

Angelic Hierarchical Planning: Optimal and Online Algorithms

Bhaskara Marthi

MIT/Willow Garage

bhaskara@csail.mit.edu

Stuart Russell

UC Berkeley

russell@cs.berkeley.edu

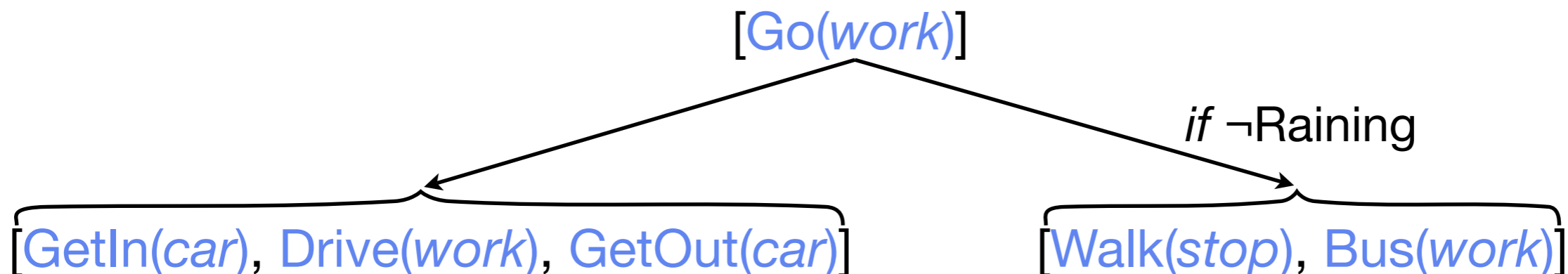
Jason Wolfe

UC Berkeley

jawolfe@cs.berkeley.edu

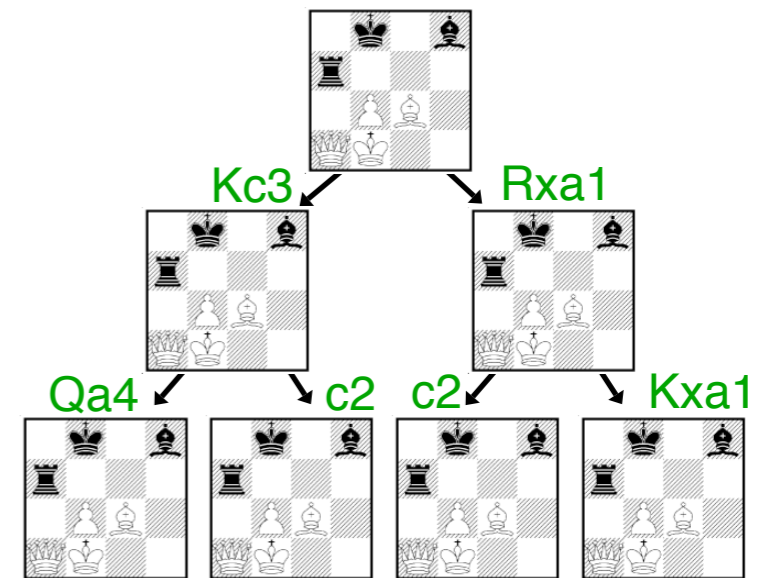
High-Level Actions (HLAs)

- Here, a **high-level action (HLA)** = a set of allowed **immediate refinements**:
 - each is a **sequence** of actions
 - may have associated preconditions
- Almost all actions we think about are high-level
 - Plan a trip
 - Vacuum the house
 - Go to work



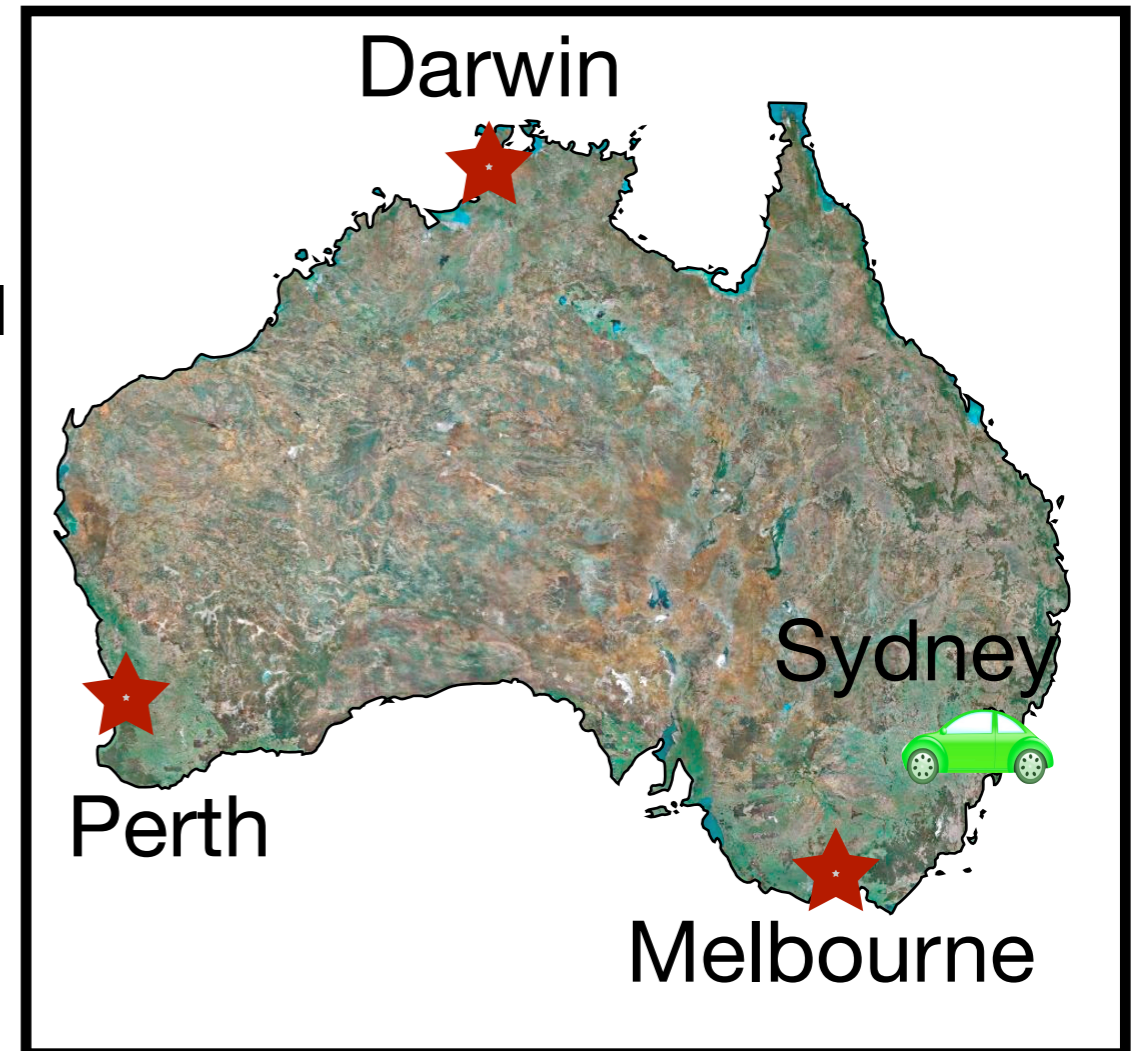
Abstract Lookahead

- k -step lookahead \gg 1-step lookahead
 - e.g., chess



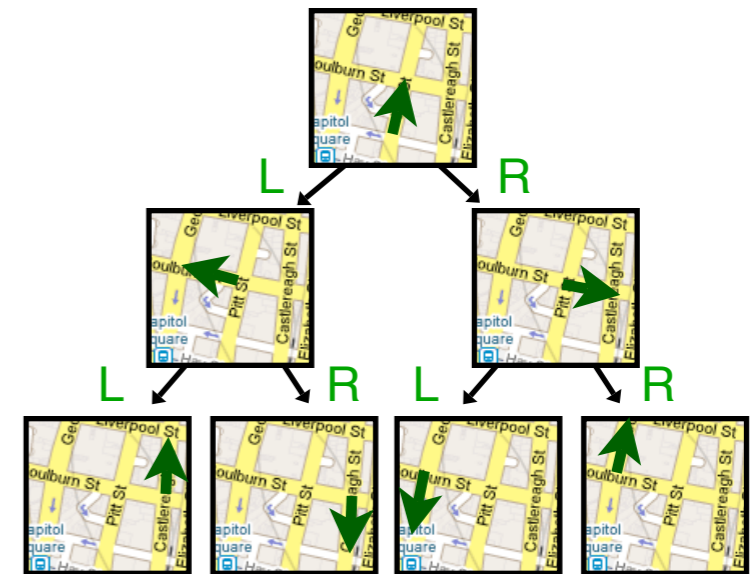
Abstract Lookahead

- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia



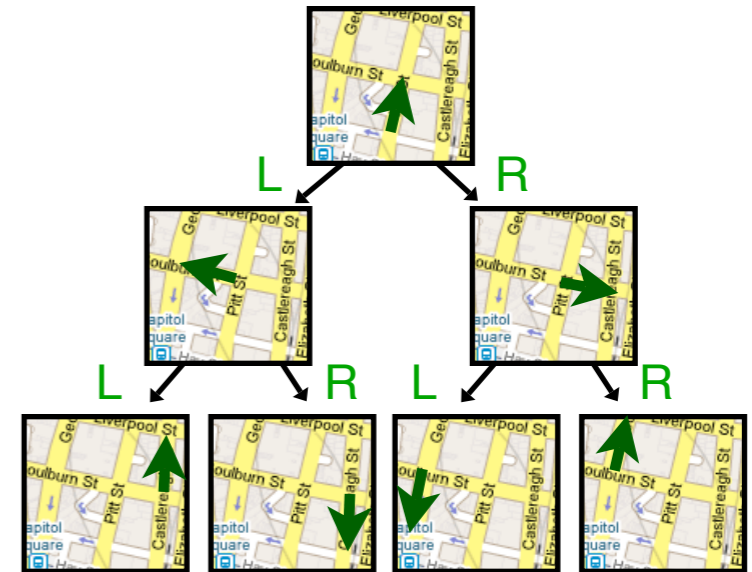
Abstract Lookahead

- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia



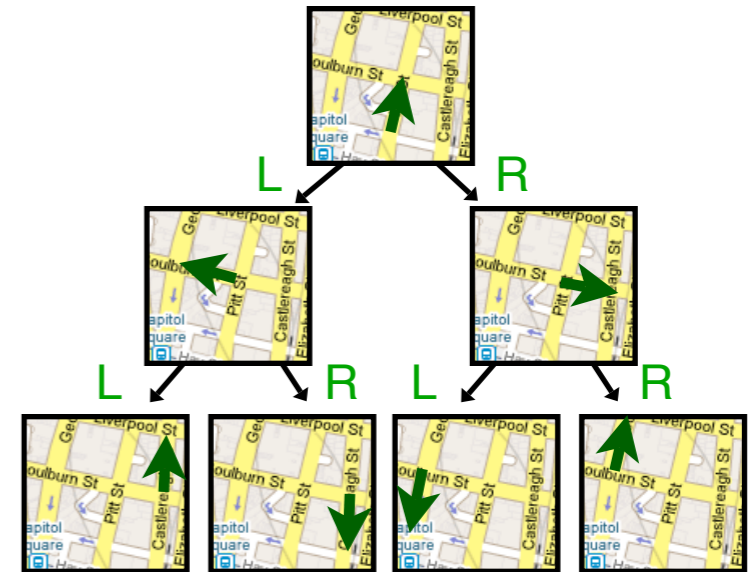
Abstract Lookahead

- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions



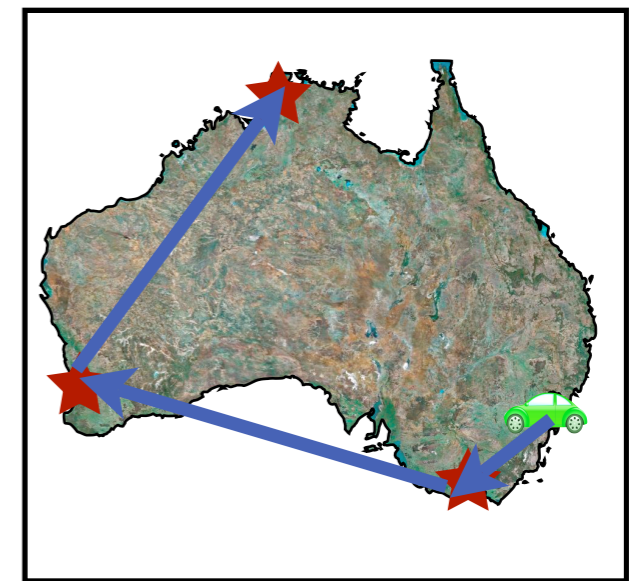
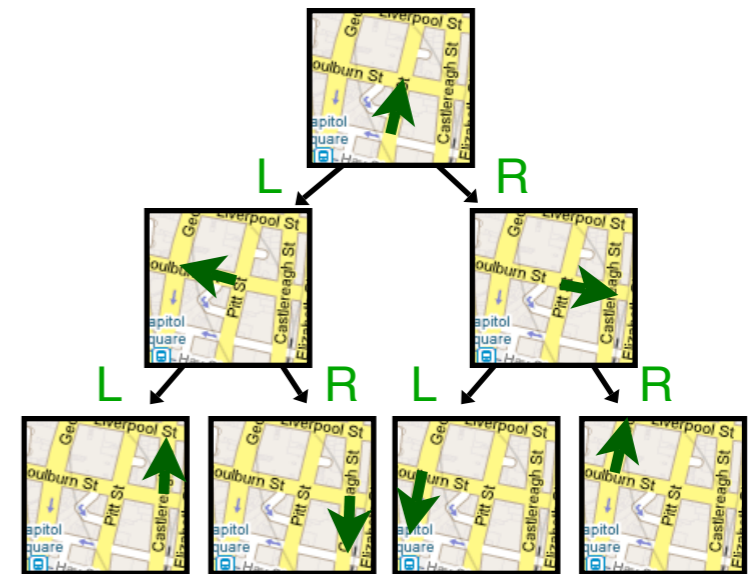
Abstract Lookahead

- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions
- **Abstract** plans with HLAs are shorter



Abstract Lookahead

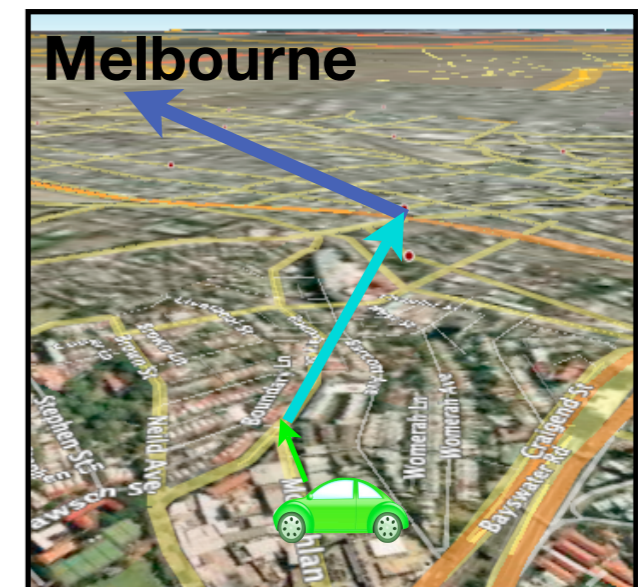
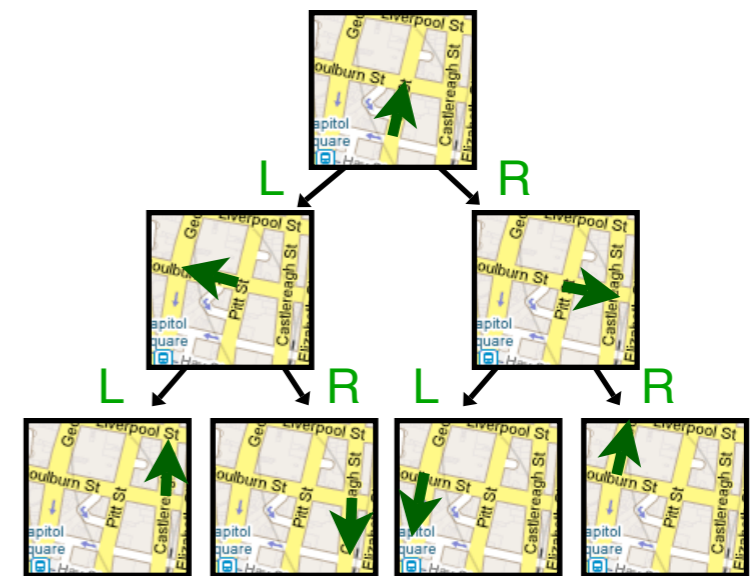
- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions
- **Abstract** plans with HLAs are shorter
 - Much shorter plans \Rightarrow exponential savings



is provably optimal

Abstract Lookahead

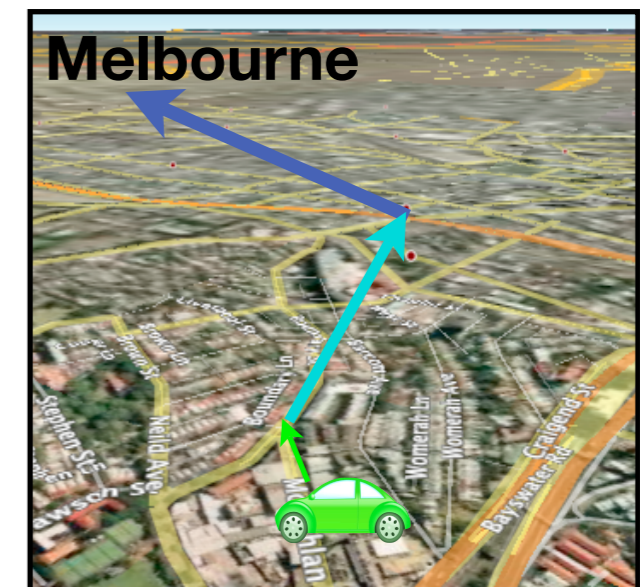
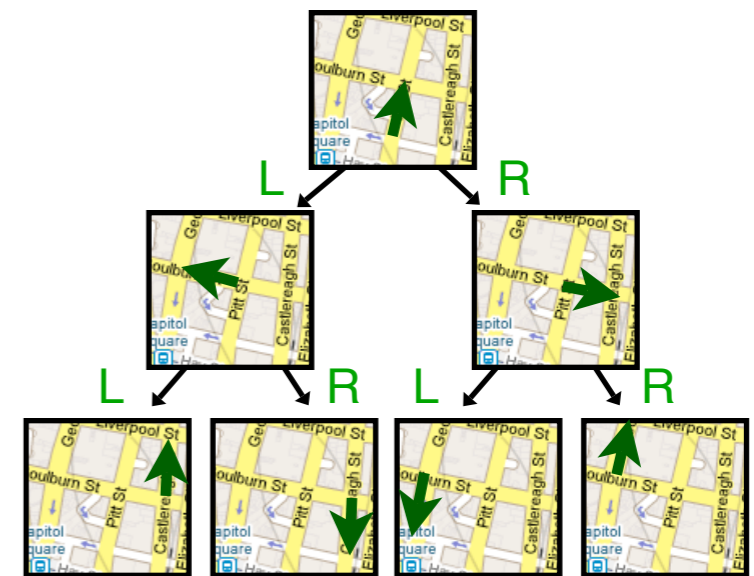
- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions
- **Abstract** plans with HLAs are shorter
 - Much shorter plans \Rightarrow exponential savings
 - Can look ahead much further



looks like a good start

Abstract Lookahead

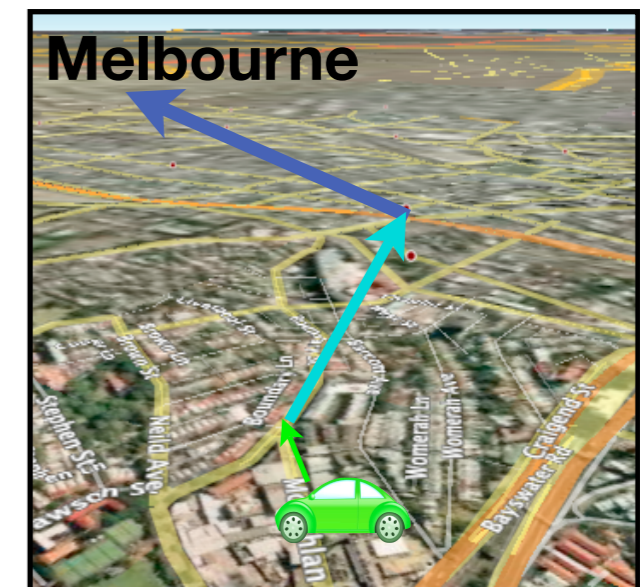
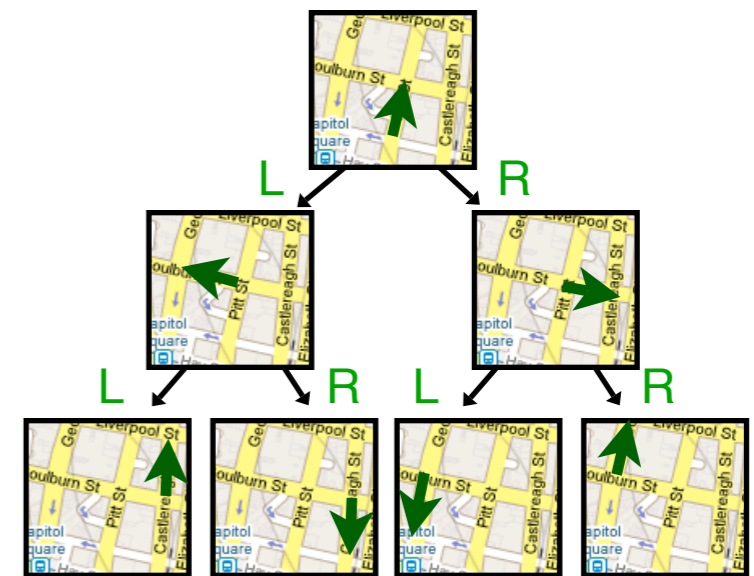
- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions
- **Abstract** plans with HLAs are shorter
 - Much shorter plans \Rightarrow exponential savings
 - Can look ahead much further
- Requires **models** for HLAs
 - i.e., transition and cost fns



looks like a good start

Abstract Lookahead

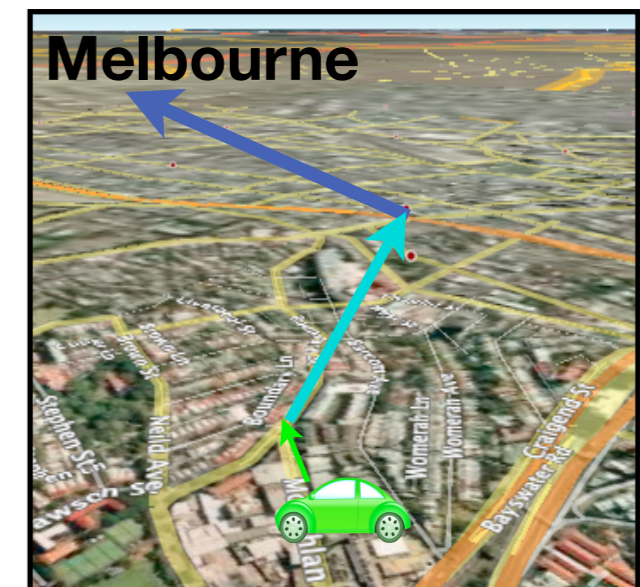
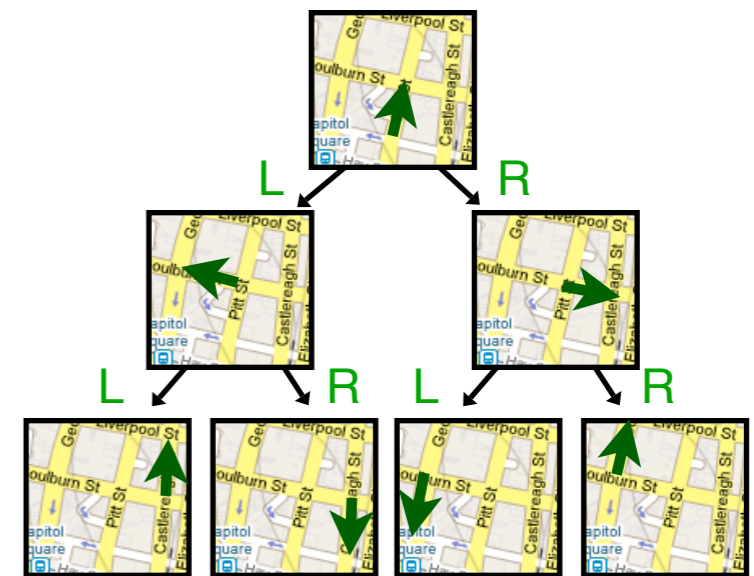
- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions
- **Abstract** plans with HLAs are shorter
 - Much shorter plans \Rightarrow exponential savings
 - Can look ahead much further
- Requires **models** for HLAs
 - i.e., transition and cost fns
 - No suitable models in literature



looks like a good start

Abstract Lookahead

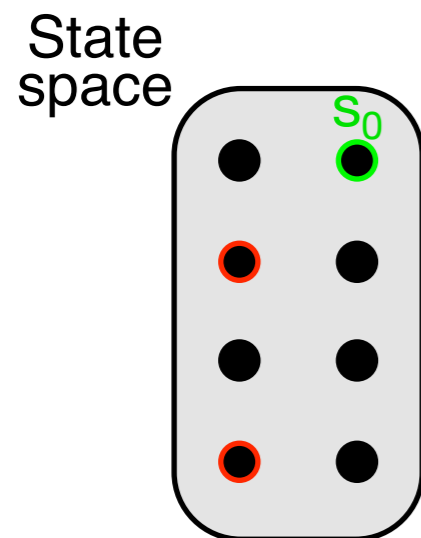
- k -step lookahead \gg 1-step lookahead
 - e.g., chess
- k -step lookahead no use if steps too small
 - e.g., first k turns in TSP of Australia
 - this is one small part of a human life,
 $\approx 20,000,000,000,000$ primitive actions
- **Abstract** plans with HLAs are shorter
 - Much shorter plans \Rightarrow exponential savings
 - Can look ahead much further
- Requires **models** for HLAs
 - i.e., transition and cost fns
 - No suitable models in literature
 - We extend our **angelic semantics**



looks like a good start

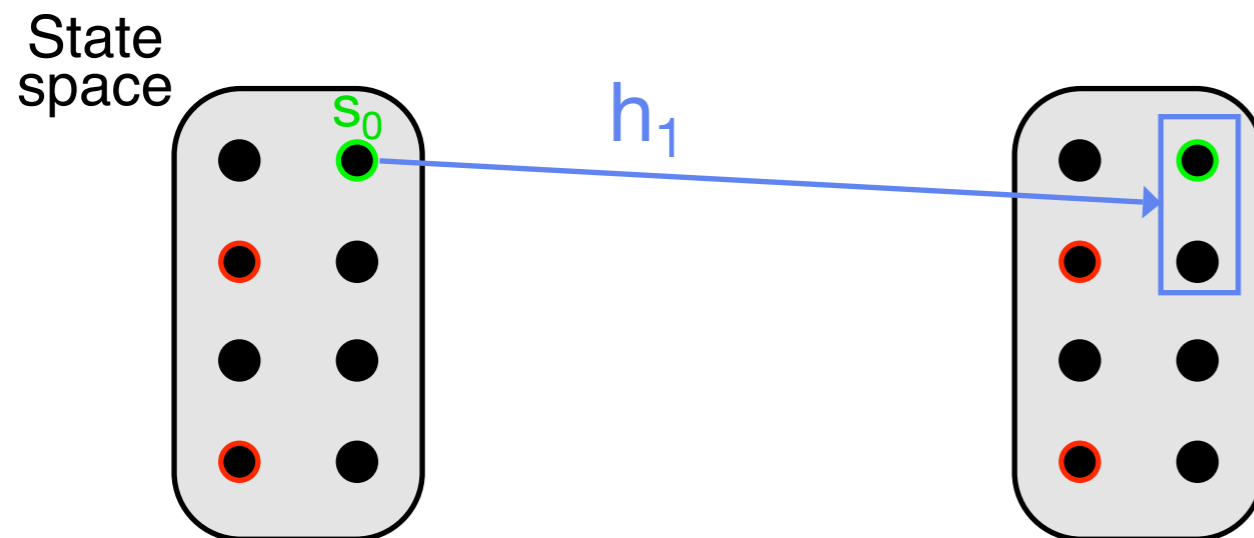
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains



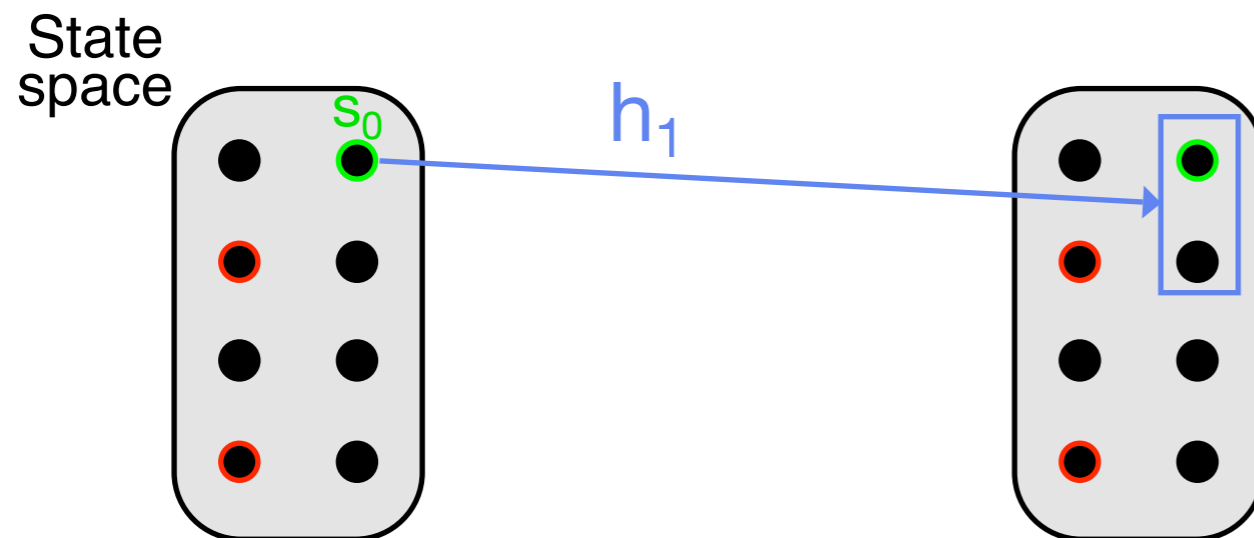
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state



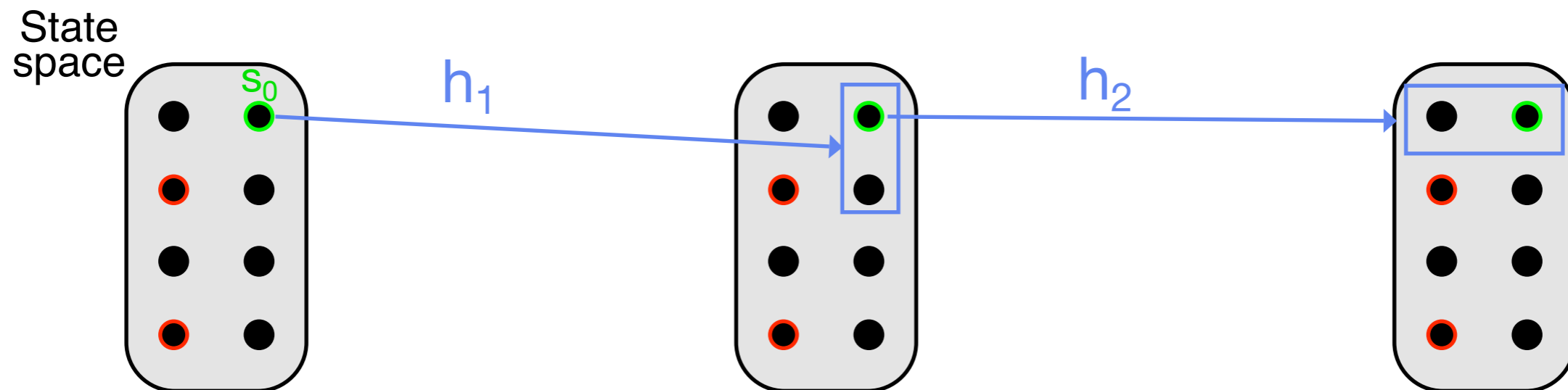
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...



