

A Relational Approach to Tool Use Learning in Robots

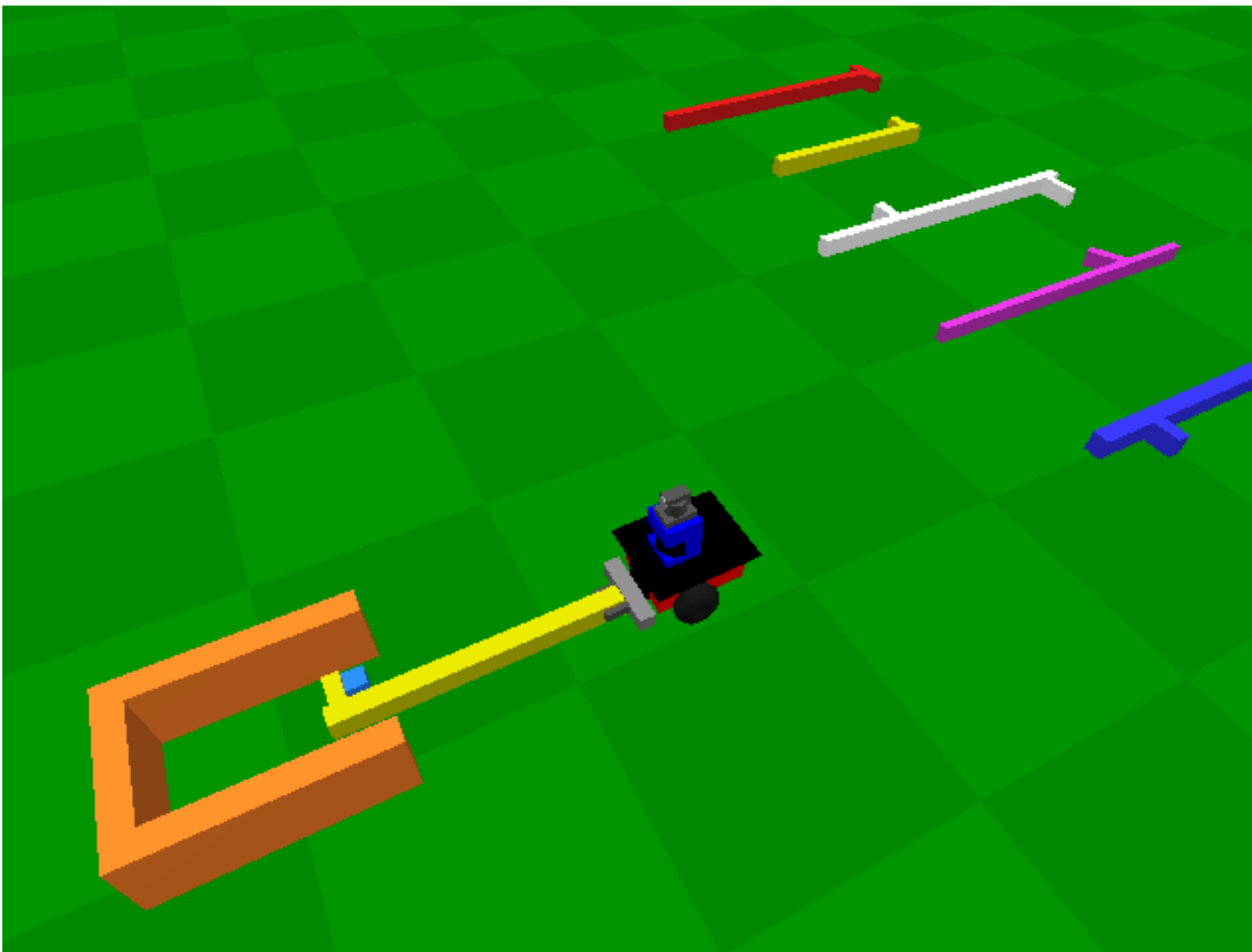
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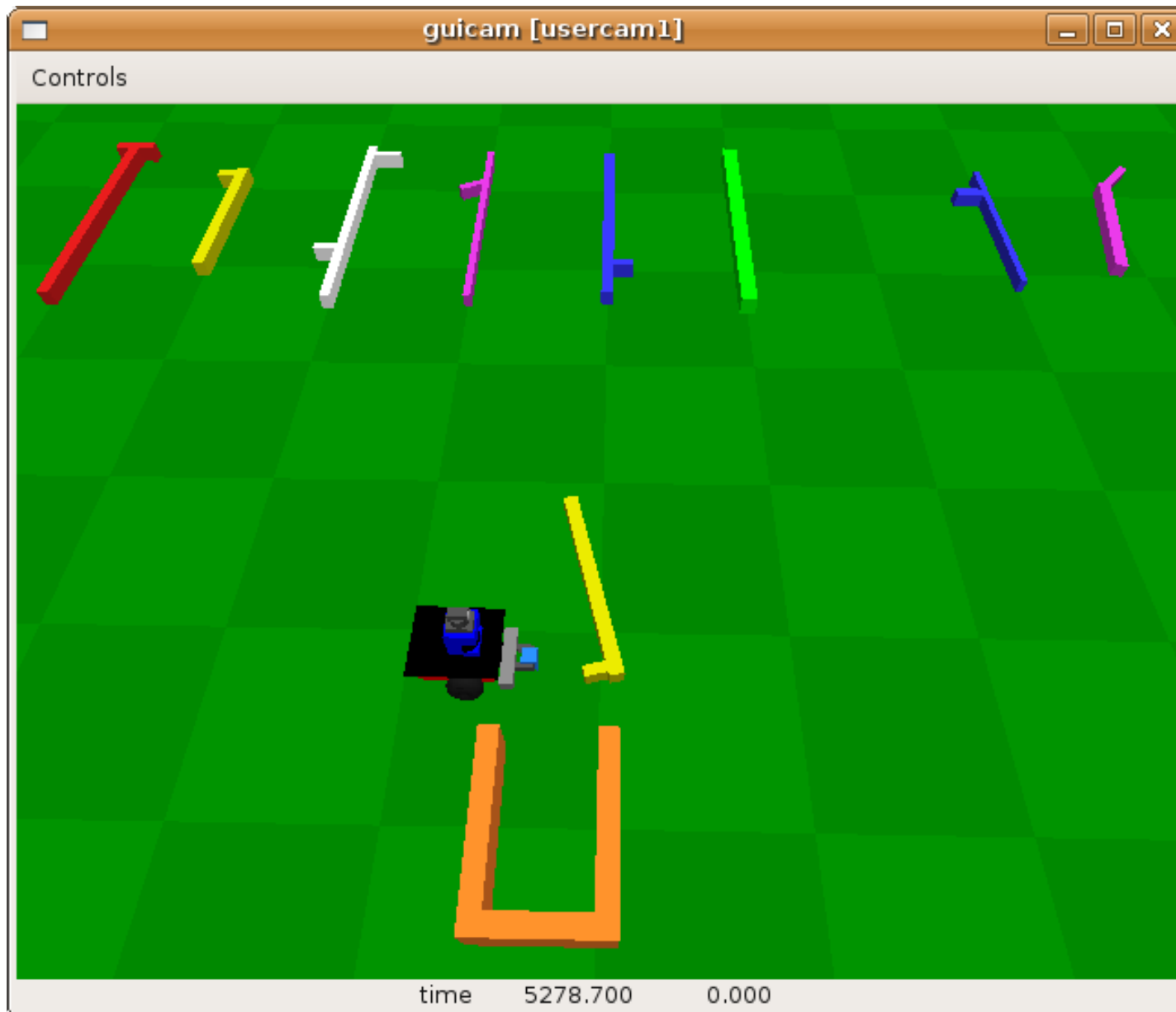
Learn to use an object as a tool by:

