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Planning in Robotics

- Since Shakey, planning evolved along two largely independent dimensions
- Solving real tasks like setting a table requires both:
- Task planning
 - Figure out long-term strategies (use tray?), stacking order
 - · Input: STRIPS-like initial state, goal, actions
- Motion planning
 - Compute a continuous path in the space of configurations of the robot
 - Input: initial & target configurations
 - Configuration: high-dimension vector
 - E.g., 20-dimensions for the PR2
 - Arms (2x8)
 - Base (3)
 - Torso height (1)



Combining Task and Motion Planners

- Task planners:
 - Effective algorithms for implicitly defined discrete transition systems
 (HSP [Bonet & Geffner '98], FF [Hoffmann & Nebel '01], FD [Helmert '06])
 - Not directly applicable to continuous state and action spaces
- · Motion planners:
 - Effective algorithms for search in high-dimensional continuous space (PRM [Kavraki et al., '96], RRT [LaValle '98], CHOMP [Ratliff et al., 2009], Trajopt [Schulman et al., 2013])
 - Not directly applicable to discrete action sequencing

Blocksworld - Pick and Place

• Planning to re-order the blocks



Assume the arm can reach for any block, can move any block if required

Blocksworld - Pick and Place

Planning to re-order the blocks



What is a state?
What is an action?

Blocksworld - Pick and Place

Planning to re-order the blocks

Actions:

Move(b,x,y) – moves block b from x to y MoveToTable(b,x) – moves block b from x to table y

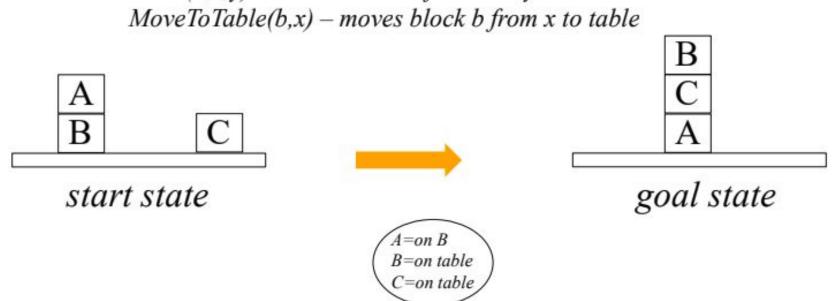


Blocksworld - Towards Graph Search

Planning to re-order the blocks

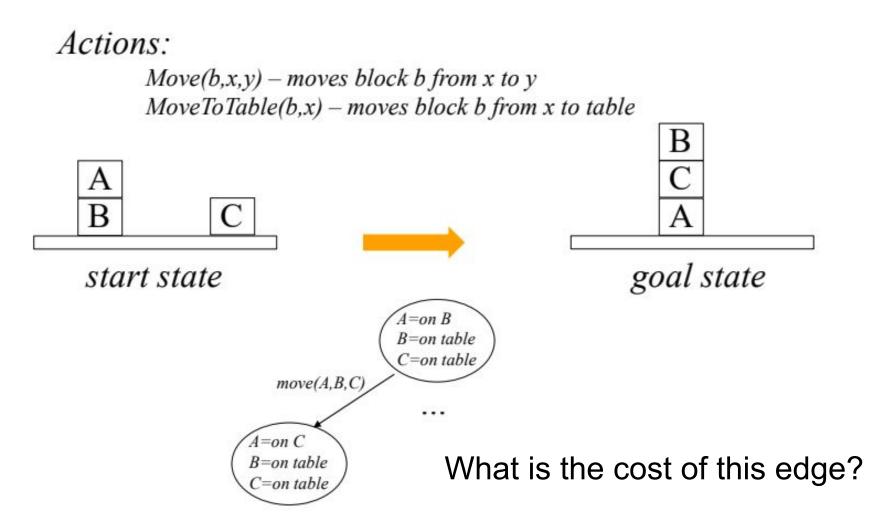
Actions:

Move(b,x,y) – moves block b from x to y



Blocksworld - Towards Graph Search

Planning to re-order the blocks



Symbolic Planning

Can we specify any task planning problem with a representational language that allows definition of states, actions and goals?

• STRIPS (=Stanford Research Institute Problem Solver)

State Representation:

Goal Representation:

Action Representation:

STRIPS (=Stanford Research Institute Problem Solver)

State Representation:

conjunction of positive(true) literals

 $(e.g, On(A,B)^On(B,Table)^On(C,Table)^Block(A)^Block(B)^Block(C)^Clear(A)^Clear(C))$

Goal Representation:

Action Representation:

Closed-World Assumption: Unspecified information is FALSE!

STRIPS (=Stanford Research Institute Problem Solver)

State Representation:

conjunction of positive(true) literals

 $(e.g, On(A,B)^On(B,Table)^On(C,Table)^Block(A)^Block(B)^Block(C)^Clear(A)^Clear(C))$

Goal Representation:

desired conjunction of positive(true) literals

Action Representation:

What is the goal representation for the Block-World Example? Can be partially specified.