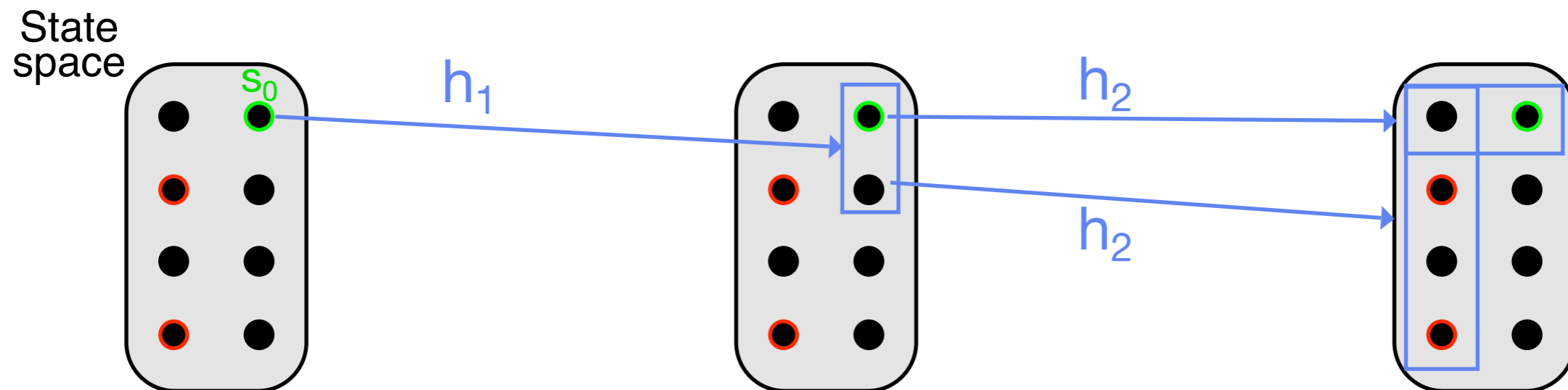
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...



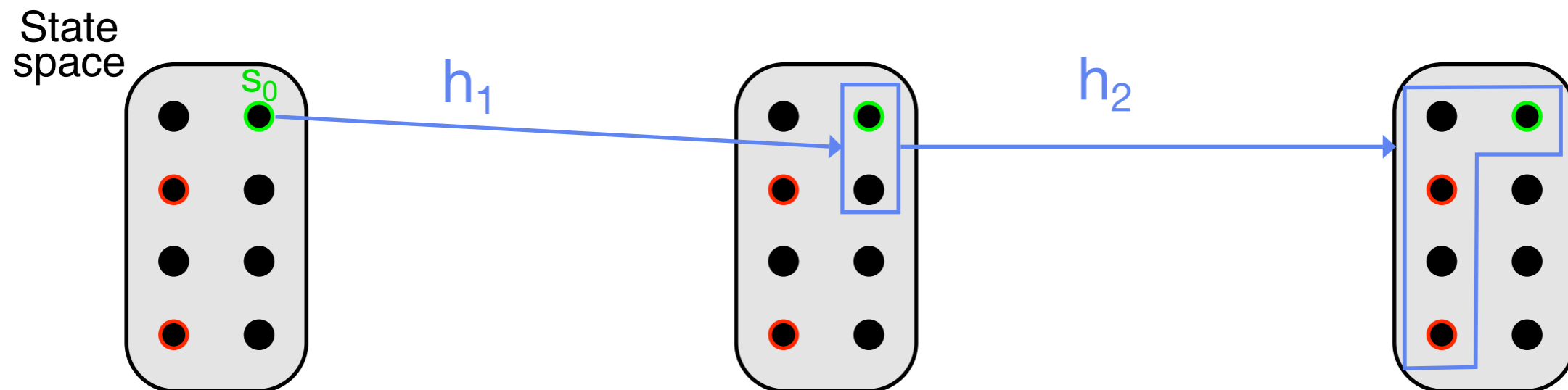
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...



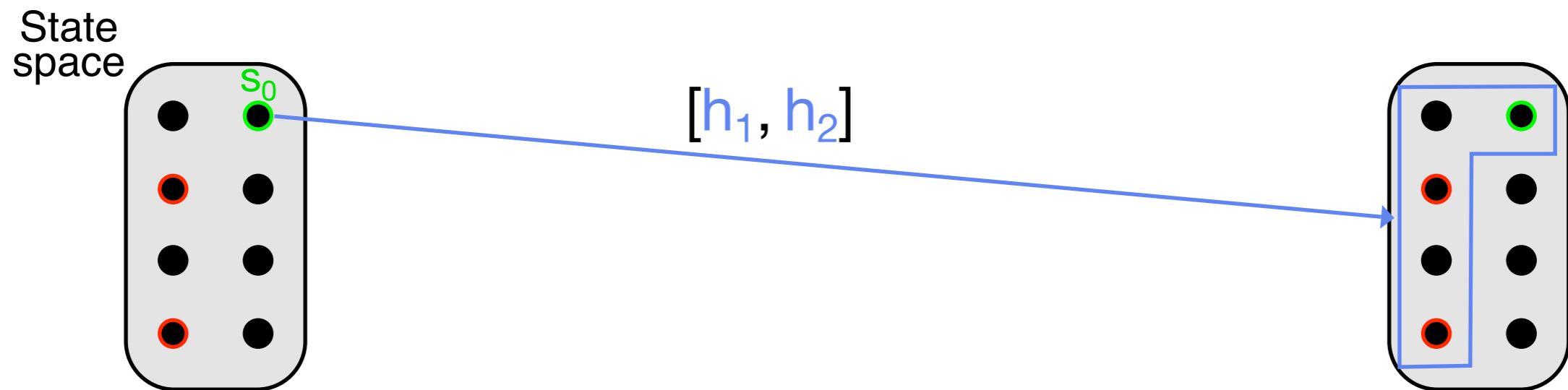
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...



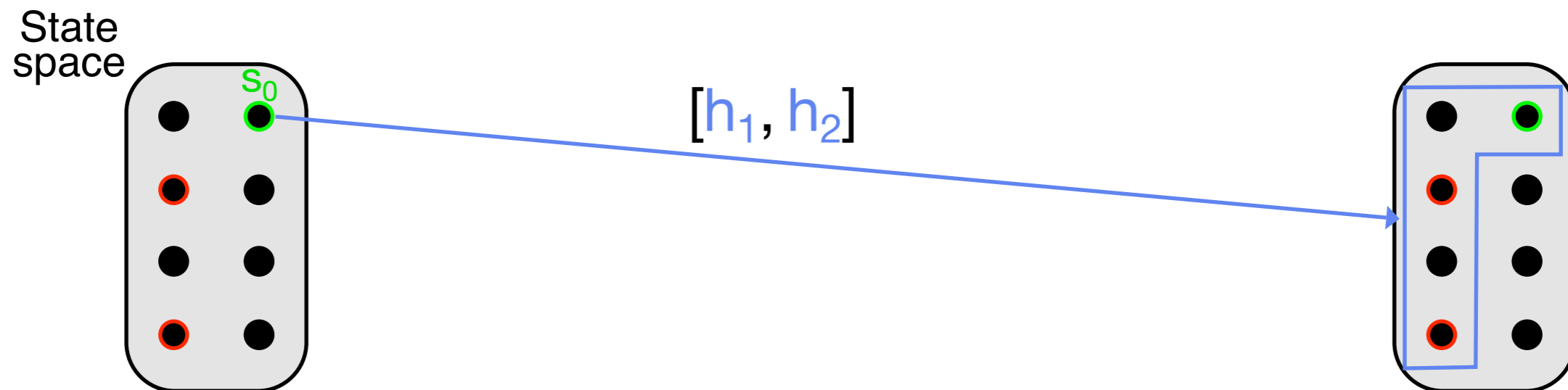
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...



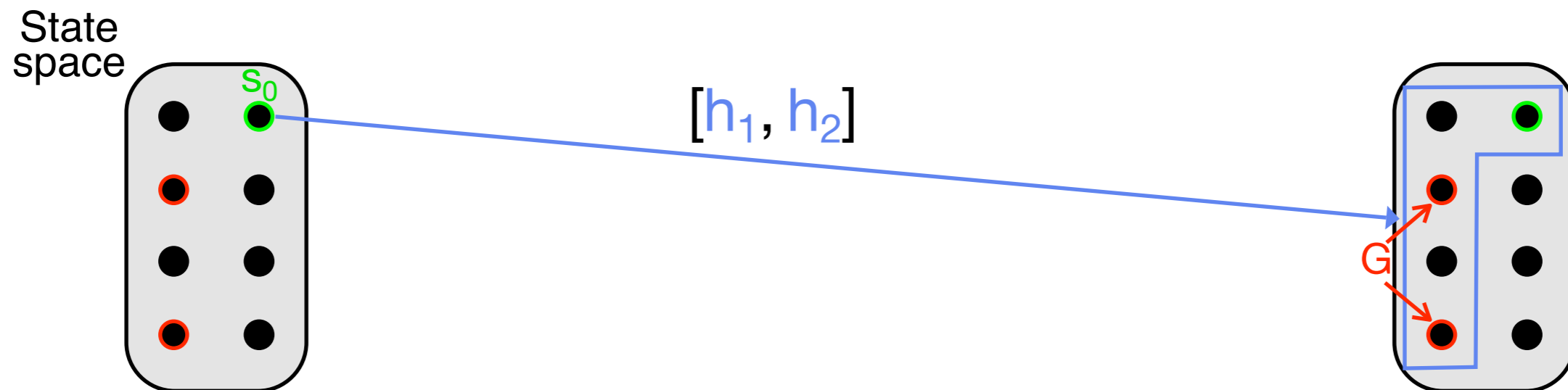
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...
 - ... allows **proving** that a plan can or cannot possibly reach the goal



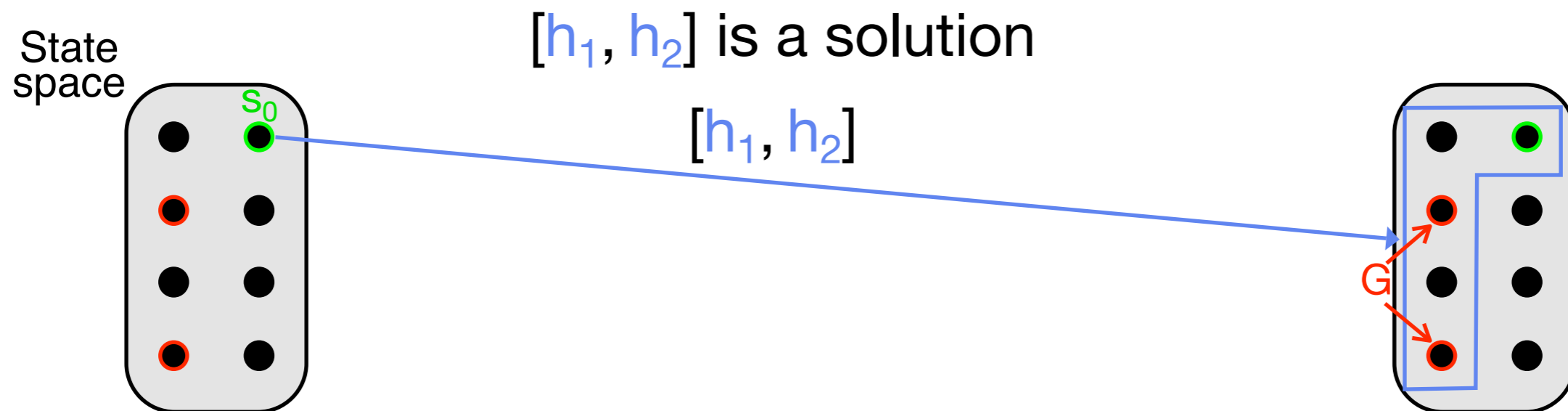
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...
 - ... allows **proving** that a plan can or cannot possibly reach the goal



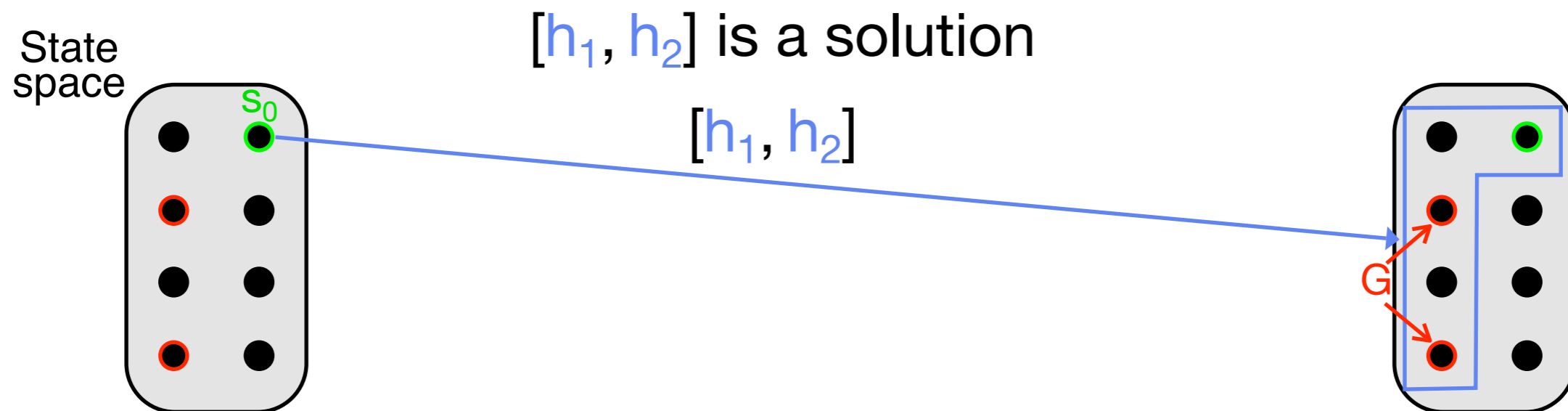
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...
 - ... allows **proving** that a plan can or cannot possibly reach the goal



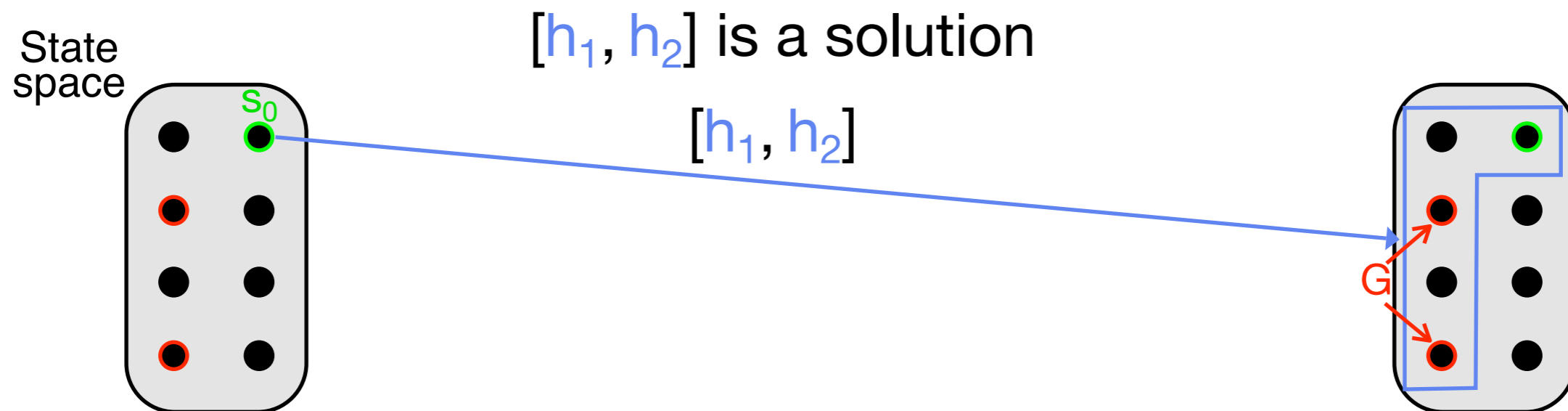
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...
 - ... allows **proving** that a plan can or cannot possibly reach the goal
- May seem related to nondeterminism ...



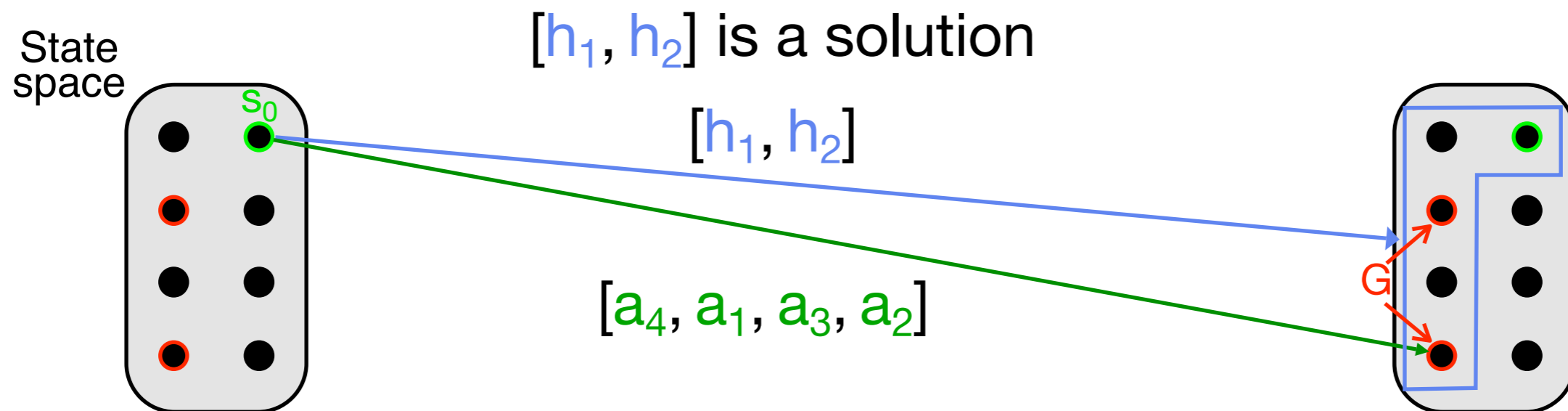
Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...
 - ... allows **proving** that a plan can or cannot possibly reach the goal
- May seem related to nondeterminism ...
 - but uncertainty is **angelic**: resolved by the agent, not an adversary



Angelic Semantics for HLAs [MRW '07]

- Models HLAs in deterministic domains
- Central idea is **reachable set** of an HLA from some state
 - When extended to sequences of actions, ...
 - ... allows **proving** that a plan can or cannot possibly reach the goal
- May seem related to nondeterminism ...
 - but uncertainty is **angelic**: resolved by the agent, not an adversary



Angelic Semantics cont.

- Approximate descriptions provide **lower & upper bounds** on reachable sets
 - Descriptions are **true**: follow logically from hierarchy

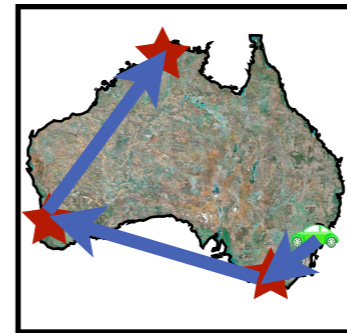
Angelic Semantics cont.

- Approximate descriptions provide **lower & upper bounds** on reachable sets
 - Descriptions are **true**: follow logically from hierarchy
- Sound & complete planning algorithm uses descriptions to
 - Commit to provably successful abstract plans:
Downward Refinement Property (DRP) automatically **satisfied**
 - potentially exponential speedup
 - **Prune** provably unsuccessful abstract plans (USP satisfied)

Contributions

- Extend angelic semantics with **action costs**
- Developed novel algorithms that do lookahead with HLAs

- **Angelic Hierarchical A*** (AHA*)



- **Angelic Hierarchical Learning Real-Time A*** (AHLRTA*)



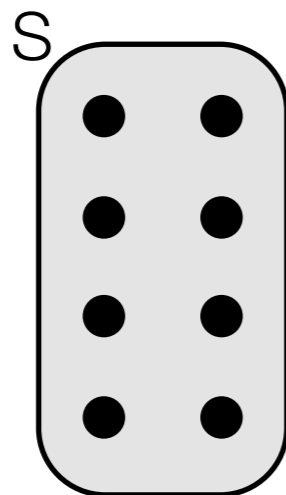
- Both require three inputs:
 - planning problem
 - action hierarchy (set of HLAs)
 - approximate models for HLAs

Deterministic Planning Problems

- Here, a planning problem =

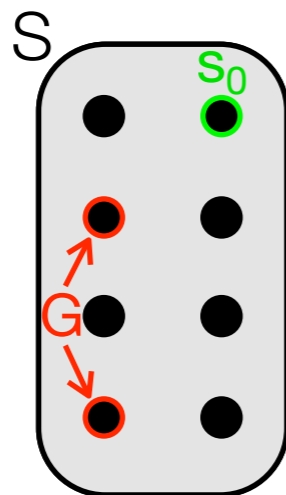
Deterministic Planning Problems

- Here, a planning problem =
 - State space S



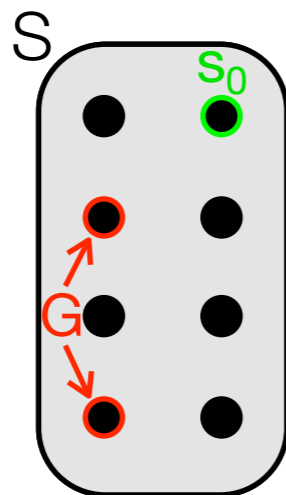
Deterministic Planning Problems

- Here, a planning problem =
 - State space S
 - Initial state s_0 , terminal set G



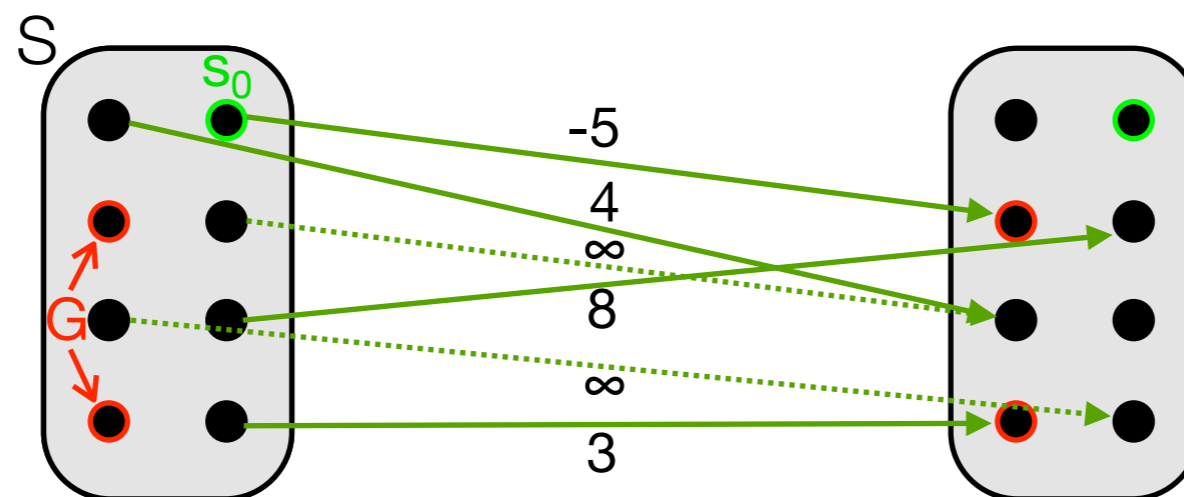
Deterministic Planning Problems

- Here, a planning problem =
 - State space S
 - Initial state s_0 , terminal set G
 - Primitive action set



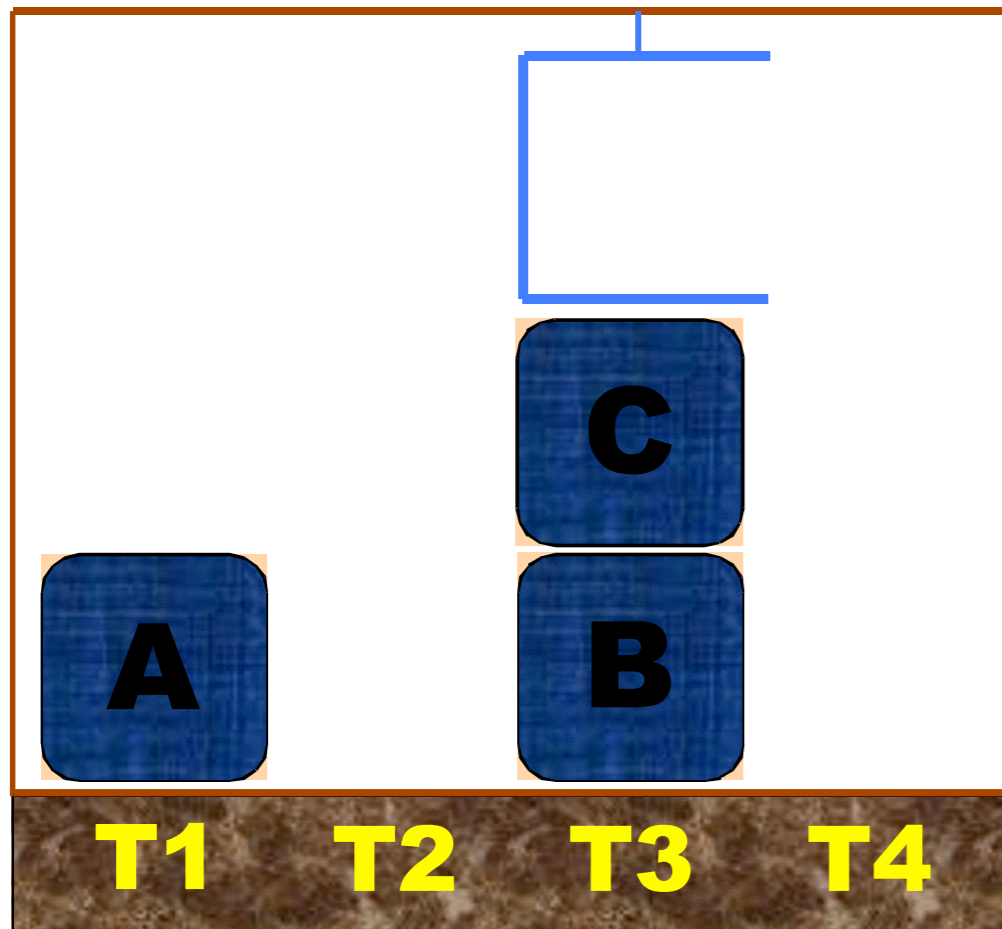
Deterministic Planning Problems

- Here, a planning problem =
 - State space S
 - Initial state s_0 , terminal set G
 - Primitive action set
 - Transition function: $S \times A \rightarrow S$
 - Cost function : $S \times A \rightarrow \mathbb{R} \cup \{\infty\}$



