
- ▶ So you can search this state space for a plan!

Searching for Primitive Solutions: Breadth First

- ▶ Start your plan P with the HLA $[Act]$,
- ▶ Take the first HLA A in P (recall that P is an *action sequence*).
- ▶ Do a breadth-first search in your hierarchical plan library, to find a refinement of A whose preconditions are satisfied by the outcome of the action in P that is prior to A .
- ▶ Replace A in P with this refinement.
- ▶ Keep going until your plan P has no HLAs and either:
 1. Your plan P 's outcome is the goal, in which case return P ; or
 2. Your plan P 's outcome is not the goal, in which case return *failure*.

Problems!

- ▶ Like forward search, you consider lots of irrelevant actions.
- ▶ The algorithm essentially refines HLAs right down to primitive actions so as to determine if a plan will succeed.
- ▶ This contradicts **common sense**!
- ▶ Sometimes you know an HLA will work *regardless* of how it's broken down!
- ▶ We don't need to know which route to take to SFOParking to know this plan works:

[*Drive(Home, SFOParking), Shuttle(SFOParking, SFO)*]

- ▶ We can capture this if we add to HLAs *themselves* a set of preconditions and effects.

Adding Preconditions and Effects to HLAs

- ▶ One challenge in specifying preconditions and effects of an HLA is that the HLA may have more than one refinement, each one with slightly different preconditions and effects!
 - ▶ If you refine *Go(Home, SFO)* with *Taxi* action: you need *Cash*.
 - ▶ If you refine it with *Drive*, you don't!
 - ▶ This difference may affect your **choice** on how to refine the HLA!
- ▶ Recall that an HLA achieves a goal if **one** of its refinements does this.
- ▶ **And you can choose the refinement!**

Getting Formal

- ▶ $s' \in \text{REACH}(s, h)$ iff s' is reachable from *at least one* of HLA h 's refinements, given (initial) state s .

$$\text{REACH}(s, [h_1, h_2]) = \bigcup_{s' \in \text{REACH}(s, h_1)} \text{REACH}(s', h_2)$$

- ▶ HLP p achieves goal g given initial state s iff $\exists s'$ st
 $s' \models g$ and $s' \in \text{REACH}(s, p)$
- ▶ So we should search HLPs to find a p with this relation to g , and then focus on refining it.
- ▶ But a pre-requisite to this algorithm is to define $\text{REACH}(s, h)$ for each h and s .
- ▶ In other words, we still need to determine how to represent effects (and preconditions) of HLAs. . .

Defining REACH

- ▶ A primitive actions makes a fluent true, false, or leaves it unchanged.
- ▶ But with HLAs you sometimes get to *choose*, by choosing a particular refinement!
- ▶ We add new notation to reflect this:
 - $\tilde{+}A$: you can possibly add A (or leave A unchanged)
 - $\tilde{-}A$: you can possibly delete A (or leave A unchanged)
 - $\tilde{\pm}A$: you can possibly add A , or possibly delete A (or leave A unchanged)
- ▶ You should now *derive* the correct preconditions and effects from its refinements!

Our SFO Example

Refinement(*Go*(*Home*, *SFO*),
PRECOND:*At*(*Car*, *Home*)
STEPS:[*Drive*(*Home*, *SFO**LongTermParking*)
Shuttle(*SFO**LongTermParking*, *SFO*)]])

Refinement(*Go*(*Home*, *SFO*),
PRECOND:*Cash*, *At*(*Home*)
STEPS:[*Taxi*(*Home*, *SFO*)]])

The 'Primitive' Actions

Action(Taxi(a, b),

PRECOND: *Cash, At(Taxi, a)*

EFFECT: \neg *Cash, \neg At(Taxi, a), At(Taxi, b)*)

Action(Drive(a, b),

PRECOND: *At(Car, a)*

EFFECT: \neg *At(Car, a), At(Car, b)*)

Action(Shuttle(a, b),

PRECOND: *At(Shuttle, a)*

EFFECT: \neg *At(Shuttle, a), At(Shuttle, b)*)

Deriving the PRECONDS and EFFECTS of the HLA

- ▶ $\neg \text{Cash}$ is EFFECT of one HLA refinement, but not the other.
- ▶ So $\approx \neg \text{Cash}$ in HLA EFFECT!

Not so Simple!

- ▶ Similar argument for $\text{At}(\text{Car}, \text{SFOParking})$
- ▶ **But you can't choose the combination:**
 $\neg \text{Cash} \wedge \text{At}(\text{Car}, \text{SFOParking})$
- ▶ Solution is to write **approximate descriptions.**

Approximate Descriptions

Optimistic Description: $\text{REACH}^+(s, h)$

- ▶ Take union of all possible outcomes from all refinements.
- ▶ So this includes $\tilde{\neg} \text{Cash}$ and $\tilde{\neg} \text{At}(\text{Car}, \text{SFOParking})$.
- ▶ This overgenerates reachable states.

Pessimistic Description: $\text{REACH}^-(s, h)$

- ▶ Only states that satisfy effects from *all* refinements survive.
- ▶ So this does *not* include $\tilde{\neg} \text{Cash}$ or $\tilde{\neg} \text{At}(\text{Car}, \text{SFOParking})$.
- ▶ This undergenerates reachable states.

$$\text{REACH}^-(s, h) \subseteq \text{REACH}(s, h) \subseteq \text{REACH}^+(s, h)$$

Algorithm for Finding a Plan

Two Important Facts:

1. If $\exists s' \in \text{REACH}^-(s, h)$ st $s' \models g$, you know h can succeed.
2. If $\neg \exists s' \in \text{REACH}^+(s, h)$ st $s' \models g$, you know h will fail!

The Algorithm:

- ▶ Do breadth first search as before.
- ▶ But now you can **stop searching** and **implement instead** when you reach an h where 1. is true.
- ▶ And you can **drop** h (and all its refinements) when 2. is true.
- ▶ If 1. and 2. are both false for the current h , then you don't know if h will succeed or fail, but you can find out by refining it.

Summary

- ▶ Execution monitoring: checking success of execution
- ▶ Replanning: repairing plans in case of failure
- ▶ HLAs and HLPs
- ▶ Using refinements and preconditions and effects of primitive actions to *approximate* which states are reachable.
- ▶ Such approximate descriptions of HLAs help to inform search and when to refine an HLP so as to reach a goal.
- ▶ Next time: **Acting under Uncertainty**