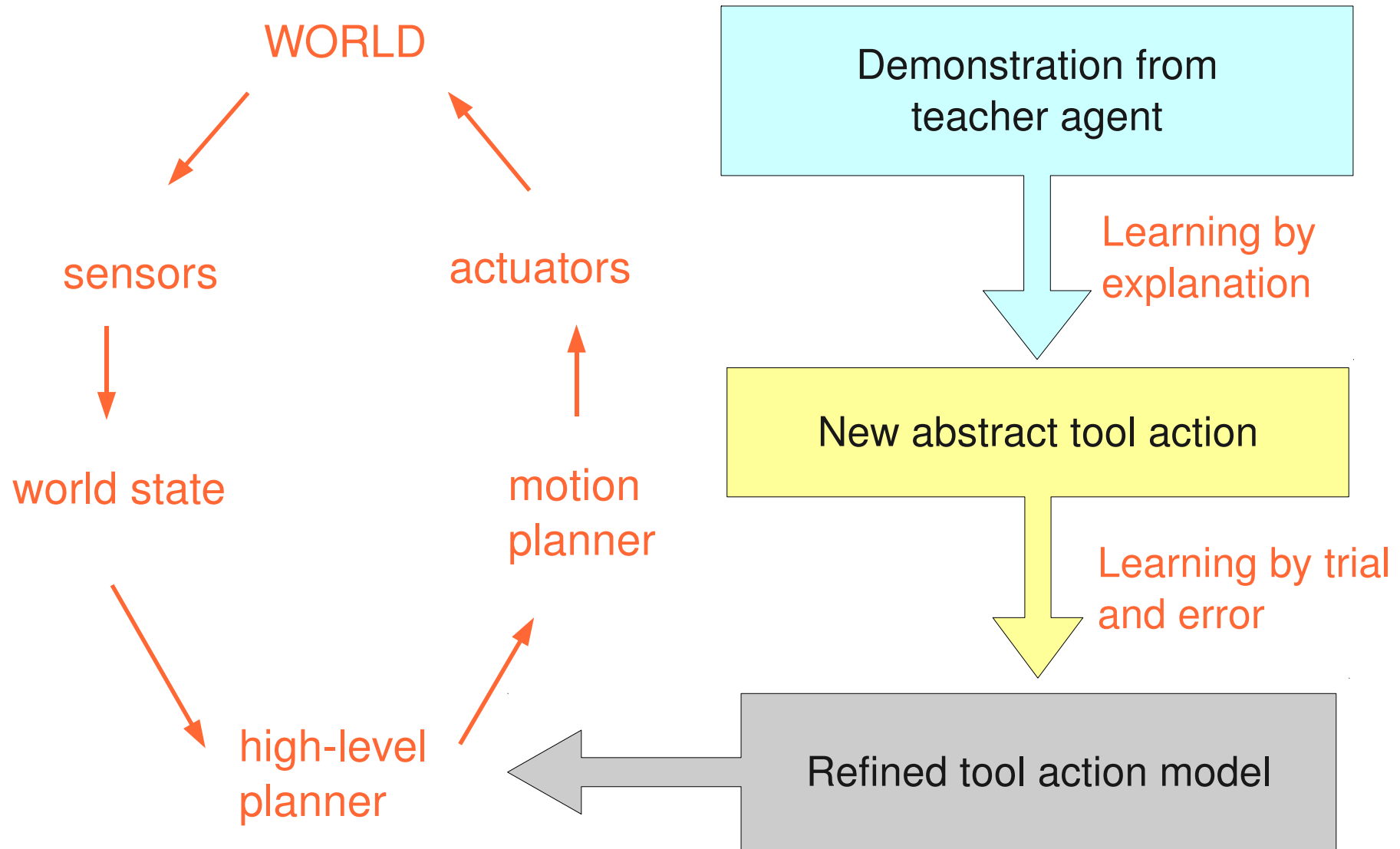
observing another agent use the tool to
achieve a goal