STRIPS (=Stanford Research Institute Problem Solver)

State Representation:

conjunction of positive(true) literals

 $(e.g, On(A,B)^On(B,Table)^On(C,Table)^Block(A)^Block(B)^Block(C)^Clear(A)^Clear(C))$

Goal Representation:

desired conjunction of positive(true) literals

Action Representation:

Preconditions: conjunction of positive(true) literals that must be held true in order for the action to be applicable **Effect**: conjunction of positive(true) literals showing how the state will change (what should be deleted and added)

MoveToTable(b,x)

Precond: $On(b,x)^Clear(b)^Block(b)^Block(x)$

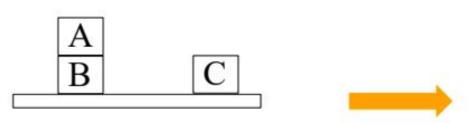
Effect: $On(b, Table)^Clear(x)^-On(b, x)$

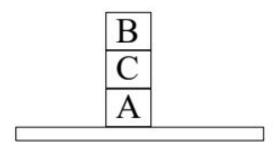
Move(b,x,y)

Precond: $On(b,x)^Clear(b)^Clear(y)^Block(b)^Block(y)^(b\sim=y)$

Effect: $On(b,y)^Clear(x)^-On(b,x)^-Clear(y)$

Representing it with STRIPS





Start state:

 $On(A,B)^On(B,Table)^On(C,Table)^Block(A)^Block(B)^Block(C)^Clear(A)^Clear(C)$

Goal state:

 $On(B,C)^On(C,A)^On(A,Table)$

Actions:

MoveToTable(b,x)

Precond: $On(b,x)^Clear(b)^Block(b)^Block(x)$

Effect: $On(b, Table) \land Clear(x) \land \neg On(b, x)$

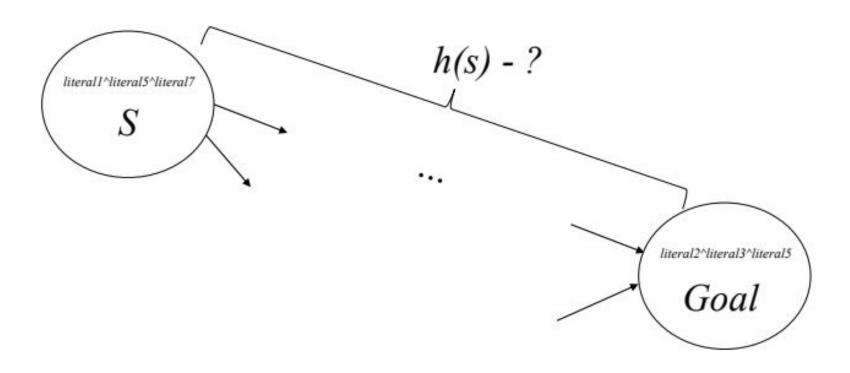
Move(b,x,y)

Precond: $On(b,x)^Clear(b)^Clear(y)^Block(b)^Block(y)^(b\sim=y)$

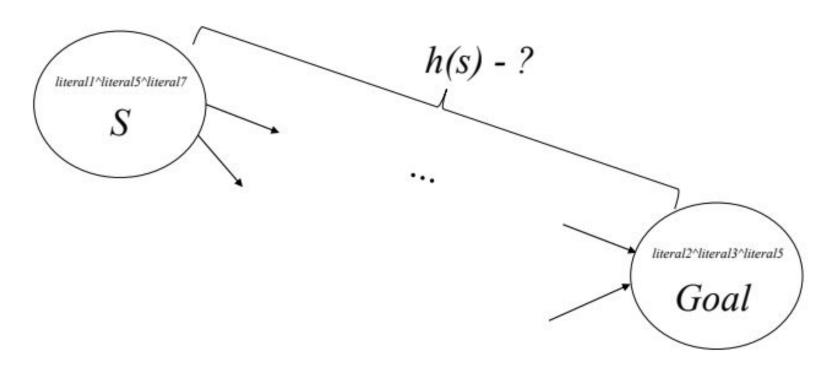
Effect: $On(b,y)^Clear(x)^-On(b,x)^-Clear(y)$

- 1. Represent Problem using STRIPS
- 2. Design a domain-independent program
- Define GetSuccessor() Function to work with Implicit Graphs
- 4. Implement A* or similar algorithms
- 5. What could be "domain-independent" heuristics?

Computing heuristics



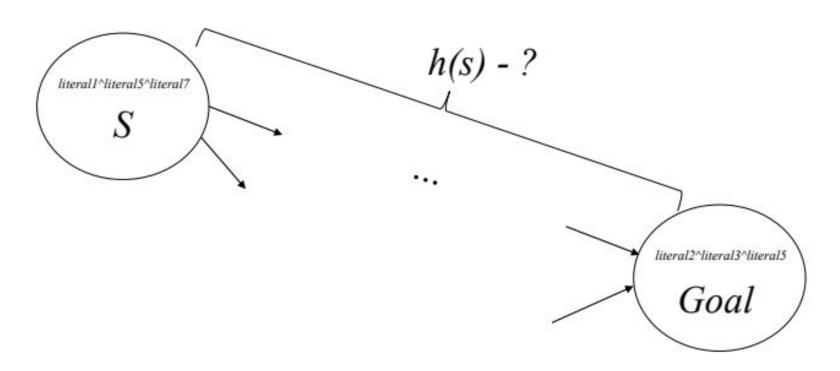
Computing heuristics



Option 1: Number of literals not yet satisfied?

