

# Planning: Representation and Example

CPSC 322 Lecture 15

February 8, 2007  
Textbook §11.1 - 11.2

# Lecture Overview

- 1 Planning
- 2 State-Based Rep
- 3 Feature-Based Rep
- 4 STRIPS

# Planning

Given:

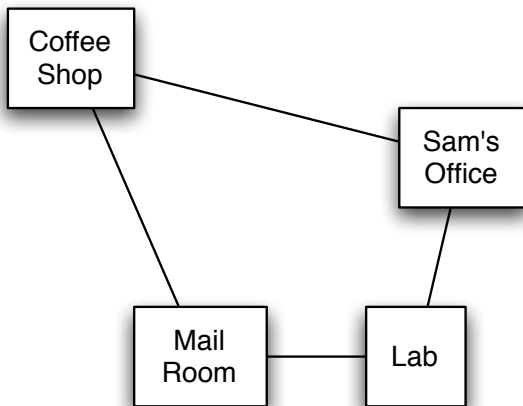
- A description of the effects and preconditions of the actions
- A description of the initial state
- A goal to achieve

find a sequence of actions that is possible and will result in a state satisfying the goal.

- recall: rather than looking for a single state that satisfies our constraints, look here for a sequence of states that gets us to a goal

# Delivery Robot Example

Consider a delivery robot named Rob, who must navigate the following environment:



# Delivery Robot Example

The **state** is defined by the following features:

*RLoc* — Rob's location

- domain: coffee shop (*cs*), Sam's office (*off*), mail room (*mr*), or laboratory (*lab*)

*RHC* — Rob has coffee

- domain: true/false. By *rhc* indicate that Rob has coffee and by  $\overline{rhc}$  that Rob doesn't have coffee.

*SWC* — Sam wants coffee (T/F)

*MW* — Mail is waiting (T/F)

*RHM* — Rob has mail (T/F)

An example state is  $\langle lab, \overline{rhc}, swc, \overline{mw}, rhm \rangle$ . How many states are there:

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*RHM* — Rob has mail (T/F)

An example state is  $\langle \overline{lab}, \overline{rhc}, \overline{swc}, \overline{mw}, \overline{rhm} \rangle$ . How many states are there:  $4 \times 2 \times 2 \times 2 \times 2 = 64$ .

# Delivery Robot Example

The robot's **actions** are:

*Move* — Rob's move action

- move clockwise (*mc*), move anti-clockwise (*mac*) not move (*nm*)

*PUC* — Rob picks up coffee

- must be at the coffee shop

*DelC* — Rob delivers coffee

- must be at the office, and must have coffee

*PUM* — Rob picks up mail

- must be in the mail room, and mail must be waiting

*DelM* — Rob delivers mail

- must be at the office and have mail

Assume that Rob can perform one action of each kind in a single step. Thus, an example action is  $\langle mc, \overline{puc}, \overline{dc}, \overline{pum}, dm \rangle$ ; we can abbreviate it as  $\langle mc, dm \rangle$ .

How many actions are there:

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How many actions are there:  $3 \times 2 \times 2 \times 2 \times 2 = 48$ .



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# State-Based Representation of a Planning Domain

- The domain is characterized by **states**, **actions** and **goals**
  - note: a given action may not be possible in all states
- **Key issue**: representing the way we transition from one state to another by taking actions
- We can't do better than a **tabular representation**:

Starting state	Action	Resulting state
:	:	:

- Problems with this representation:
  - too big
  - hard to modify
  - doesn't capture underlying structure

# Example State-Based Representation

State	Action	Resulting State
$\langle lab, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle mc \rangle$	$\langle mr, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$
$\langle lab, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle mac \rangle$	$\langle off, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$
$\langle lab, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle nm \rangle$	$\langle lab, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$
$\langle off, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle mac, \overline{dm} \rangle$	$\langle cs, \overline{rhc}, \overline{swc}, \overline{m\overline{w}}, \overline{rhm} \rangle$
$\langle off, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle mac, \overline{dm} \rangle$	$\langle cs, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$
$\langle off, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle mc, \overline{dm} \rangle$	$\langle lab, \overline{rhc}, \overline{swc}, \overline{m\overline{w}}, \overline{rhm} \rangle$
$\langle off, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$	$\langle mc, \overline{dm} \rangle$	$\langle lab, \overline{rhc}, swc, \overline{m\overline{w}}, rhm \rangle$
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# Feature-Based Representation

- **Features** helped us to represent CSPs more compactly than states could. Can they help us here?
- Note that our problem is **trickier** than the original CSP problem:
  - we used to be looking for a single variable assignment, out of the space of all variable assignments, that satisfies our constraints
- Now we are looking for a **sequence of variable assignments** that
  - begins at the initial state
  - proceeds from one state to another by taking valid actions
  - ends up at a goal
- This means that instead of having one variable for every feature, we must instead have one variable for every feature at each time step, indicating the value taken by that feature at that time step!

