

**COSC76/276 Artificial Intelligence**

**Fall 2022**

**Informed Search & Deriving Heuristics**

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# Reminders

- Thank you for participating in Slack and joining office hours!
  - Please continue to feel free to reach out
- SA-2 due today Oct 1<sup>st</sup> 11:59pm ET (completion based)
- PA-1 due Sep 29 at 11:59pm ET (1 day blanket extension for everyone\_
- PA-2 will be out today. Due Oct 7<sup>th</sup> at 11:59 ET.

# Recap: Search problem

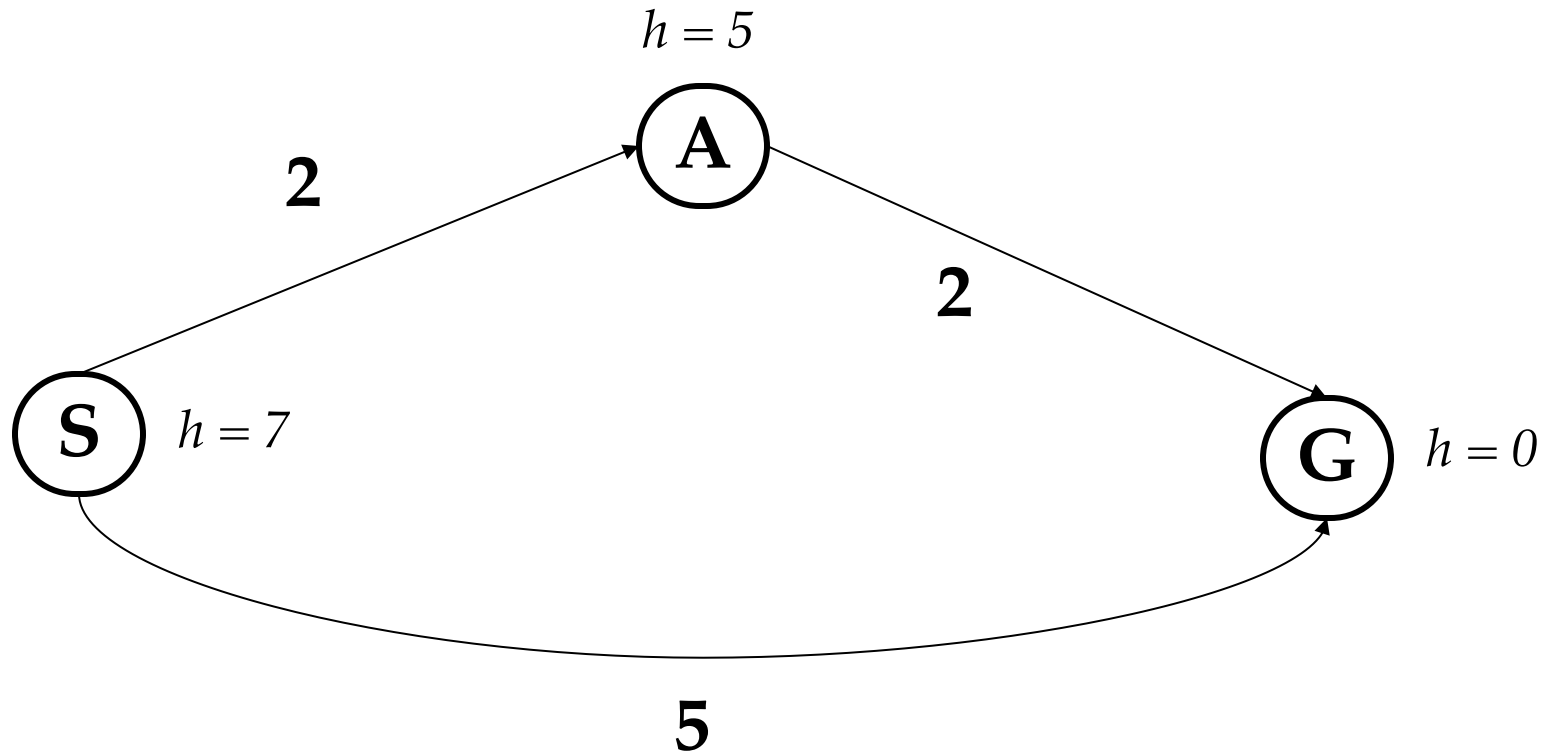
- Uninformed search
  - BFS, DFS, variants, and their properties
  - Do not consider any information about the goal
  - Uninformed search with cost: UCS
- Informed search
  - Heuristic
  - Greedy and  $A^*$

# A\* search

- Combine UCS and greedy
- Evaluation function (cost + heuristic)
  - $f(n) = g(n) + h(n)$

Goes to the goal but backtracks as needed.

# Optimality of A\*



- Which solution will be found by A\*?
  - S,G, with cost 5 instead of S,A,G with cost 4, because of overestimated heuristic

# Optimality of A\* for tree-search

- A\* *tree* search produces shortest paths if the heuristic is **optimistic** (also called **admissible**): it underestimates cost of path to goal from any node on the tree.

