

*16-350*

*Planning Techniques for Robotics*

*Interleaving Planning and Execution:  
Incremental Heuristic Search*

*Maxim Likhachev*

*Robotics Institute*

*Carnegie Mellon University*

# Planning during Execution

- Planning is a repeated process!
  - partially-known environments
  - dynamic environments
  - imperfect execution of plans
  - imprecise localization
- Need to be able to re-plan fast!
- Several methodologies to achieve this:
  - anytime heuristic search: return the best plan possible within  $T$  msec
  - **incremental heuristic search: speed up search by reusing previous efforts**
  - real-time heuristic search: plan few steps towards the goal and re-plan later

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  - partially-known environments → *edgcost changes*
  - dynamic environments → *edgcost changes, goal changes*
  - imperfect execution of plans → *robot pose changes/deviates off the path*
  - imprecise localization → *robot pose changes/deviates off the path*
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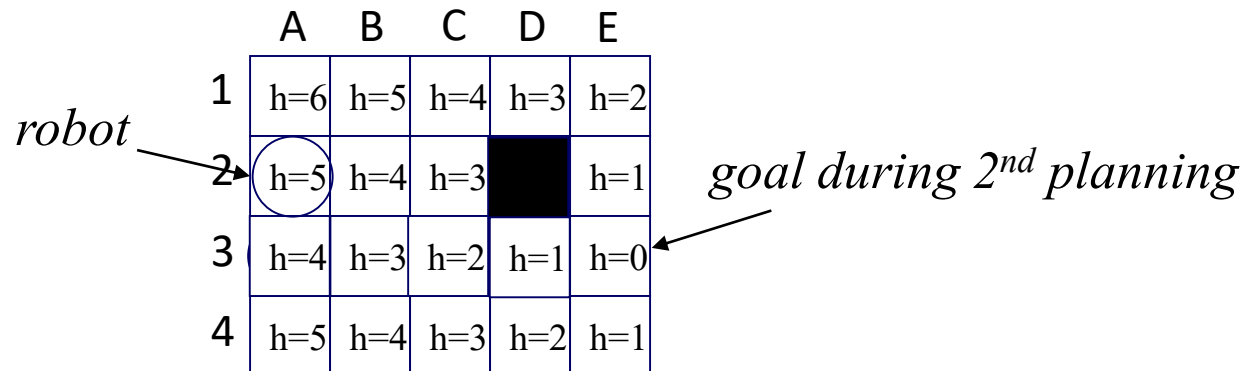
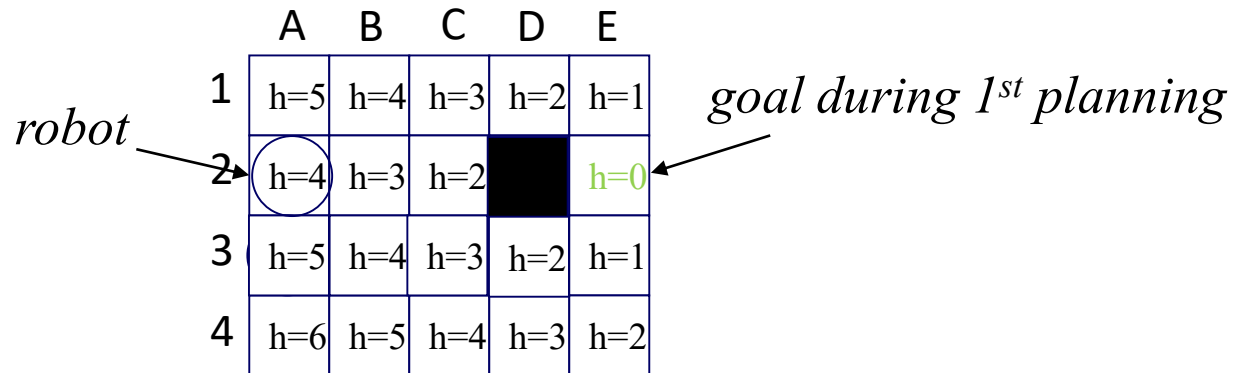
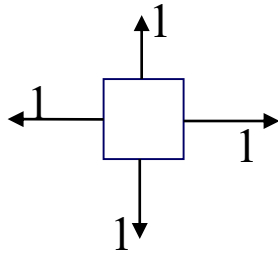
*Re-compute heuristics with respect to the **new** goal, and  
continue searching until the **new** goal state is expanded*

# Only Goal Changes

- Example on the board!

$$h(\text{cell } \langle x,y \rangle) = |x-x_{\text{goal}}| + |y-y_{\text{goal}}| \text{ (Manhattan Distance)}$$

4-connected grid



# Only Robot Pose Changes

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***Do the search backwards:***