Trial-and-error to refine a theory about
how to use the tool



Trainer's Demonstration





Assumptions

All objects are graspable

i.e. no control problems

The robot has background knowledge of some actions

Unknown action sequence consists of

tool positioning actions

application of tool resulting in goal state

Action Models

ACTION `put_under(Cup, Tap)`

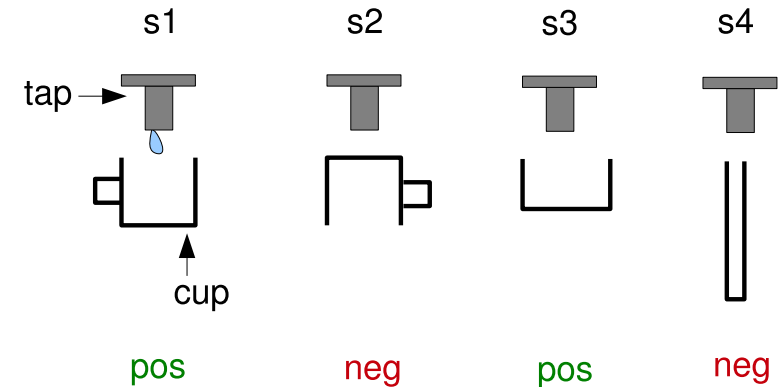
PRE `in_gripper(Cup),`
 `gripping(Cup),`
 `clear_underneath(Tap),`
 `orientation(Cup, vertical-up)`

ADD `below(Cup, Tap),`
 `near(Cup, Tap),`
 `aligned_vertically(Cup, Tap)`

DEL `clear_underneath(Tap)`

MOVING `robot-arm, Cup` `// manipulated objects`

PRIMITIVES `fwd, back, left, right, up, down, rotatecw, rotateccw`



Background Knowledge

grip(Obj)

PRE ¬gripping,
 in_gripper(Obj)
ADD gripping
DEL -
PRIMTV closeGrip
MOVING -

ungrip(Obj)

PRE gripping,
 in_gripper(Obj)
ADD -
DEL gripping
PRIMTV openGrip
MOVING -

remove_from_gripper(Obj)

PRE in_gripper(Obj),
 ¬gripping
ADD empty gripper
DEL in_gripper(Obj)
PRIMTV back
MOVING robot

recognise_goto

PRE empty_gripper
MOVING robot

put_in_gripper(Obj)

PRE forall(Tube:tube, ¬in(Obj,Tube)),
 empty_gripper,
 ¬gripping,
 forall(Obstacle:obj, ¬obstructing(Obstacle,Obj))
ADD in_gripper(Obj)
DEL empty_gripper
PRIMTV fwd, back, rotleft, rotright
MOVING robot

move_obstacle(ObjA,ObjB)

PRE moveable obj(ObjA),
 obstructing(ObjA,ObjB),
 in_gripper(ObjA),
 gripping
ADD -
DEL obstructing(ObjA,ObjB)
PRIMTV fwd, back, rotleft, rotright
MOVING robot, ObjA

recognise_carry_obj(Obj)

PRE in_gripper(Obj),
 gripping
MOVING robot, Obj

Dynamic predicates:

in_gripper(+obj,+state)
touching(+obj,-obj,-side,+state)
at_right_angles(+obj,+obj,+state)
at_oblique_angle(+obj,+obj,+state)
parallel(+obj,+obj,+state)
on_axis(+obj,+obj,+state)
on_perp_axis(+obj,+obj,+state)
in_tube(+obj,-tube,+state)
in_tube_end(+obj,-tube,-end,+state)
in_tube_side(+obj,-tube,-side,+state)
obstructing(+obj,+obj,+state)