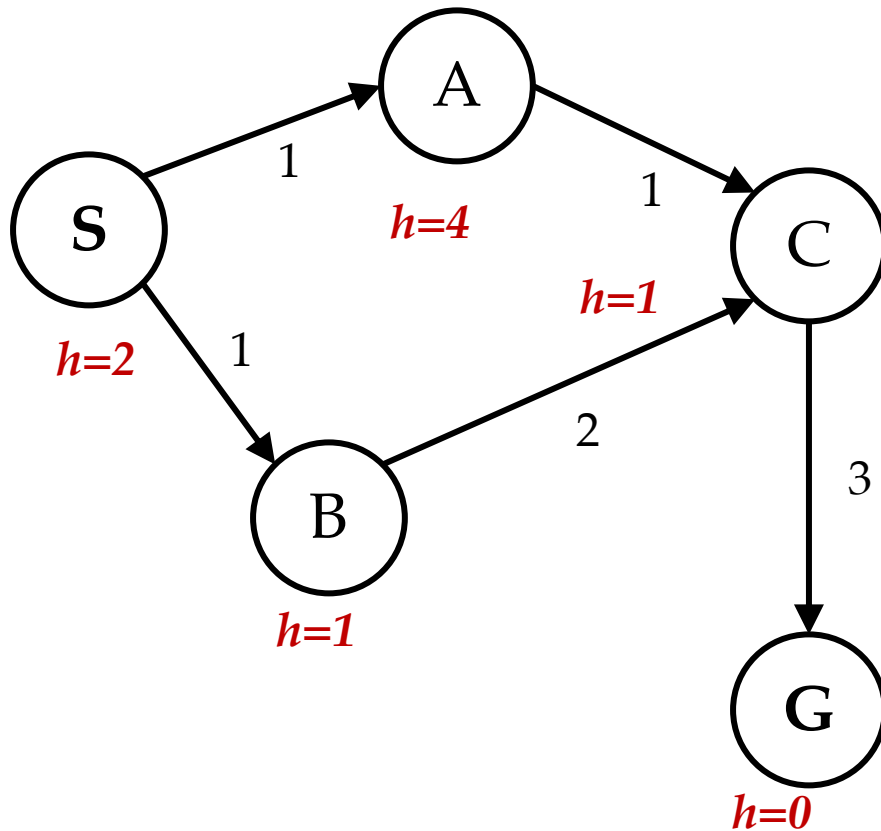
$$0 \leq h(n) \leq h^*(n)$$

where  $h^*(n)$  is the true cost from node  $n$ .

Admissible (optimistic) heuristics slow down bad plans but never outweigh true costs

Can you think of an example?

# Optimality of A\* (graph-search)



- **Priority queue:**
- S2
- (pop S2, push B2, A5)
- B2 A5
- (pop B2, push C4)
- C4 A5
- (pop C4, push G6)
- A5, G6
- (pop A5, C3 not pushed as already explored)
- G6
- (pop G6)

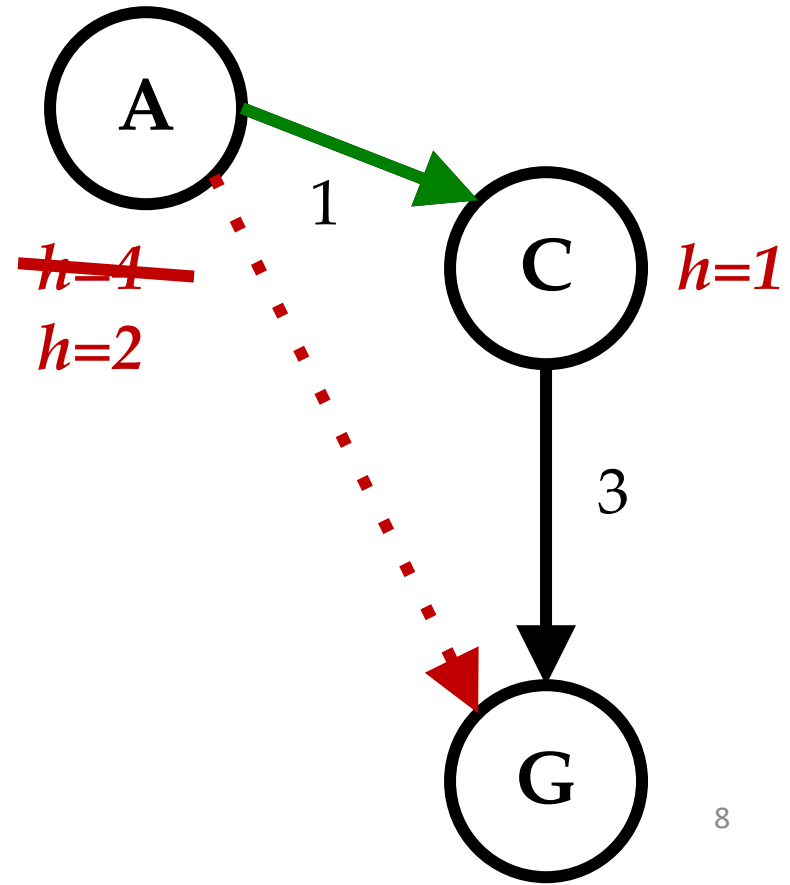
Solution: S,B,C,G with cost 6,  
instead of S,A,C,G with cost 5

```
for each child of current_state:
    if child not in explored:
        add child to explored
        pack child state into a node, with backpointer to current_node
        add the node to the frontier
    else if child is in frontier with higher f
        replace that frontier node with child node
```

# Optimality of A\* for graph-search

- A\* search on a graph produces shortest paths if the heuristic is **consistent** (also called **monotone**).