*Then, the problem becomes “Only Goal Changes” that we know how to solve already*



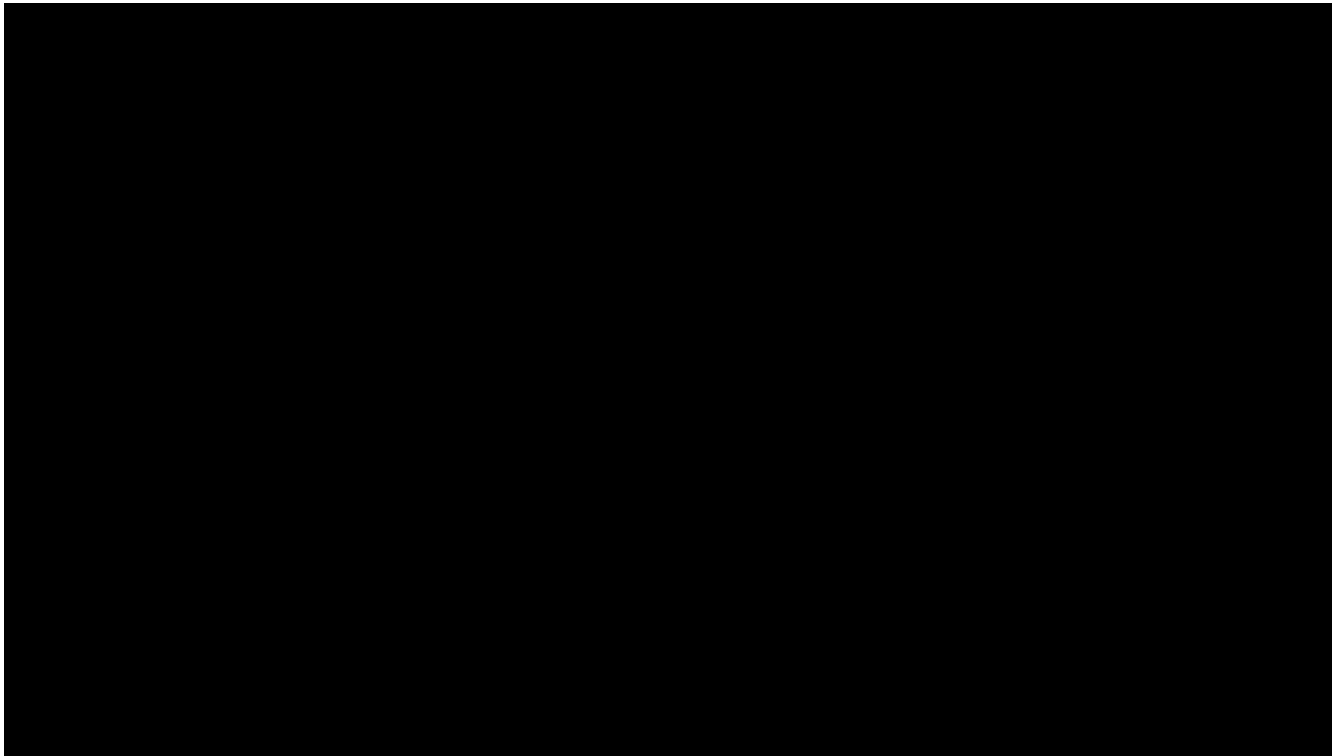
# What if both Robot Pose and its Goal change?

*Too bad!*

*Typically, you are better off re-planning from scratch then.*

# Changes to Edgecosts

- Two main reasons
  - Noisy perception (e.g., flickering obstacles, sensed position of obstacles is shifting, robot localization is noisy, etc.)
  - Partially-known environment