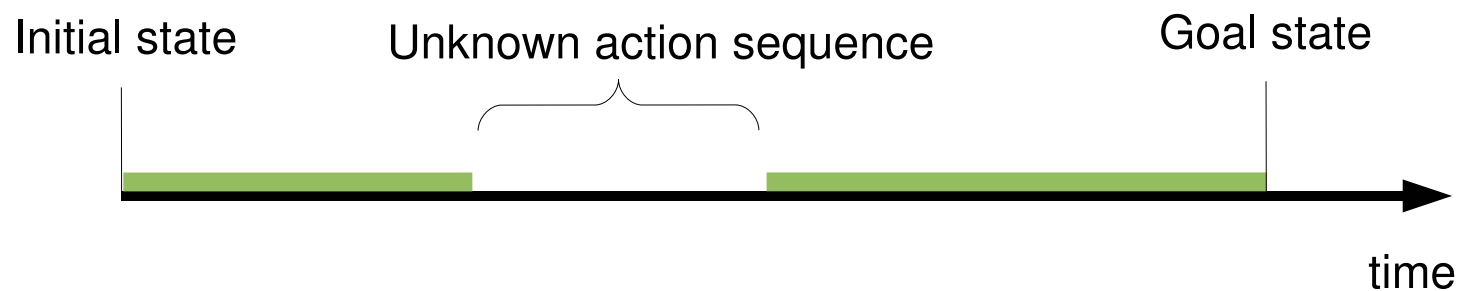
Static predicates:

attached_side(+obj,-obj,-side)
attached_end(+obj,-obj,-disttype)
attached_angle(+obj,-obj,-angletype)
num_attachments(+obj,-number)
longest_component(+obj)
attached_type(+obj,-obj,-attachtype)
narrower(+obj,+obj)
shorter(+obj,+obj)
shape(+obj,-shape)
closed_tube(-tube, -obj, -obj, -obj)

The world is continuous

First step is to discretise time-series observations of trainer

Convert to action sequence



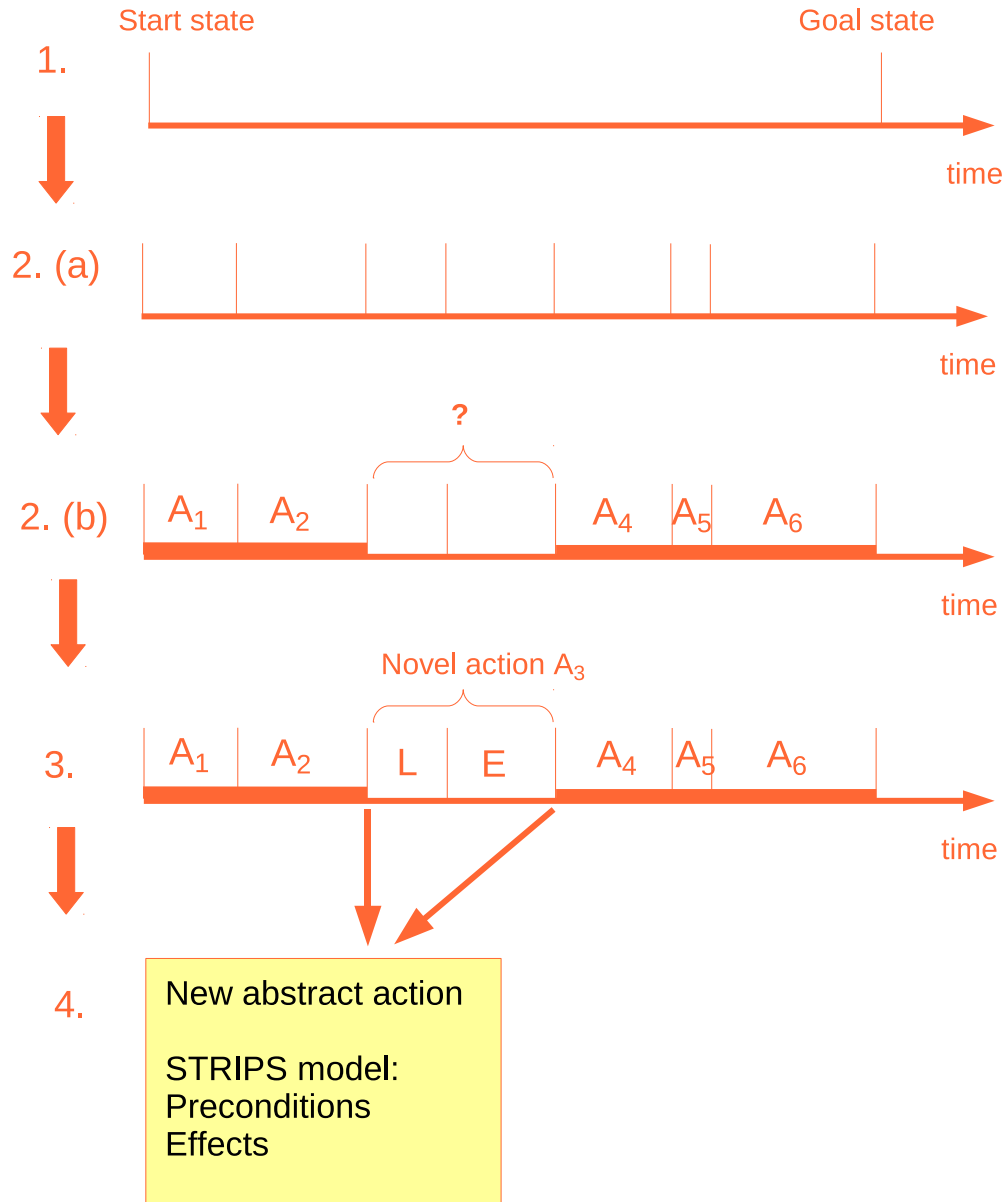
Object Motion

A distinct action begins or ends each time an object or agent starts or stops moving

