

Heuristics in Planning; Uncertainty in Planning

Manuela M. Veloso

Carnegie Mellon University
Computer Science Department

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Readings:

- Chapter 10, Russell & Norvig

Heuristics in Planning

- State-space planning
 - Goal selection
 - Not domain-independent
 - Action selection
 - Number of preconditions true in state
 - Total number of preconditions
- Plan-space planning
 - No need for goal selection
 - Action with less open (unlinked) conditions

How to Generate Heuristics

- Domain-Specific
 - Program in (or *learn*) heuristics specifically for that domain
- Domain Analysis
 - Preprocess domain to generate meta-control knowledge
- On-Line
 - Solve a *relaxed* form of the problem
 - Use as estimate for original problem

Heuristics

- Assume complete *subgoal independence*
- Assume *no or limited negative* interactions
- Identify *landmarks*

Sources of Uncertainty

- Initial State
 - Unknown predicates
- Action Models
 - Non-deterministic effects
- Sensor Noise
 - Partially hidden state

Uncertainty

- State uncertainty
 - Possible worlds
- Action uncertainty
 - Preconditions
 - Effects
 - Conditional effects

Conditional Effects

- Action Sprinkle-grass
- Pre:
 - turned-on water
- Effects:
 - wet grass
 - If object on grass
 - Then wet object
- Result of applying action
- Use of conditional effects as goals

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Conditional Planning

- When applying an action, need “to sense” to verify conditions
- Determine state in which action will be applied
 - State-space planners
 - Plan-state planners
 - Other planners

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Conformant Planning

- Create a plan that achieves the goal *in spite of uncertainty* in state and action
- Examples:
 - Tray tilting

Conformant Graphplan (Smith & Weld, 1998)

- Extends Graphplan
 - Conditional effects
 - Multiple possible worlds
 - Uncertainty represented using disjunction
- Basic Approach:
 - Create parallel plan graphs for each possible world
 - Search backwards through each plan graph in parallel
 - **Confront** undesired effects in one possible world caused by choosing action in another possible world

BDD Representation

- Compact representation of Boolean formulae
 - Invented by Dean Randy Bryant (1985)

- A BDD is a DAG with:
 - Terminal nodes labeled 0 or 1
 - A set of variable nodes with two edges:
low (false) and *high* (true)
 - A linear ordering of variables.
 - Bad variable ordering leads to exponential size
 - Reductions:
 - Uniqueness of nodes associated to the same variable
 - No redundant tests

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Summary

- Heuristics
- Uncertainty
 - Conditional planning
 - Conformant planning
 - Nondeterministic planning

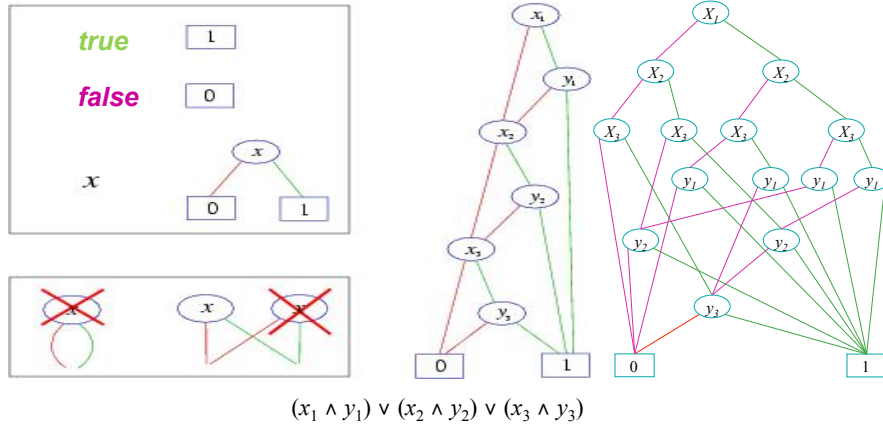
- Next
 - Probabilistic representations
 - Planning under probabilistic uncertainty

