
Hierarchical task and motion planning

Tomas Lozano-Perez
Leslie Pack Kaelbling
MIT CSAIL

Real robot meets real world



Real robot meets real world: large spaces



Configuration space

- joint angles of robot
- base pose
- positions and orientations of all objects in house
- are the dishes clean?
- is the stove on?
- are the leftovers edible?

Real robot meets real world: long horizon



Steps to clean kitchen:

- put away food (10 items)
- wash dishes (40 items)
- put away dishes (40 items)
- clean surfaces (10 items)

Or:

- 1,000 pick, place, or wipe

Or:

- 100,000 linearly interpolated joint motions

Real robot meets real world: uncertainty



Current-state uncertainty

- What is inside the tupperware?
- Is the dishwasher clean?
- What is the exact pose of the pot?
- What's the friction of a wet dish?

Predictive uncertainty

- What will happen when the robot lifts the cookie sheet?
- What is the error in the motor control?
- When will the inhabitants come home?

Addressing real world challenges

Large continuous
and discrete state
space



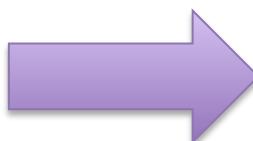
Symbolic and geometric
descriptions of state sets

Long planning and
execution horizon



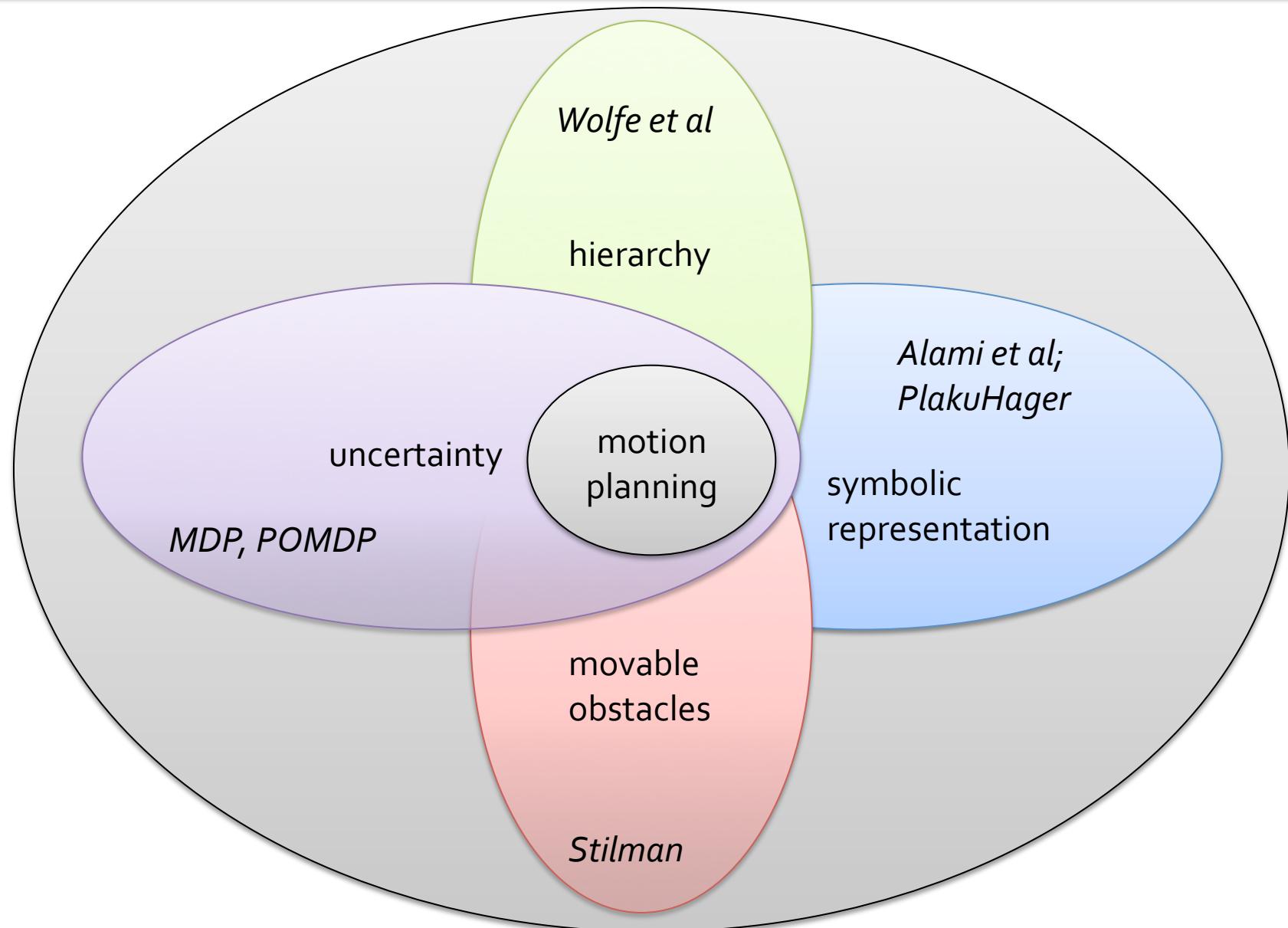
Temporal hierarchical
decomposition

Present and
predictive
uncertainty



Replanning with
determinized models in
belief space

Some related approaches



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Symbolic and geometric descriptions of state sets

Symbols for non-geometric properties

- are the dishes clean? *Clean(dish24)*
- is the stove on? *On(stove)*
- are the leftovers edible? *Edible(lasagna)*

Symbols for geometric and physical abstractions

- is the beer in the refrigerator? *In(beer, refrigerator)*
 - region containment
- is the robot holding the pot? *Holding(pot)*
 - contacts, force closure, ...

Short descriptions
for large sets

Classical symbolic action descriptions

```
Put(Block, Target):  
exists: Source  
pre: On(Block, Source),  
      Clear(Block),  
      Clear(Target)  
result:  
      On(Block, Target),  
      not On  
      (Block, Source),  
      Clear(Source),  
      not Clear(Target)
```

- Requires complete symbolic description of world state
- No explicit reasoning: requires all results to be asserted
- Requires enumeration of all possible values of variables

Symbolic planning in the real world

Classical formulation

- Requires complete symbolic description of world state
- Requires enumeration of all possible values of variables
- No explicit reasoning: requires all results to be asserted

Real-world formulation

- Symbolic description incomplete in infinite domain (regions, ...)
- Generators for values of existential variables
- Extra reasoning about consistency and entailment for geometric properties

Specific start state; abstract goal

Initial state known in geometric detail



Goal set is abstract, symbolic

$tidy(house) \wedge charged(robot)$

Planning operator: Place

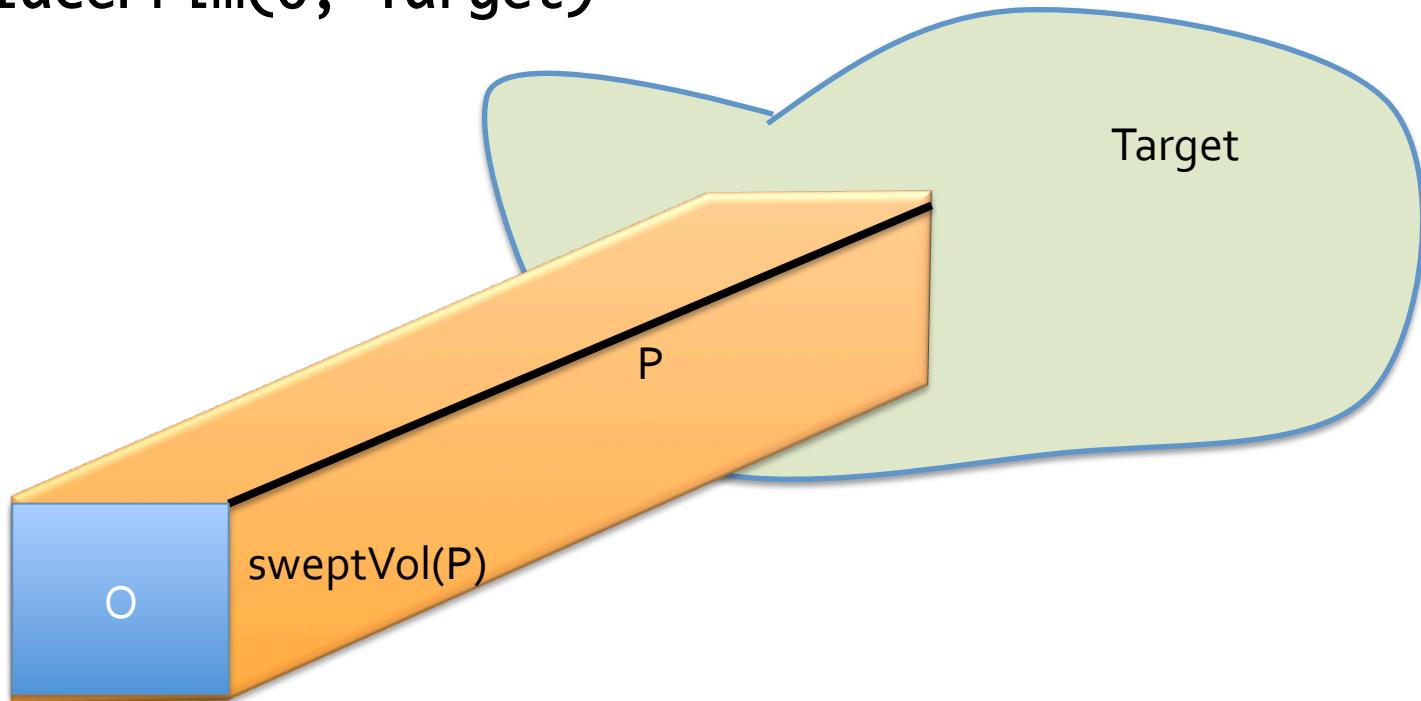
Place(0, Target):

exists: $P \in \text{generatePlacePaths}(0, \text{Target})$

pre: ClearX(sweptVol(P), 0), Holding(0)

result: In(0, Target)

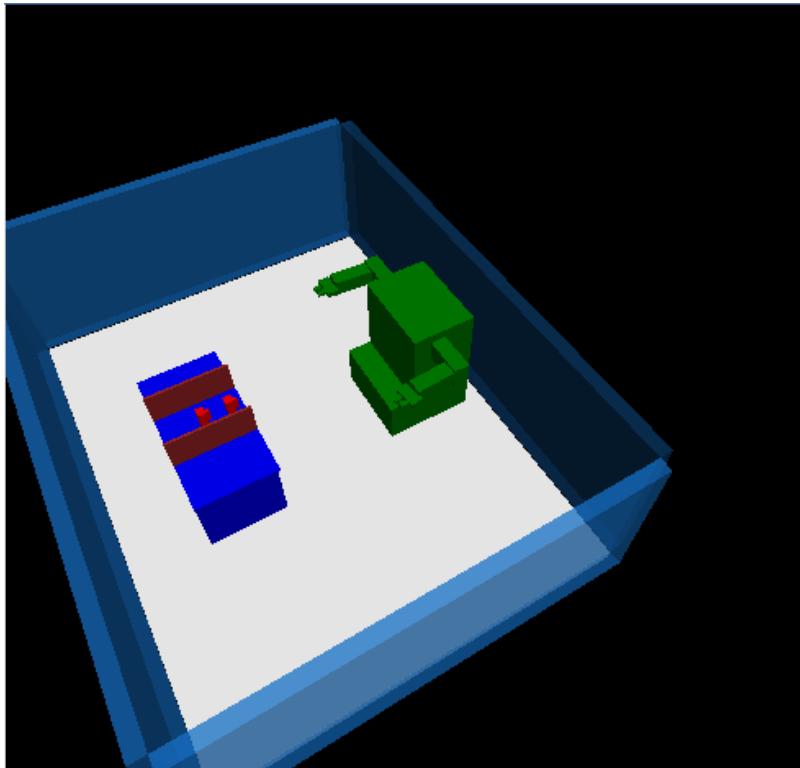
prim: PlacePrim(0, Target)