Object Contact

An action starts or stops when two objects come into contact or break contact

Segmenting Observations



Segment on changes

Identify known action

Extract novel actions

Matching Actions

A segment is matched with an abstract action if:

Objects manipulated in the segment can be matched to MOVING list in action model

Preconditions are true at beginning of segment

Effects are true at end of segment

If more than one action matches:

choose action with most specific preconditions

Segment	Moving objects	Teacher's action (unknown)
1	robot	Put tool in gripper
2	gripper	Close gripper
3	robot, tool	Put tool in pulling pose
4	robot, tool, box	Pull box with tool
5	robot, tool	Put tool aside
6	gripper	Open gripper
7	robot	Put box in gripper
8	gripper	Close gripper
9	robot, box	Carry away box

Explanation of Observed Actions

Segment	Moving objects	Explanation	Match Type
1	robot	put in gripper(hookstick)	exact
2	gripper	grip(hookstick)	exact
3	robot, tool	recognise_carry_obj(hookstick)	partial
4	robot, tool, box	??	none
5	robot, tool	move obstacle(hookstick,box)	exact
6	gripper	ungrip(hookstick)	exact
7	robot	remove_from_gripper(hookstick), put in gripper(box)	exact
8	gripper	grip(box)	exact
9	robot, box	recognise_carry_obj(box)	partial

position_tool(Tool, Box)

PRE:

in_gripper(Tool),
gripping

ADD:

tool_pose(Tool, Box),
obstructing(Tool, Box)

DEL: -

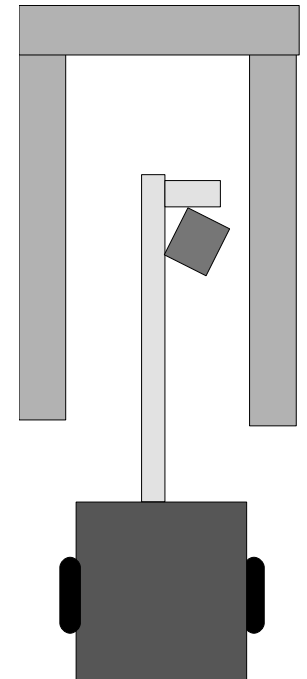
pull_from_tube(Tool, Box, Tube)

PRE:

tool_pose(Tool, Box),
in_gripper(Tool),
gripping,
in_tube(Box, Tube)

ADD: -

DEL: in_tube(Box, Tube)



Initial version space boundaries:

```
tool_poseG(Tool, Box, State) :- true.
```

```
tool_poseS(Tool, Box, State) :- saturation(s1).
```

Most specific hypothesis is constructed by saturating trainer's example

Bottom Clause

tool poseS(Tool, Box, State):-

% static properties

attached(Tool, Hook),
 num attachments(Tool, 1),
 num attachments(Box, 0),
 longest component(Tool),
 narrower(Tool, Box),
 shorter(Box, Tool),
 shape(Tool, sticklike),
 shape(Box, boxlike),
 closed tube(Tool, TubeLeft, TubeRight, TubeBack),
 attached side(Tool, Hook, right),
 attached side(TubeLeft, TubeBack, right),
 attached side(TubeRight, TubeBack, left),
 attached end(Tool, Hook, front),
 attached end(TubeLeft, TubeBack, front),
 attached end(TubeRight, TubeBack, front),
 attached angle(Tool, Hook, right angle),
 attached angle(TubeLeft, TubeBack, right angle),
 attached angle(TubeRight, TubeBack, right angle),
 attached type(Tool, Hook, end to end),
 attached type(TubeLeft, TubeBack, end to end),
 attached type(TubeRight, TubeBack, end to end),
 num attachments(Hook, 0),
 num attachments(TubeLeft, 1),
 num attachments(TubeRight, 1),
 num attachments(TubeBack, 0),
 narrower(Hook, Tool),
 narrower(Hook, Box),