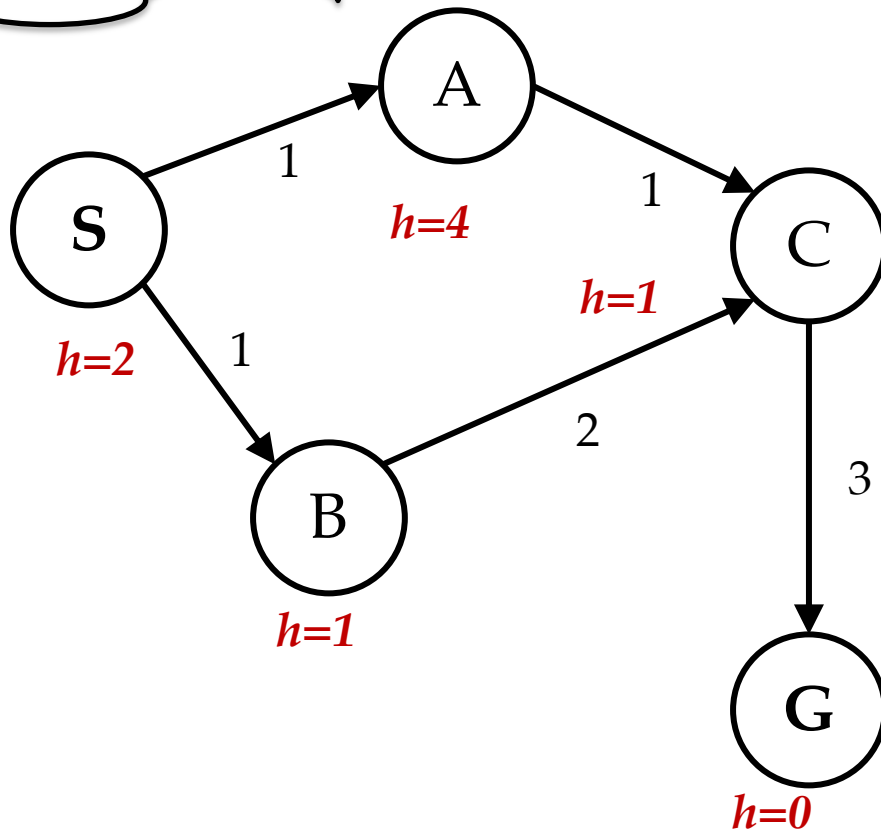
$$h(n) \leq c(n, n') + h(n')$$





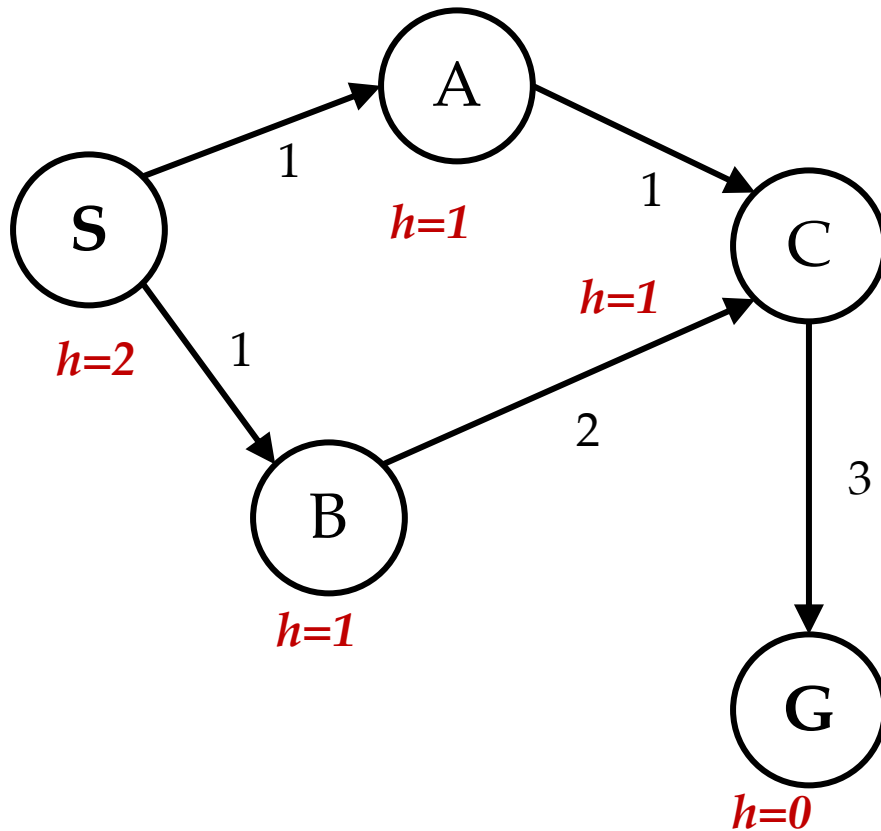
Discussion. Which heuristic value and how should be changed?