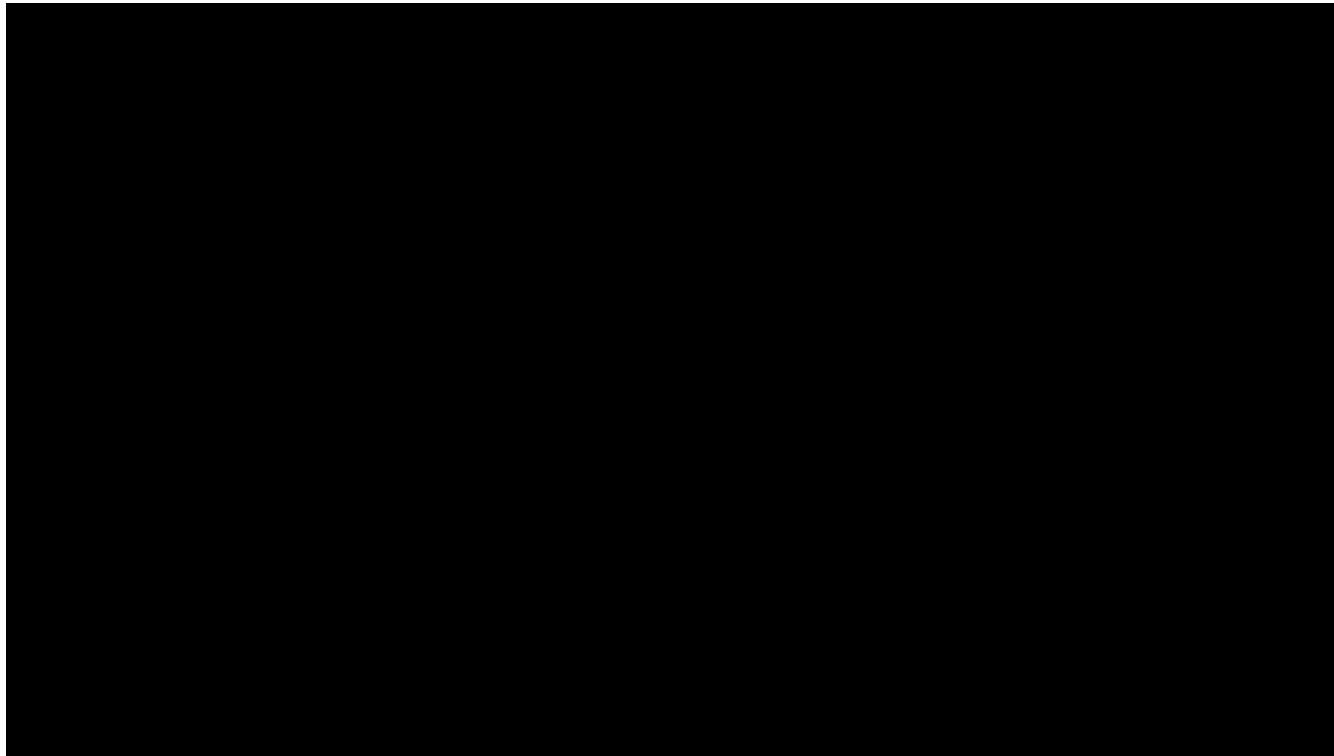


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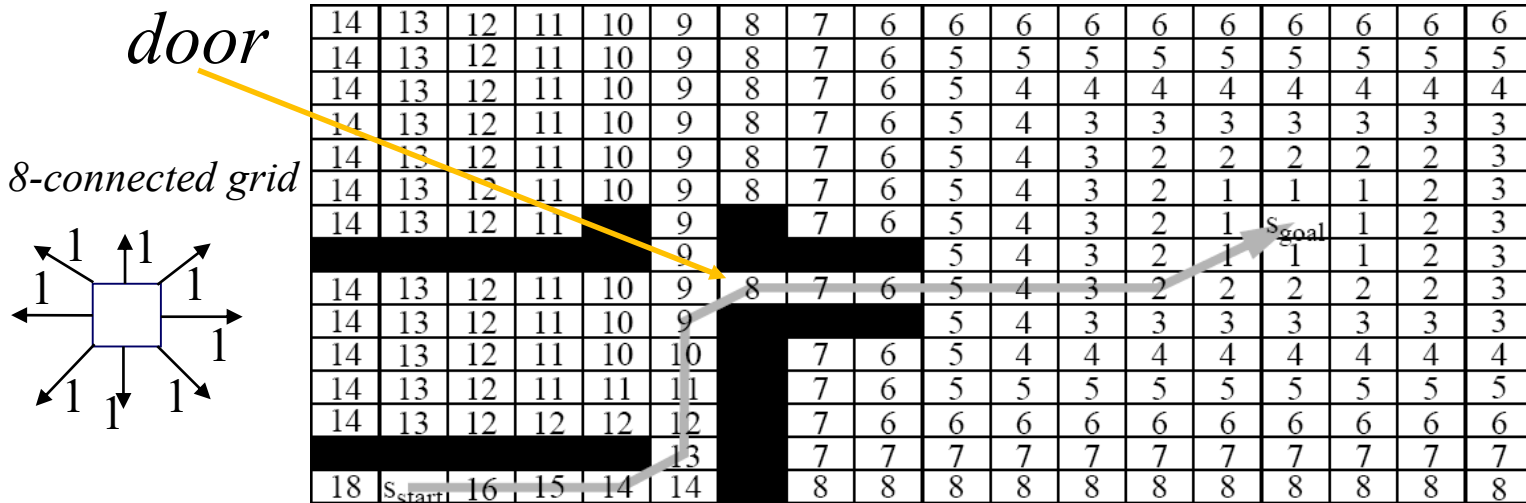
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*What should we assume about unknown space?*