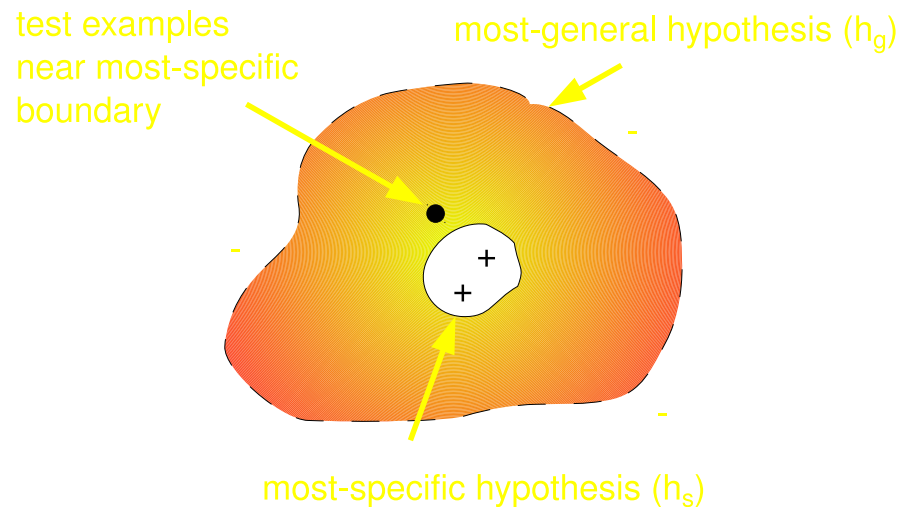
% fluents

in gripper(Tool, State),
 touching(Tool, Box, right, State),
 at oblique angle(Tool, Box, State),
 in tube(Box, Tube, State),
 in tube end(Box, Tube, front, State),
 in tube side(Box, Tube, right, State),
 touching(Hook, Box, back, State),
 at right angles(Hook, TubeLeft, State),
 at right angles(Hook, TubeRight, State),
 at right angles(Tool, TubeBack, State),

narrower(Tool, TubeLeft),
 narrower(TubeLeft, Box),
 narrower(Hook, TubeLeft),
 narrower(Tool, TubeRight),
 narrower(TubeRight, Box),
 narrower(Hook, TubeRight),
 narrower(Tool, TubeBack),
 narrower(TubeBack, Box),
 narrower(Hook, TubeBack),
 narrower(TubeBack, TubeLeft),
 narrower(TubeBack, TubeRight),
 shorter(Hook, Tool),
 shorter(Tool, TubeLeft),
 shorter(Tool, TubeRight),
 shorter(TubeBack, Tool),
 shorter(Box, Hook),
 shorter(Box, TubeLeft),
 shorter(Hook, TubeLeft),
 shorter(TubeBack, TubeLeft),
 shorter(Box, TubeRight),
 shorter(Hook, TubeRight),
 shorter(TubeBack, TubeRight),
 shorter(TubeBack, TubeLeft),
 shorter(Box, TubeBack),
 shorter(Hook, TubeBack),
 shape(TubeLeft, sticklike),
 shape(TubeRight, sticklike),
 shape(TubeBack, sticklike),

at oblique angle(Box, Hook, State),
 at oblique angle(Box, TubeLeft, State),
 at oblique angle(Box, TubeRight, State),
 at oblique angle(Box, TubeBack, State),
 parallel(Tool, TubeLeft, State),
 parallel(Tool, TubeRight, State),
 parallel(Hook, TubeBack, State),
 in tube(Hook, Tube, State),
 in tube end(Hook, Tube, front, State),
 in tube side(Hook, Tube, right, State).

Generating an Experiment



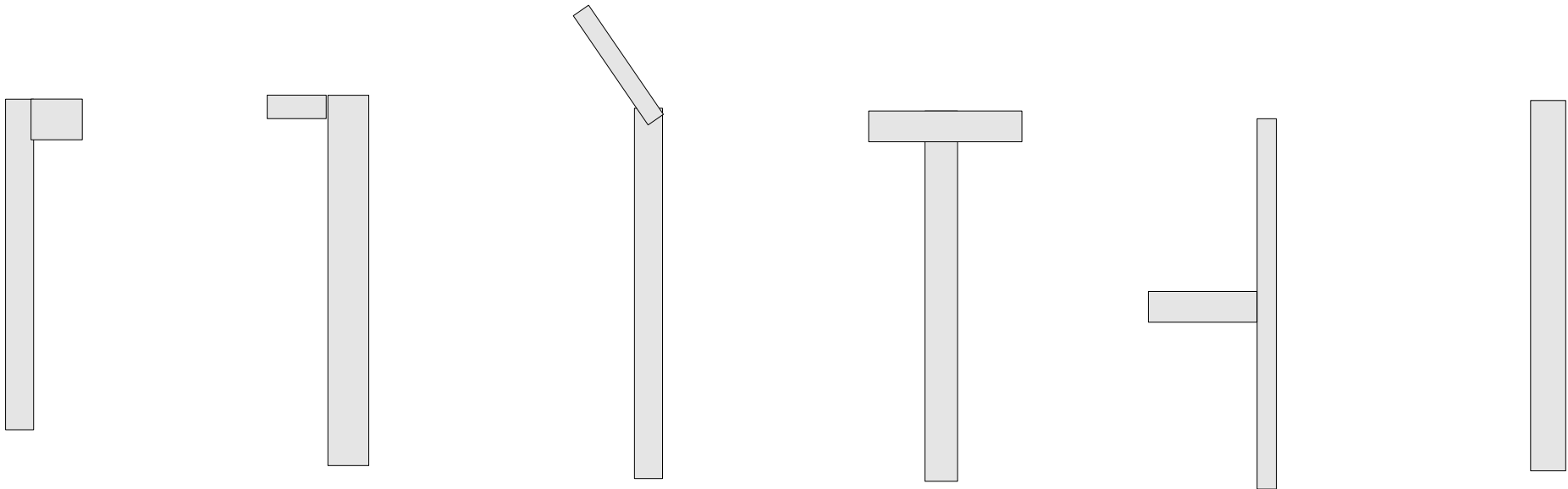
Start with H_G and add literals from H_S but ...