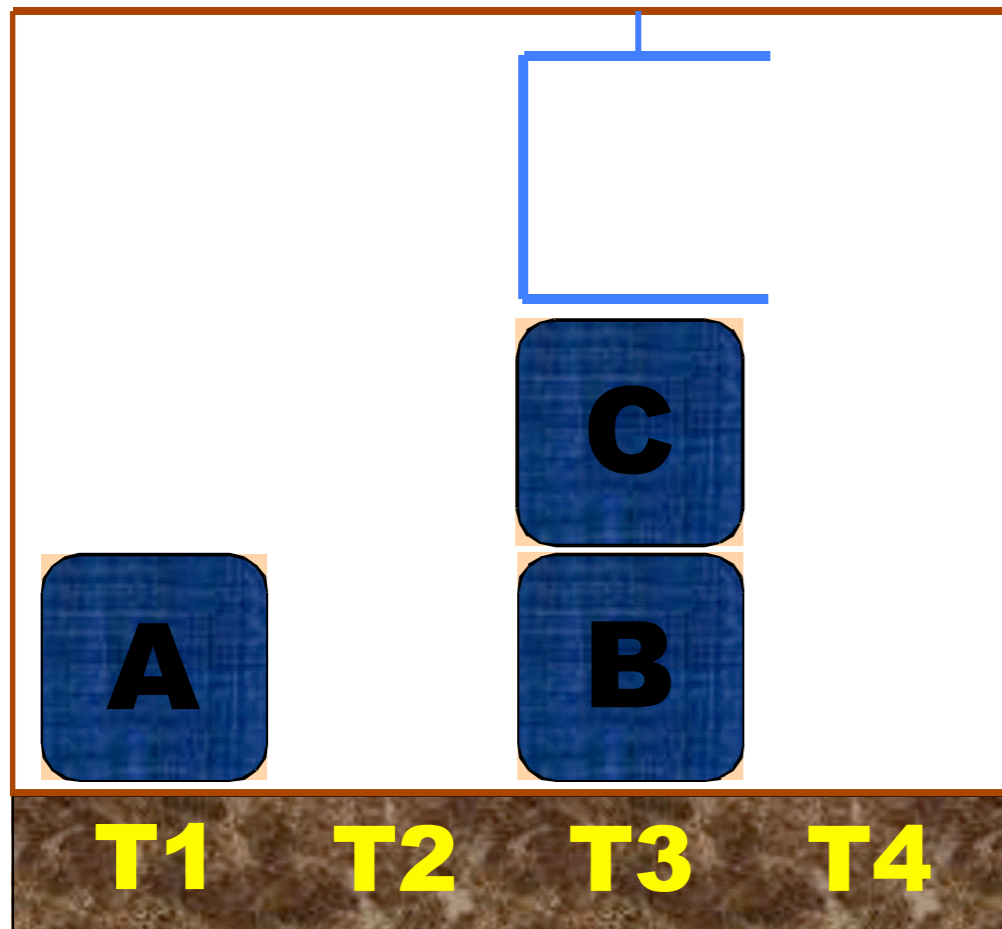
Transitions & costs for action a_1

Running Example: *Warehouse World* Domain



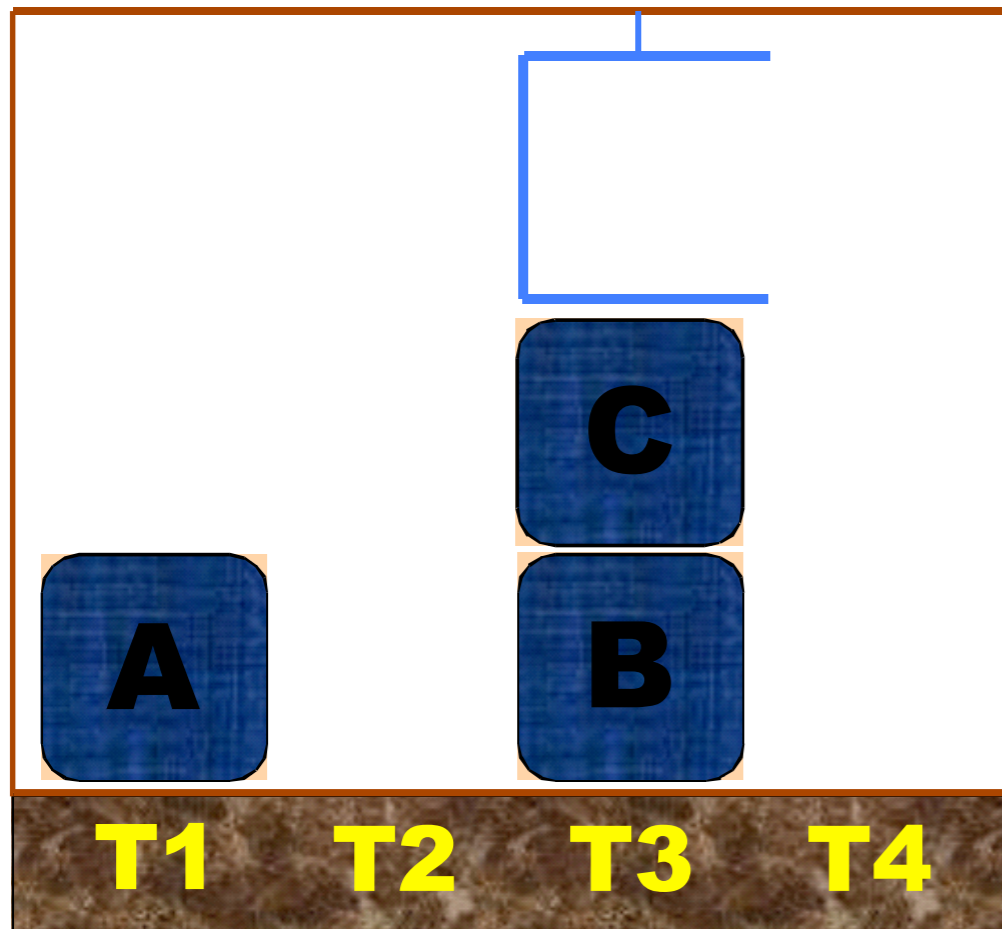
- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row

Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1

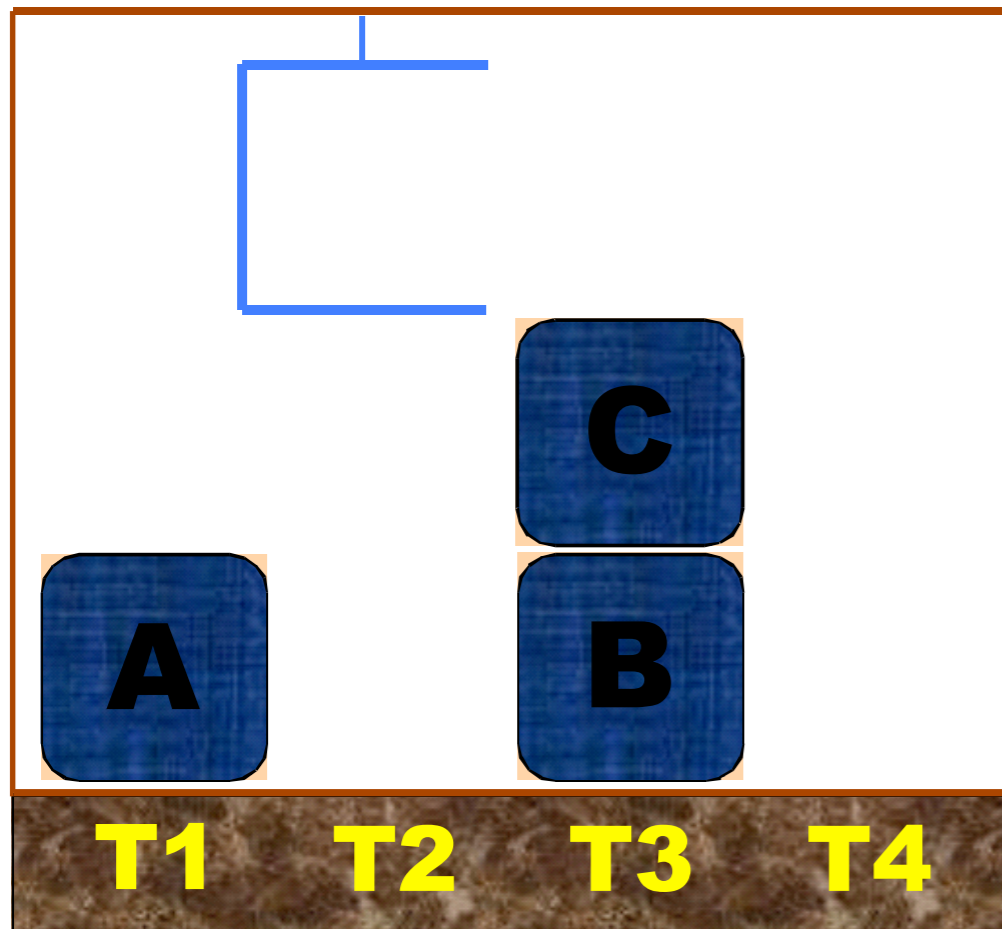
Running Example: *Warehouse World* Domain



L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

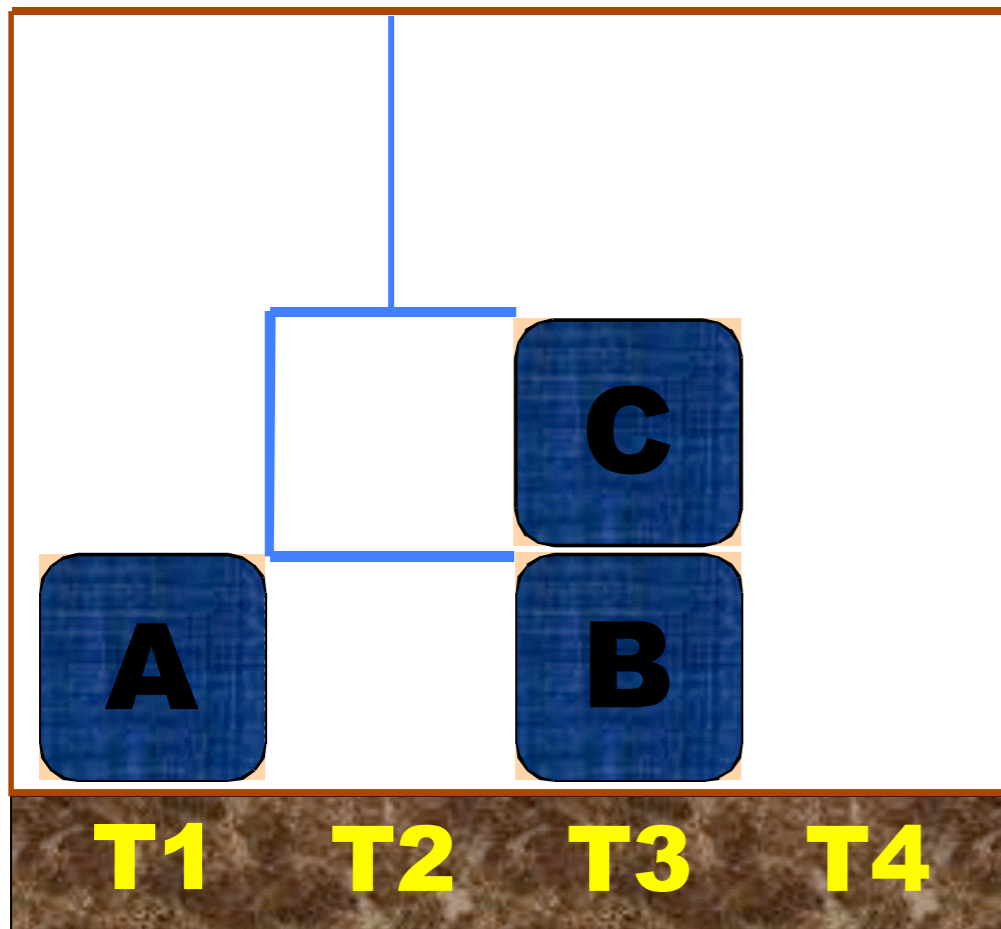
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

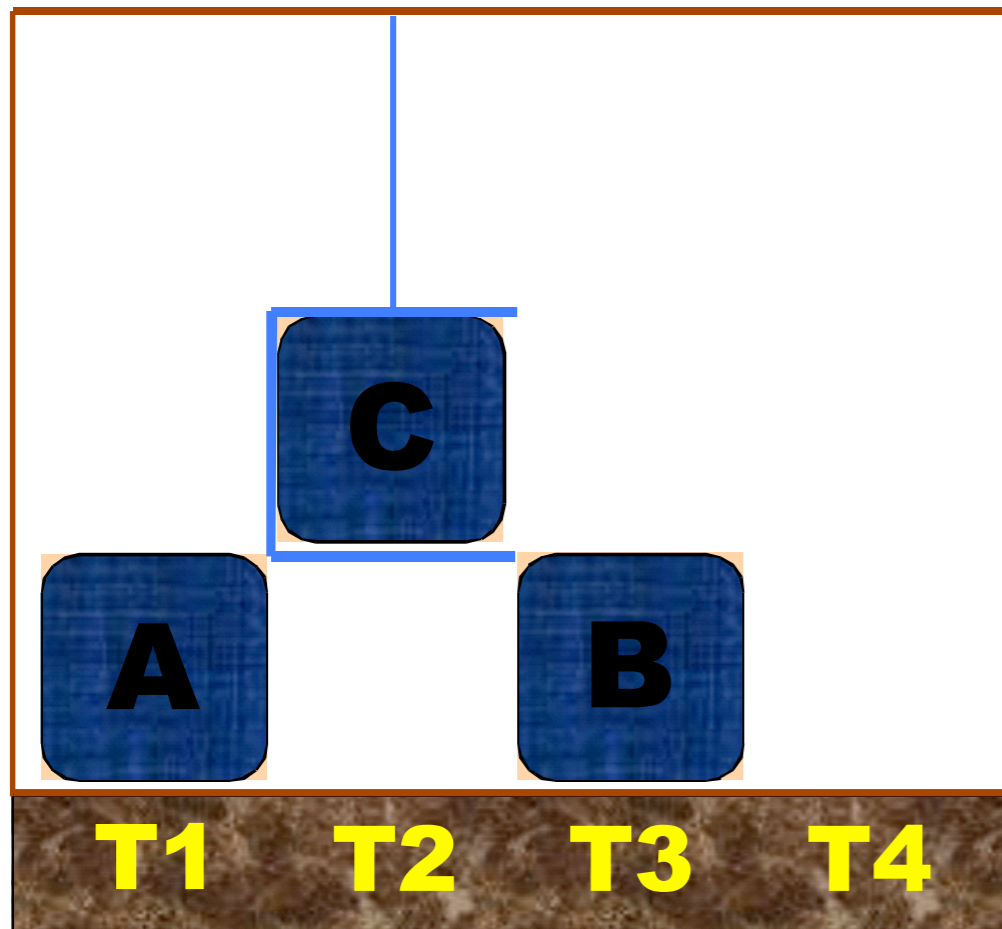
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

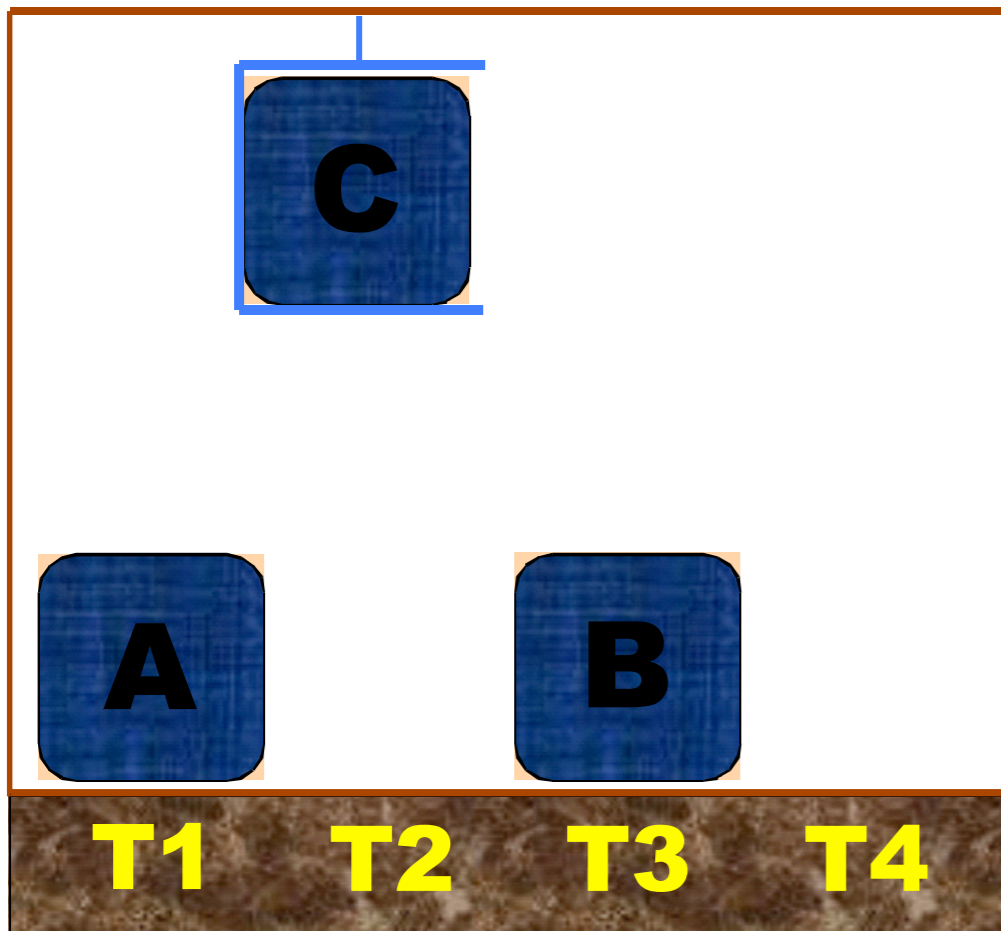
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

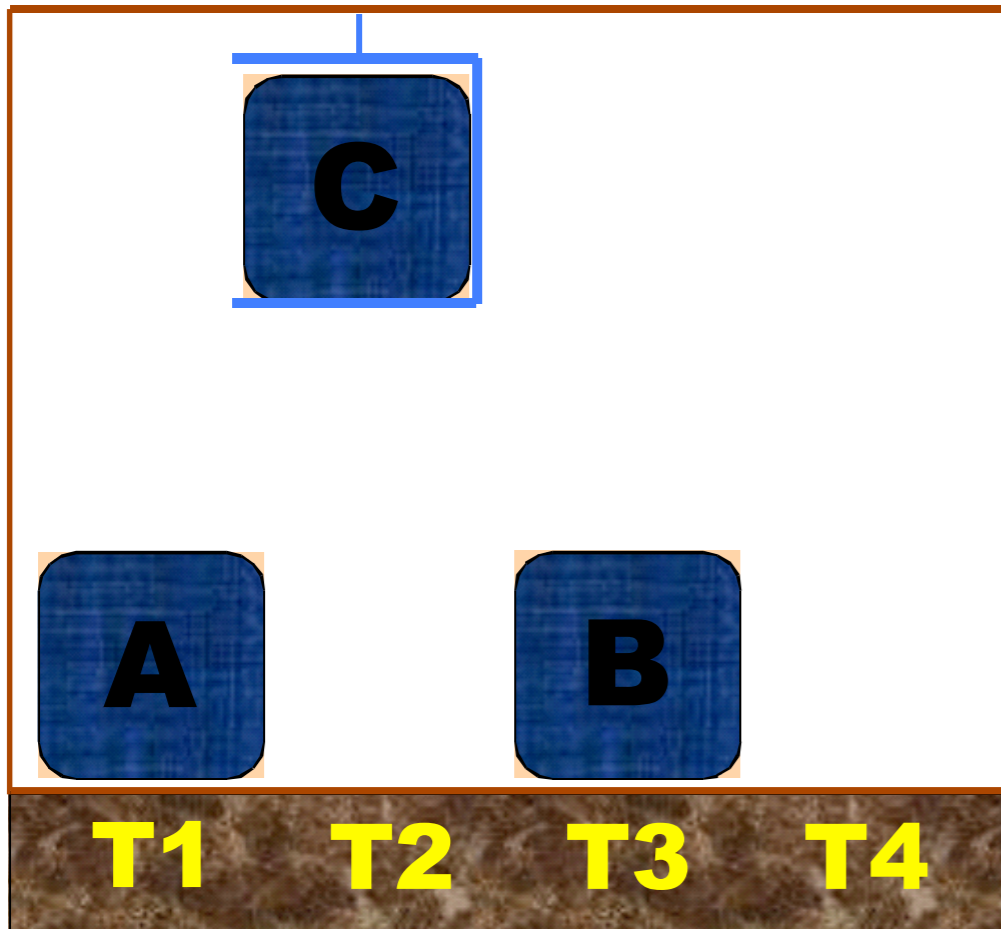
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

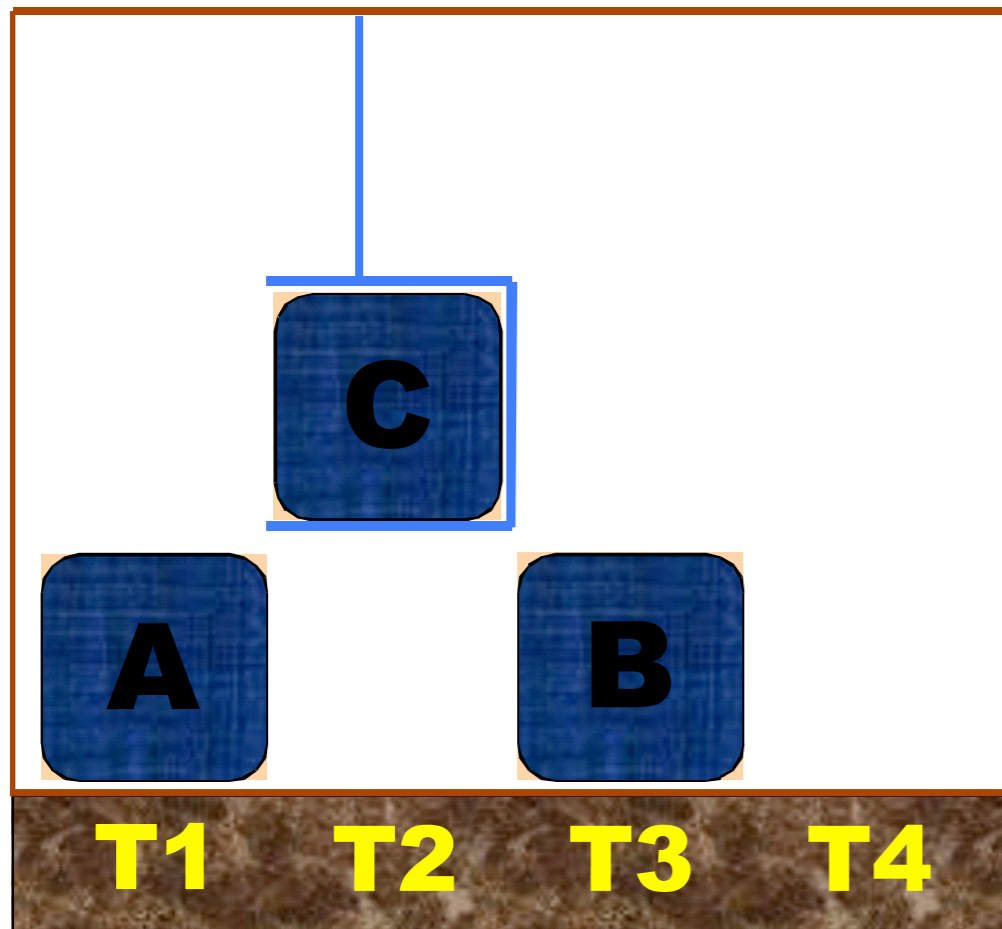
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

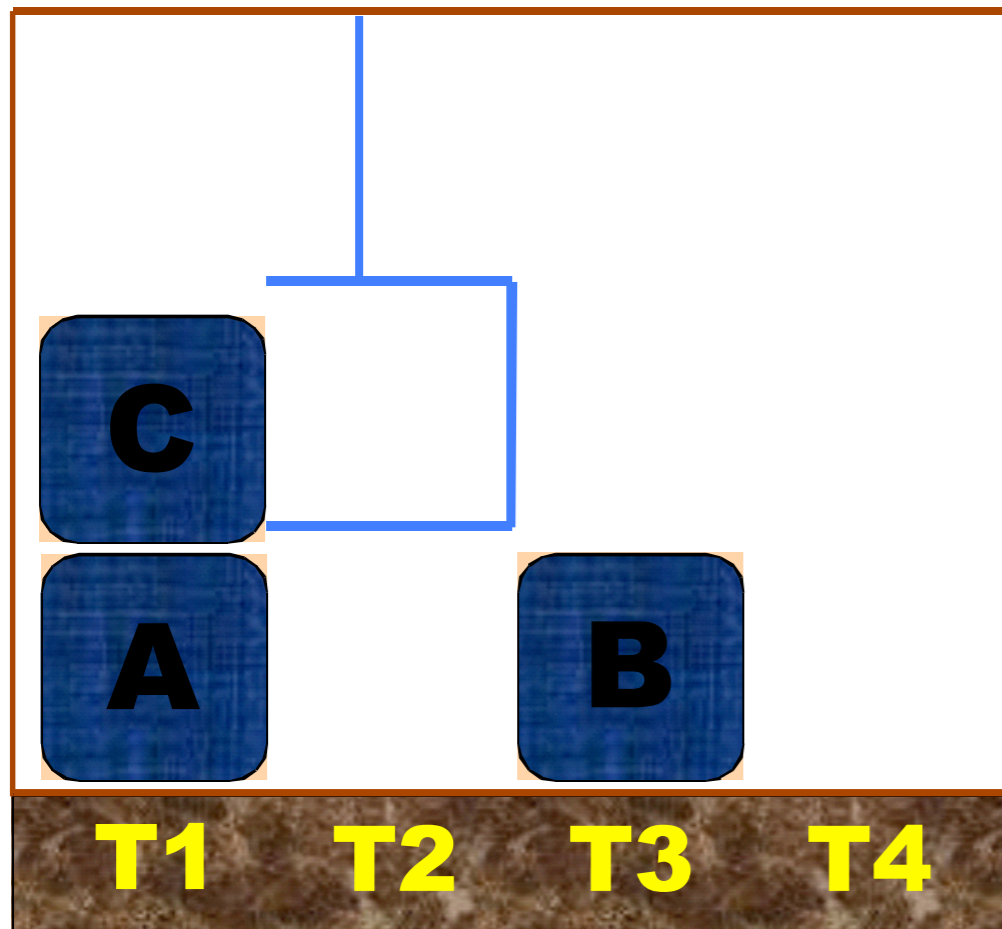
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