Generators

Efficient conservative planner

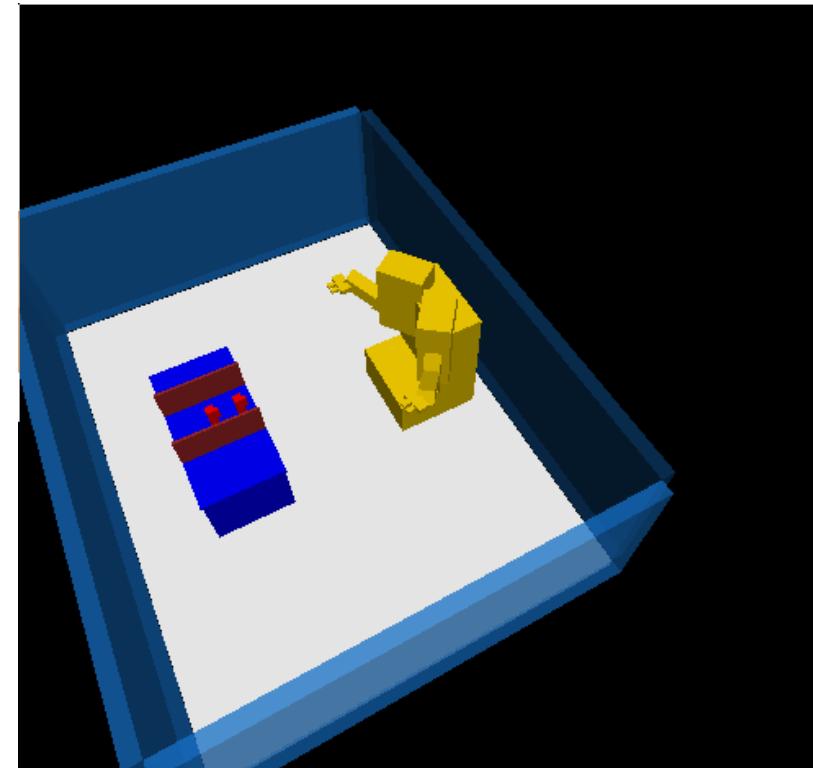
- Simplified robot model
- Objects: x, y, z, th
- Grasp selection



Primitives

Your favorite motion planner

- Accurate robot model
- No reasoning about objects
- Grasps from generator

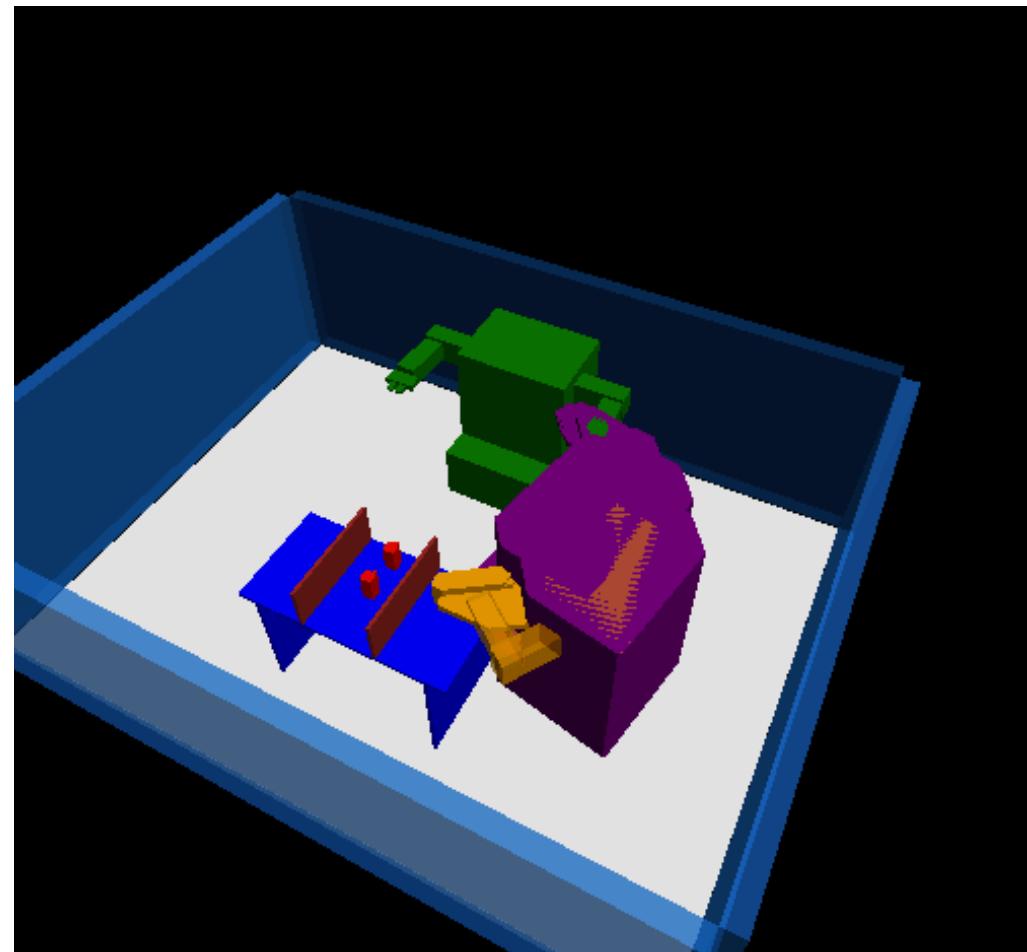


Generators

Place(0, Target):

exists: $P \in \text{generatePlacePaths}(0, \text{Target})$

- visibility graph planner
- fail fast
- conservative
- not the path we'll use:
certificate that one exists

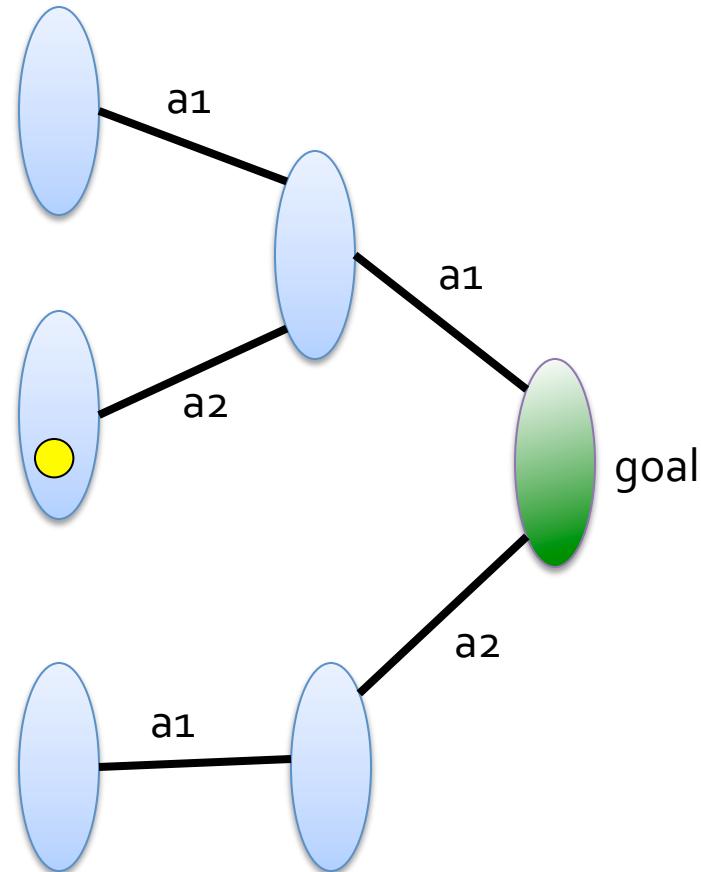


Goal Regression / Pre-image backchaining

Weakest precondition of goal set under each action sequence

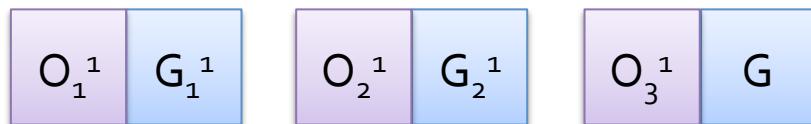
Test whether start state is in a pre-image

Represent goal and pre-images as conjunctions of fluents



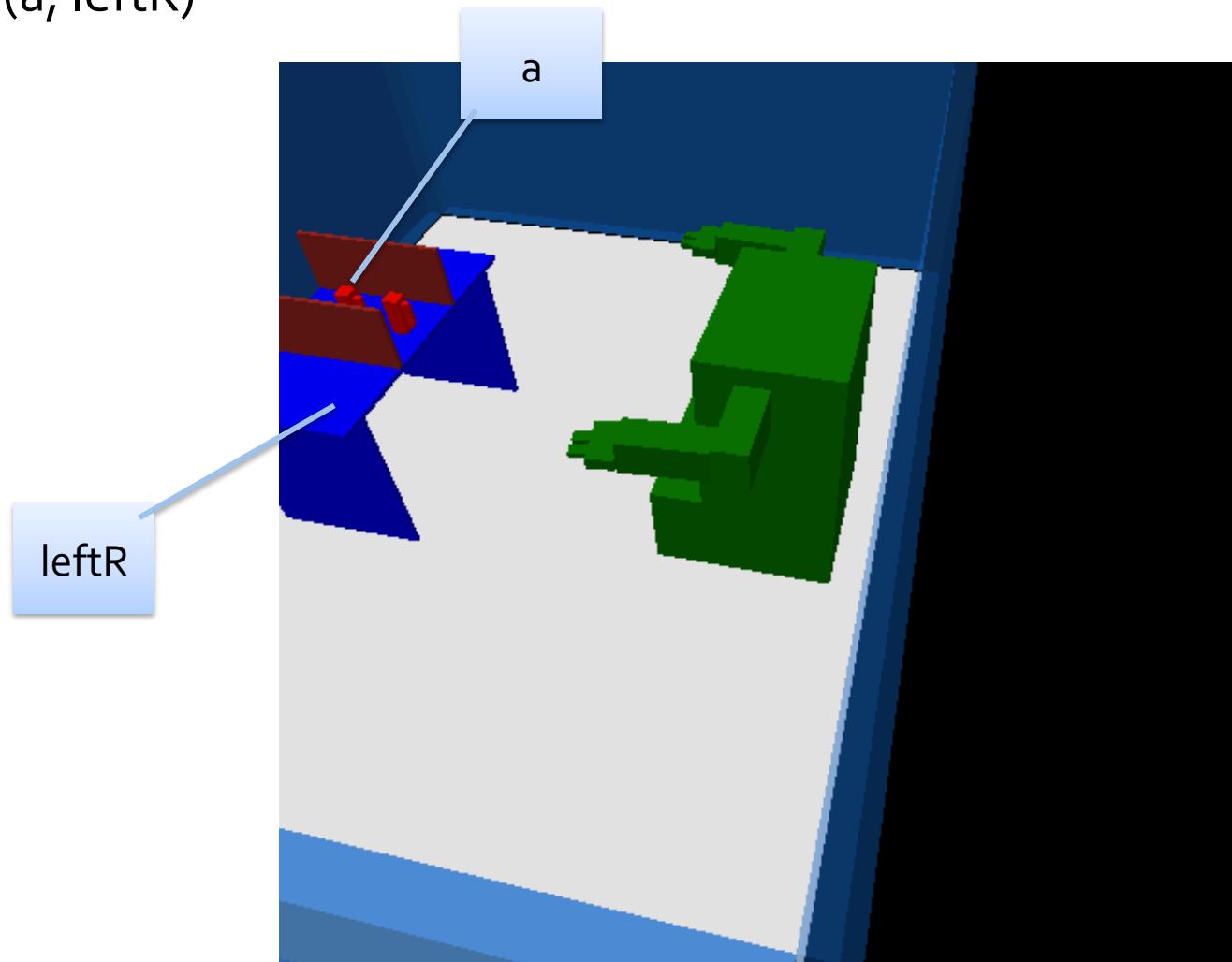
Why regression?