select tool that satisfies the most structural constraints in H_S

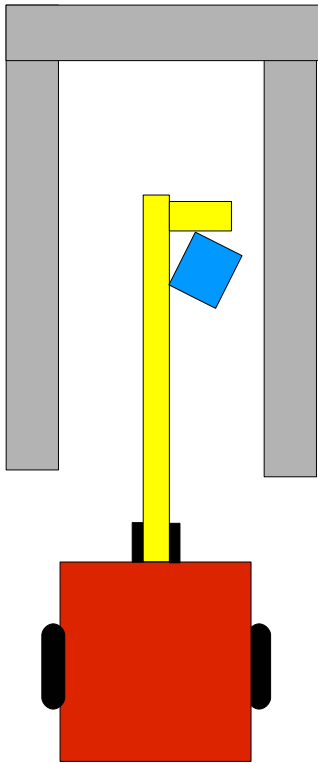
select pose to maximise the spatial literals satisfied in H_S

Bias specific-to-general search results by trying to generate experiments near most specific boundary

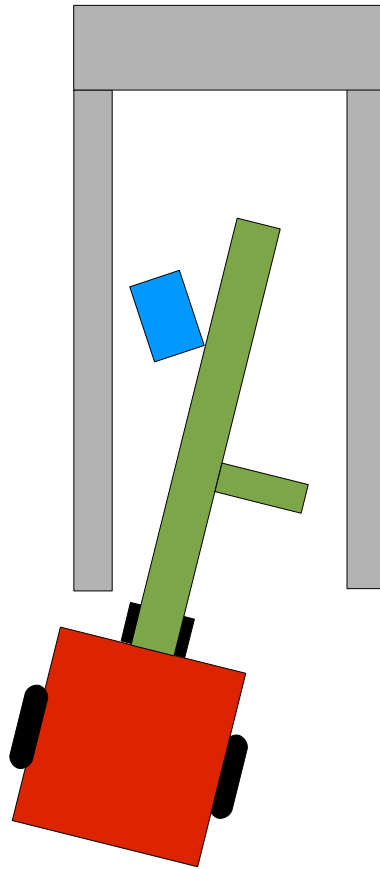
Tools Available for Experimentation



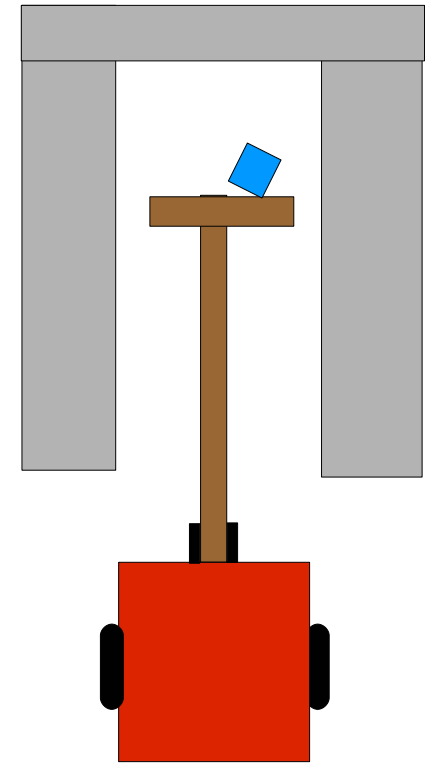
Positive and Negative Examples of Tool Use



pos



neg



neg

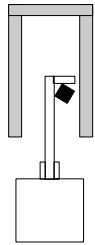
An uncovered positive example generalises H_S

Like GOLEM, find RLGG of old hypothesis and new example

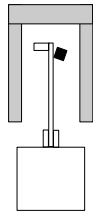
A covered negative example specialises H_G

Use negative-based reduction to recreate H_G

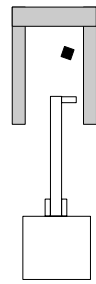
Find most general H_G by removing literals from H_S as long as negative examples are not covered



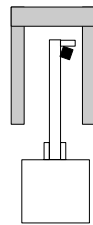
pos: s1



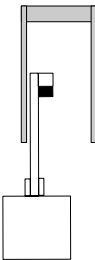
neg: s2



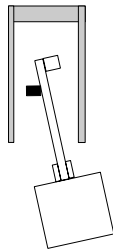
neg: s3



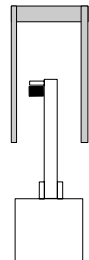
pos: s4



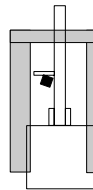
neg: s5



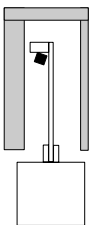
neg: s6



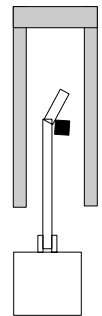
pos: s7



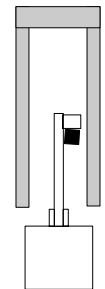
neg: s8



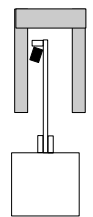
pos: s9



neg: s10



pos: s11



pos: s12

Evolution of most general hypothesis:

`tool_poseG(Tool, Box, State) :-
attached side(Tool, Hook, right).`

...