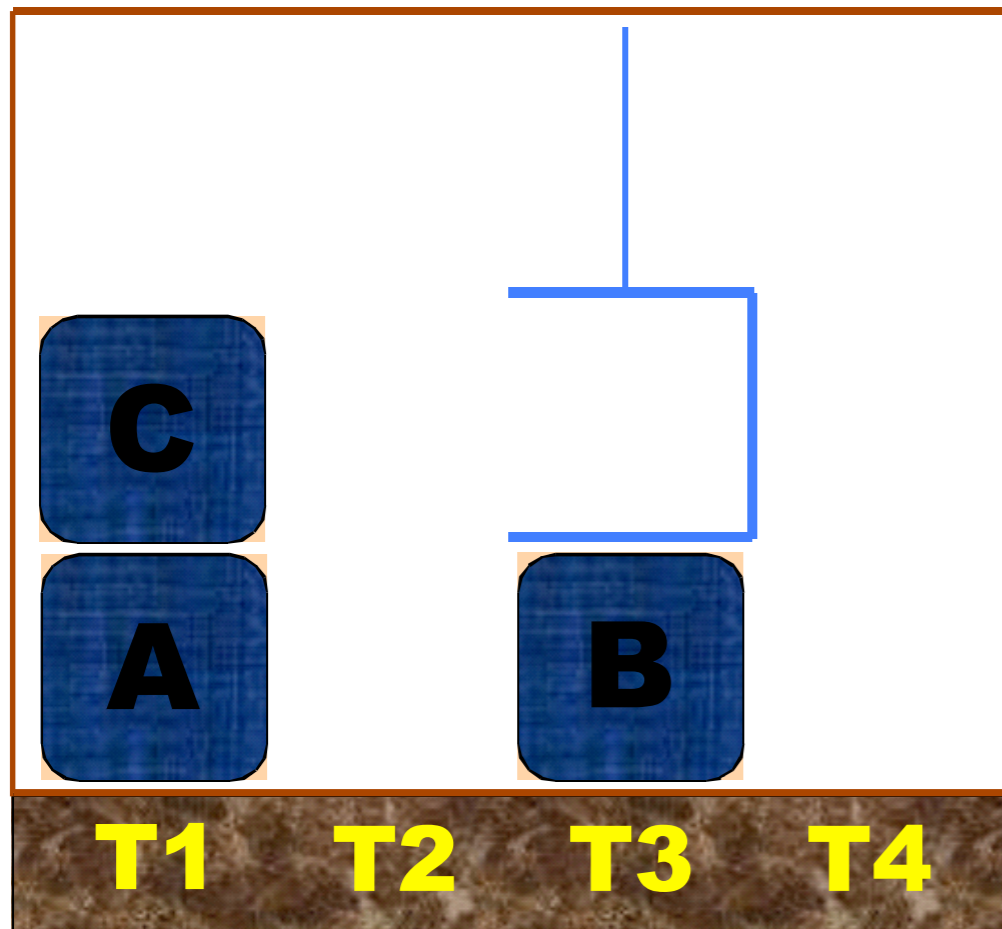
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

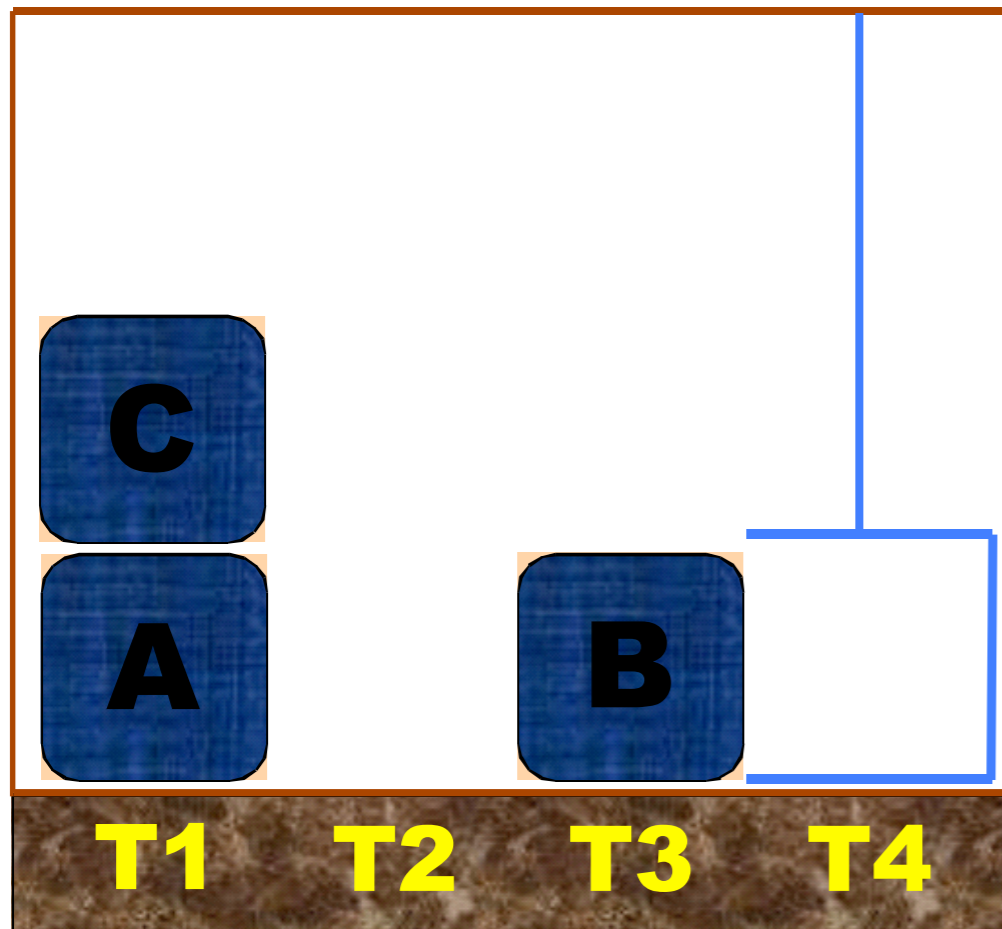
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

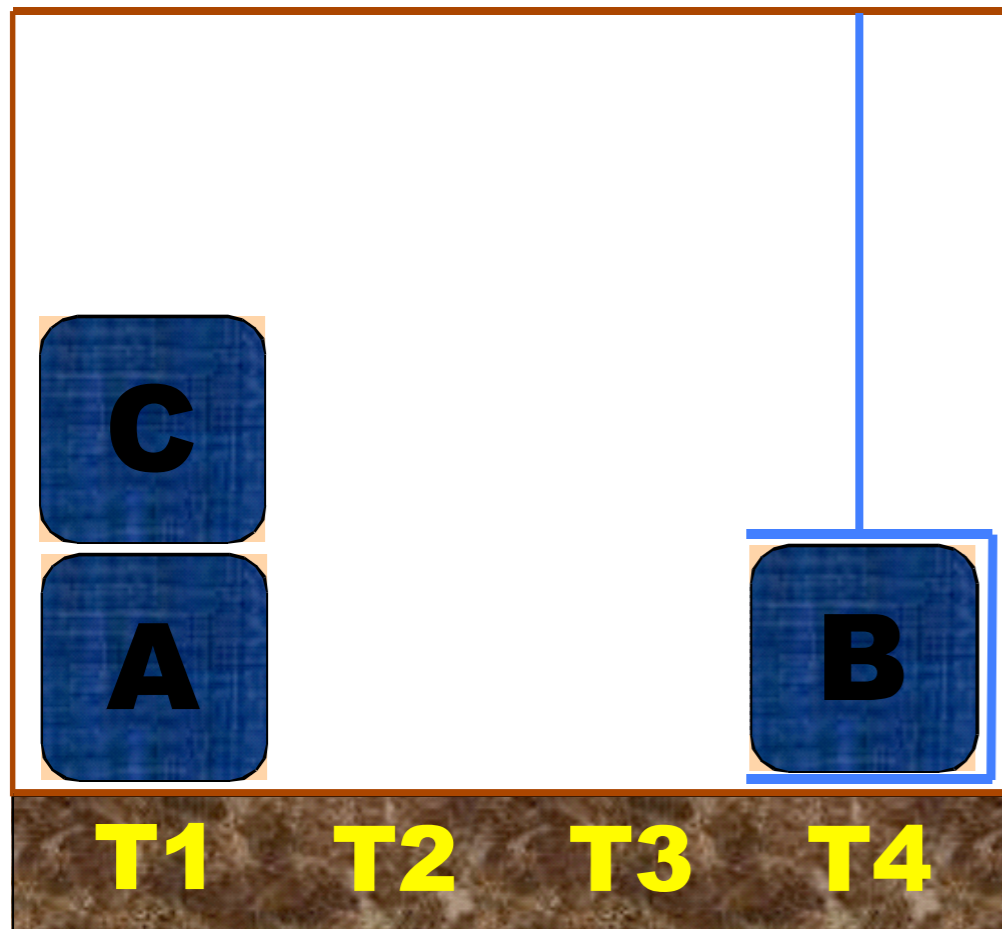
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

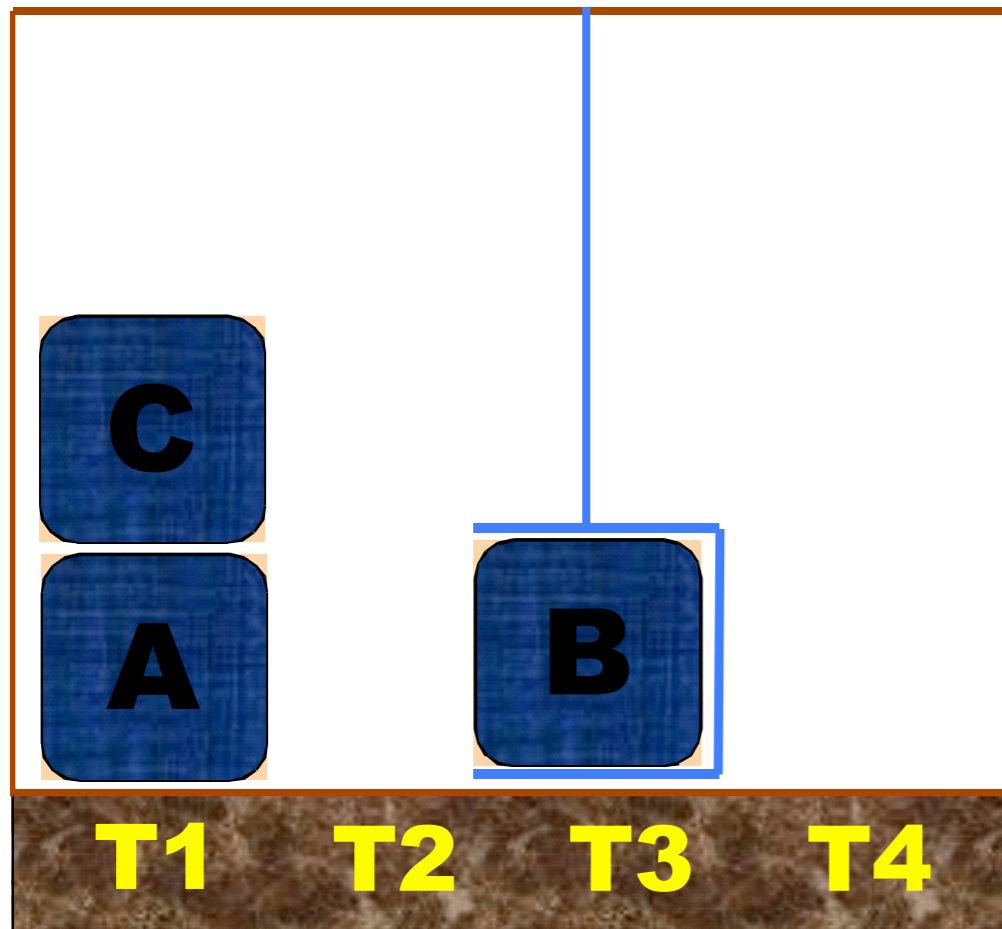
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

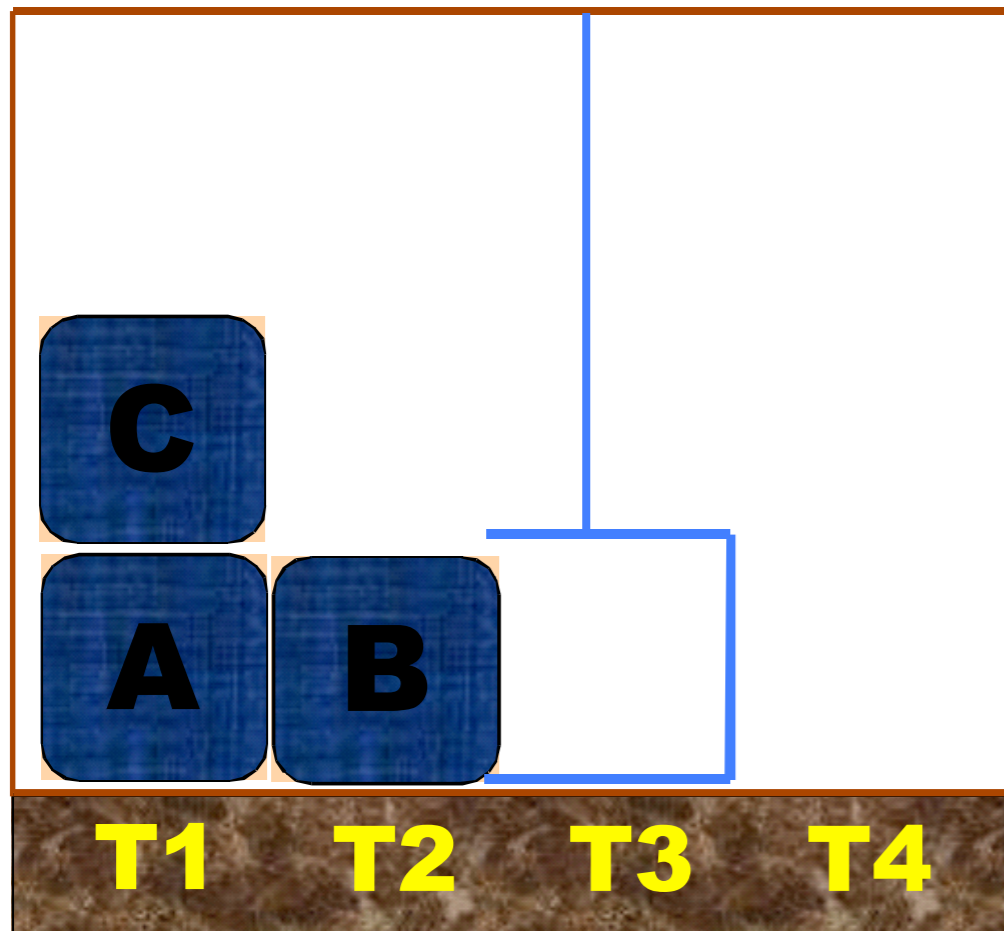
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

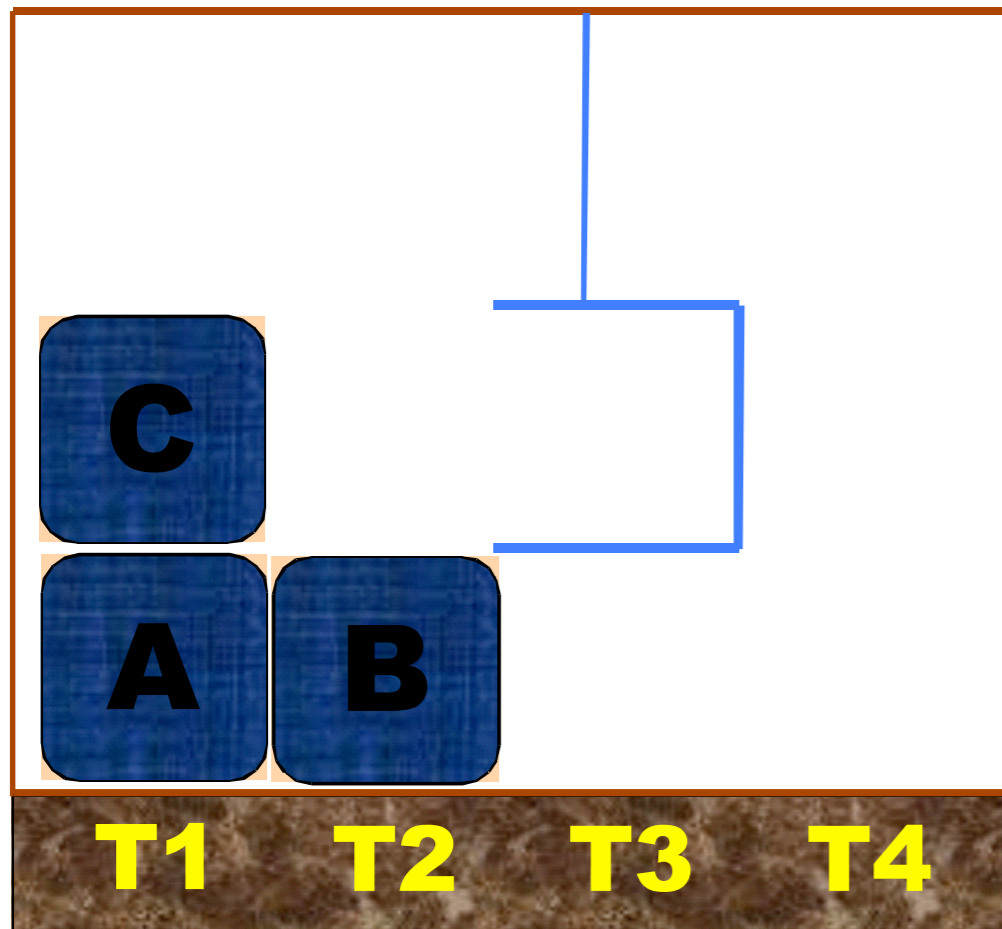
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

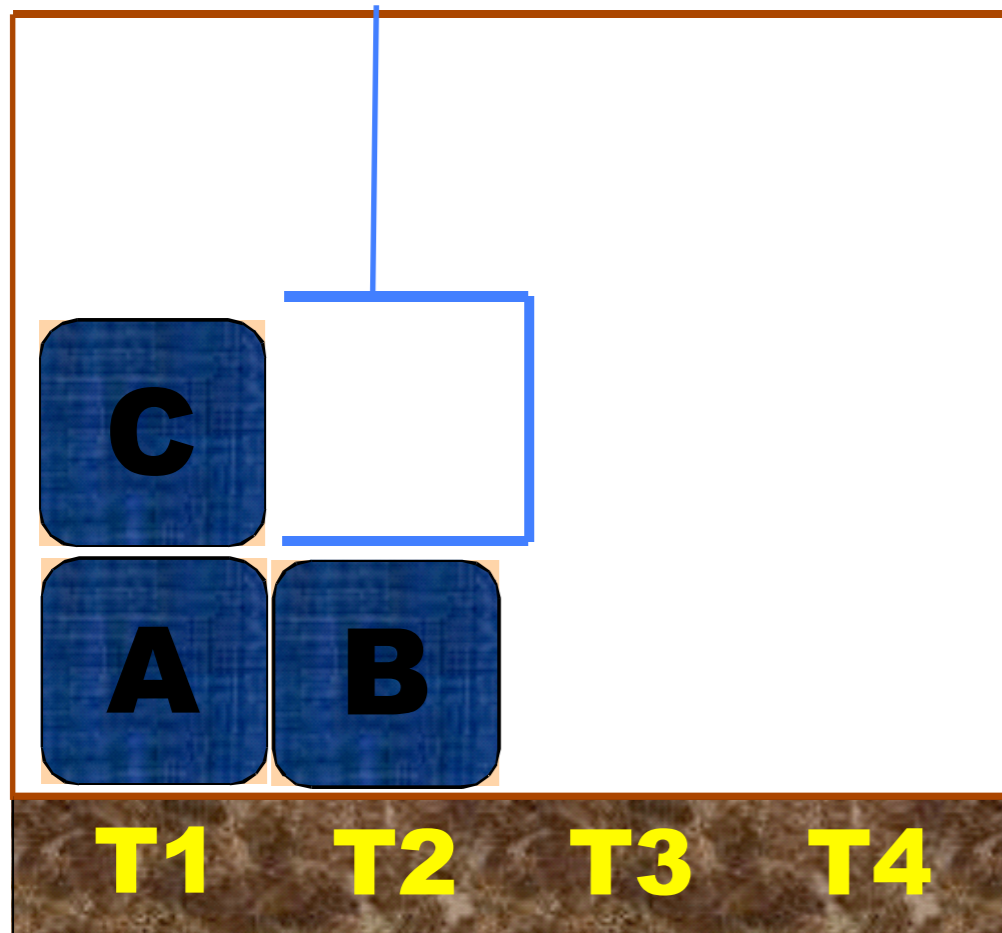
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

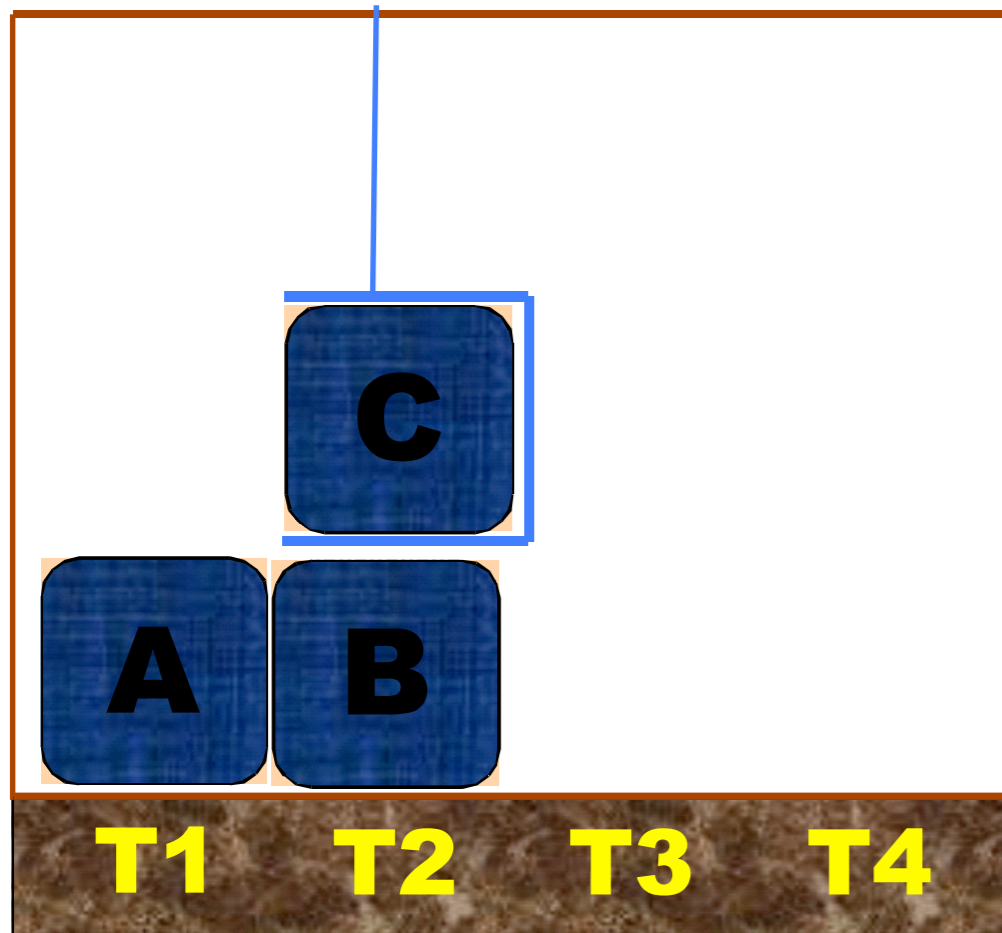
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

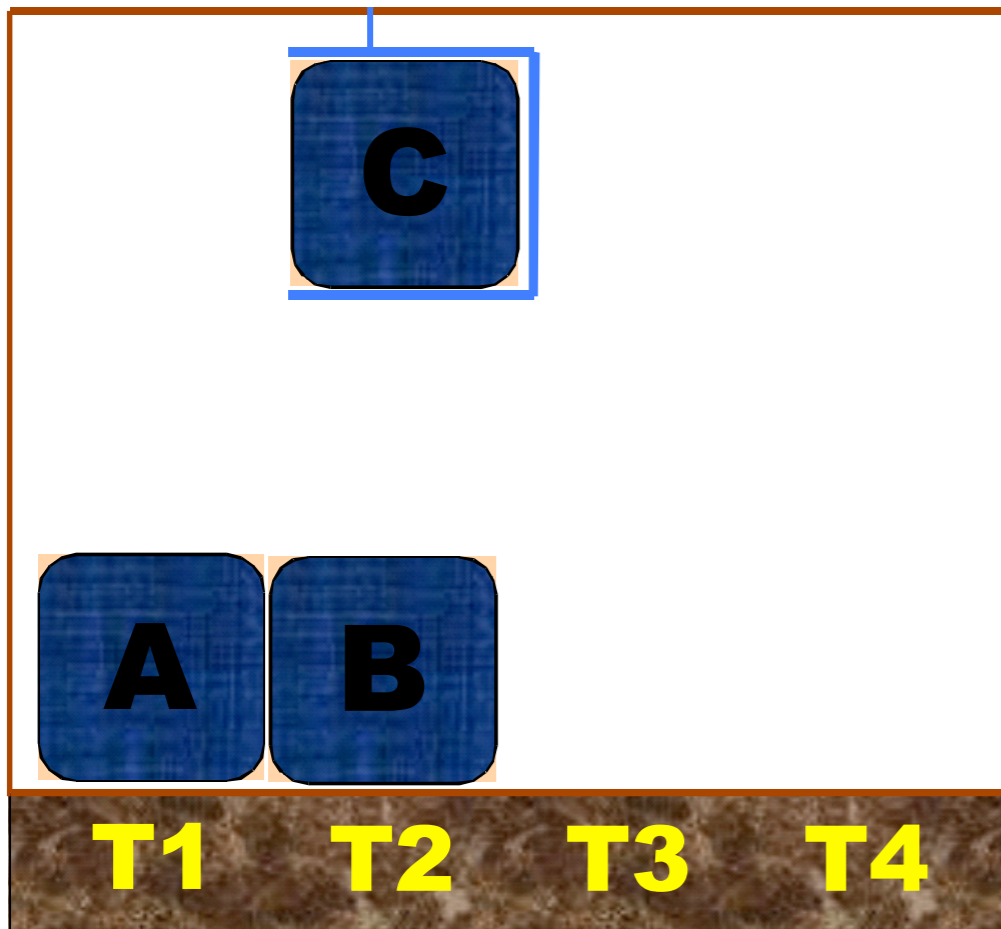
Running Example: *Warehouse World* Domain



L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

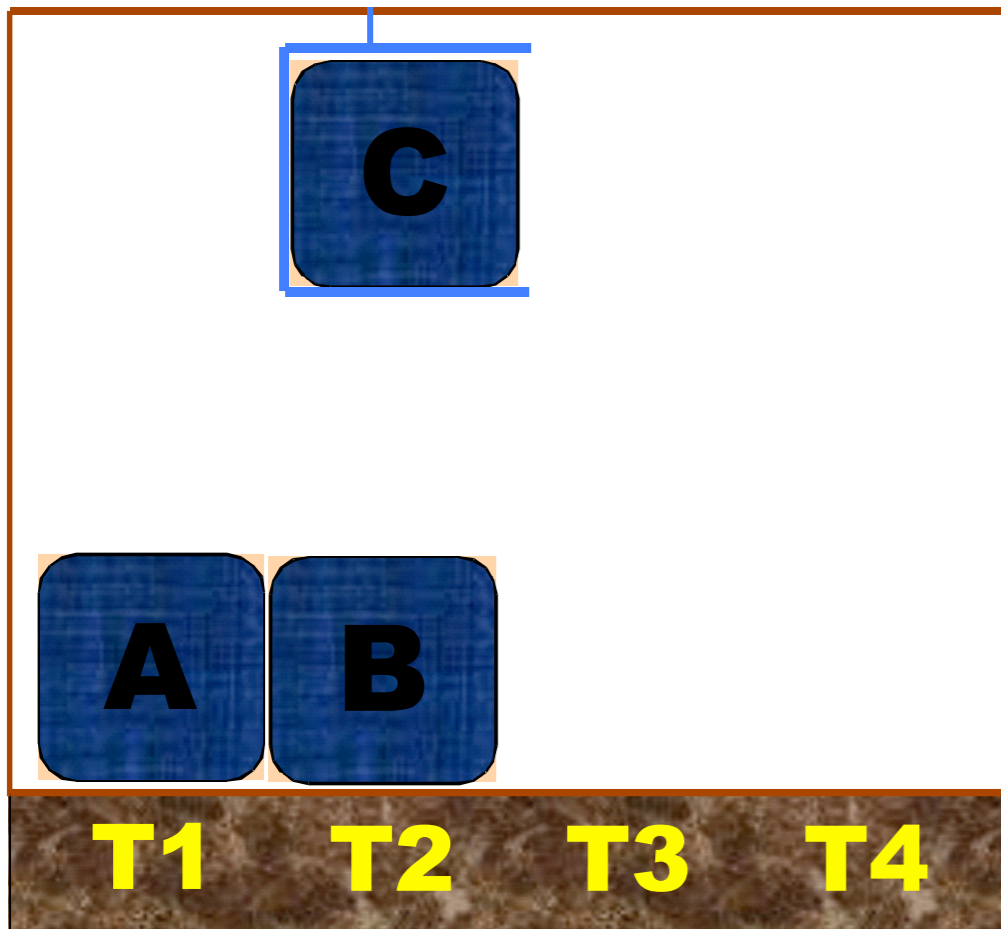
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