Admissible? Can we use it?

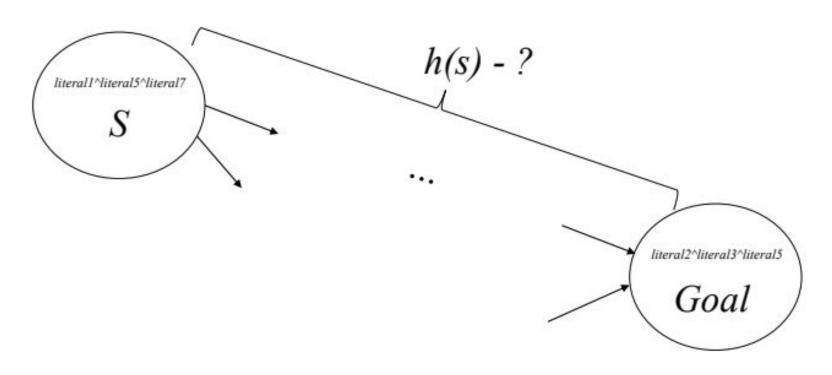
Computing heuristics



Option 2: Compute heuristic using a relaxed problem.

Ex: Actions don't have -ve effects (empty-delete-list heuristic)

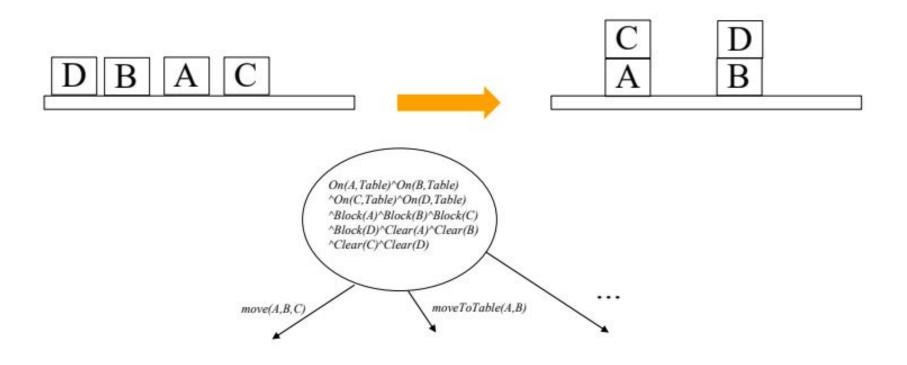
Computing heuristics



Option 2: Compute heuristic using a relaxed problem.

Admissible? Pros and Cons?

Challenges in Naive Graph Search



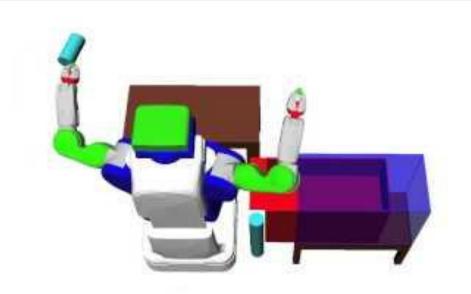
Graphs can blow up!!

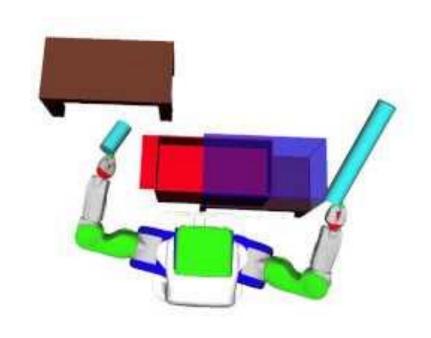
Possible solution: Planning Graphs! [See the other slide deck]

Where does Motion Planning come into play?

Where does Motion Planning come into play?

- 1. Action Verification Collisions. Robot feasibility.
- 2. Sequence Verification Order feasibility.
- 3. Sequence Optimization Edges have non-unit costs.





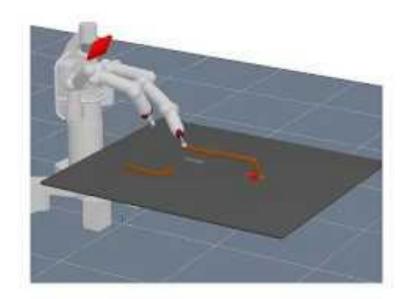
Time STREET,

Found Solutions

The only goal specification is to touch the rediball with either hand, or to let the blue ball touch the green patch.

The system has full knowledge of the scene, including the geometric shapes of all objects, but knows of no further semantics specific to objects.

The double-hook, in analogy to Betty-the-Crow



Toussaint, Allen, Smith, Tenenbaum: Differentiable Physics and Stable Modes for Tool-Use and Manipulation Planning (R.SS. 2018)