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BDD Example



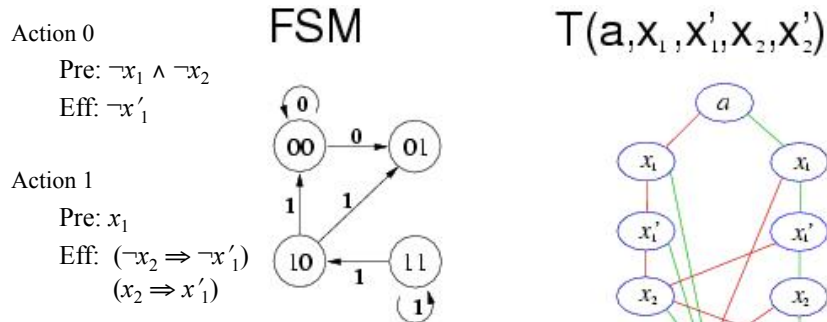
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Representing Actions

- Representing the Transition Relation
 - Boolean formula describing relationship between variables in current and next state: $T(a, x, x')$



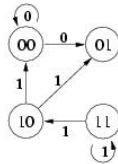
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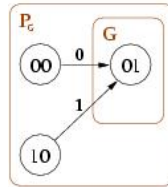
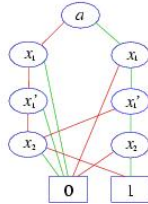
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Pre-Image Computation

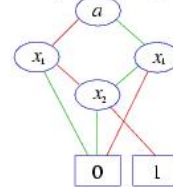
FSM



$T(a, x_1, x_1', x_2, x_2')$



$P_G(a, x_1, x_2)$



$$\text{Pre-Image}(a, x) = \exists x'. T(a, x, x') \wedge V(x')$$

$$P_G(a, x_1, x_2) = \exists x_1', x_2'. T(a, x_1, x_2, x_1', x_2') \wedge (\neg x_1' \wedge x_2')$$

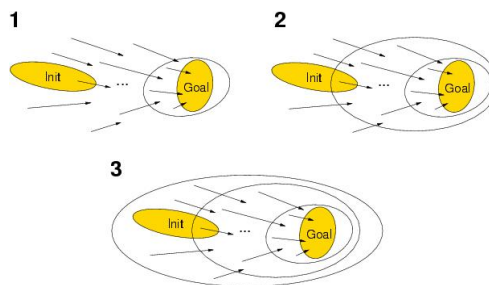
$$\text{Post-Image}(a, x) = \exists x. T(a, x, x') \wedge V(x)$$

BDD-Based Planning

- Key Idea: Solve for **Sets** of States

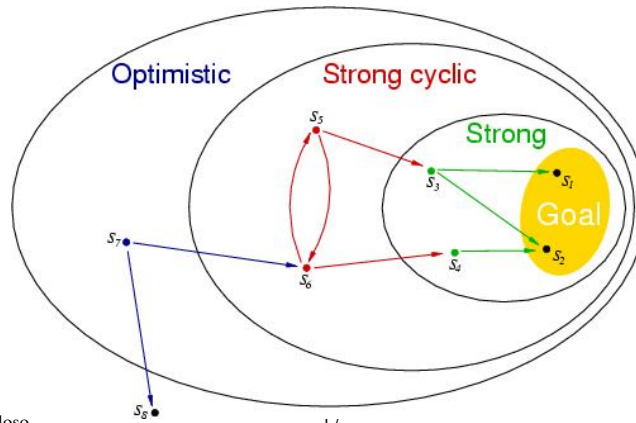
```

function Plan(T, I, G)
  U := 0; V := G
  while I not subset V
    U_c := PreComp(T, V)
    if U_c = 0 then
      return failure
    else
      U := U union U_c
      V := V union states(U_c)
  return U
  
```



BDD-Based Planning

- Different “Guarantees” for Generated Policies
 - Depends on strictness of *PreComp* rule



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