`tool_poseG(Tool, Box, State) :-
attached side(Tool, Hook, right),
touching(Hook, Box, back, State).`

...

`tool_poseG(Tool, Box, State) :-
attached(Tool, Hook),
narrower(Hook, Box),
touching(Hook, Box, back, State).`

...

`tool_poseG(Tool, Box, State) :-
in_tube_side(Box, Tube, Side, State),
attached_side(Tool, Hook, Side),
touching(Hook, Box, back, State),
attached_angle(Tool, Hook, rightangle),
attached_end(Tool, Hook, back).`

To run experiment:

learned actions must be incorporated into a plan

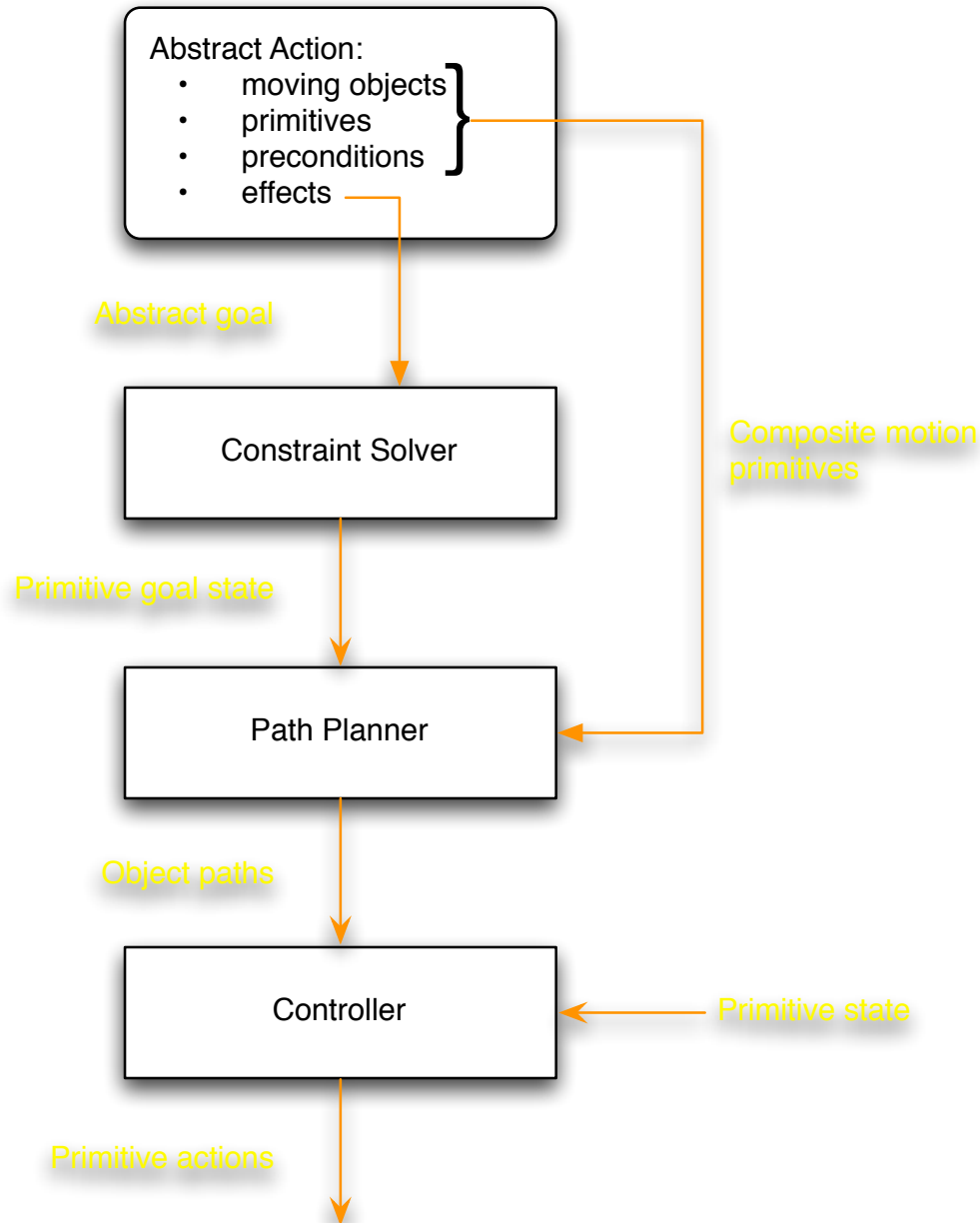
sequence of abstract actions generated by planner must be turned into action primitives

i.e. real numbers needed!

Fast Forward
(Hoffman & Nebel)

ECLiPSe
(Apt & Wallace)

Rapid Random Trees
(Kuffer & LaValle)



Extract the spatial sub-goals from action model

Constraint solver finds a primitive world state satisfying spatial subgoals

Moveable objects in MOVING list are treated as a single geometric object to be manipulated by the motion planner. Allowed movement primitives are given by PRIMITIVES list.

Motion planner uses composite object motion primitives to generate a path from current world state to ground goal state generated in step 2

Track path using a generic controller

Follow until the primitive goal state is reached

Failures detected by timeout cause re-planning