# Optimality of A\* (graph-search)



```
for each child of current_state:
  if child not in explored:
    add child to explored
    pack child state into a node, with backpointer to current_node
    add the node to the frontier
  else if child is in frontier with higher f
    replace that frontier node with child node
```

# Optimality of A\* (graph-search)



With consistent heuristic

- **Priority queue:**
- S2
- (pop S2, push B2, A2)
- B2 A2
- (pop B2, push C4)
- A2 C4
- (pop A2, push C3, C4 marked as to be removed)
- C3, C4
- (pop C3, push G5)
- G5
- (pop G5)

Solution: S,A,C,G with cost 5,

```
for each child of current_state:
    if child not in explored:
        add child to explored
        pack child state into a node, with backpointer to current_node
        add the node to the frontier
    else if child is in frontier with higher f
        replace that frontier node with child node
```

# Properties of A\*

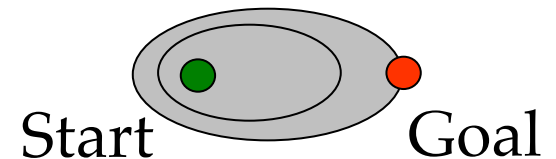
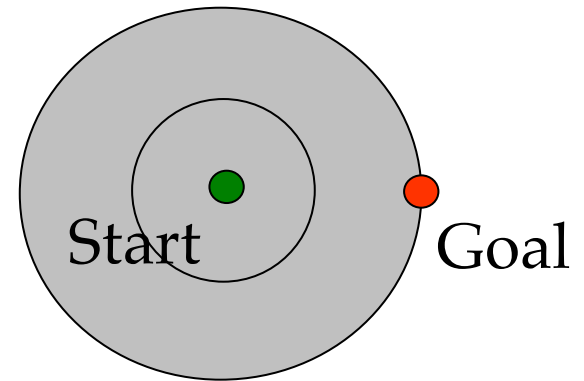
- Space - Keeps all nodes in memory
- Optimal – Yes\*
- Complete – Yes
- Time –  $b^d$  Worst case scenario

# Properties of A\*

- Space - Keeps all nodes in memory
- Optimal – Yes\*
- Complete – Yes
- Time – In practice, depends on goodness of  $h$ .  
A good heuristic reduces the effective branching factor:  $b$  times the relative error in  $((h^* - h)/h^*)$

# Comparison with UCS and A\*

- Uniform-cost expands equally in all “directions”
- A\* expands mainly toward the goal, but with some limits to ensure optimality



# A\* applications

- Video games
- Pathing / routing problems
- Resource planning problems
- Robot motion planning
- ...

# Summary

- To include cost: Uniform cost search
  - Use priority queue
- To include information about the goal: use heuristic
  - Greedy search
  - A\*: combines both UCS and greedy
- Heuristic should be admissible and consistent to guarantee optimality for tree and graph search, respectively

# Next

- Deriving heuristics



# Deriving heuristics

- Most of the work in solving hard search problems optimally is in coming up with admissible heuristics
- Often, admissible heuristics are solutions to *relaxed problems*, where the problem has fewer restrictions.
  - The relaxed problem may have better solutions as there might be shortcuts

# Example: 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

- State?
- Action?
- Goal test?
- Action cost?

# of distinct states  
reachable?

# Example: 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

- State: integer locations of tiles
- Action: move blank tile in an adjacent location in the grid
- Goal test: check with given goal state
- Action cost: 1 per move

# Example: 8-puzzle

7	2	4
5		6
8	3	1

**Start State**

1	2	3
4	5	6
7	8	

**Goal State**

- State: integer locations of tiles
- Action: move blank tile in an adjacent location in the grid
- Goal test: check with given goal state
- Action cost: 1 per move
- Average branching factor?

Can you think of a heuristic?

# Example: 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

- Heuristics?

# Example: 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

- $h_1(n)$  = number of misplaced tiles
  - A tile can “teleport” from square A to square B.

# Example: 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

- $h1(n)$  = number of misplaced tiles
  - A tile can “teleport” from square A to square B.
- $h2(n)$  = total Manhattan distance
  - A tile can move from square A to square B if A is adjacent to B regardless of whether it is blank or not.