- Compact logical description of infinite pre-image sets
- Just need to test whether starting geometry is in the set
- **Goal-directedness:**
 - In natural settings, the initial state is incredibly detailed
 - Forward branching factor is huge!!
 - Goal has simple symbolic description
- Clear criteria for execution monitoring



Regression and generation

Goal: $\text{In}(a, \text{leftR})$

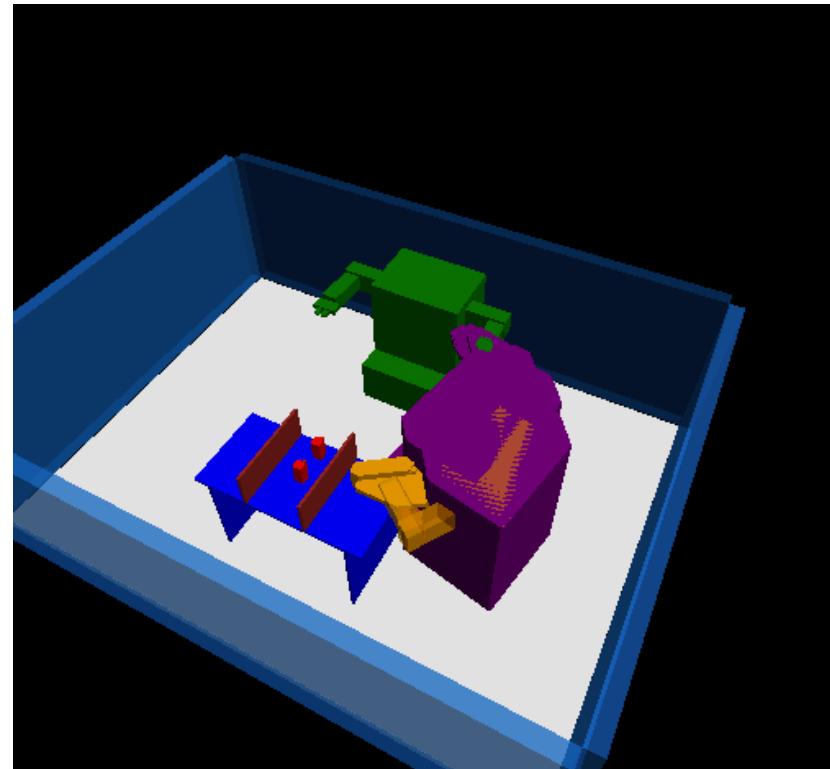


Regression and generation

Goal: $\text{In}(a, \text{leftR})$

O: $\text{Place}(a, \text{leftR})$

G: **Holding(a)**, $\text{ClearX}(\text{Swept}(\text{PlacePath}(a, \text{leftR})))$



Regression and generation

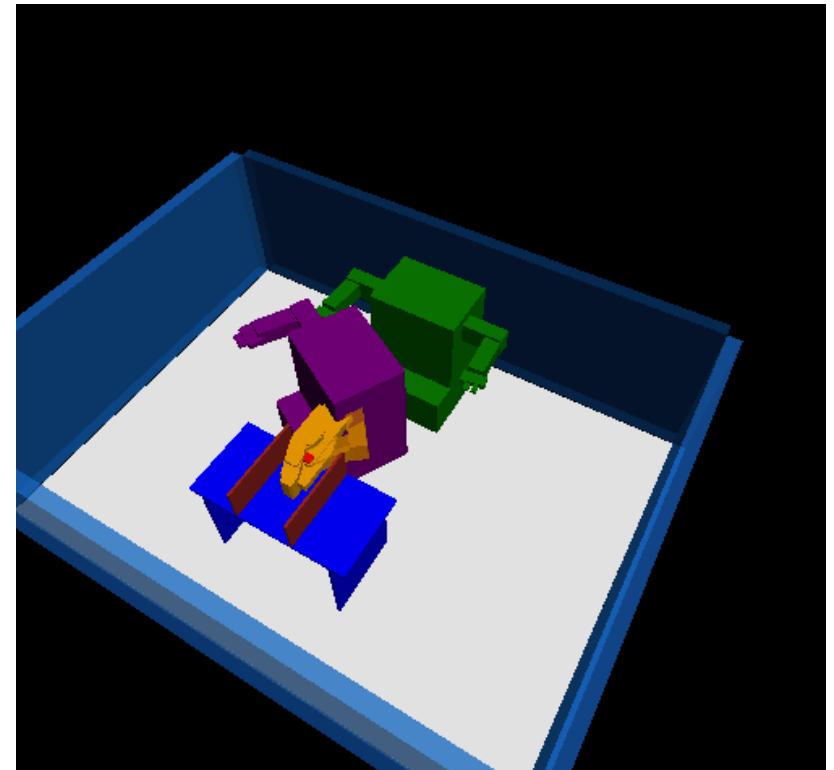
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O: $\text{Place}(a, \text{leftR})$

G: **Holding(a)**, $\text{ClearX}(\text{Swept}(\text{PlacePath}(a, \text{leftR})))$

O: $\text{Pick}(a)$

G: **ClearX(Swept(PickPath(a)),**
ClearX(Swept(PlacePath(a, leftR))



Regression and generation

Goal: $\text{In}(a, \text{leftR})$

O: $\text{Place}(a, \text{leftR})$

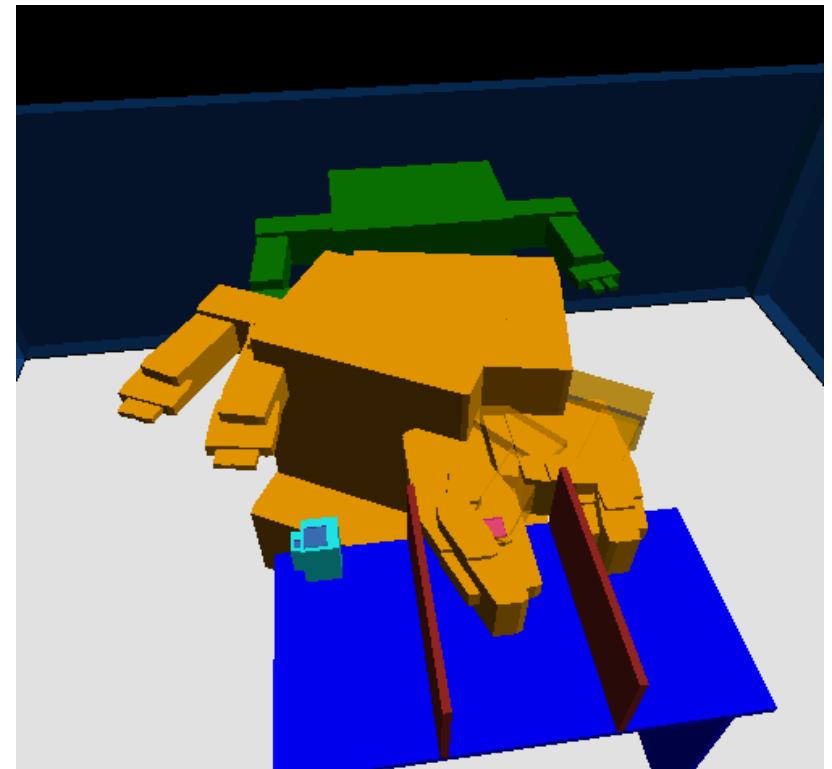
G: **Holding(a)**, $\text{ClearX}(\text{Swept}(\text{PlacePath}(a, \text{leftR})))$

O: $\text{Pick}(a)$

G: **ClearX(Swept(PickPath(a)),**
ClearX(Swept(PlacePath(a, leftR)))

O: $\text{Remove}(b, \text{Swept}(\text{PickPath}(a)))$

G: **In(b, Parking(b))**,
 $\text{ClearX}(\text{Swept}(\text{PlacePath}(a, \text{leftR})))$



Regression and generation

Goal: In(a, leftR)

O: **Place(a, leftR)**

G: Holding(a), ClearX(Swept(PlacePath(a, leftR)))

O: **Pick(a)**

G: ClearX(Swept(PickPath(a))), ClearX(Swept(PlacePath(a, leftR)))

O: Remove(b, Swept(PickPath(a)))

G: In(b, Parking(b)), ClearX(Swept(PlacePath(a, leftR)))

O: **Place(b, Parking(b))**

G: Holding(b), ClearX(Sw(PP(b, Park(b)))), ClearX(Sw(PlaceP(a, leftR)))

O: **Pick(b)**

G: ClearX(Sw(PickP(b))), ClearX(Sw(PP(b, Park(b)))),
ClearX(Sw(PlaceP(a, leftR)))

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Large continuous
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Symbolic and geometric
descriptions of state sets

Long planning and
execution horizon



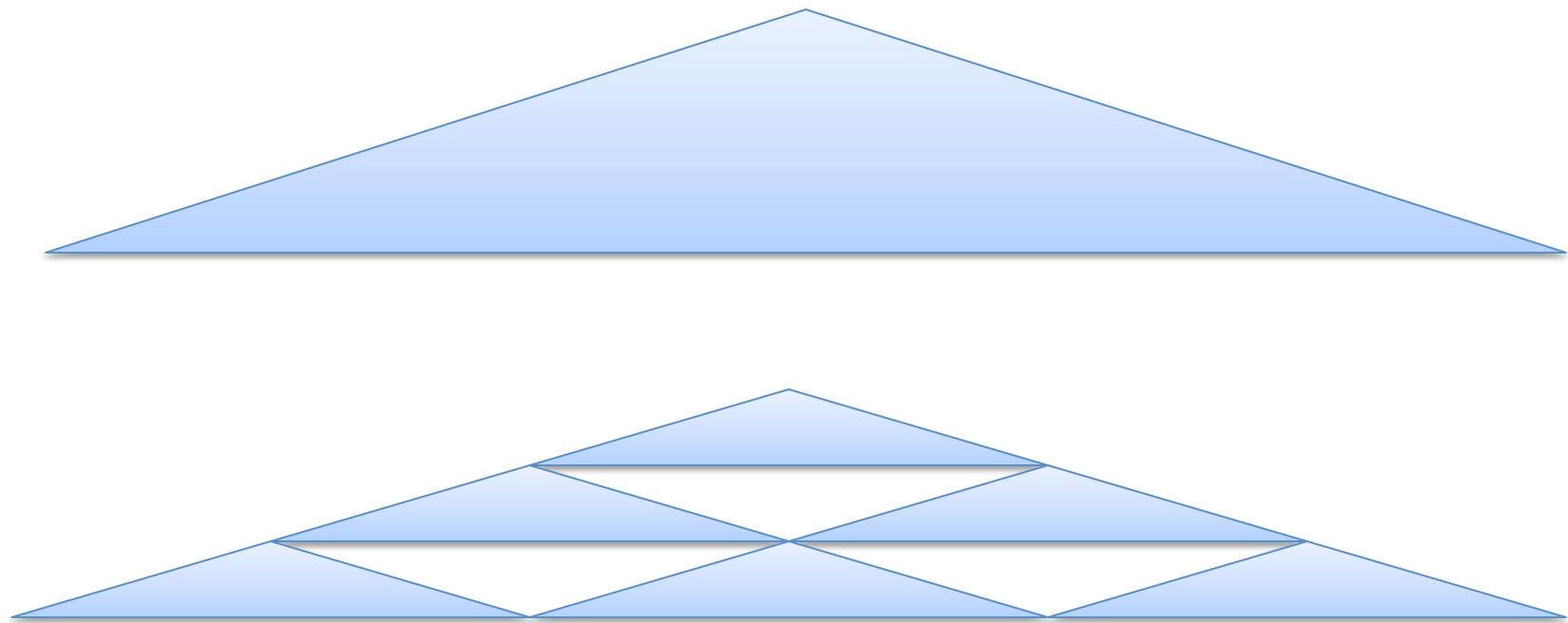
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Replanning with
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Hierarchy crucial for large problems