# Feature-Based Representation

- 1 Model when the **actions are possible**, in terms of the values of the features of the current state
- 2 Model the **state transitions** in a “factored” way:

Why might this be more tractable/managable than the tabular representation?

# Feature-Based Representation

- 1 Model when the **actions are possible**, in terms of the values of the features of the current state
- 2 Model the **state transitions** in a “factored” way:

Why might this be more tractable/managable than the tabular representation?

- if some action doesn't depend on or modify some feature of the state, we can achieve some representational savings here
- The representation can be easier to modify/update.

# Feature-Based Representation

We need two things to replace the tabular representation:

- 1 Modeling when **actions are possible**:
  - Provide a function that indicates when an action can be executed
  - **Precondition** of an action: a function (proposition) of the state variables that is true when the action can be carried out



# Feature-Based Representation

We need two things to replace the tabular representation:

- 1 Modeling when **actions are possible**
- 2 Modeling **state transitions** in a “factored” way:
  - **causal rules**: explain how the value of a variable describing a feature at time step  $t$  depends on the action taken at time  $t - 1$ 
    - things that are changed in the world
    - example: *act* causes  $V_1 = v_1$

# Feature-Based Representation

We need two things to replace the tabular representation:

- ① Modeling when **actions are possible**
- ② Modeling **state transitions** in a “factored” way:
  - **causal rules**
  - **frame rules**: explain how the value of a variable describing a feature at time step  $t$  depends on the value of the variable that describes the same feature at time step  $t - 1$ 
    - things that are **not changed** in the world
    - example:  $V_4 = v_7$  is maintained when condition  $c$  holds
    - the need for frame rules is counter-intuitive: but remember, a computer doesn't know that the variables describing the same feature at different time steps have anything to do with each other, unless they are related to each other using appropriate constraints.

# Example

When is  $RLoc_{t+1} = cs$ ?

- $RLoc_{t+1} = cs$  when  $RLoc_t = cs$  and  $Move_t = nm$
- $RLoc_{t+1} = cs$  when  $RLoc_t = off$  and  $Move_t = mac$
- $RLoc_{t+1} = cs$  when  $RLoc_t = mr$  and  $Move_t = mc$

When is  $rhc$  true?

- $RHC_{t+1} = rhc$  when  $RHC_t = rhc$  and  $DelC_t = \overline{dc}$
- $RHC_{t+1} = rhc$  when  $PUC_t = puc$

Which of these rules are frame rules?

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# The STRIPS Representation

- The previous representation was **feature-centric**:
  - for every feature, where does its value come from?
- STRIPS is an **action-centric** representation:
  - for every action, what does it do?
- This leaves us with no way to state frame rules.
- **The STRIPS assumption**:
  - all variables not explicitly changed by an action stay unchanged

# STRIPS Actions

- In STRIPS, an action has **two parts**:
  - 1 **Precondition**: a logical test about the features that must be true in order for the action to be legal
  - 2 **Effects**: a set of assignments to variables that are caused by the action
- If the feature  $V$  has value  $v$  after the action  $a$  has been performed, what can we conclude about  $a$  and/or the state of the world?

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- If the feature  $V$  has value  $v$  after the action  $a$  has been performed, what can we conclude about  $a$  and/or the state of the world?
  - either  $V = v$  was true in the state of the world immediately preceding execution of action  $a$ , or  $a$  sets  $V = v$ , or both.

# STRIPS vs. Feature Representation

- How can we write **causal rules and frame rules** for an action *act* with effects list  $[V_1 = v_1, \dots, V_k = v_k]$ ?



# STRIPS vs. Feature Representation

- How can we write **causal rules and frame rules** for an action  $act$  with effects list  $[V_1 = v_1, \dots, V_k = v_k]$ ?
  - For each variable  $V_i$  in the effects list, write the causal rule " $V_i = v_i$  when  $act$ "
  - For each variable  $V_j$  not in the effects list, and every one of  $V_j$ 's values  $v_k$ , write the frame axiom " $V_{j,t+1} = v_k$  when  $act_t$  and  $V_{j,t} = v_k$ "

# Example

STRIPS representation of the action **pick up coffee**,  $PUC$ :

- **preconditions**  $Loc = cs$  and  $RHC = \overline{rhc}$
- **effects**  $RHC = rhc$

STRIPS representation of the action **deliver coffee**,  $DelC$ :

- **preconditions**  $Loc = of$  and  $RHC = rhc$
- **effects**  $RHC = \overline{rhc}$  and  $SWC = \overline{swc}$

Note that Sam doesn't have to want coffee for Rob to deliver it; one way or another, Sam doesn't want coffee after delivery.