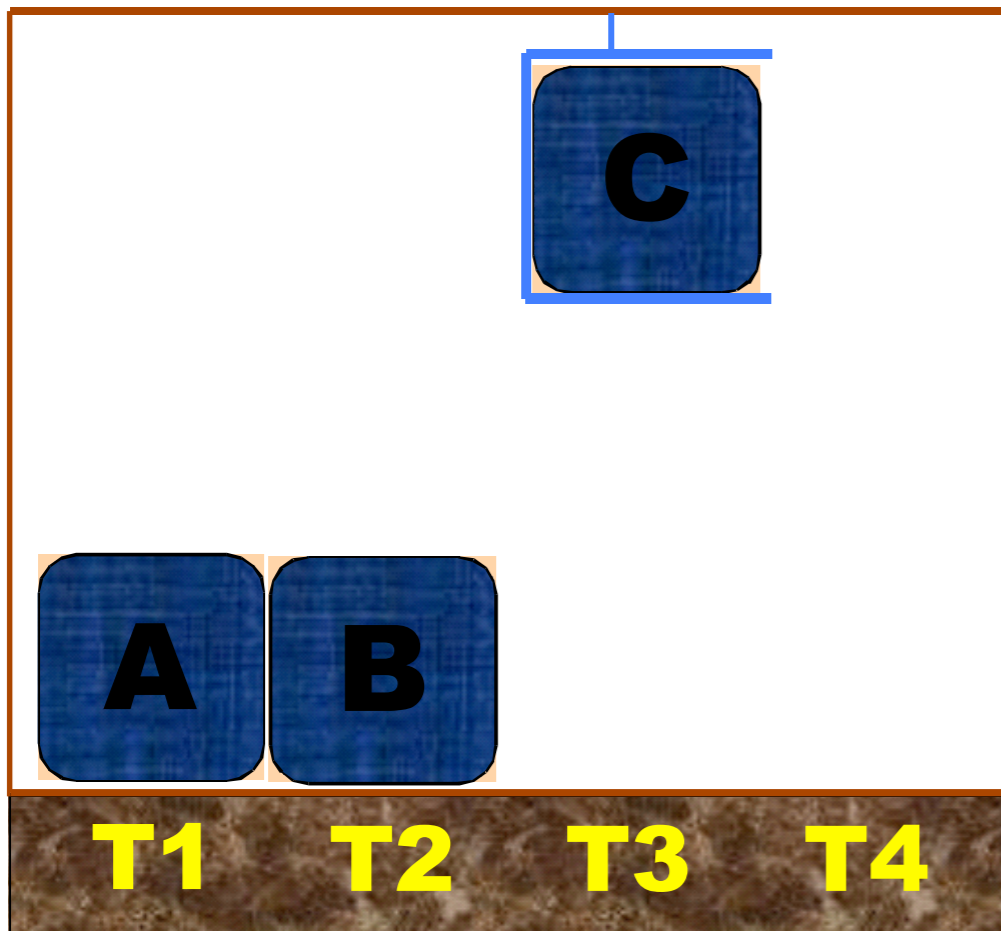
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

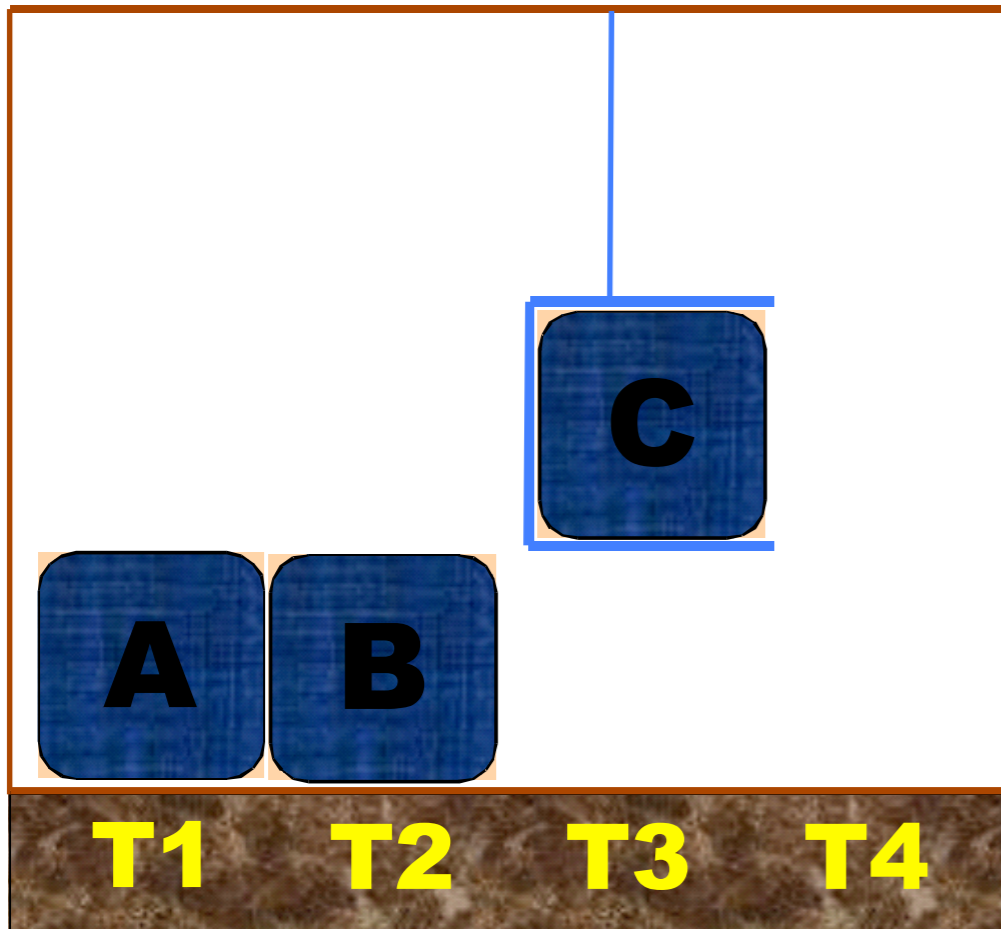
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

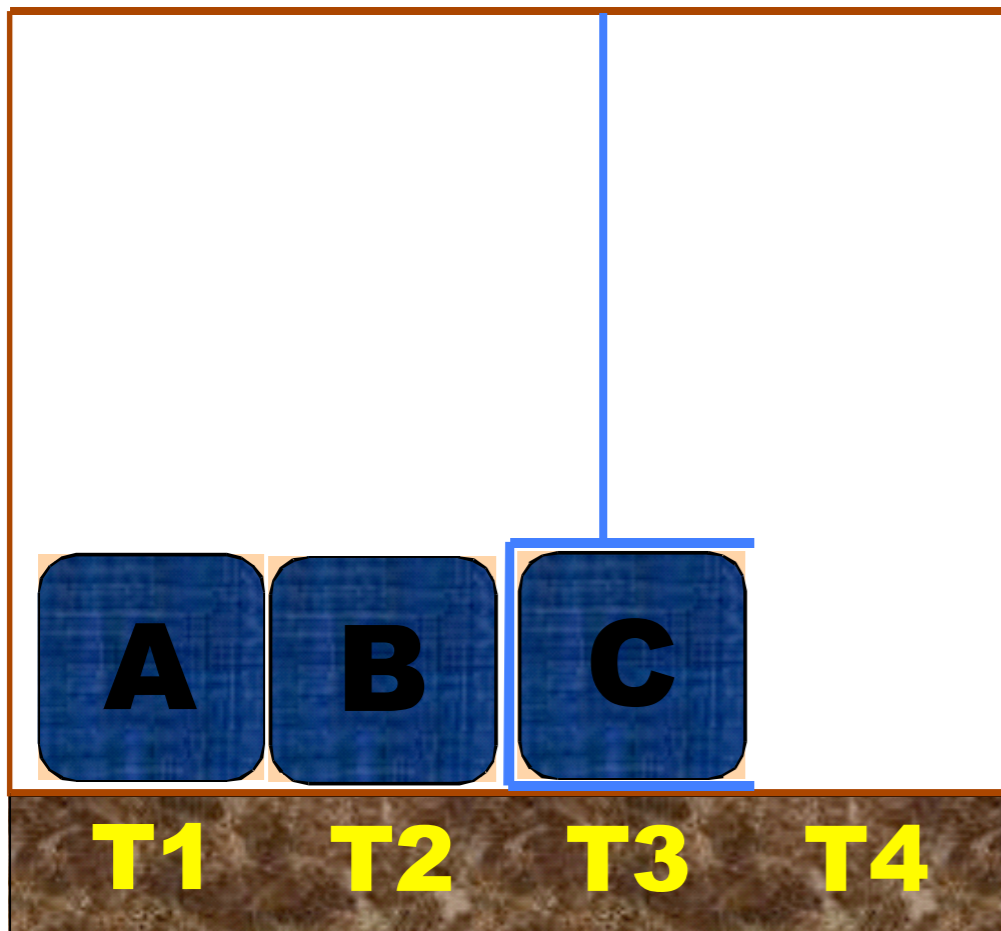
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

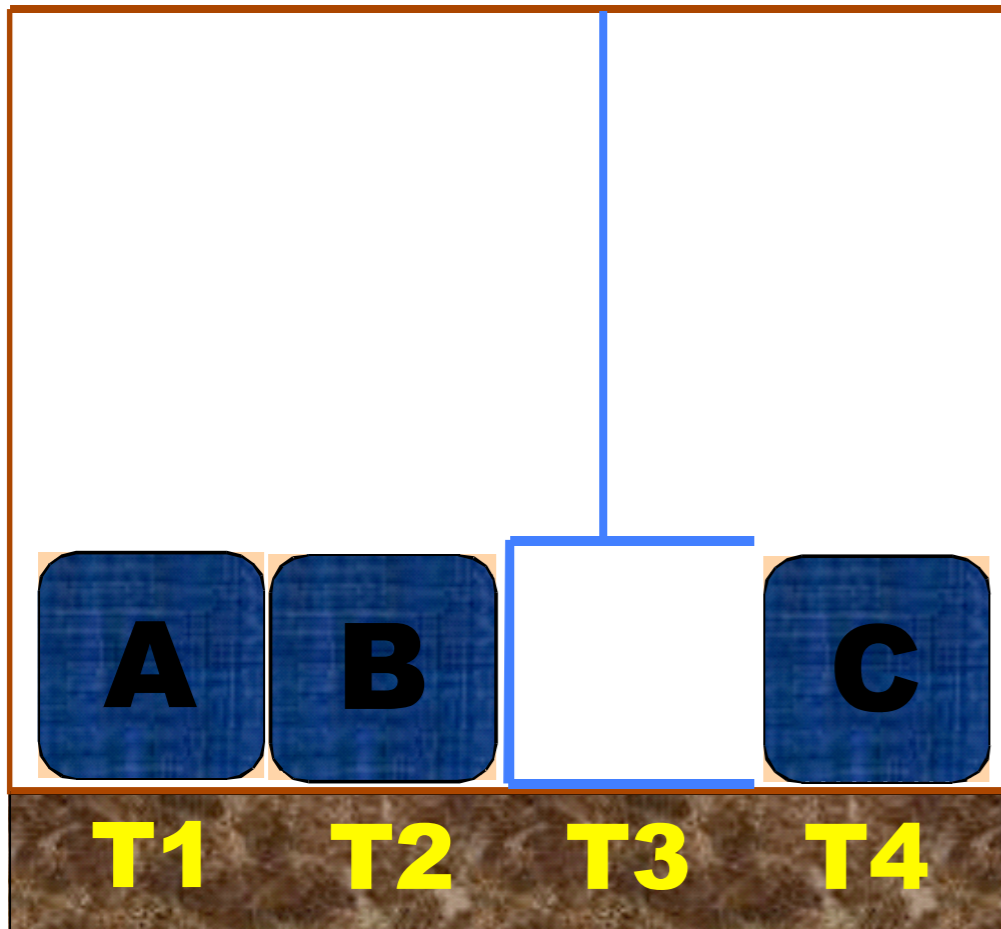
Running Example: *Warehouse World* Domain



- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

Running Example: *Warehouse World* Domain



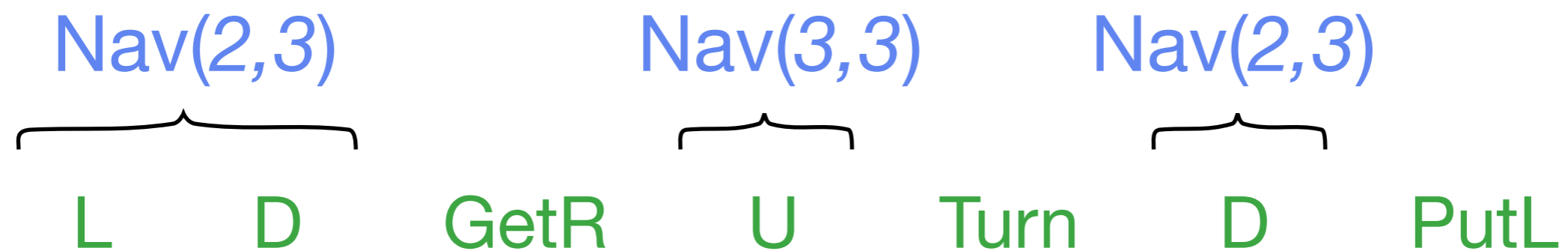
- Elaborated *Blocks World* with discrete spatial constraints
 - Gripper must stay in bounds
 - Can't pass through blocks
 - Can only turn at top row
- All actions have cost 1
- Goal: have C on T4
 - Can't just move directly
 - Final plan has 22 steps

L, D, GetR, U, Turn, D, PutL,
R, R, D, GetL, L, PutL, U, L,
GetL, U, Turn, R, D, D, PutR

Running Example: *Warehouse World* HLAs

L D GetR U Turn D PutL

Running Example: *Warehouse World* HLAs



Running Example: *Warehouse World* HLAs

NavT(2,3)



Nav(2,3)



L

D

GetR

NavT(2,3)



Nav(3,3)



U

Turn

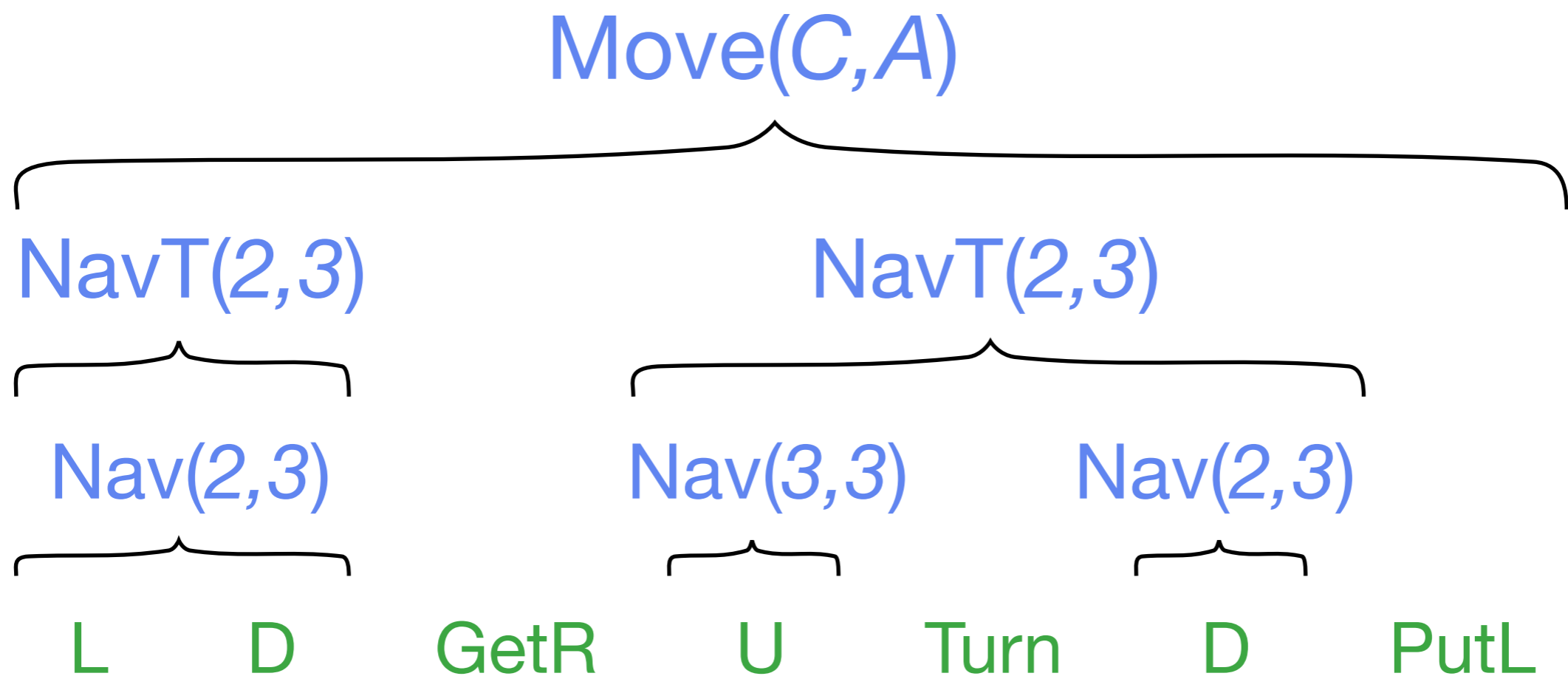
Nav(2,3)



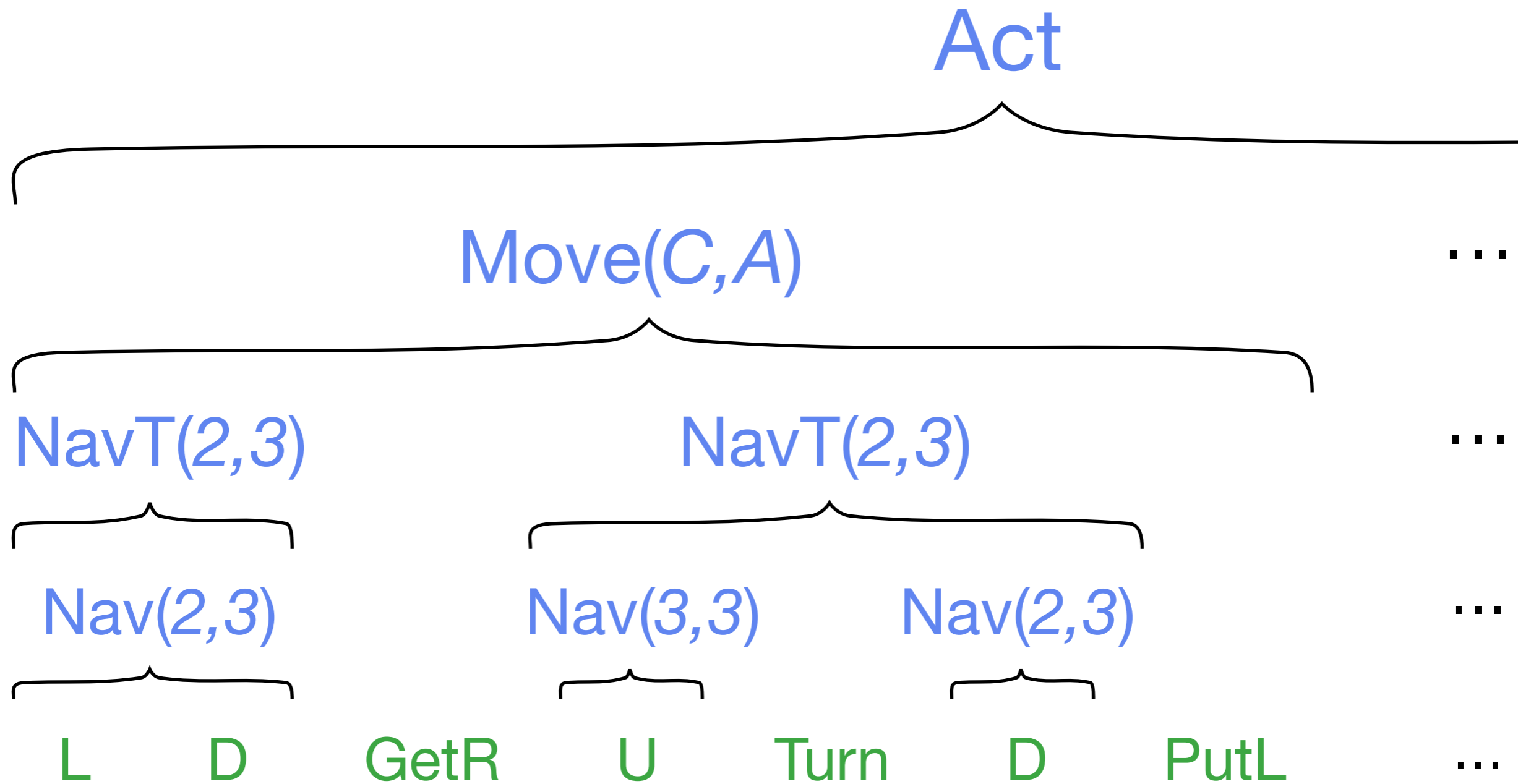
D

PutL

Running Example: *Warehouse World* HLAs



Running Example: *Warehouse World* HLAs



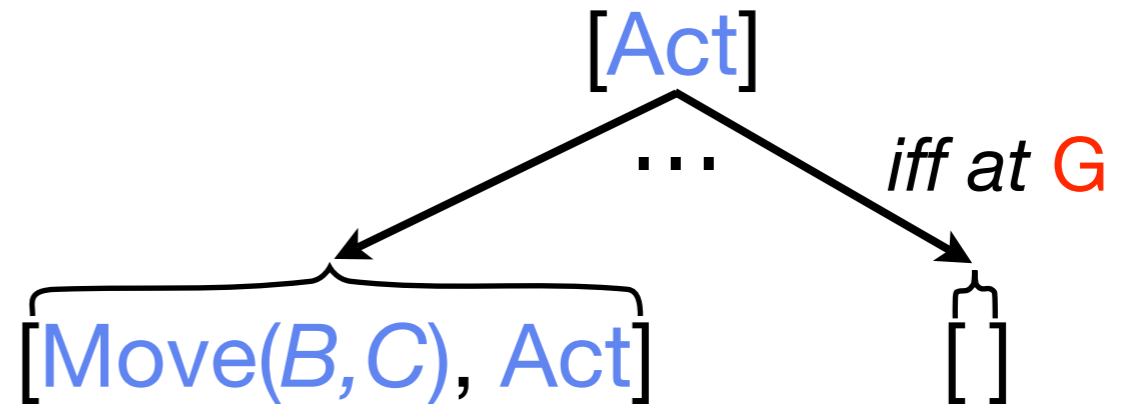
Running Example: *Warehouse World* HLAs

- Plans of interest are **primitive refinements** of special HLA **Act**

[Act]

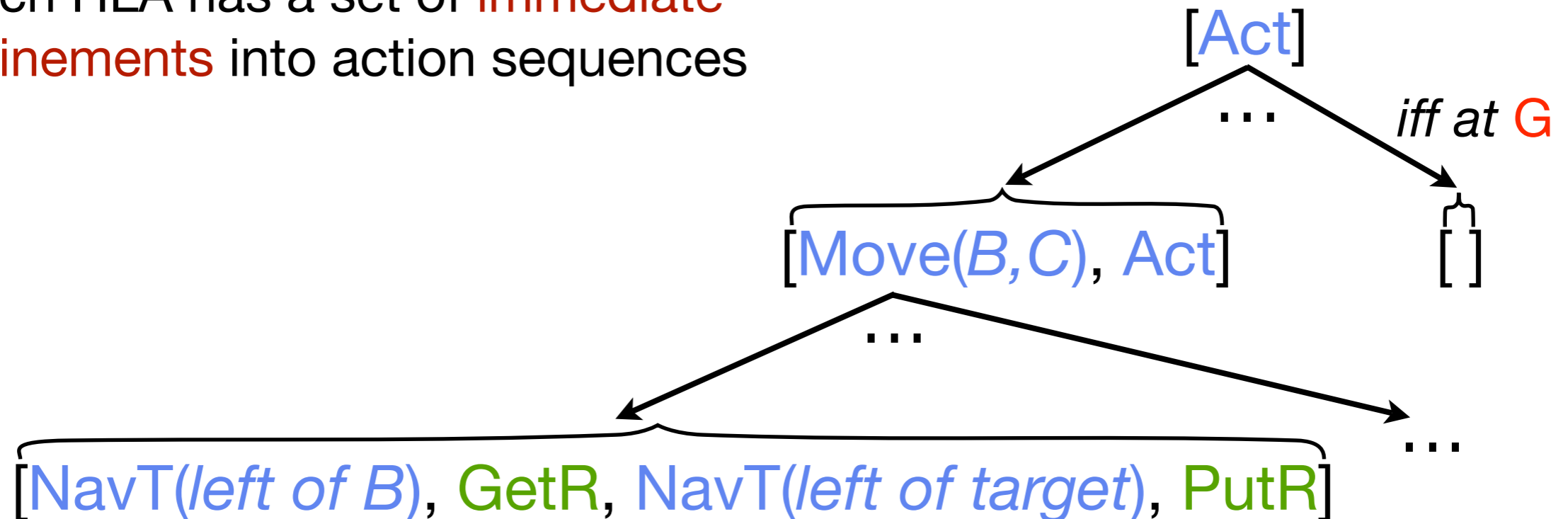
Running Example: *Warehouse World* HLAs

- Plans of interest are **primitive refinements** of special HLA **Act**
- Each HLA has a set of **immediate refinements** into action sequences



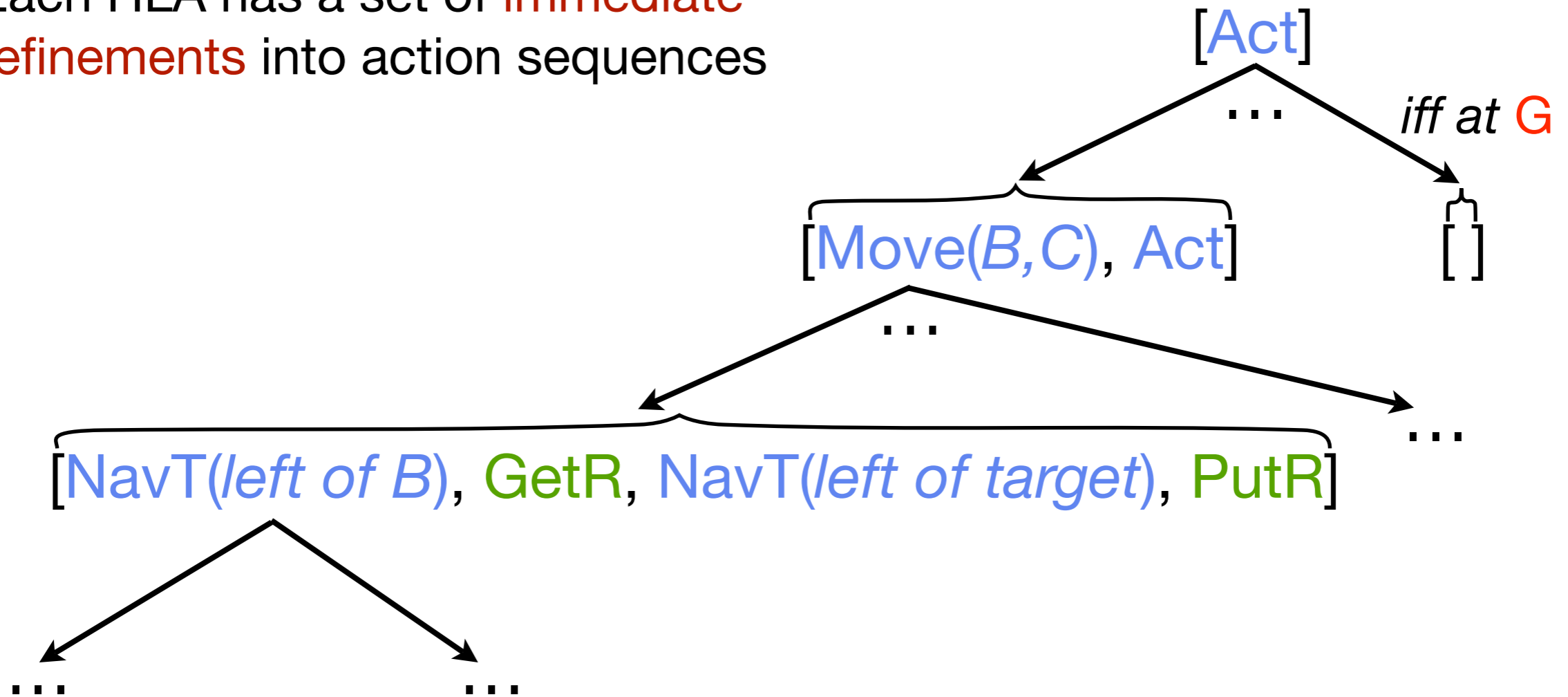
Running Example: Warehouse World HLAs

- Plans of interest are **primitive refinements** of special HLA **Act**
- Each HLA has a set of **immediate refinements** into action sequences