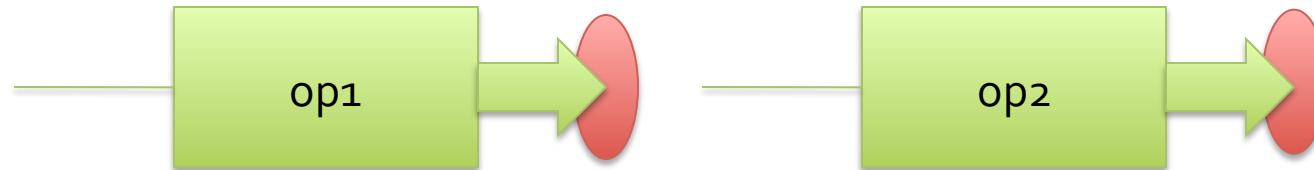


Hierarchical semantics

Subgoal is an abstract operator:



What does it mean to sequence two subgoals?



Depends on who gets to choose the outcome:



us



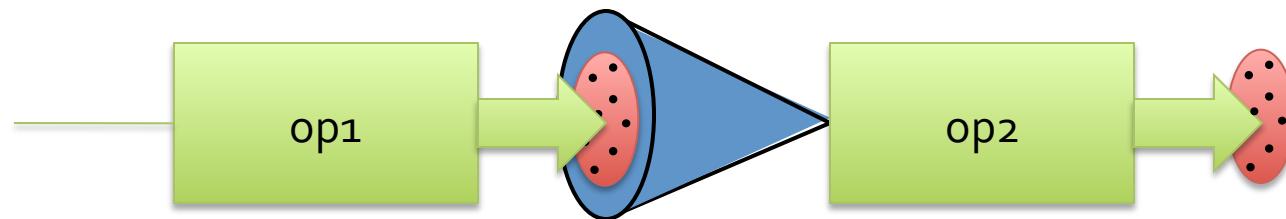
nature

Wolfe, Marthi, Russell

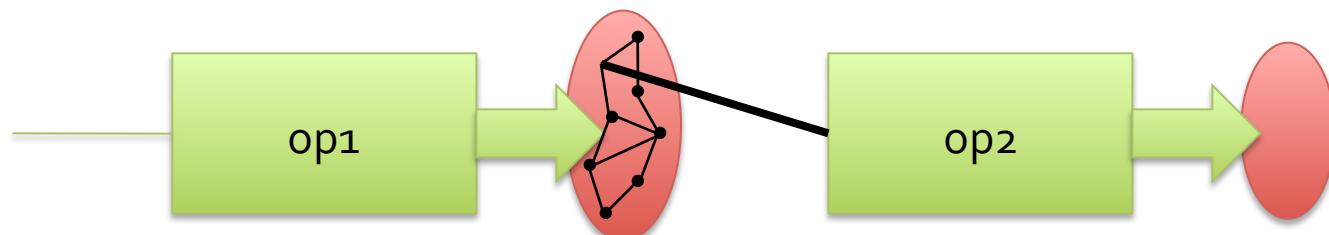
Satanic Semantics



We have to handle any outcome the devil picks



Okay if: Preconditions of op2 can be achieved from
any state resulting from op1



Postponing preconditions creates hierarchy

Pick(0):

pre: ...

res: Holding(0)

Place(0, R):

pre: 0.Holding(0)

1.InRobot(room(R))

res: In(0, R)



Postponing preconditions creates hierarchy

Pick(0):

pre: ...

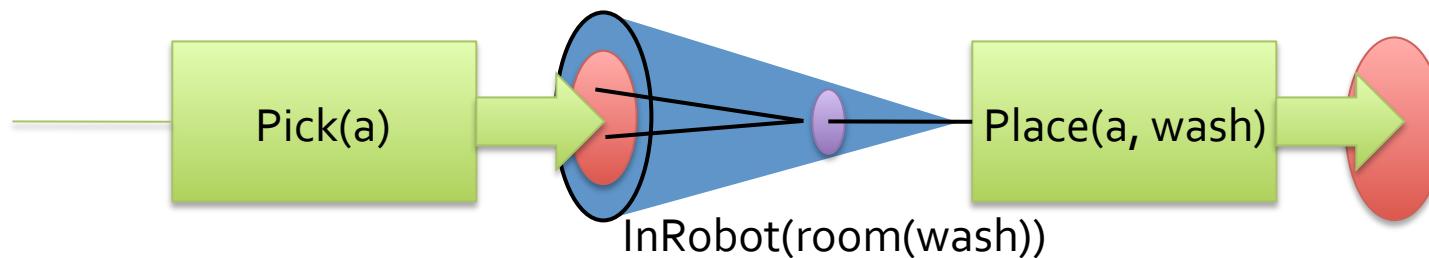
res: Holding(0)

Place(0, R):

pre: 0.Holding(0)

~~1.InRobot(room(R))~~

res: In(0, R)



Precondition `InRobot(room(R))` is serialized with the place primitive

Hierarchical plan

Pick(0):

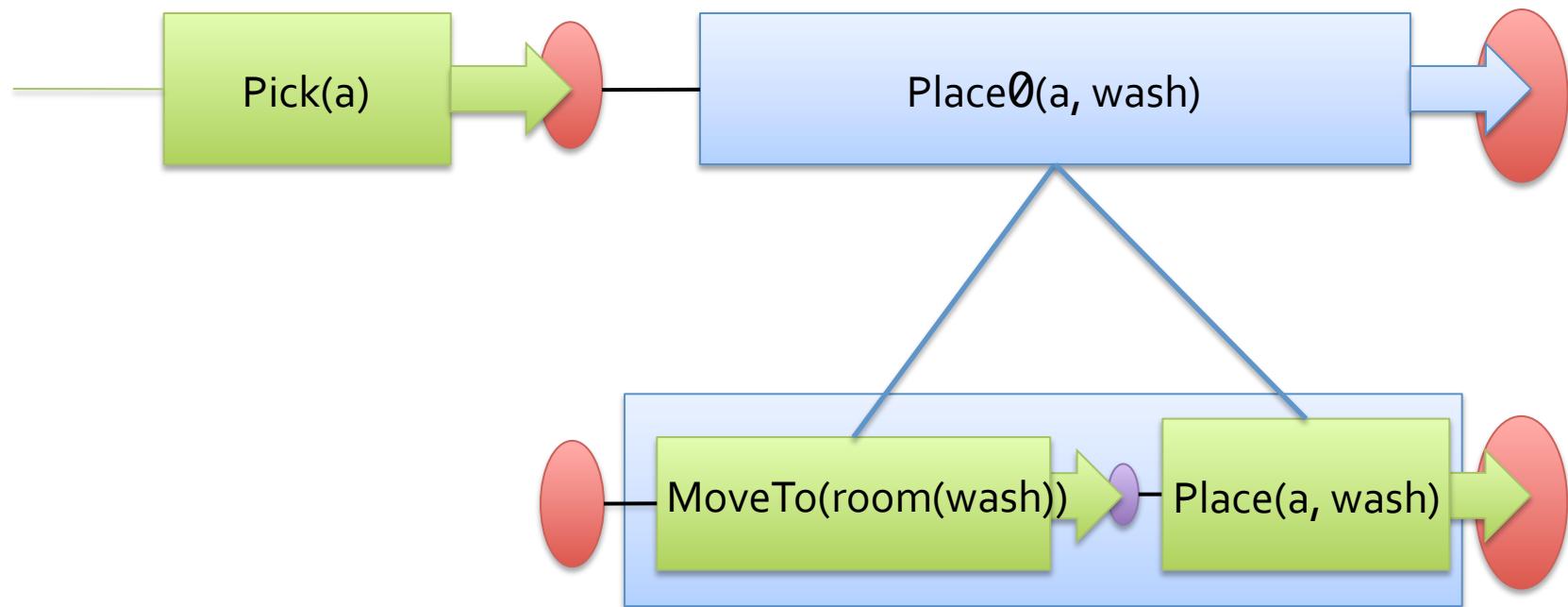
pre: ...

res: Holding(0)

Place0(0, R):

pre: Holding(0)

res: In(0, R)



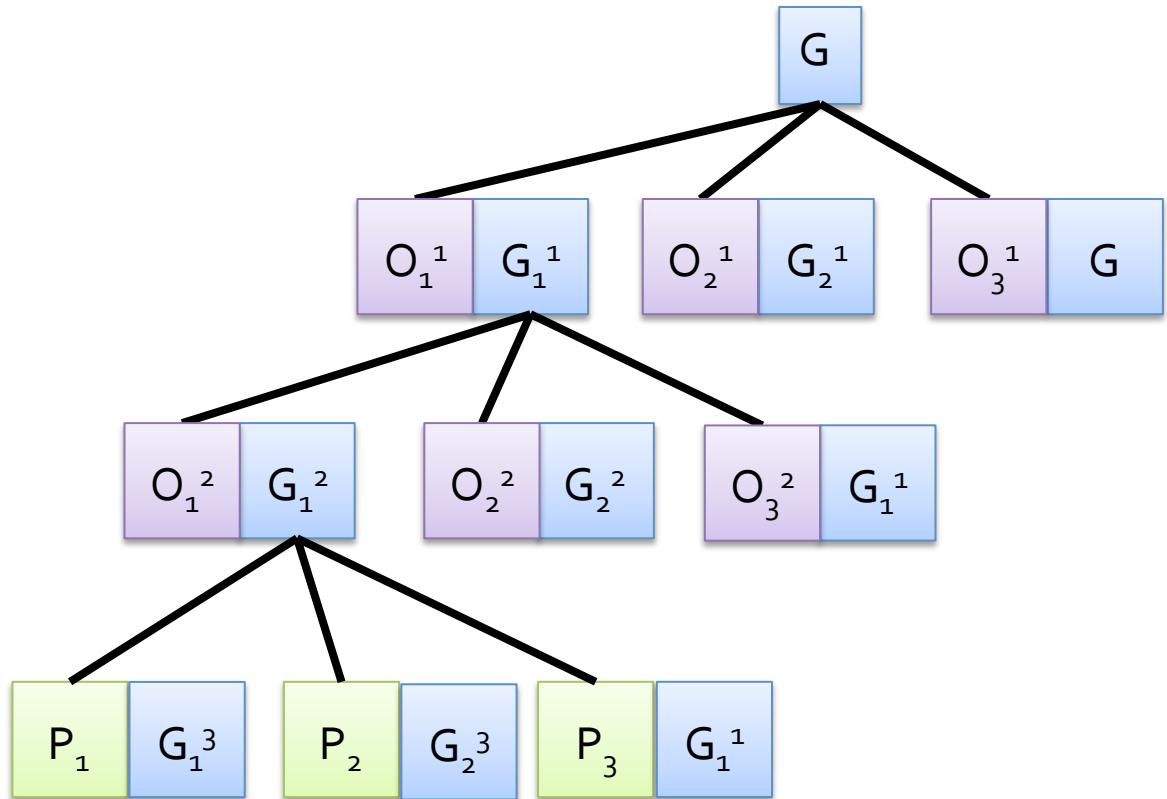
In the now



- Don't consider all the ways pick(a) could terminate: grasp of object, location of robot,
- When it is time to plan for Place0(a, wash), actual world state is start state