# Example: 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

- $h1(n) = 6$
- $h2(n) = 4+0+3+3+1+0+2+1$

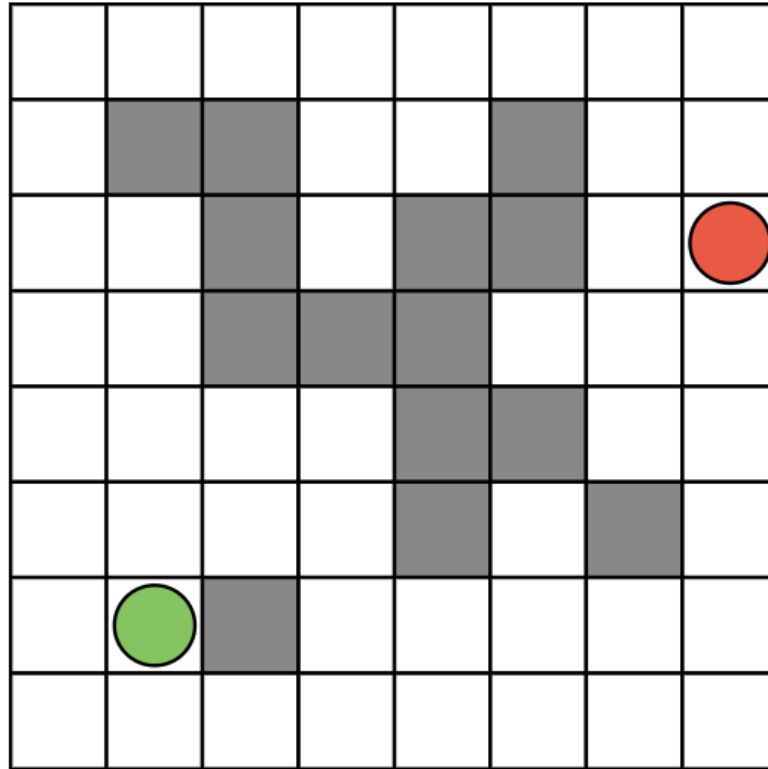


# Dominance of heuristics

- If  $h_2(n) \geq h_1(n)$  for all  $n$  (both admissible) then  $h_2$  dominates  $h_1$  and is better for search
- If the relation between heuristics is not known, given any admissible heuristics, take the max

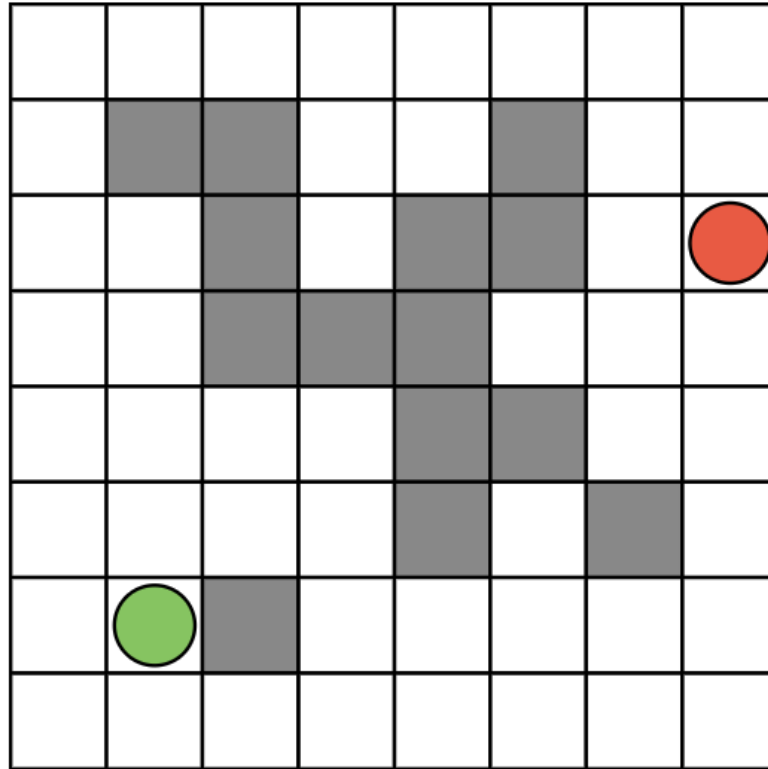
Heuristics? Write  
down on paper

# PA2: Mazeworld problem



Robot can move N, S, E, W onto empty squares.