Running Example: Warehouse World HLAs

- Plans of interest are **primitive refinements** of special HLA **Act**
- Each HLA has a set of **immediate refinements** into action sequences



Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)

Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans



Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



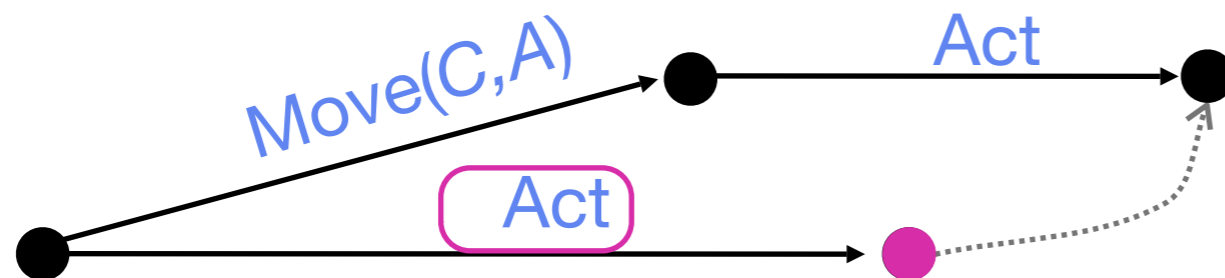
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



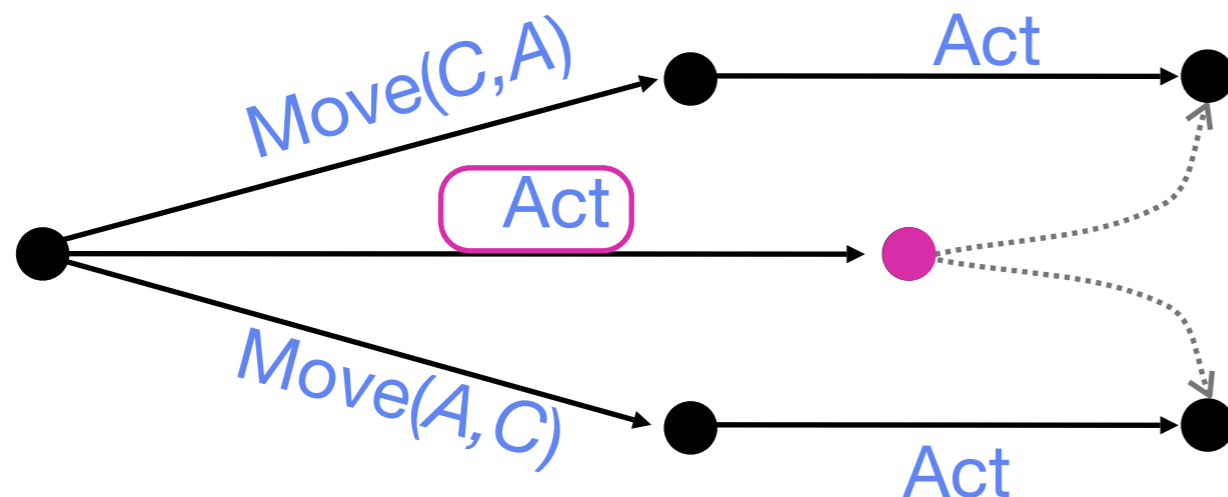
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



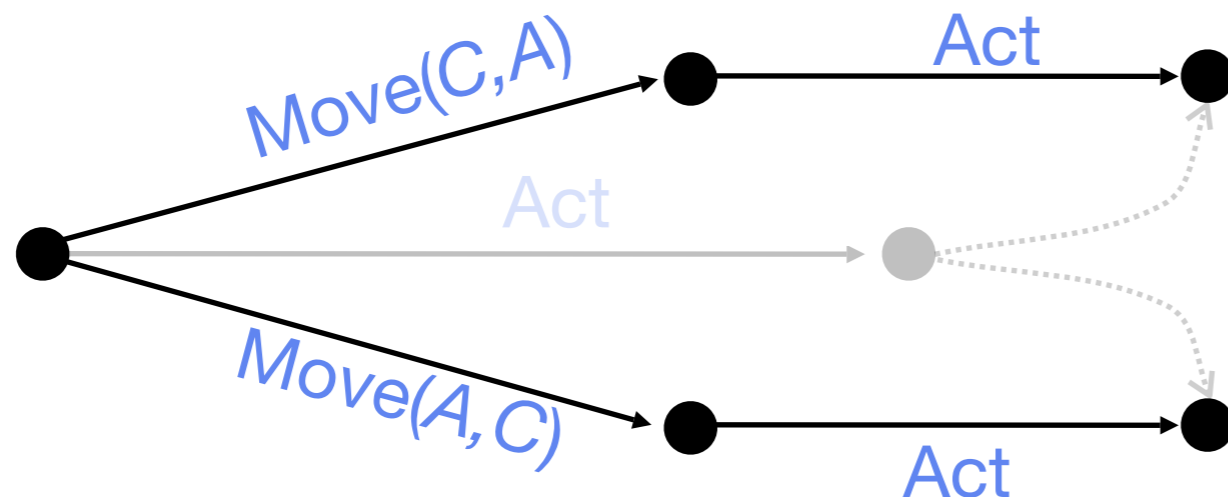
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



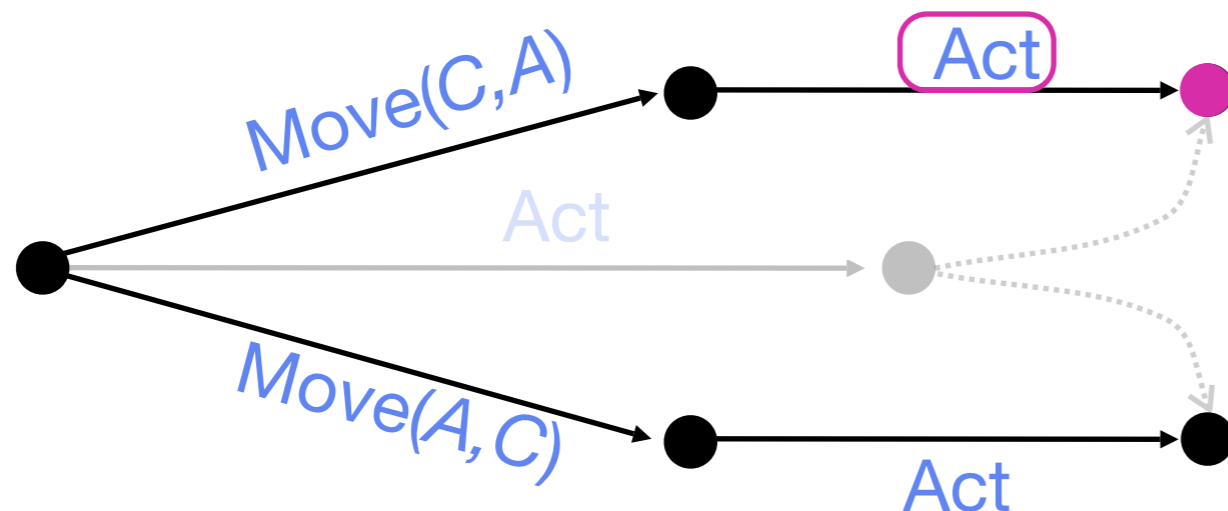
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



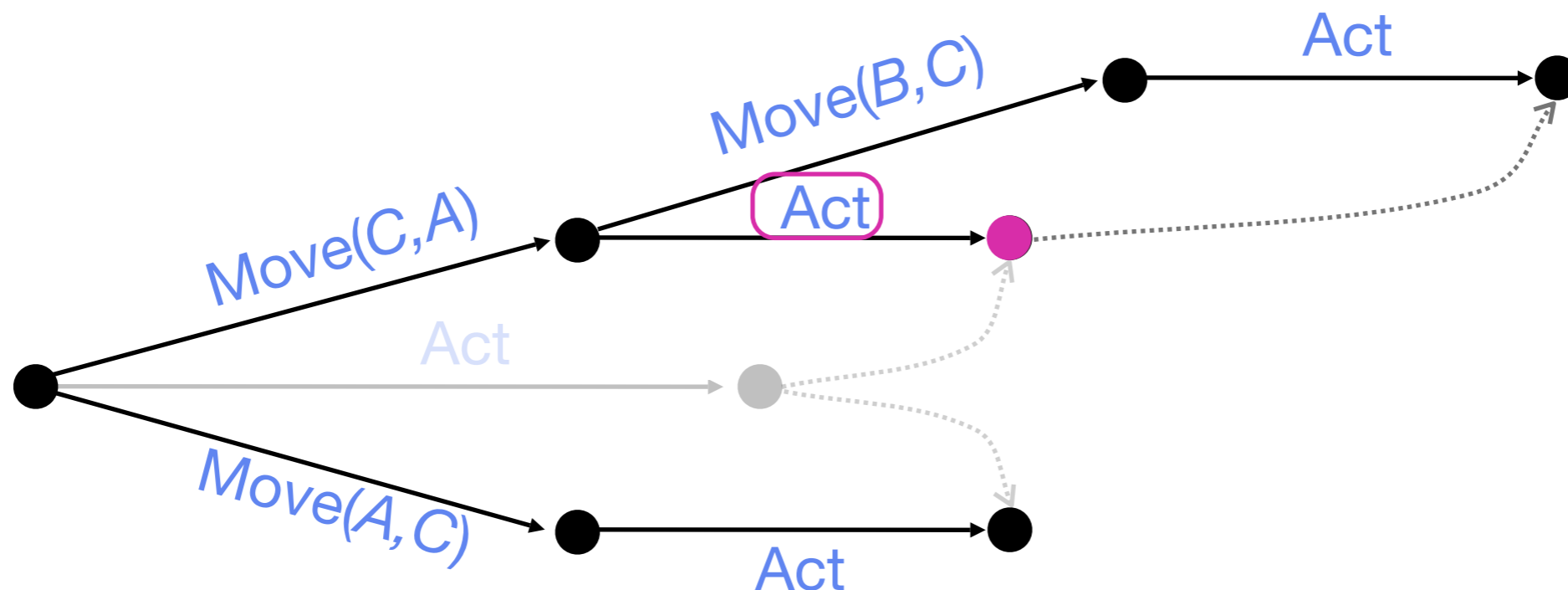
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



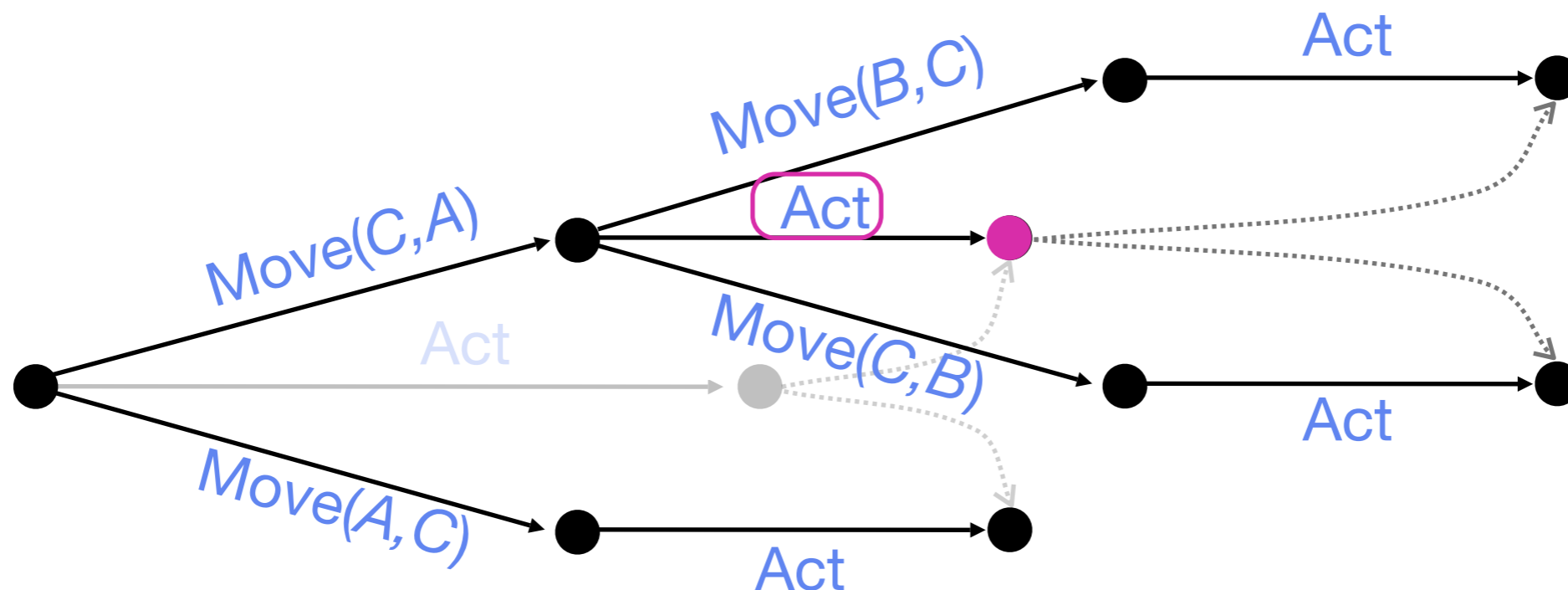
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



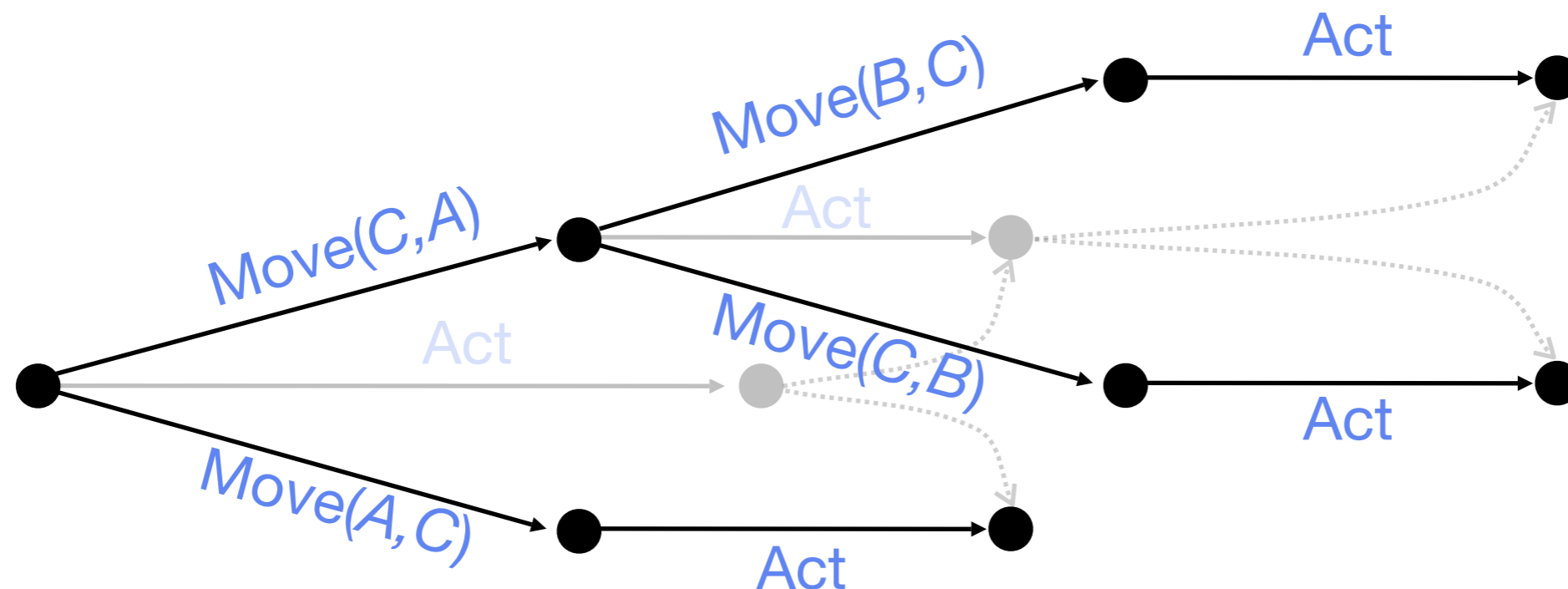
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



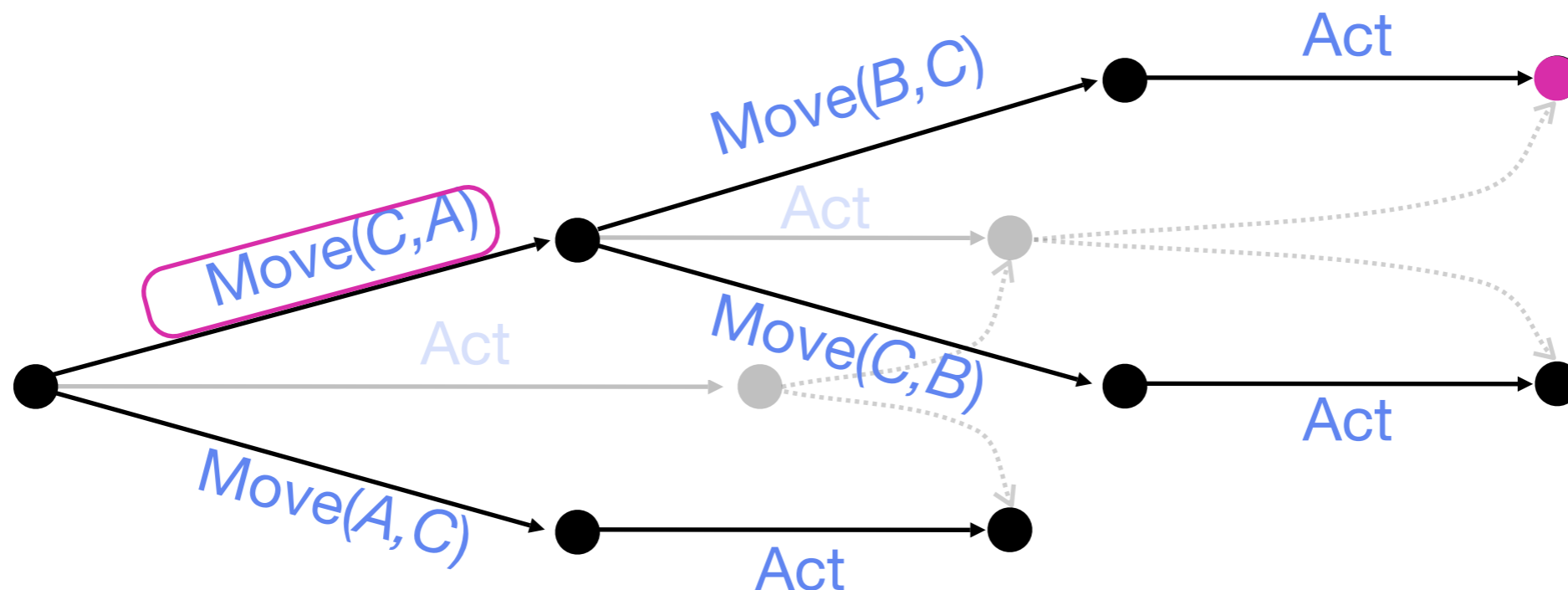
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



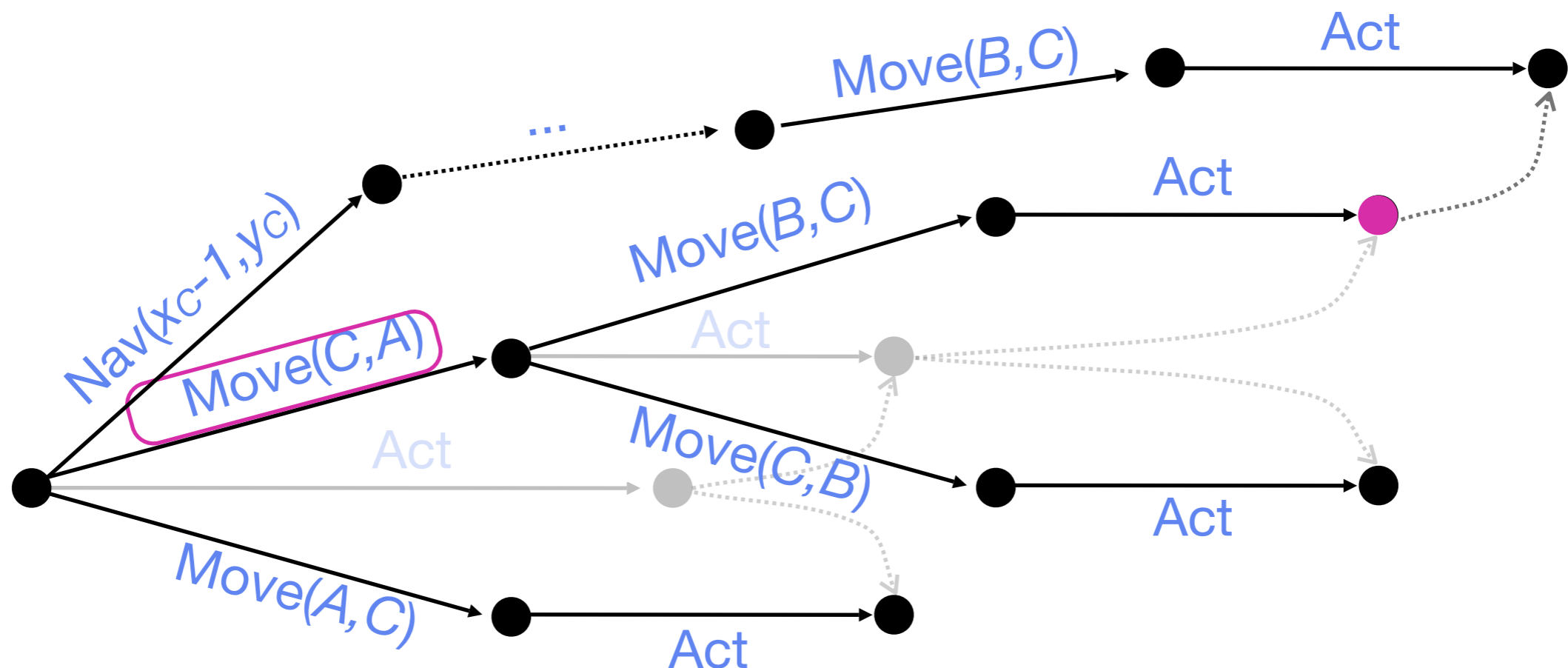
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



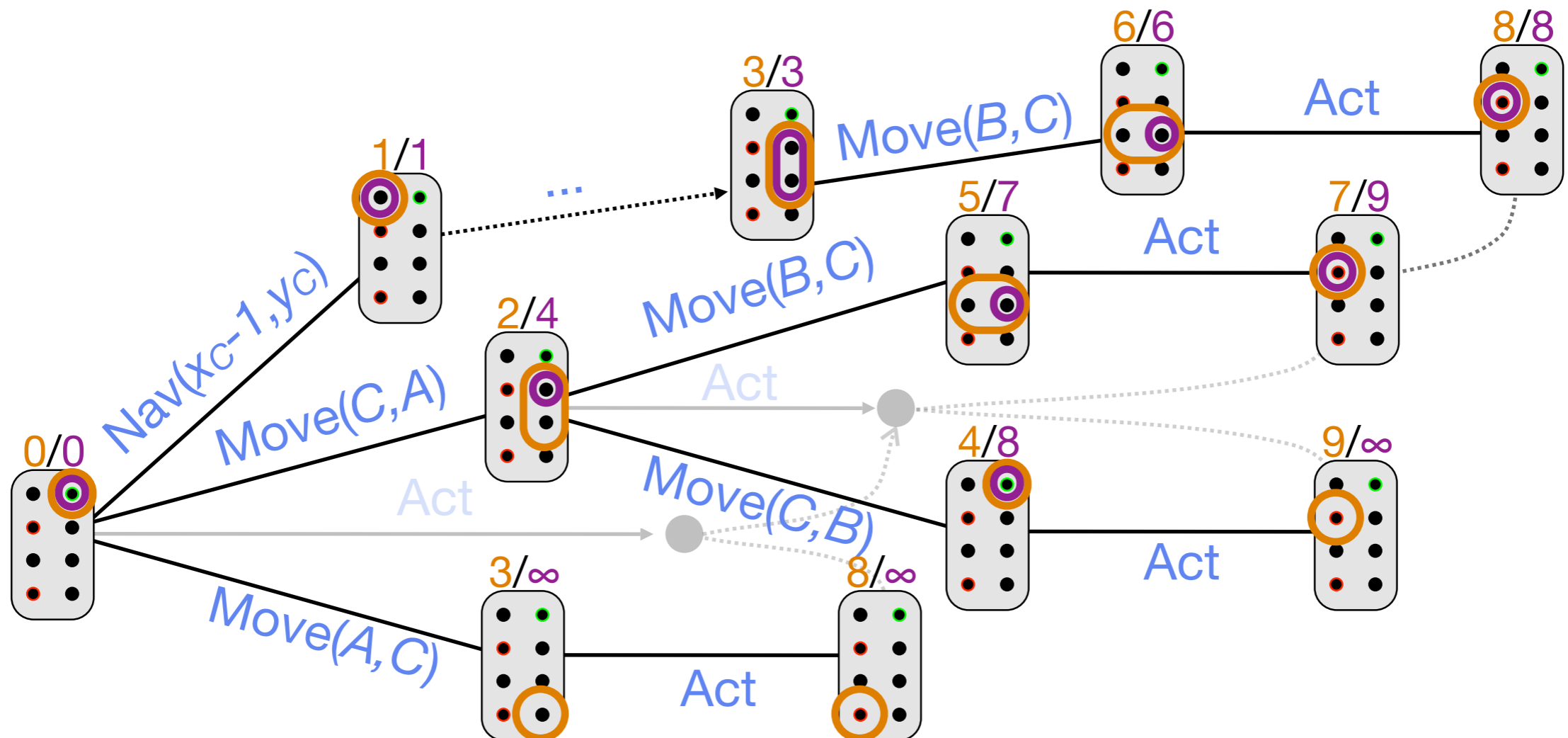
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)



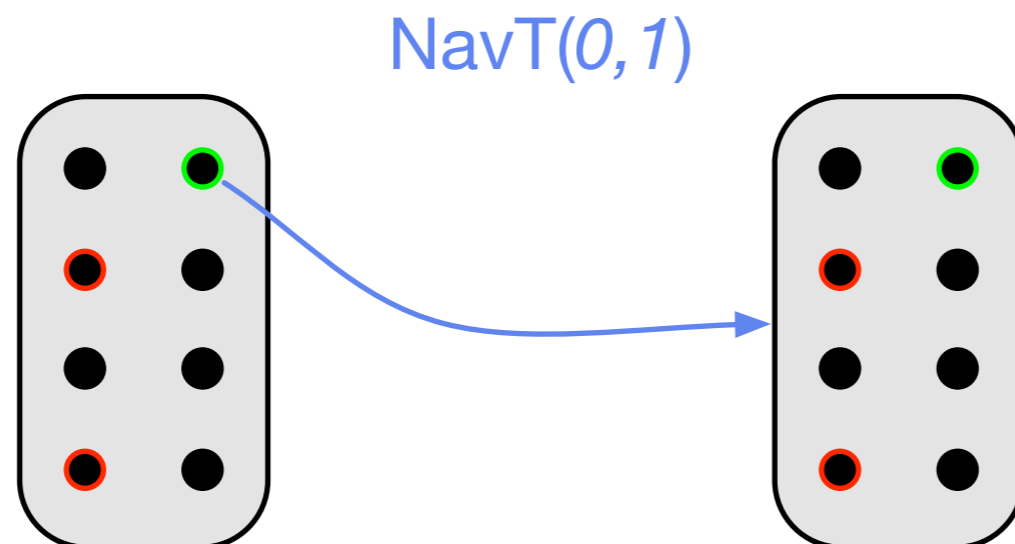
Abstract Lookahead Trees (ALTs)

- ALTs generalize lookahead trees for flat algs (e.g., A^*)
 - Represent a set of potential plans
 - Basic operation: **refine** a plan (replace with all refs. at some HLA)
 - Nodes have **optimistic** & **pessimistic valuations**