Planning in the now

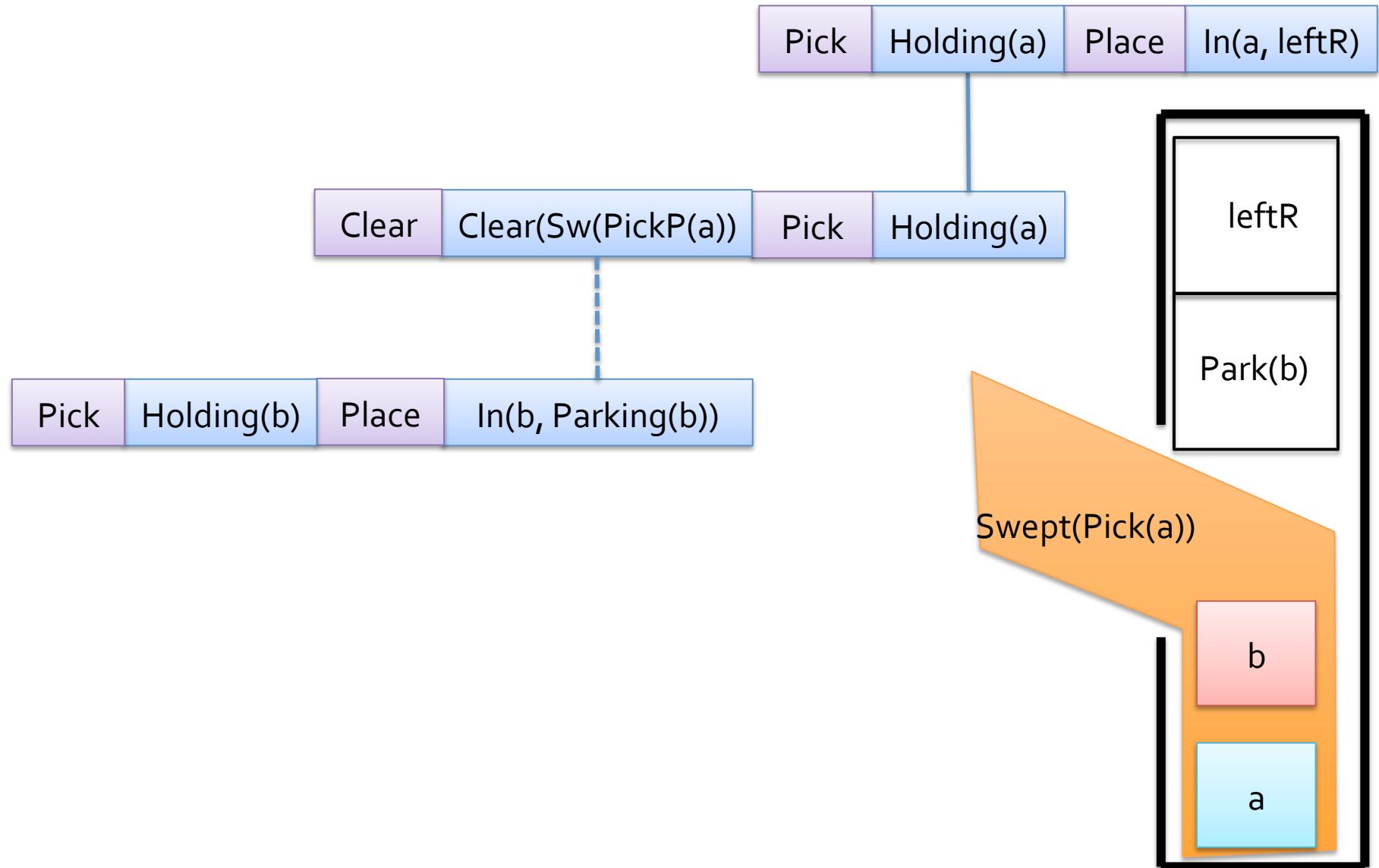
- maintain left expansion of plan tree
- each level uses a higher-fidelity model
- keep track of pre-image for each operation
- recursively plan to achieve those preconditions
- execute primitives



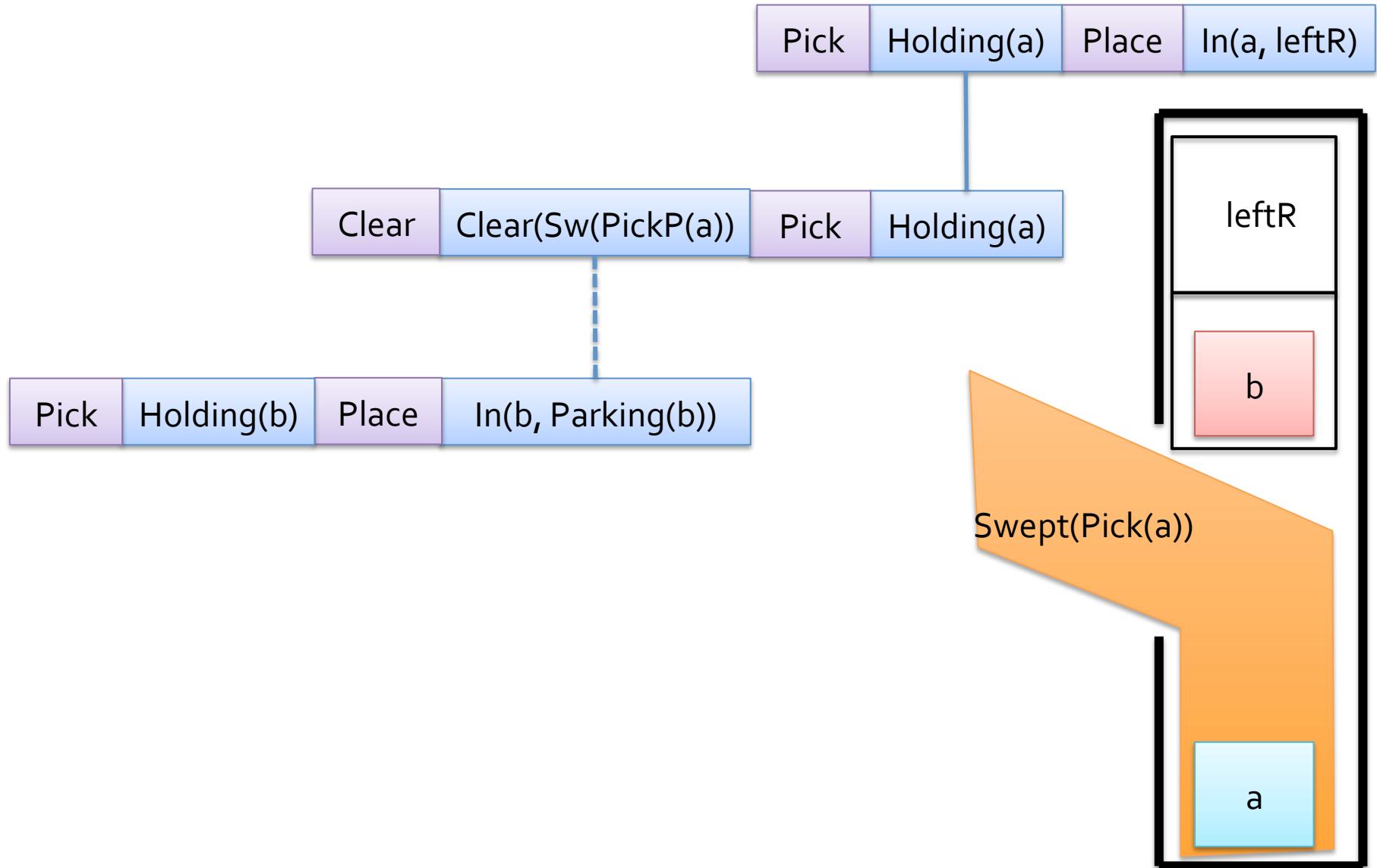
HPN Algorithm

```
HPN(currentState, goal, operators, absLevel, world):  
    if holds(goal, currentState):  
        return TRUE  
    else p = PLAN(currentState, goal, operators, absLevel)  
        for (oi, gi) in p  
            if prim(oi):  
                currentState = world.execute(oi)  
            else HPN(currentState, gi, operators,  
                      NEXTLEVEL(absLevel, oi), world)
```