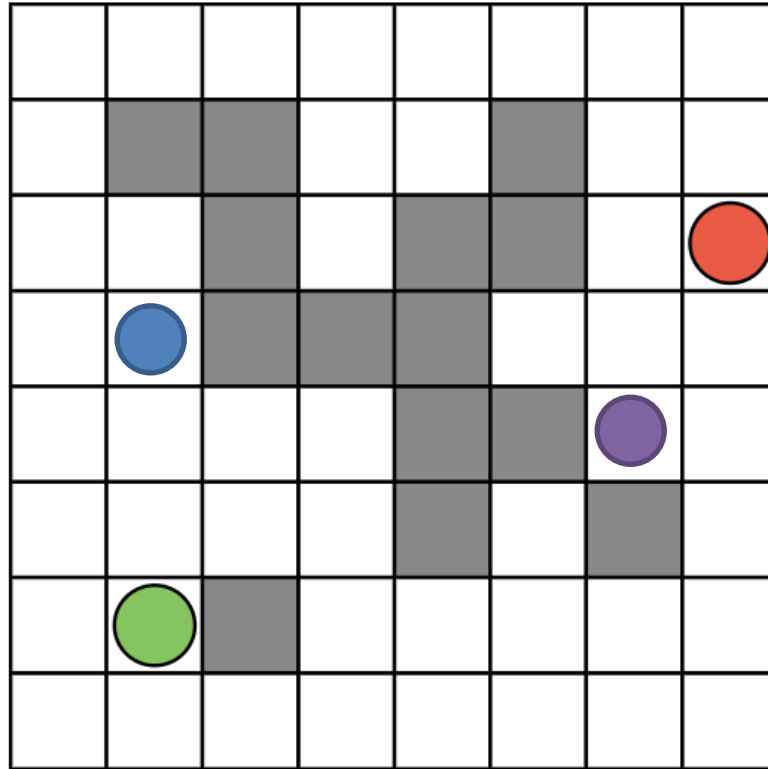
# PA2: Mazeworld problem



Robot can move N, S, E, W onto empty squares.

- $h1(n)$  = Manhattan distance
- $h2(n)$  = Euclidean distance

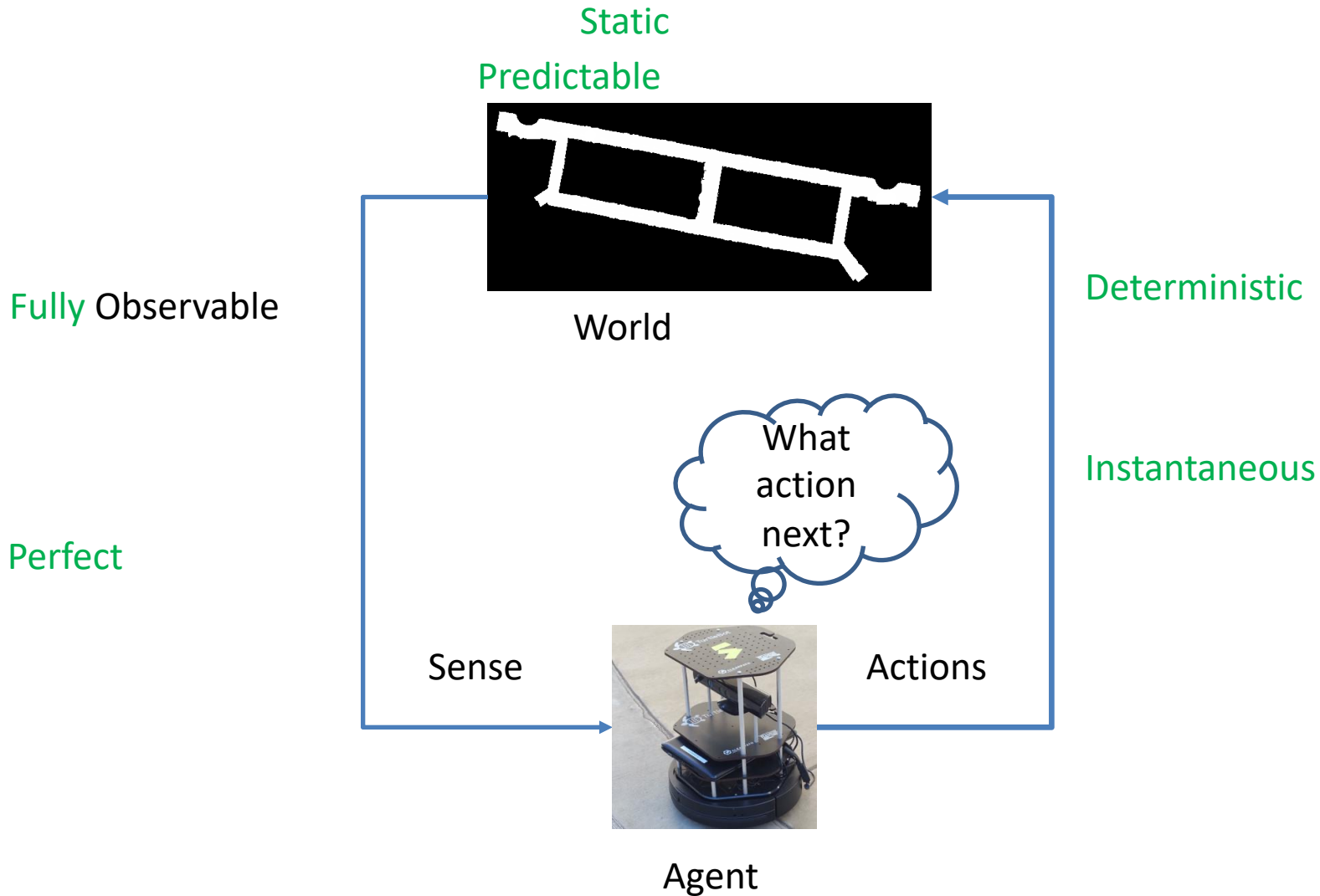
# PA2: Mazeworld problem



Robots can move N, S, E, W onto empty squares.