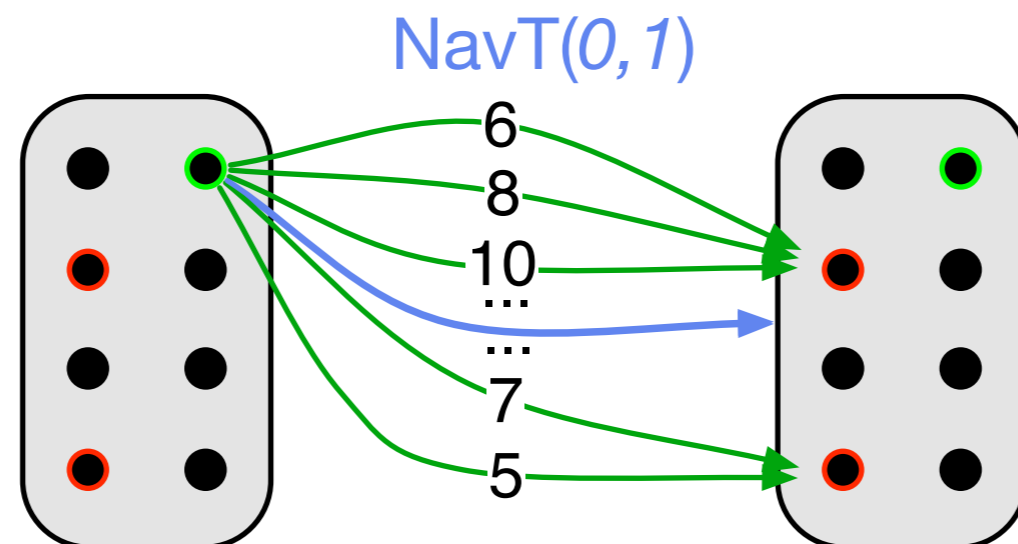
Modeling HLAs

- An HLA is fully characterized by planning problem + hierarchy



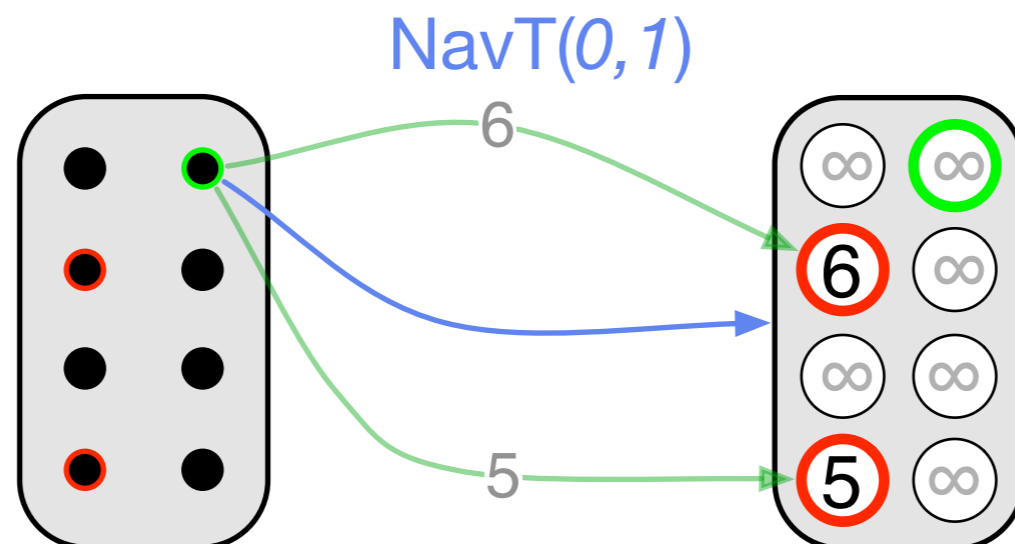
Modeling HLAs

- An HLA is fully characterized by planning problem + hierarchy
 - But without **abstraction**, lose benefits of hierarchy



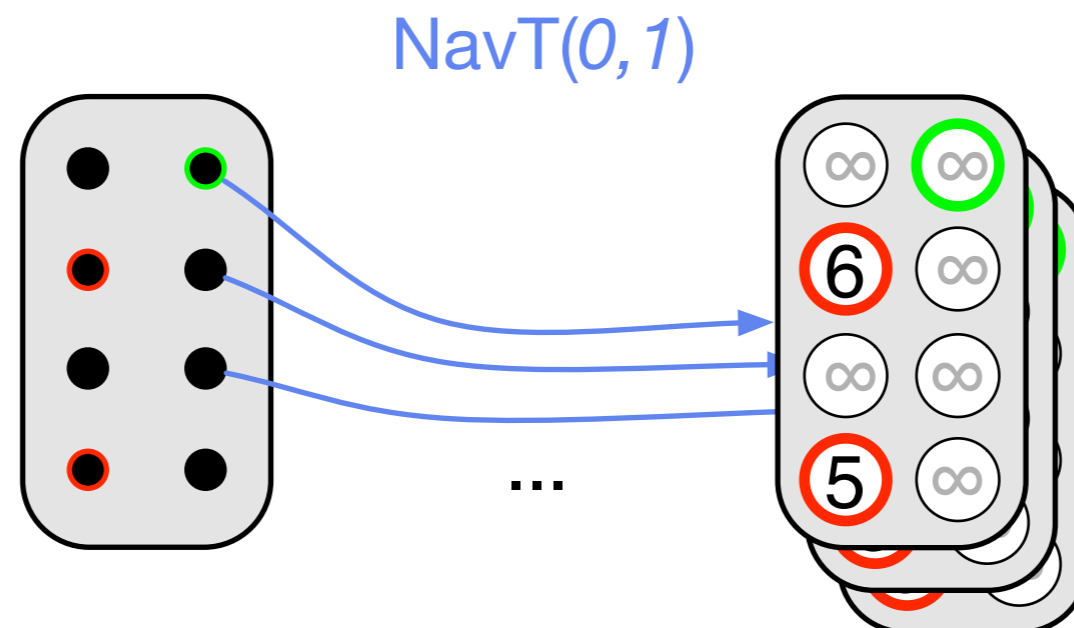
Modeling HLAs

- An HLA is fully characterized by planning problem + hierarchy
 - But without **abstraction**, lose benefits of hierarchy
- Extension of idea from “Angelic Semantics for HLAs” [MRW ‘07]:
 - **Valuation** of HLA h from state s :
 - For each s' , **min cost** of any primitive refinement of h that takes s to s'



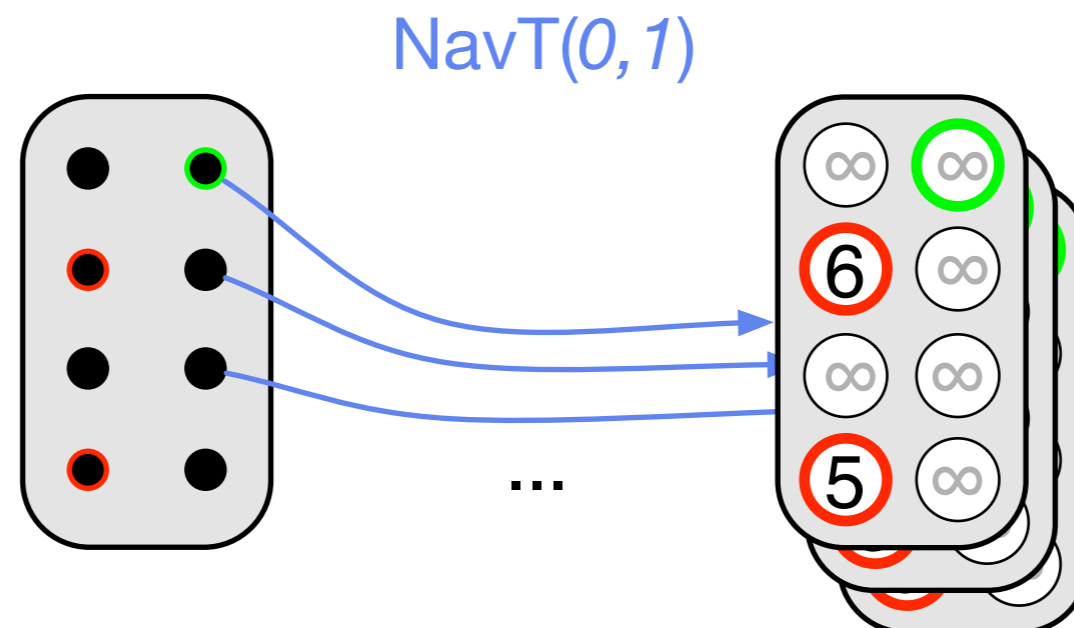
Modeling HLAs

- An HLA is fully characterized by planning problem + hierarchy
 - But without **abstraction**, lose benefits of hierarchy
- Extension of idea from “Angelic Semantics for HLAs” [MRW ‘07]:
 - **Valuation** of HLA h from state s :
 - For each s' , **min cost** of any primitive refinement of h that takes s to s'
 - **Exact description** of h = valuation of h from each s



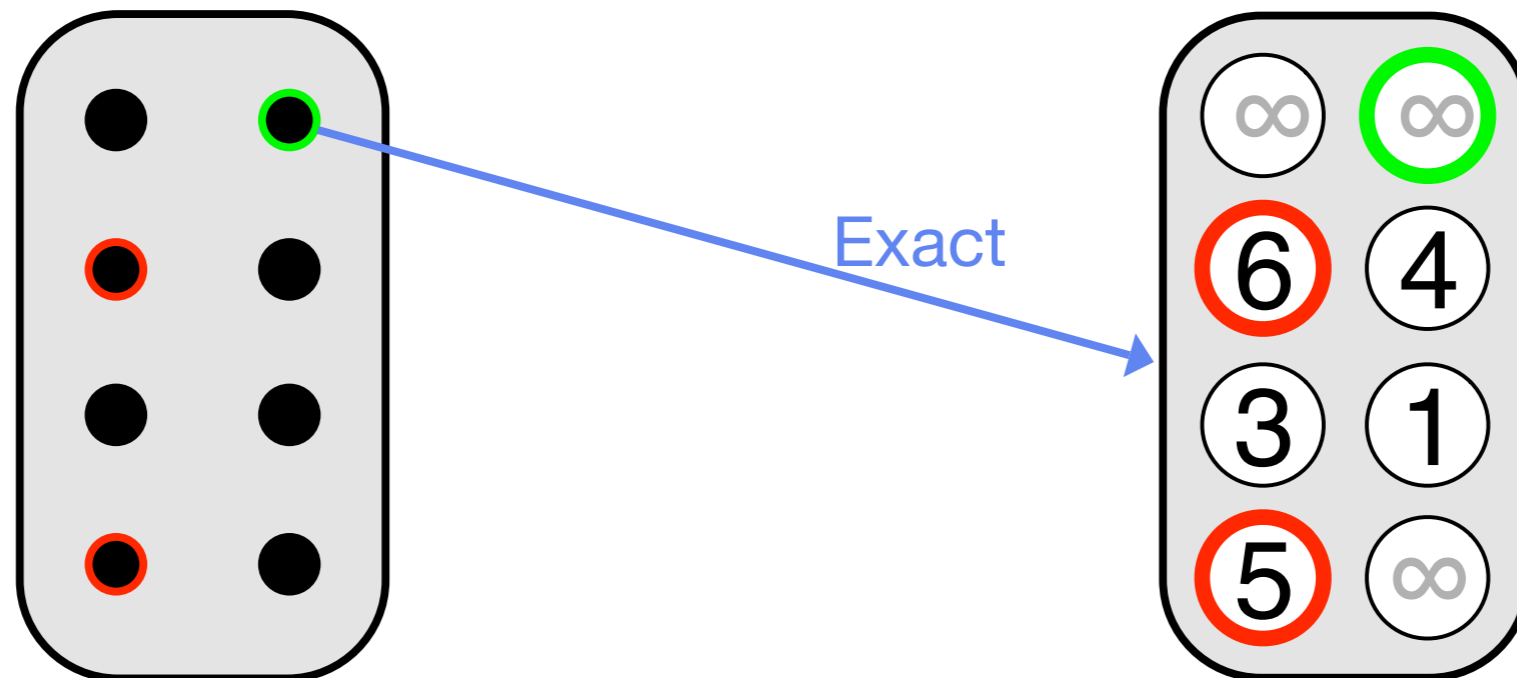
Modeling HLAs

- An HLA is fully characterized by planning problem + hierarchy
 - But without **abstraction**, lose benefits of hierarchy
- Extension of idea from “Angelic Semantics for HLAs” [MRW ‘07]:
 - **Valuation** of HLA h from state s :
 - For each s' , **min cost** of any primitive refinement of h that takes s to s'
 - **Exact description** of h = valuation of h from each s
 - But this description has no compact, efficient representation in general



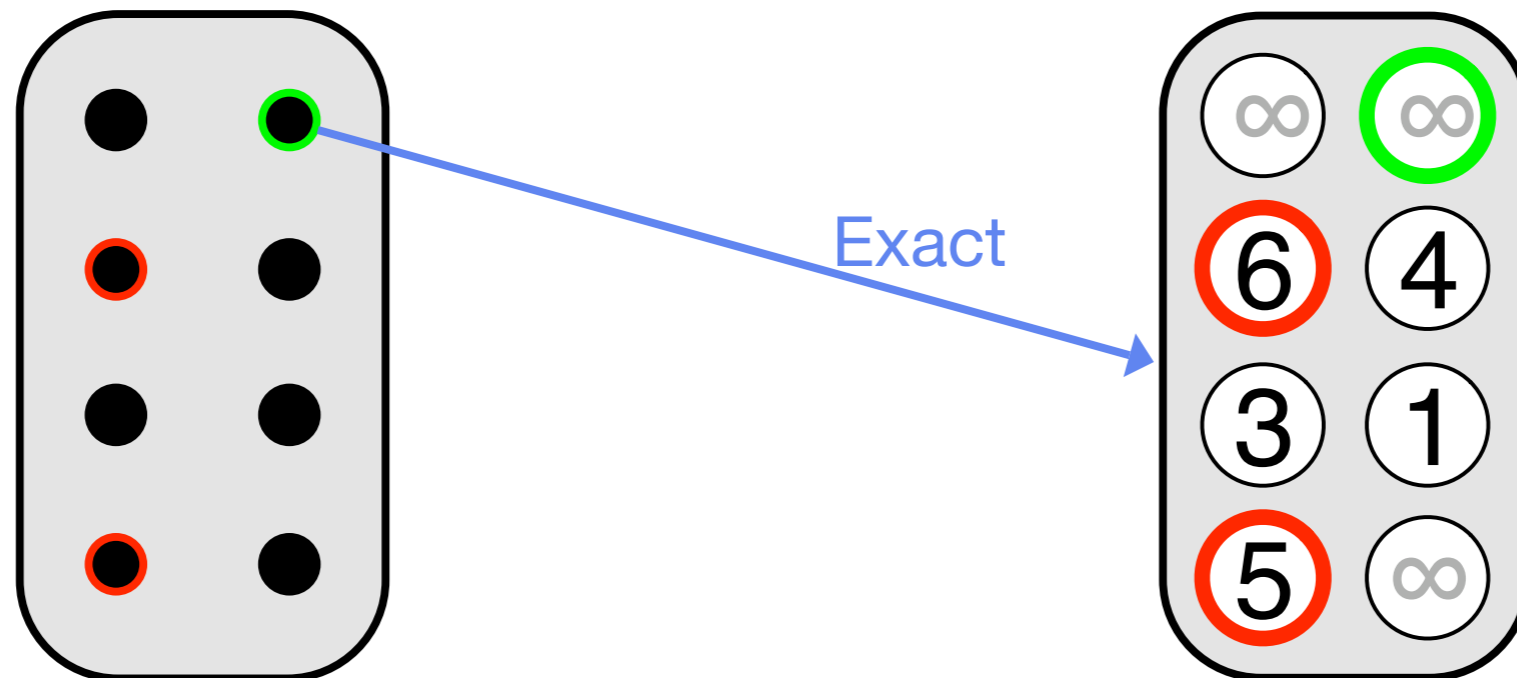
Optimistic and Pessimistic Valuations

- Instead, use **approximate** valuations



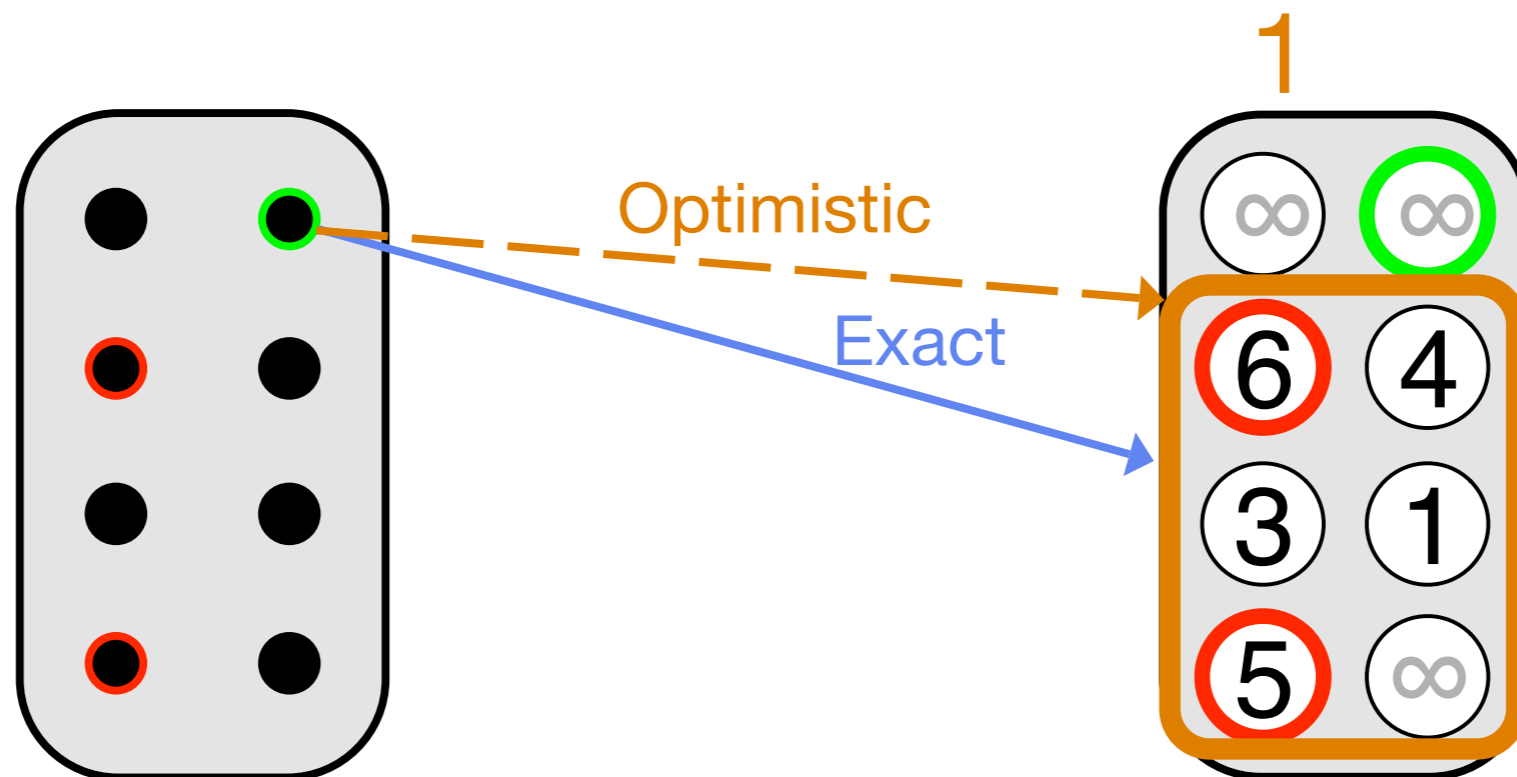
Optimistic and Pessimistic Valuations

- Instead, use **approximate** valuations
- We choose a simple form: **reachable set** + **cost bound** on set



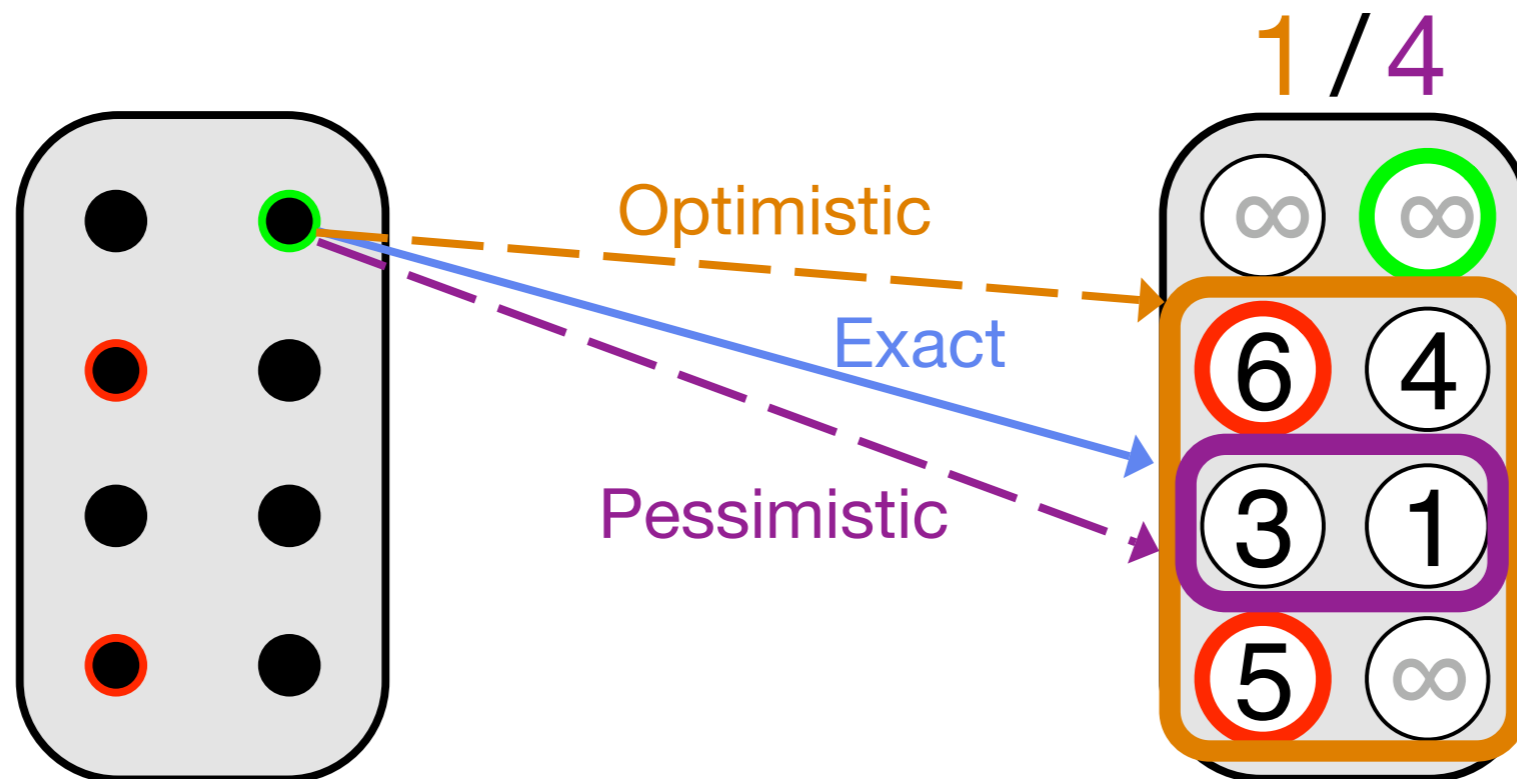
Optimistic and Pessimistic Valuations

- Instead, use **approximate** valuations
- We choose a simple form: **reachable set** + **cost bound** on set
- **Optimistic** valuations **never overestimate** best achievable cost



Optimistic and Pessimistic Valuations

- Instead, use **approximate** valuations
- We choose a simple form: **reachable set** + **cost bound** on set
- **Optimistic** valuations **never overestimate** best achievable cost
- **Pessimistic** valuations **never underestimate** best achievable cost



Representing Descriptions: NCSTRIPS

Representing Descriptions: NCSTRIPS

- Descriptions specify propositions (**possibly**) added/deleted by HLA

Representing Descriptions: NCSTRIPS

- Descriptions specify propositions (**possibly**) added/deleted by HLA

NavT(x_t, y_t)

(Pre: At(x_s, y_s))

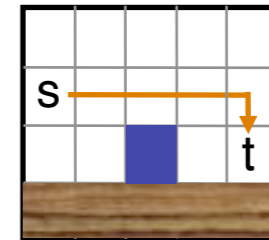
Representing Descriptions: NCSTRIPS

- Descriptions specify propositions (**possibly**) added/deleted by HLA
 - Also include a cost bound

NavT(x_t, y_t)

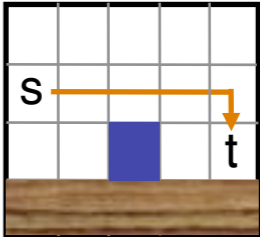
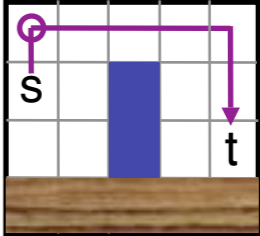
(Pre: $At(x_s, y_s)$)

Opt: $-At(x_s, y_s), +At(x_t, y_t), \tilde{\neg}FaceR, \tilde{\neg}FaceR$
 $cost \geq |x_s - x_t| + |y_s - y_t|$



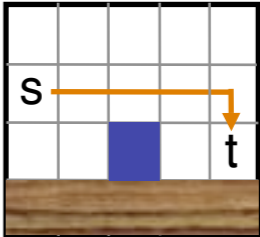
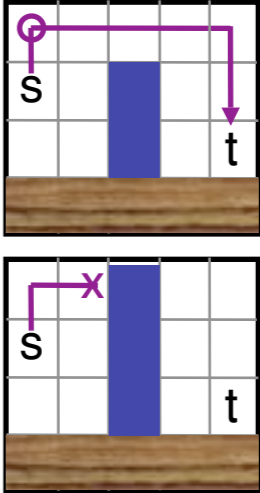
Representing Descriptions: NCSTRIPS

- Descriptions specify propositions (**possibly**) added/deleted by HLA
 - Also include a cost bound
 - Can condition on features of initial state

$NavT(x_t, y_t)$	(Pre: $At(x_s, y_s)$)
<p>Opt: $-At(x_s, y_s), +At(x_t, y_t), \sim FaceR, \tilde{+}FaceR$ $cost \geq x_s - x_t + y_s - y_t$</p>	
<p>Pess: IF $Free(x_t, y_t) \wedge \forall x Free(x, y_{max})$: $-At(x_s, y_s), +At(x_t, y_t), \sim FaceR, \tilde{+}FaceR$ $cost \leq x_s - x_t + 2 y_{max} - y_t - y_s + 1$</p>	

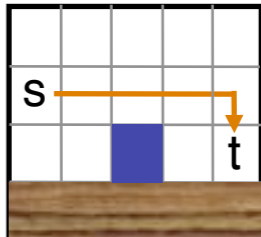
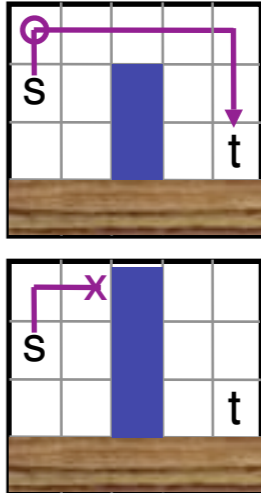
Representing Descriptions: NCSTRIPS

- Descriptions specify propositions (**possibly**) added/deleted by HLA
 - Also include a cost bound
 - Can condition on features of initial state

NavT(x_t, y_t)	(Pre: At(x_s, y_s))
<p>Opt: -At(x_s, y_s), +At(x_t, y_t), $\tilde{\text{FaceR}}$, $\tilde{\text{FaceR}}$ $cost \geq x_s - x_t + y_s - y_t$</p>	
<p>Pess: IF Free(x_t, y_t) \wedge $\forall x$ Free(x, y_{max}) : -At(x_s, y_s), +At(x_t, y_t), $\tilde{\text{FaceR}}$, $\tilde{\text{FaceR}}$ $cost \leq x_s - x_t + 2 y_{max} - y_t - y_s + 1$ ELSE: nil</p>	

Representing Descriptions: NCSTRIPS

- Descriptions specify propositions (**possibly**) added/deleted by HLA
 - Also include a cost bound
 - Can condition on features of initial state
- An simple algorithm **progresses** a valuation (DNF + #) through an NCSTRIPS description to produce next valuation

$NavT(x_t, y_t)$	(Pre: $At(x_s, y_s)$)
<p>Opt: $-At(x_s, y_s), +At(x_t, y_t), \tilde{FaceR}, \tilde{FaceR}$ $cost \geq x_s - x_t + y_s - y_t$</p>	
<p>Pess: IF $Free(x_t, y_t) \wedge \forall x Free(x, y_{max})$: $-At(x_s, y_s), +At(x_t, y_t), \tilde{FaceR}, \tilde{FaceR}$ $cost \leq x_s - x_t + 2 y_{max} - y_t - y_s + 1$</p> <p>ELSE: nil</p>	

Angelic Hierarchical A* (AHA*)