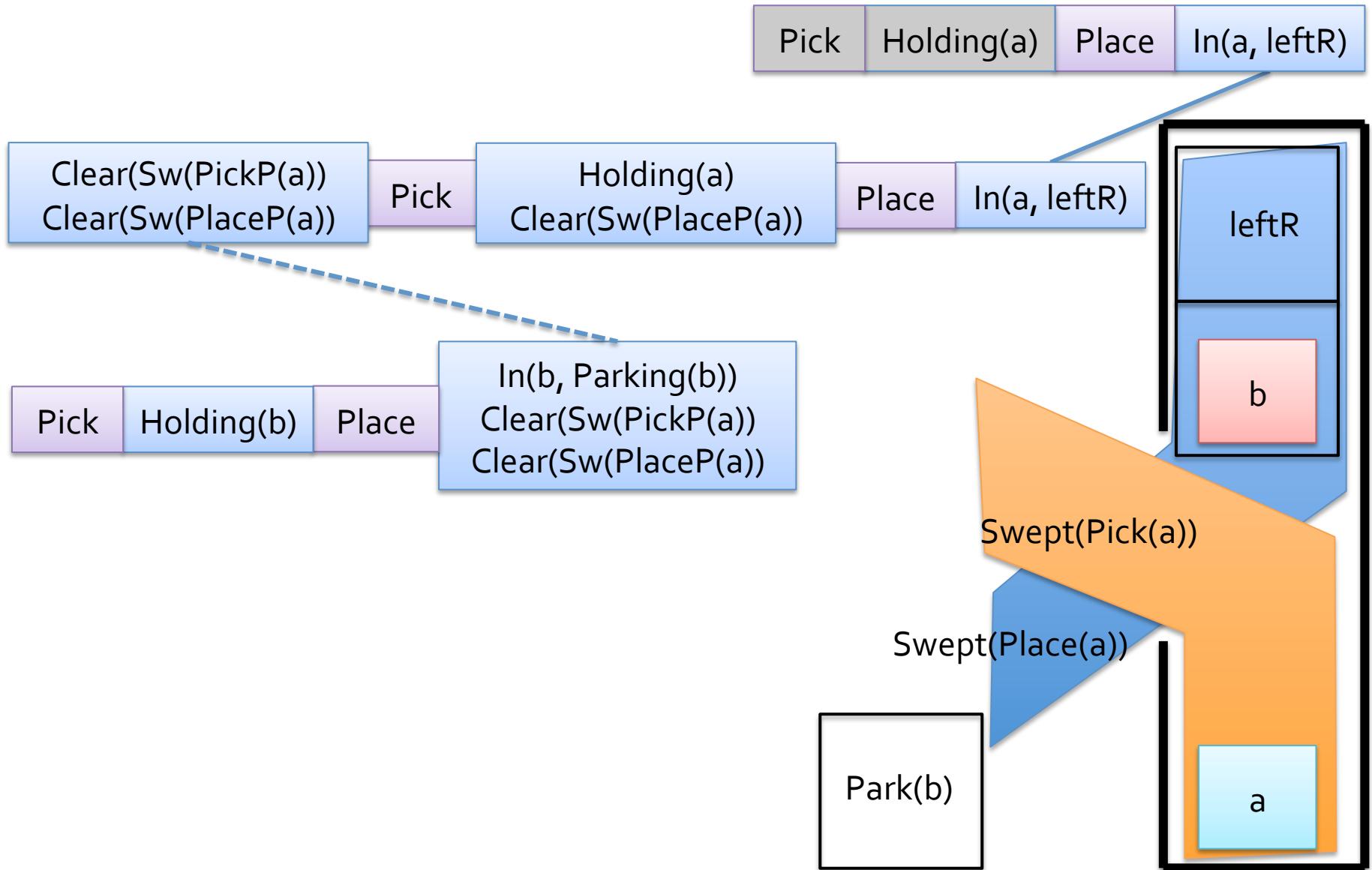
Hierarchical pick and place



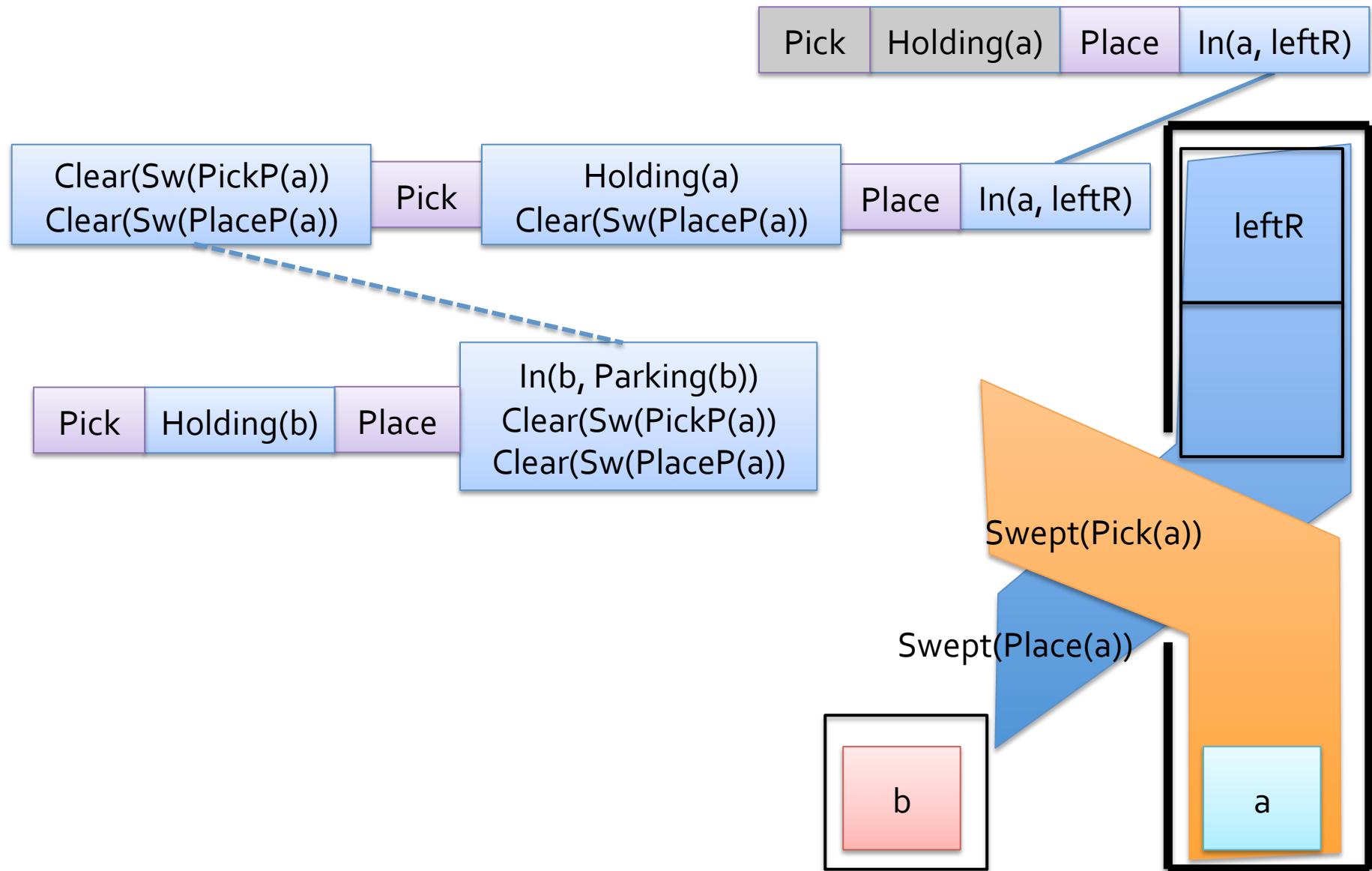
Hierarchical pick and place: too abstract!!



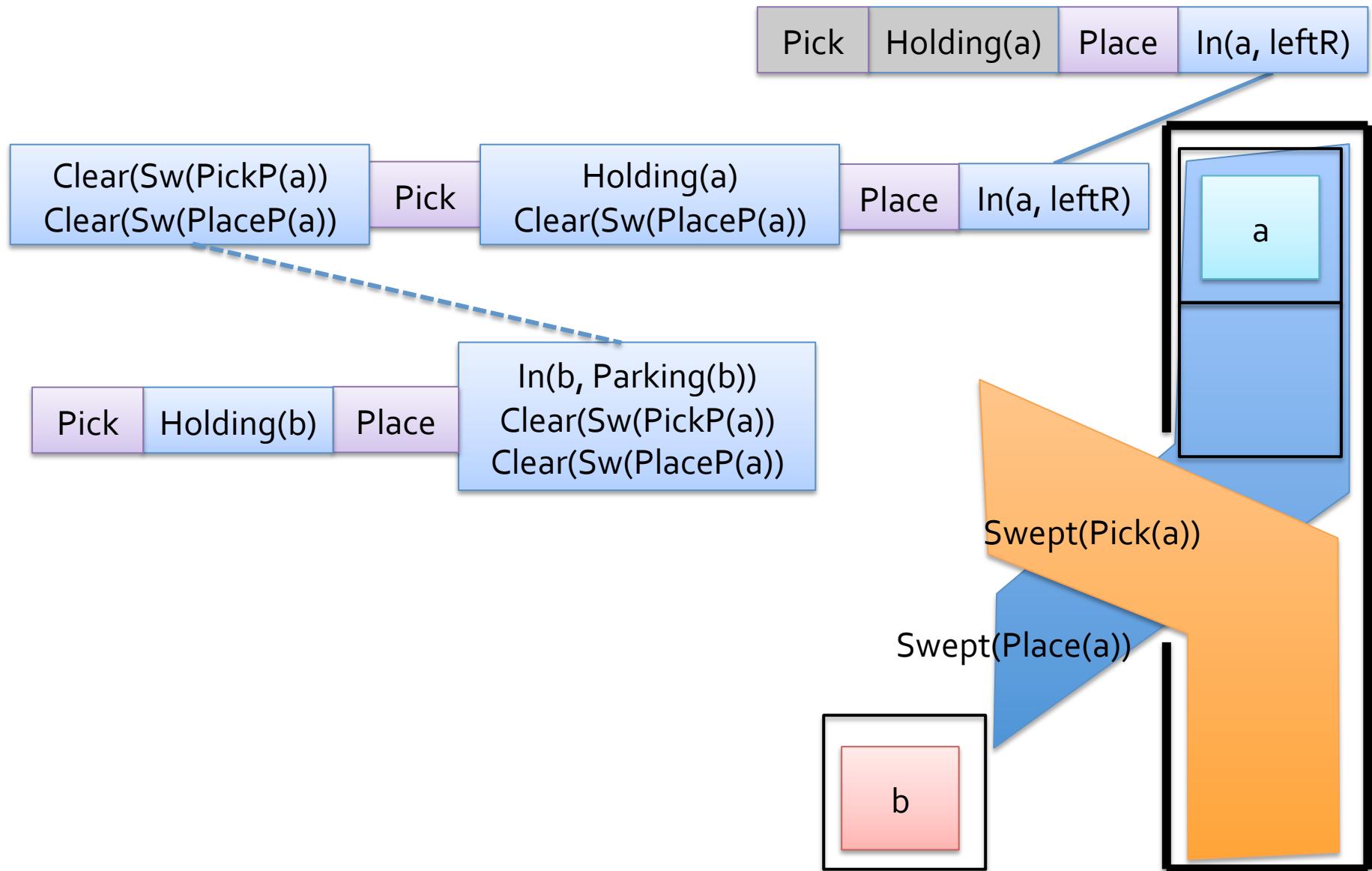
Decrease abstraction level



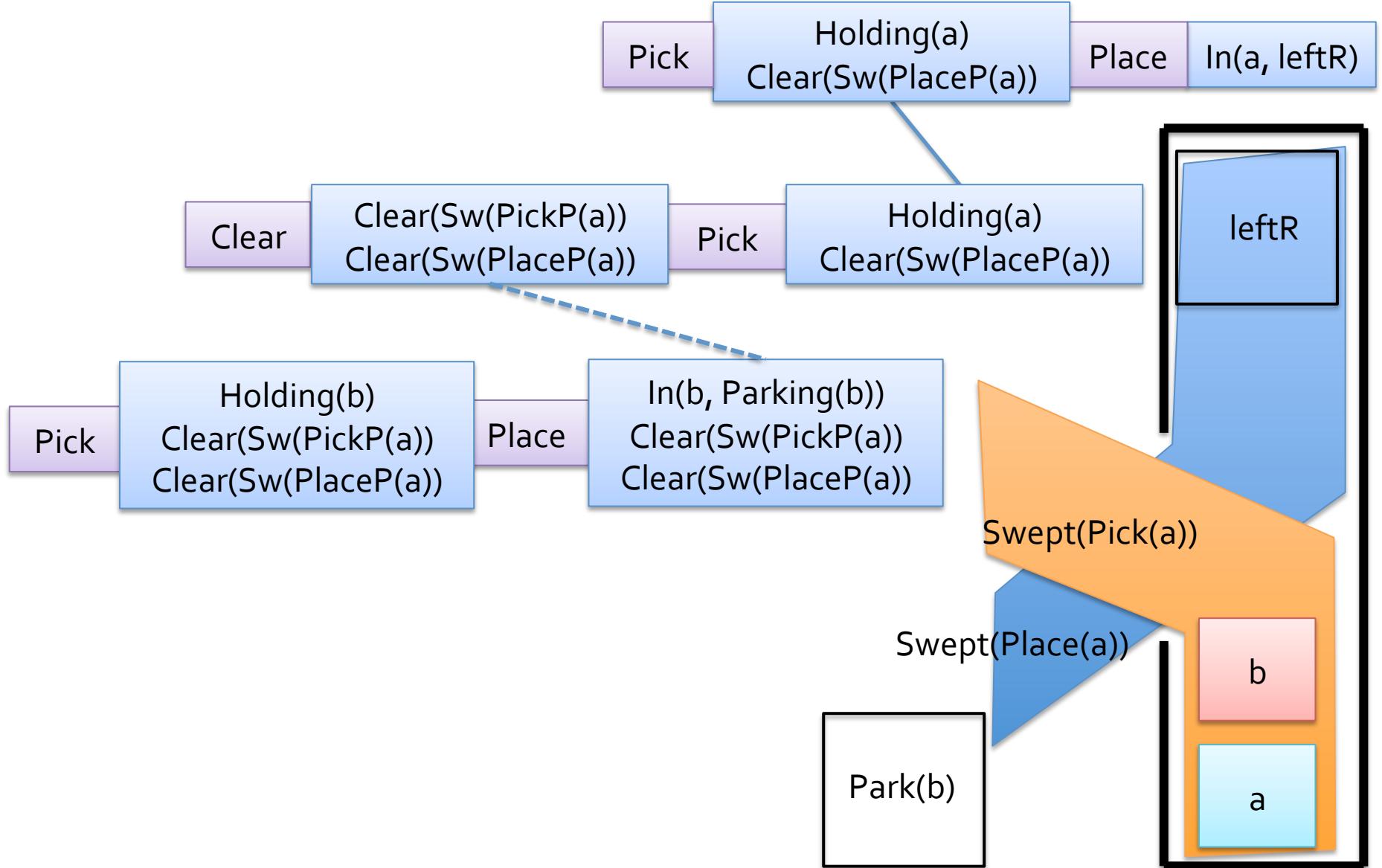
Decrease abstraction level:



Decrease abstraction level:



Hierarchical pick and place: consider clearX



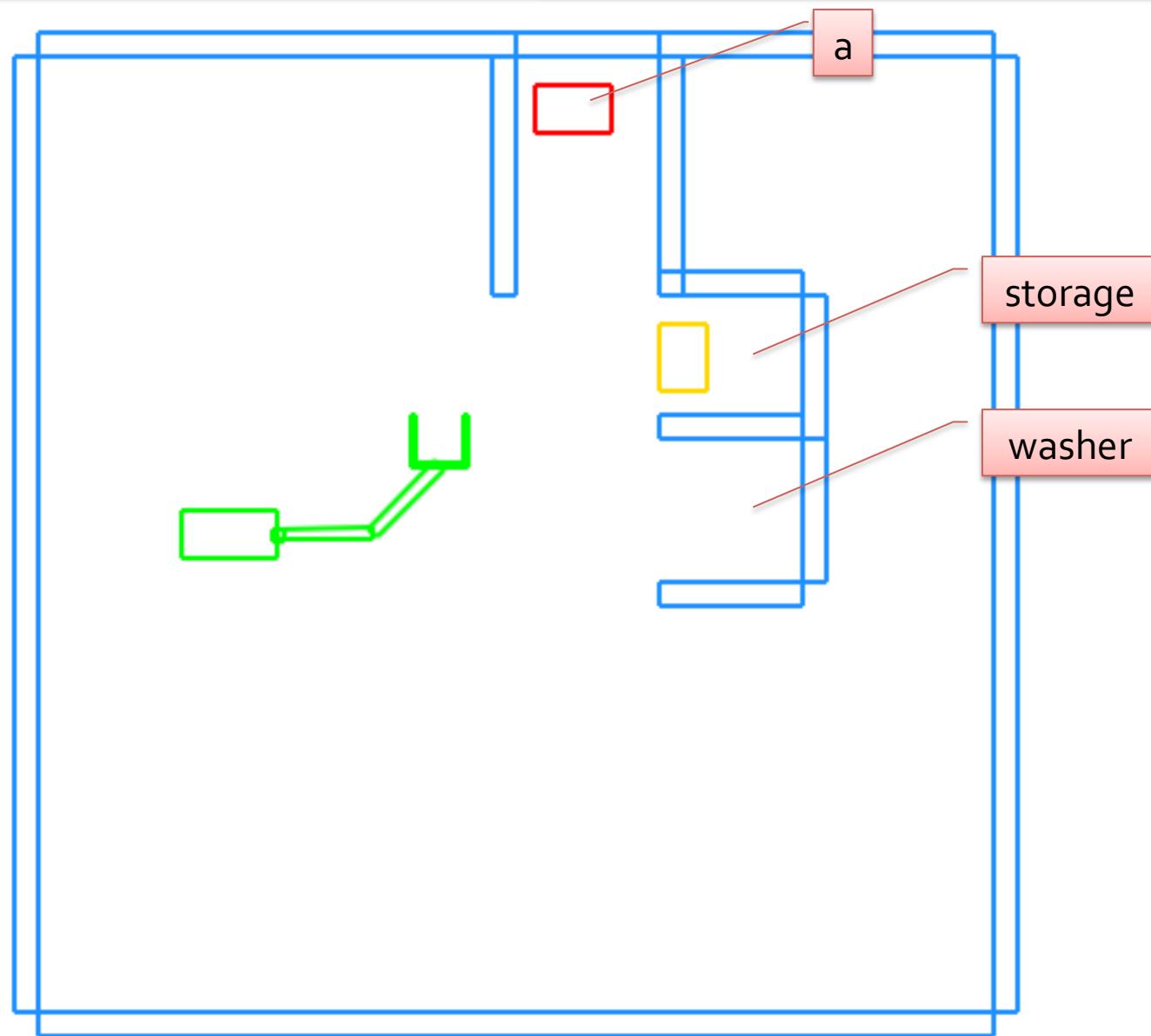
Correctness

If

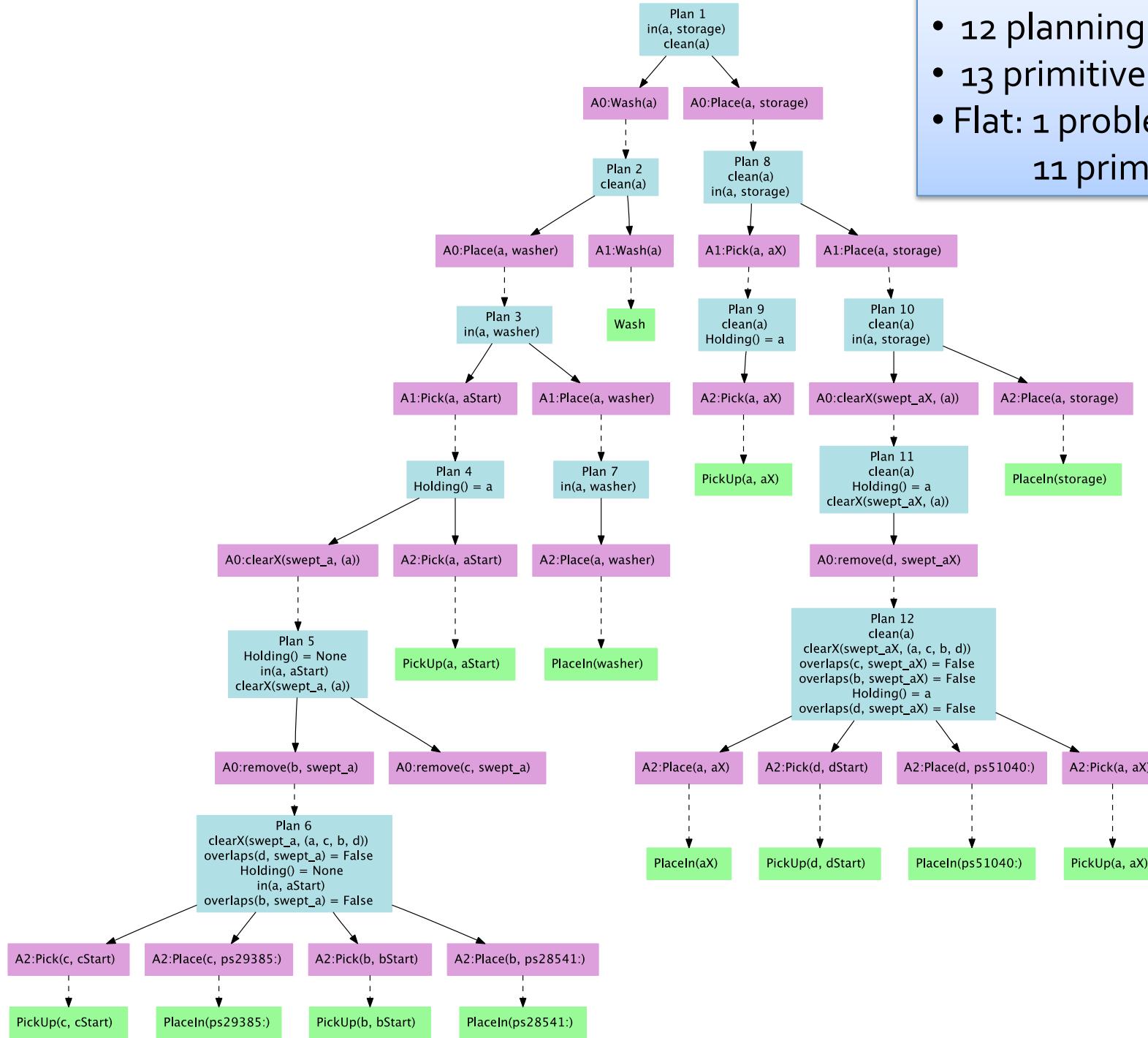
- the domain specified by operators *ops* at the most concrete abstraction level is a complete and correct formalization of the primitive actions of domain *w*
- any state reachable from *start* is reachable from any other state reachable from *start*, and
- some state in *G* is reachable from *start*,

then executing $HPN(start, G, ops, H, w)$ will cause domain *w* to be in a state $s \in G$.

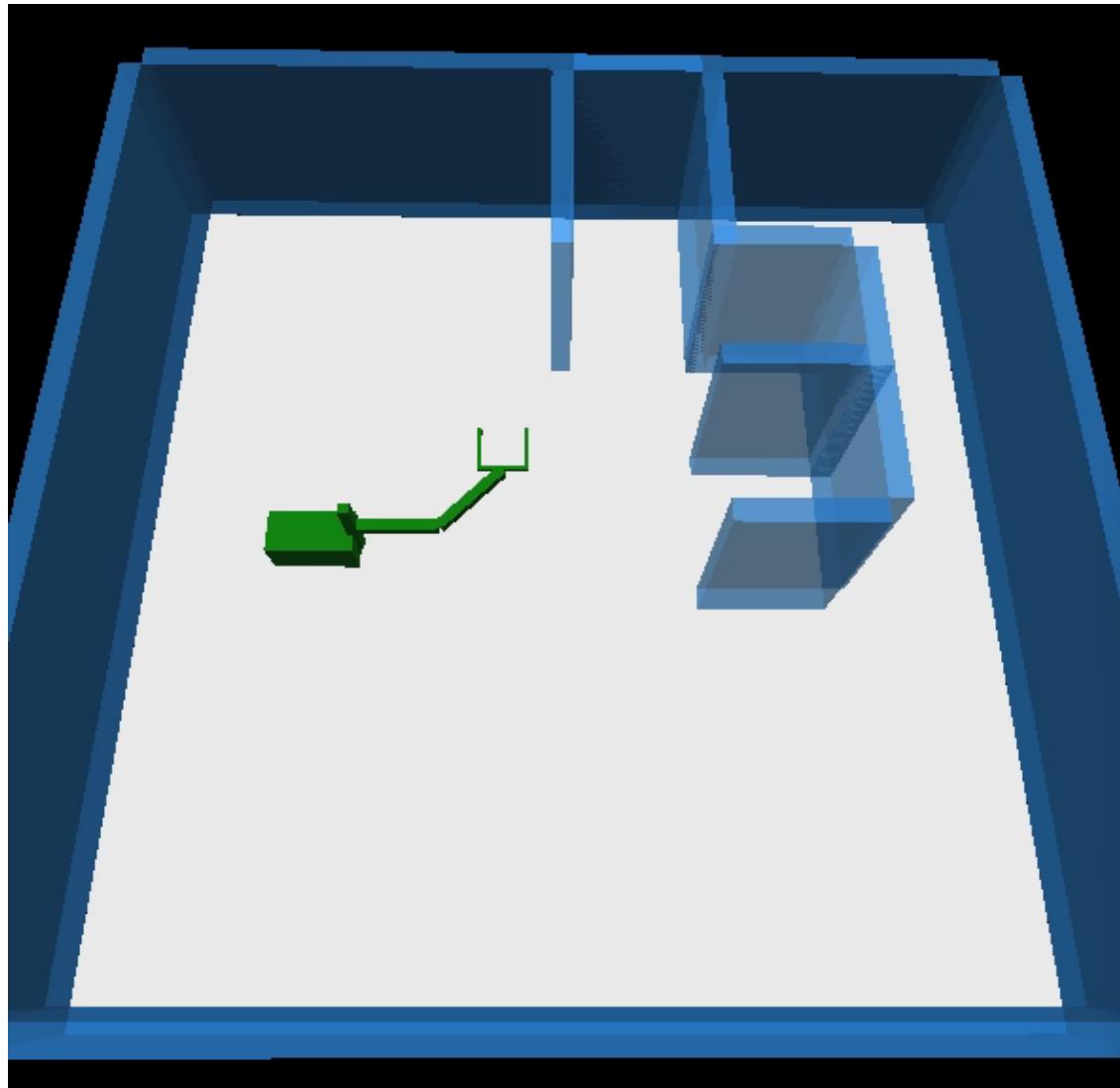
Wash a block and put it away



- 12 planning problems
- 13 primitive steps
- Flat: 1 problem, 11 primitive steps