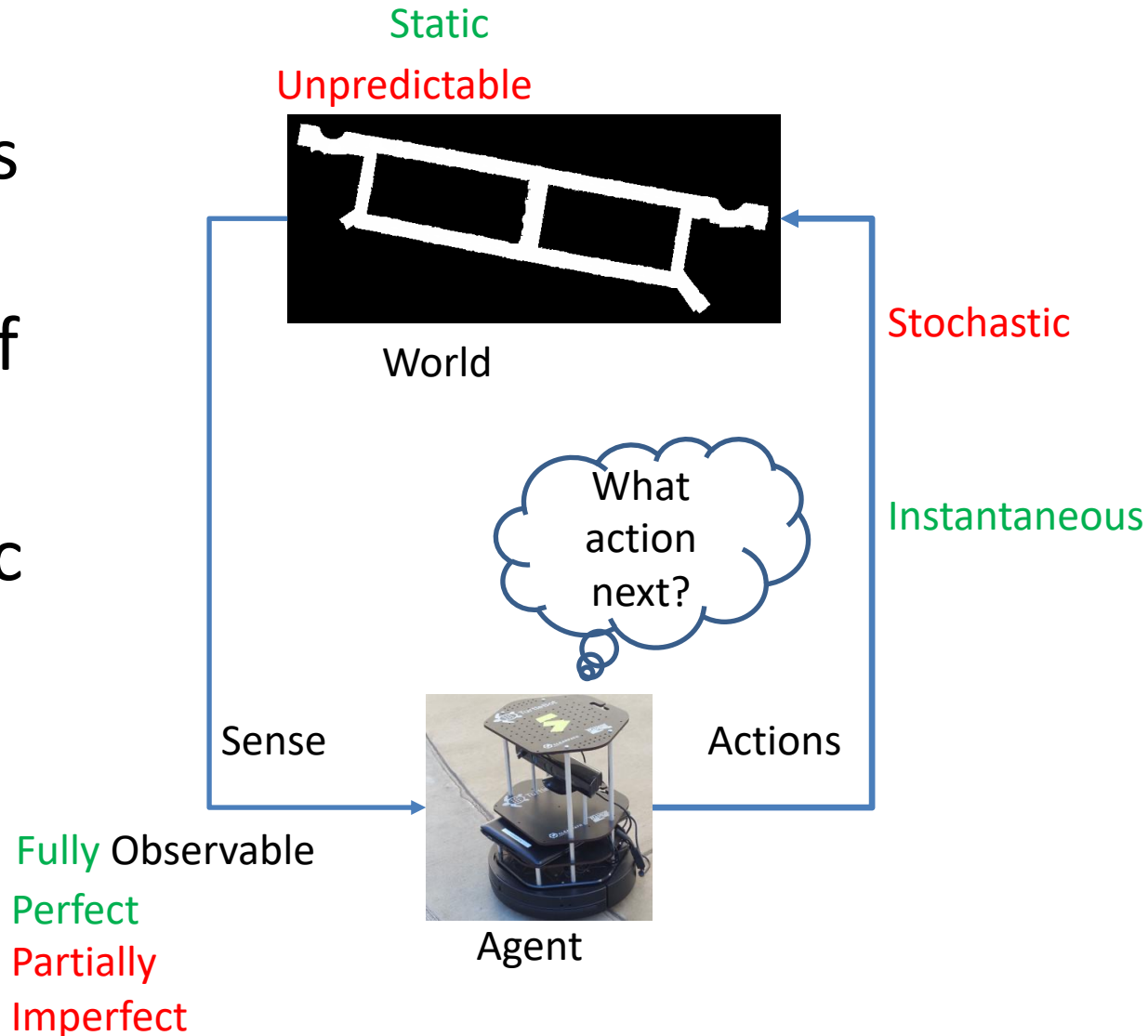
# **Non-deterministic actions and partial observations**