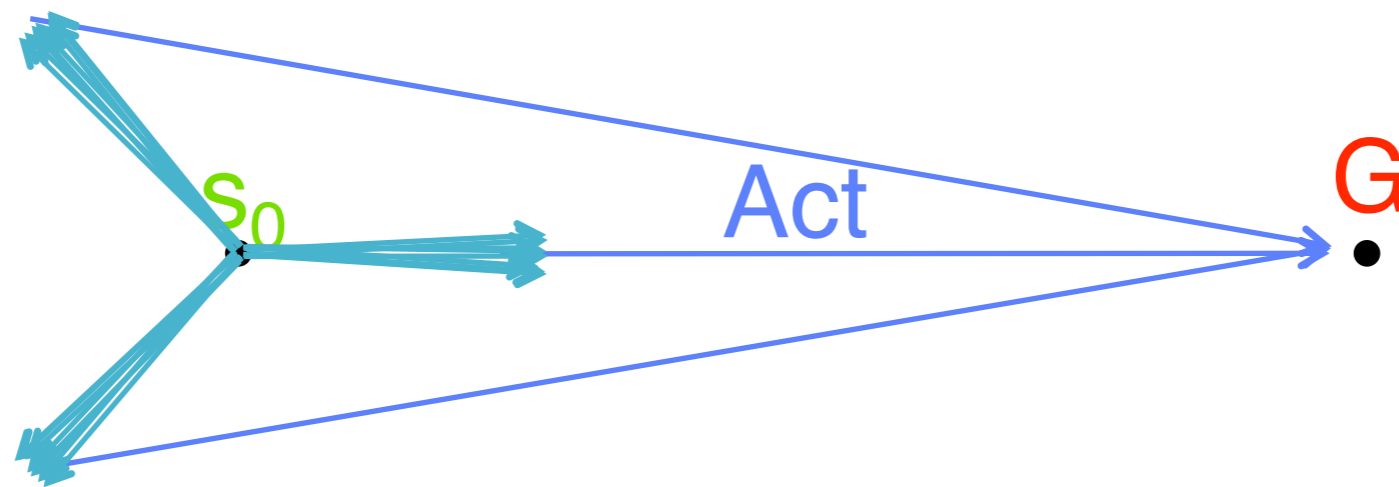
- Construct an ALT with the single plan [Act]
- Loop
 - Select a plan with **minimal optimistic cost** to **G**
 - If primitive, return it
 - Otherwise, **refine** one of its HLAs
 - **Prune dominated** refinements

AHA*: Intuitive Picture



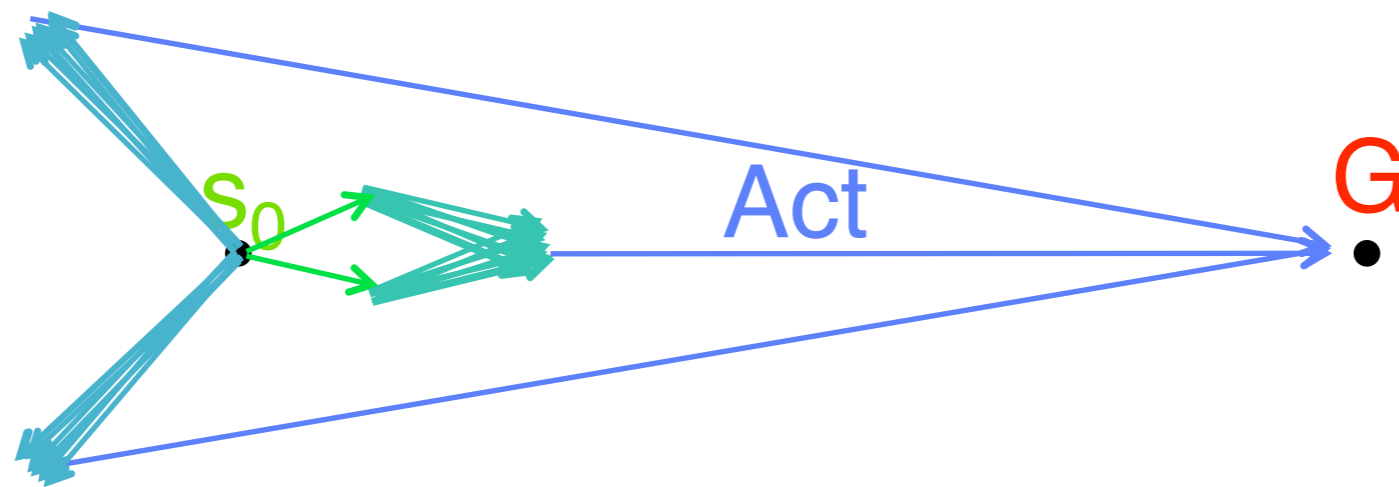
highest-level
primitive

AHA*: Intuitive Picture



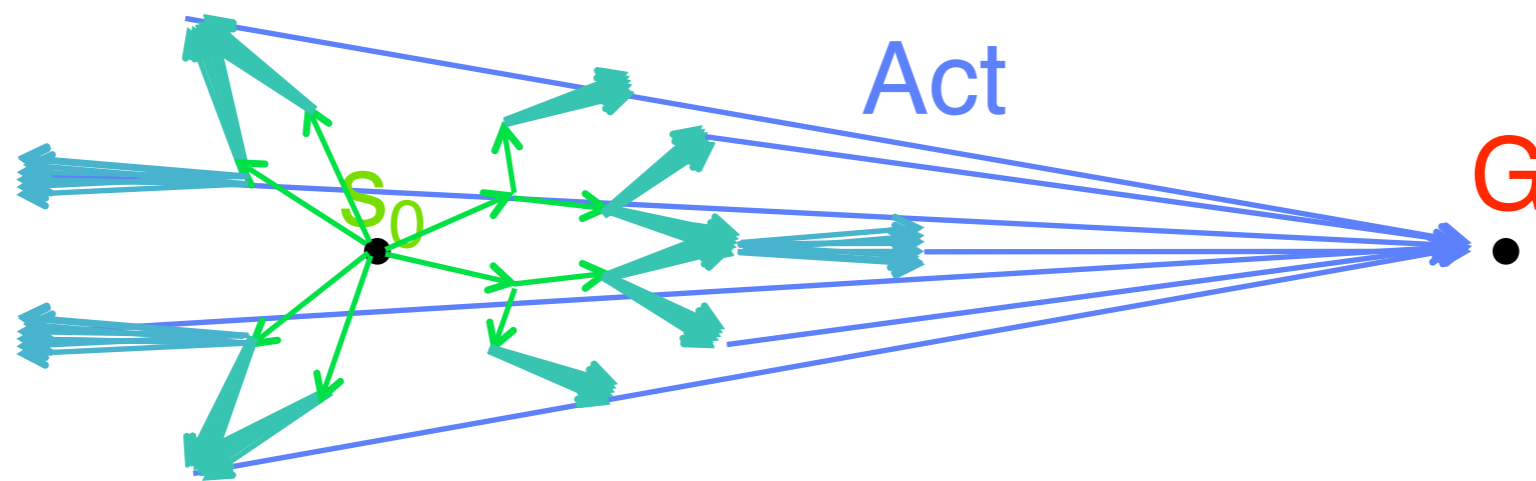
highest-level
primitive

AHA*: Intuitive Picture



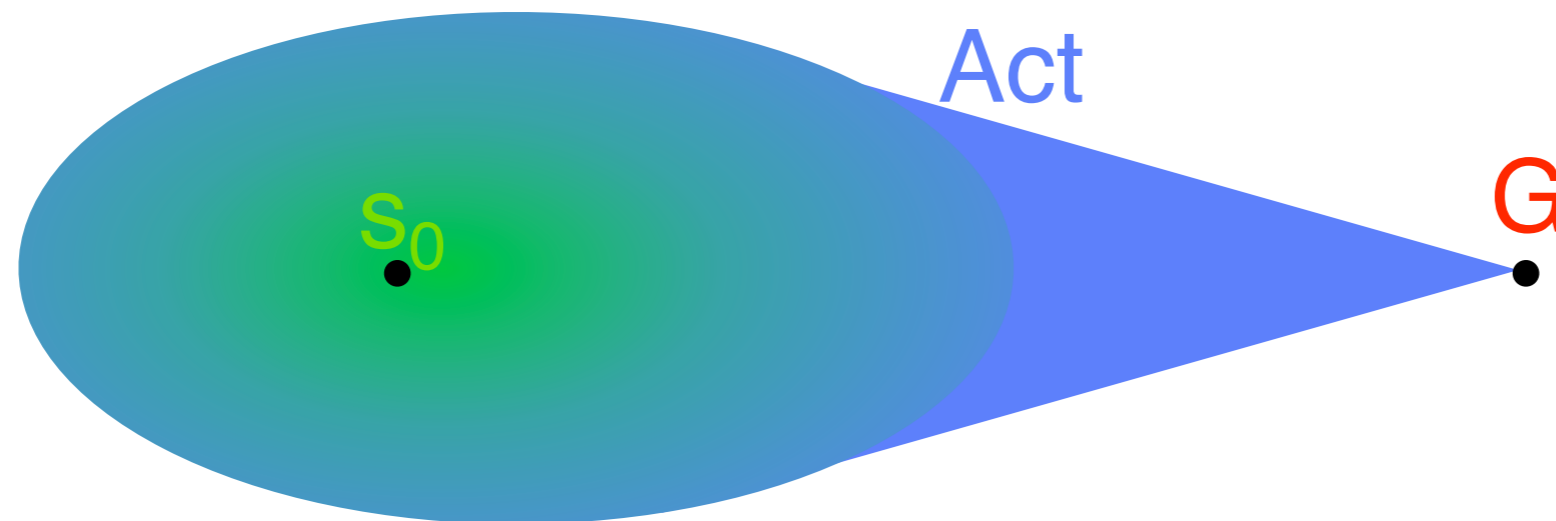
highest-level
primitive

AHA*: Intuitive Picture



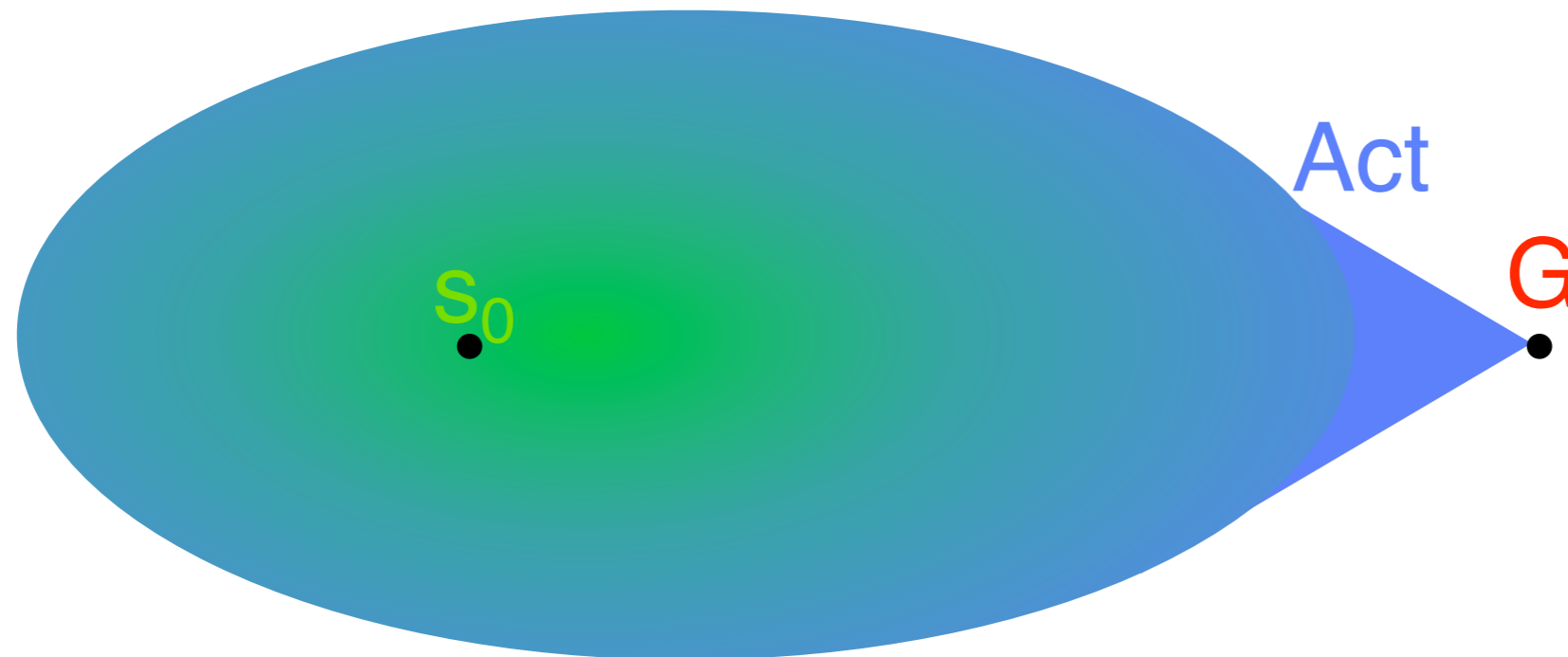
highest-level
primitive

AHA*: Intuitive Picture



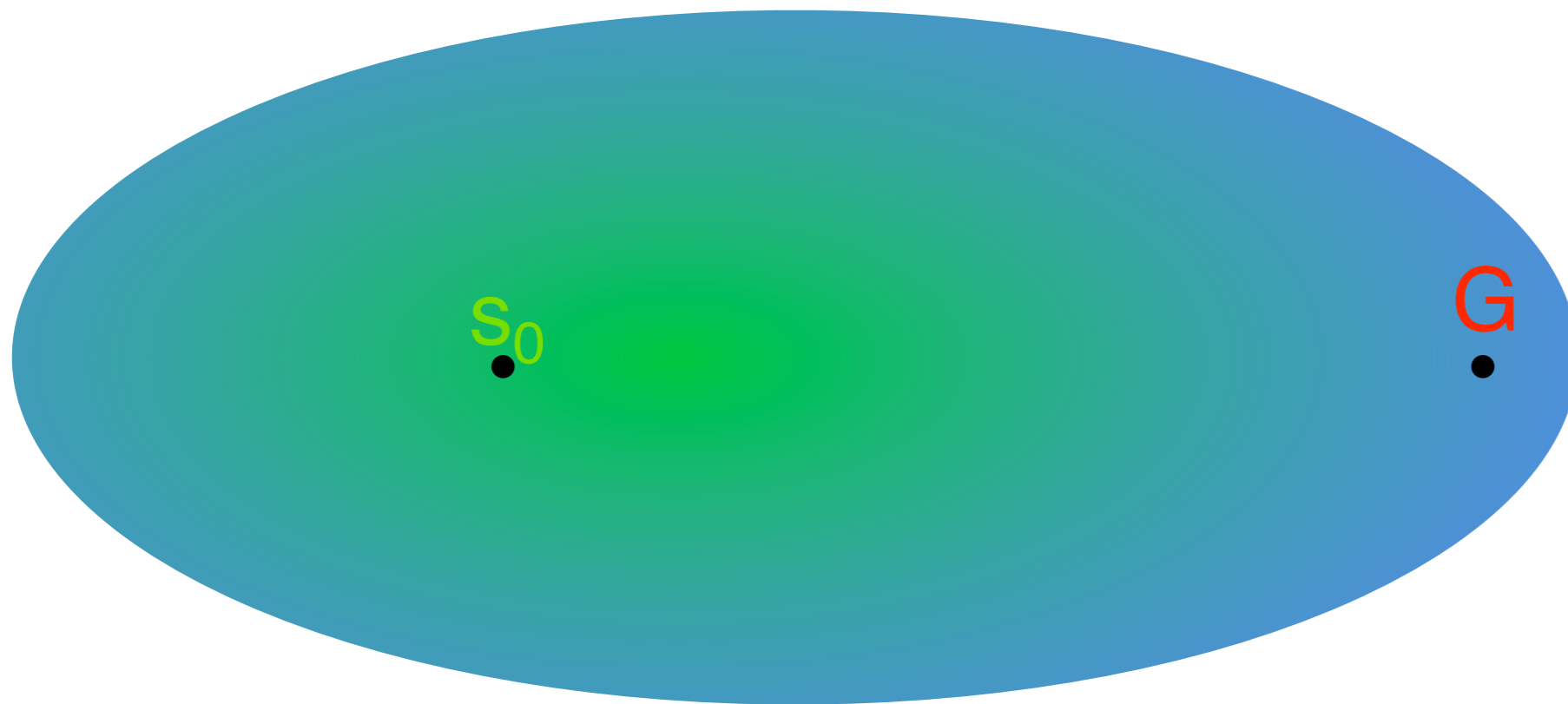
highest-level
primitive

AHA*: Intuitive Picture



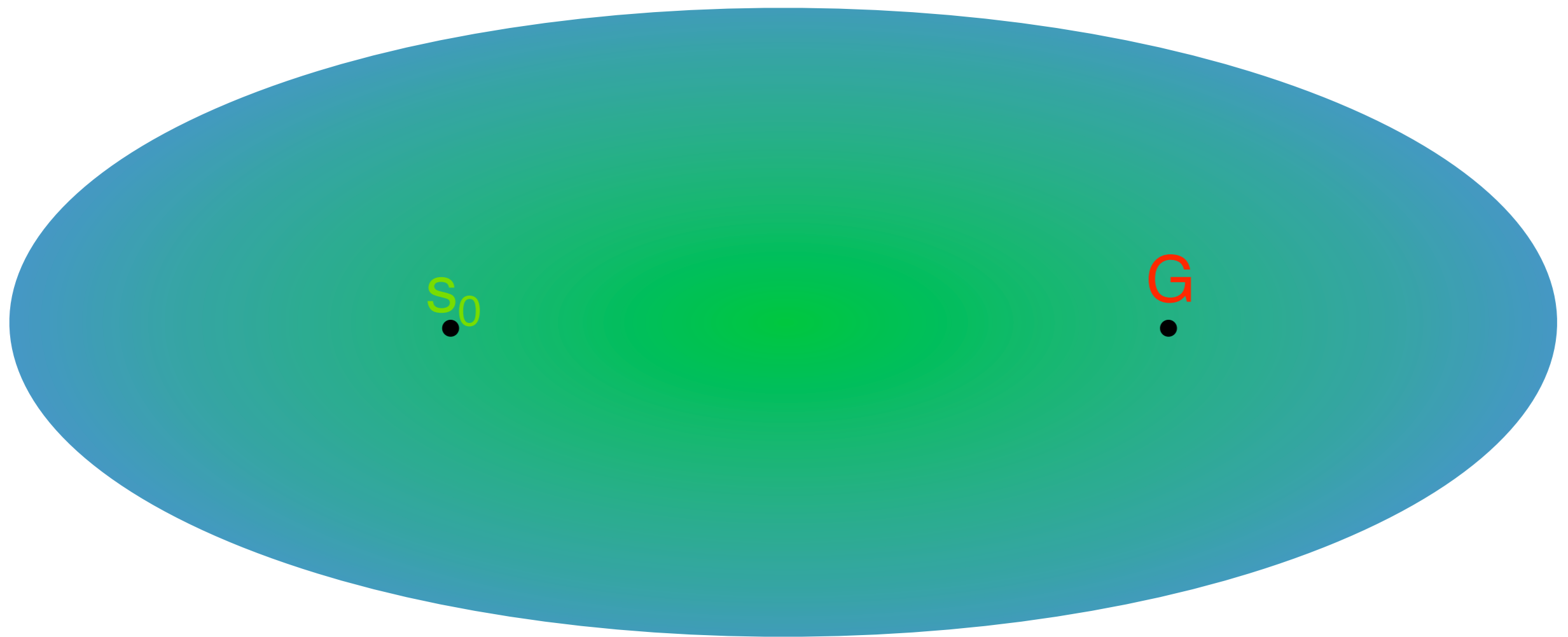
highest-level
primitive

AHA*: Intuitive Picture



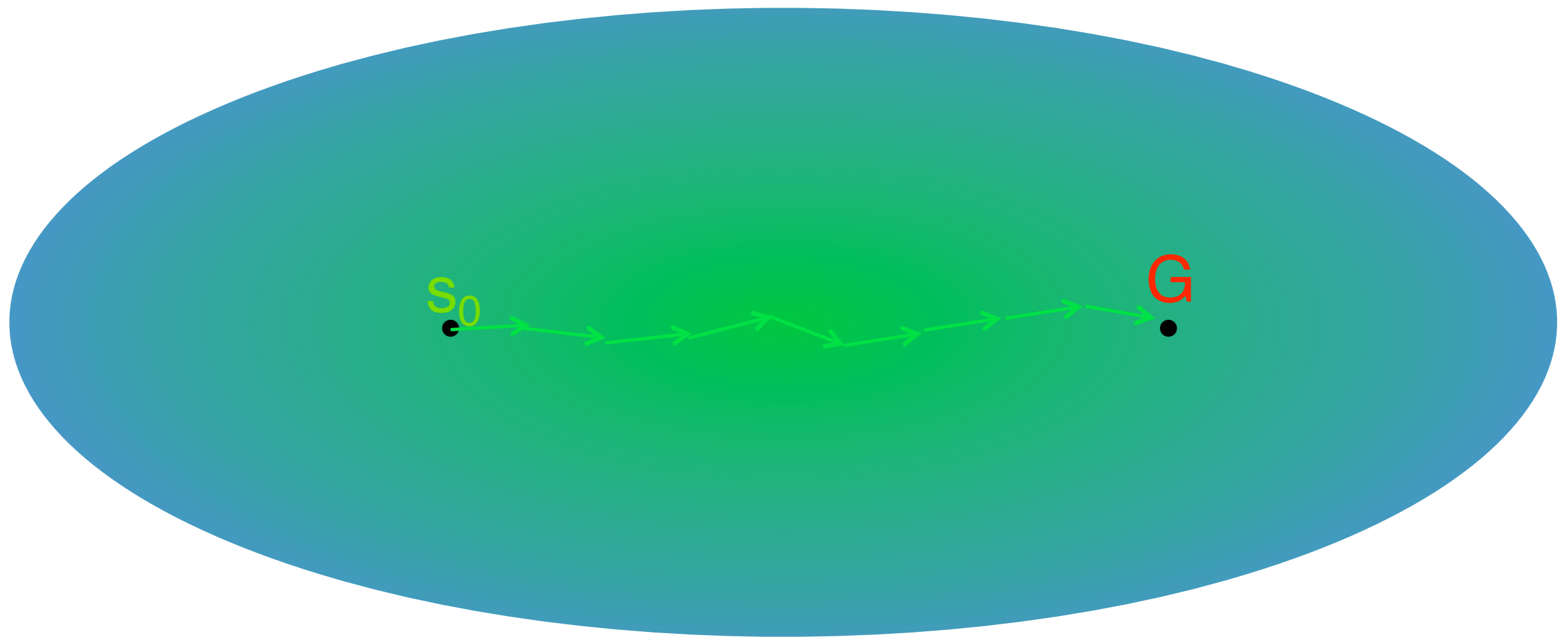
highest-level
primitive

AHA*: Intuitive Picture



highest-level
primitive

AHA*: Intuitive Picture



highest-level
primitive

Analysis of AHA*

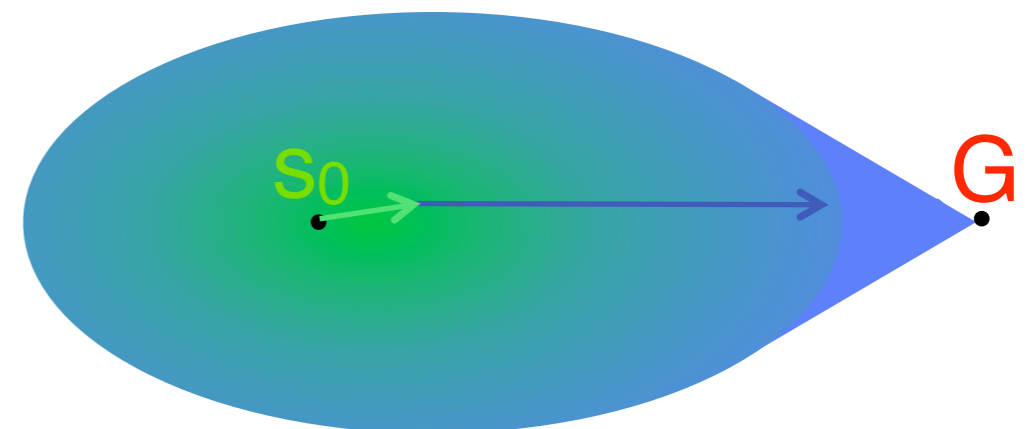
- AHA* is **hierarchically optimal** (HO)
 - Optimistic valuation → **admissible heuristic**
 - Pruning never rules out all HO plans
- Better descriptions lead to lower runtime
 - optimistic → directed search
 - pessimistic → pruning (refine HO plans **w/o backtracking**)
- Reduces to A* given “flat” hierarchy: **Act** → [**Prim**, **Act**]

Solution Length	A*	AHA*
7	0.9	0.6
16	10	4.7
25	40	11
37	550	30
44	> 10000	68

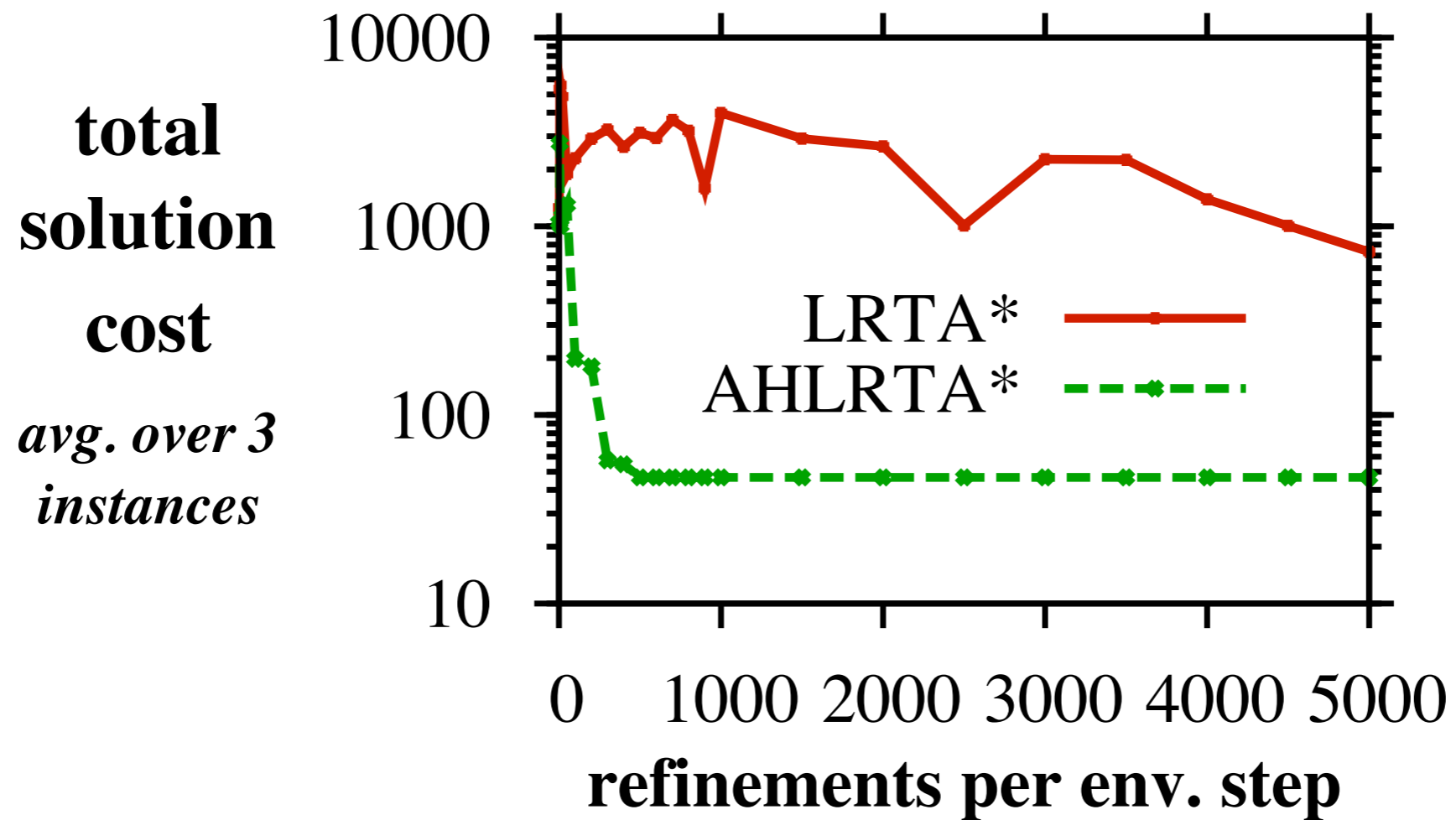
runtimes in seconds on five warehouse world instances of increasing solution length

Online Search

- Situated agents must cope with passage of time
 - **offline** planning rarely feasible
 - common alternative: **real-time search**
- Korf's Learning Real-Time A* (LRTA*):
 - Combines limited **lookahead** + **learning**
 - Always reaches goal, converges to optimal
- **Angelic Hierarchical LRTA*** (AHLRTA*)
 - Performs **hierarchical lookahead**
 - Shares LRTA*'s guarantees
 - Reduces to LRTA* given "flat" hierarchy



Online Results



1 AHLRTA* refinement \approx 5 LRTA* refinements

Summary

Model-based hierarchical planning is theoretically interesting, shows promising empirical performance