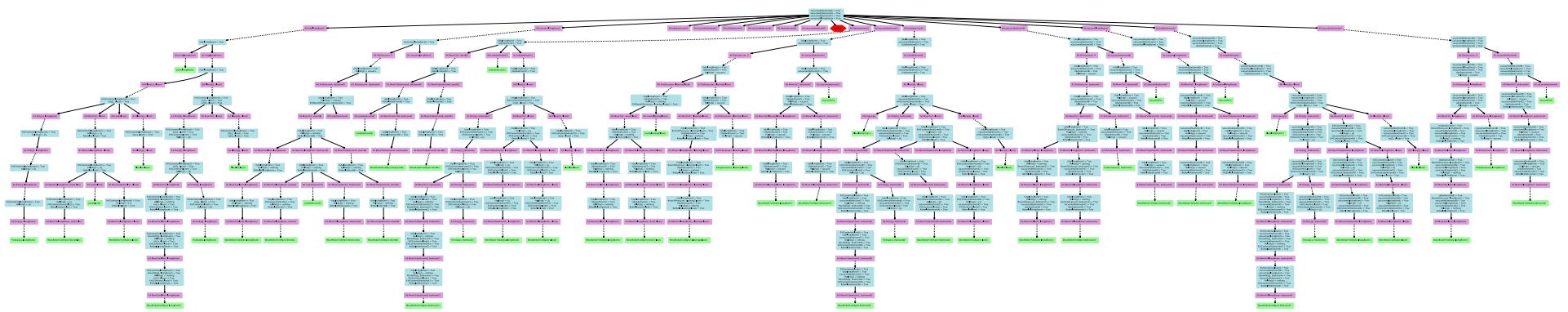
Wash a block and put it away



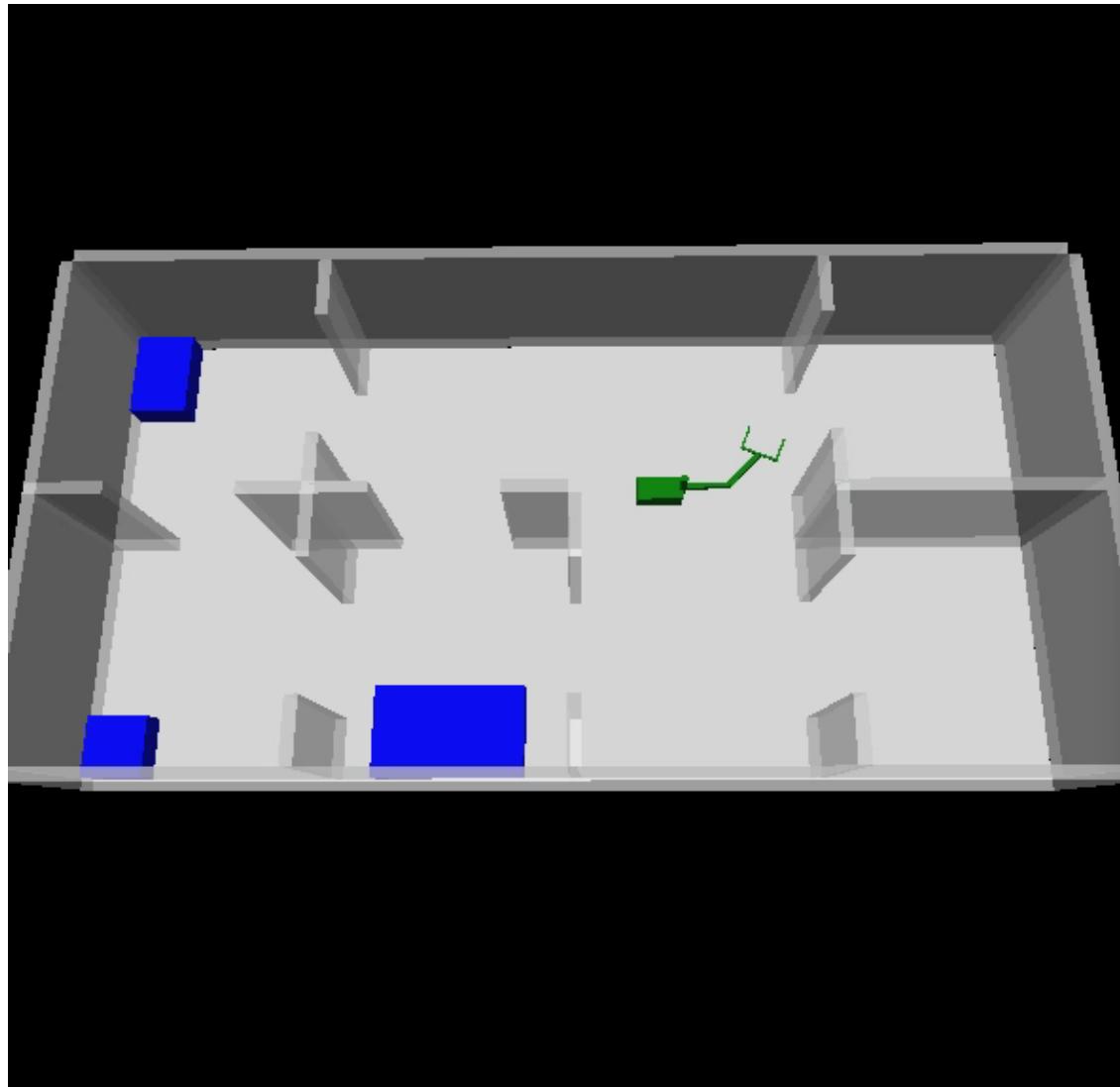
Cleaning house

Goal: mop four of the rooms in the house

- have to vacuum before mopping
- have to put away junk items before vacuuming



Cleaning house



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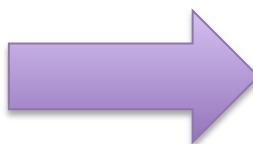
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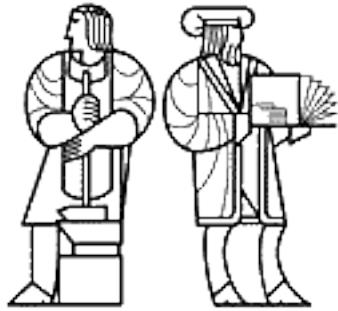
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Mens et Manus



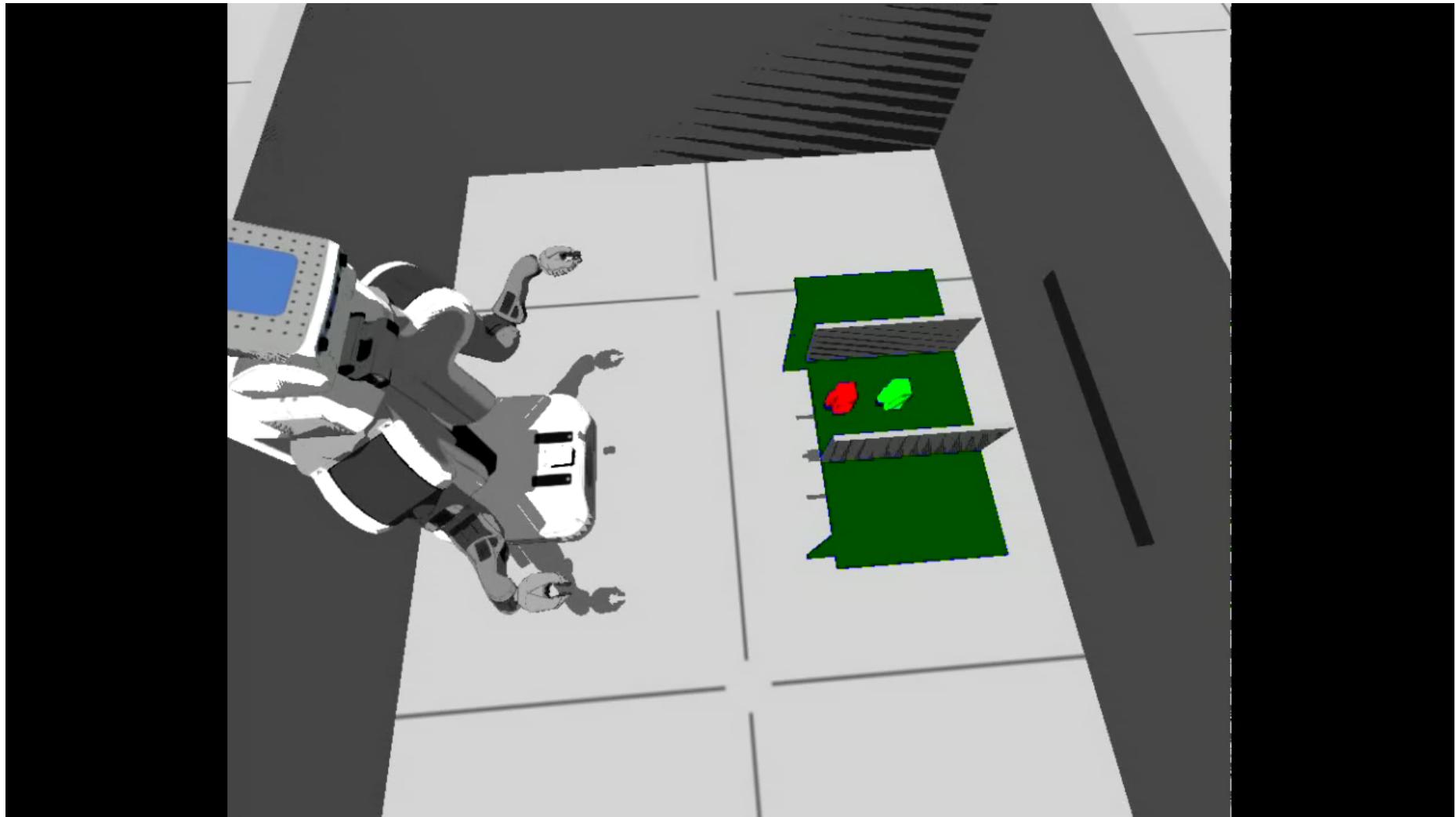
M and M



Eminem



M and M dreams of swapping two cups



M and M video