# Classical Planning view



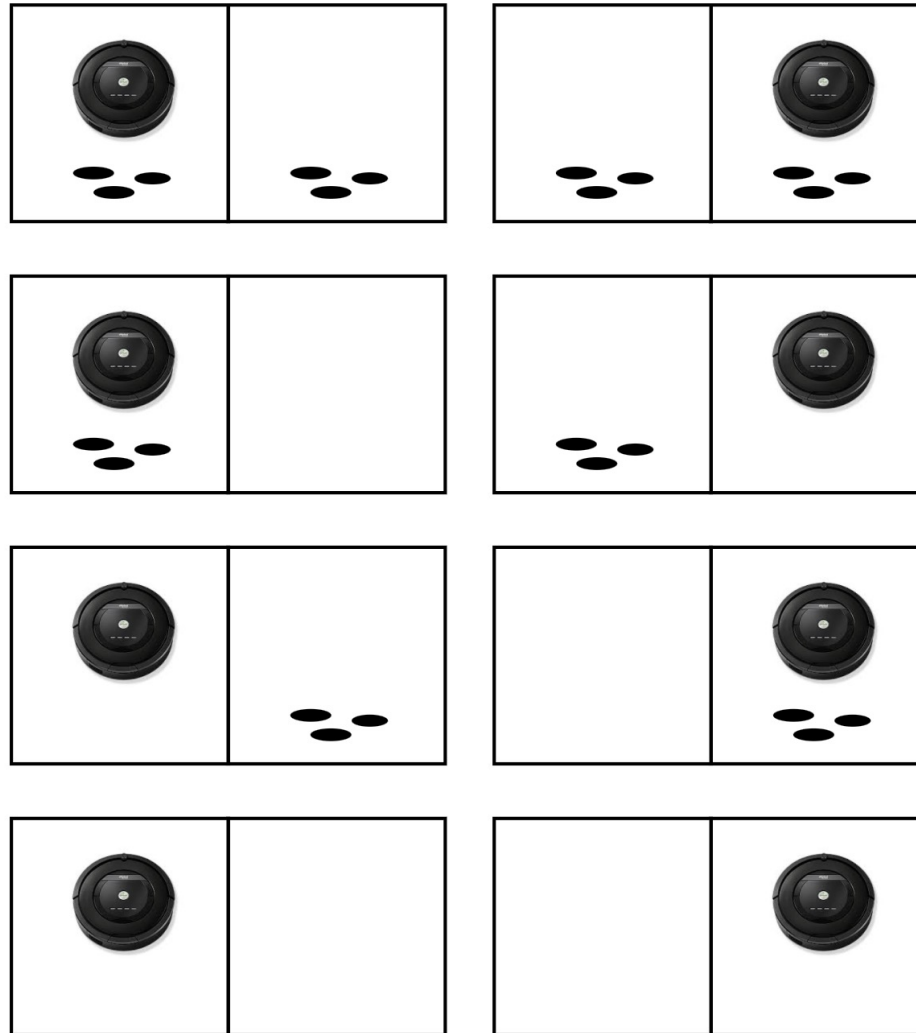
# Planning views

- Different planning views which involve different set of techniques
- E.g., Stochastic planning



# Vacuum world

- Clean 2 rooms

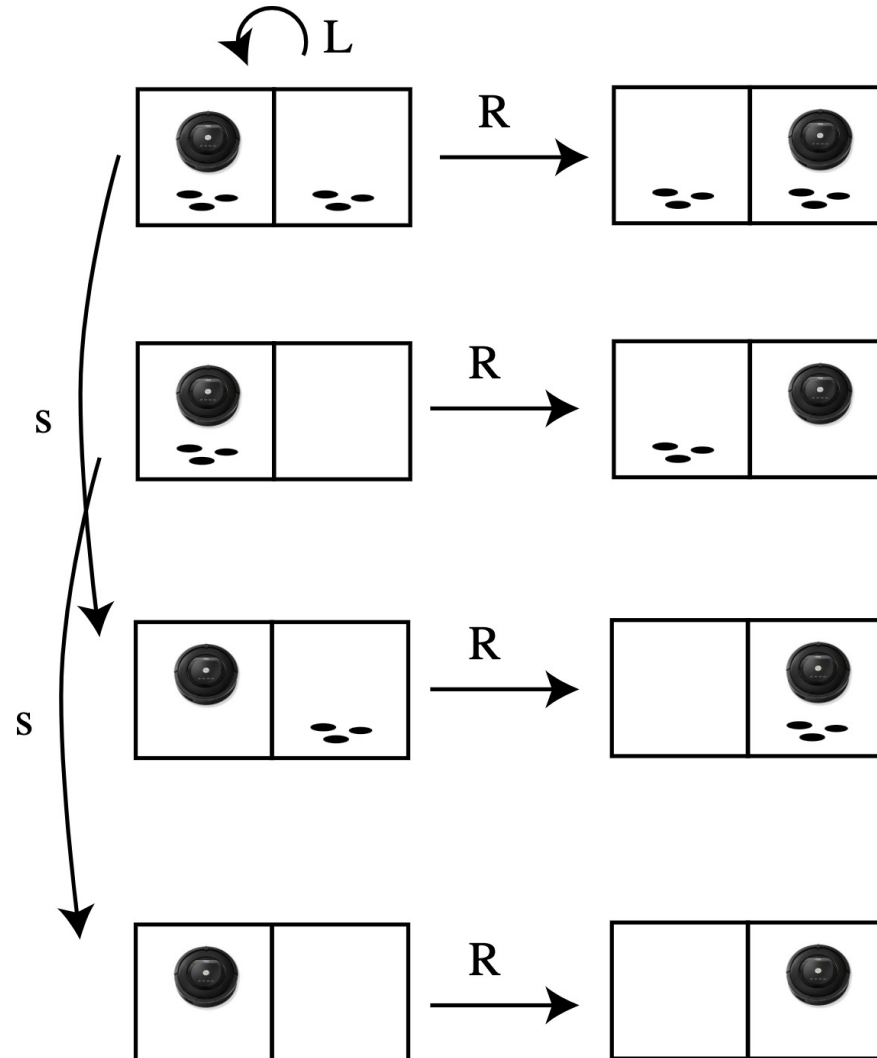




# Vacuum world: State space

- L go left
- R go right
- S sweep

(not all edges  
are marked)



# Non-deterministic actions

Sweep action:

- If dirty square, then the action cleans the square and sometimes cleans up dirt in an adjacent square
- If applied to a clean square, it might deposit dirt