***Freespace Assumption:** Assume that any “unknown” space is traversable until sensed otherwise!*

# Changes to Edgecosts

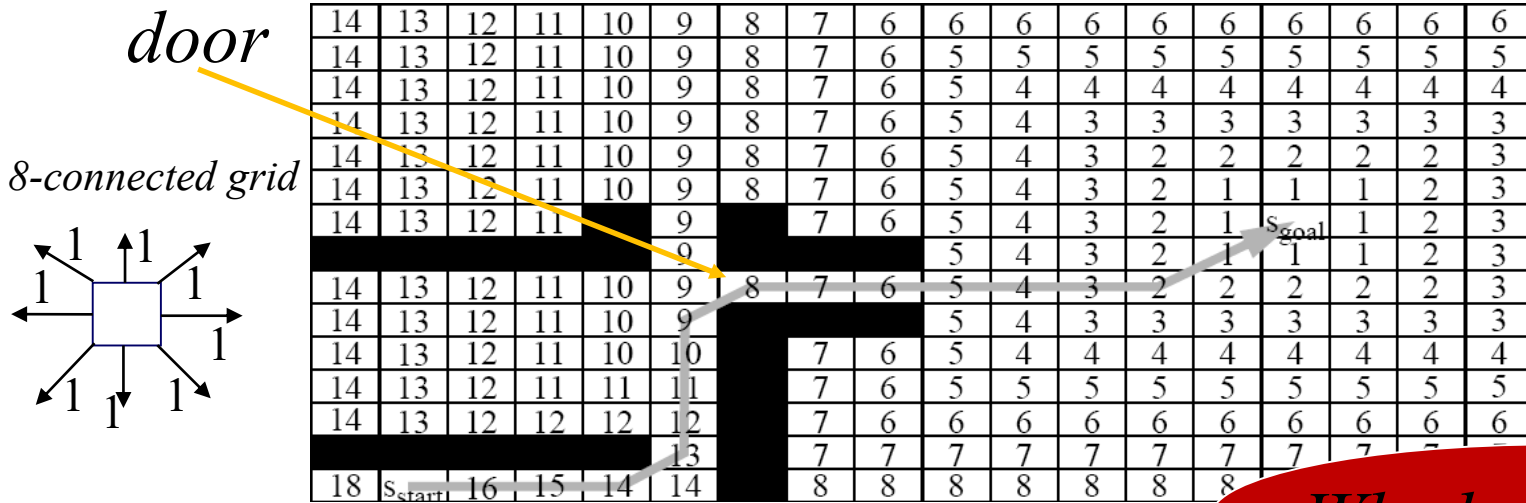
- The robot doesn't initially know the status of the door



*We ran an uninformed  $A^*$  search backwards  
(that is, all  $g$ -values are costs to  $s_{goal}$ )*

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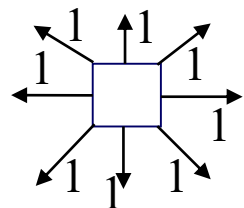


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# Changes to Edgecosts

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8-connected grid



14	13	12	11	10	9	8	7	6	6	6	6	6	6	6	6	6	6
14	13	12	11	10	9	8	7	6	5	5	5	5	5	5	5	5	5
14	13	12	11	10	9	8	7	6	5	4	4	4	4	4	4	4	4
14	13	12	11	10	9	8	7	6	5	4	3	3	3	3	3	3	3
14	13	12	11	10	9	8	7	6	5	4	3	2	2	2	2	2	3
14	13	12	11		9		7	6	5	4	3	2	1	1	1	2	3
					9				5	4	3	2	1	1	1	2	3
14	13	12	11	10	9	8	7	6	5	4	3	2	2	2	2	2	3
14	13	12	11	10	9				5	4	3	3	3	3	3	3	3
14	13	12	11	10	10		7	6	5	4	4	4	4	4	4	4	4
14	13	12	11	11	11		7	6	5	5	5	5	5	5	5	5	5
14	13	12	12	12	12		7	6	6	6	6	6	6	6	6	6	6
					13		7	7	7	7	7	7	7	7	7	7	7
18	S <sub>start</sub>	16	15	14	14		8	8	8	8	8	8	8	8	8	8	8

States with changed g-values

during execution, the robot found out that the door is closed

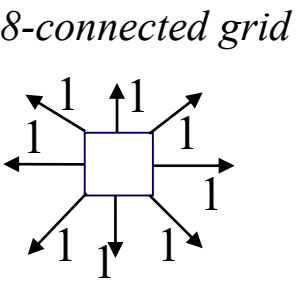
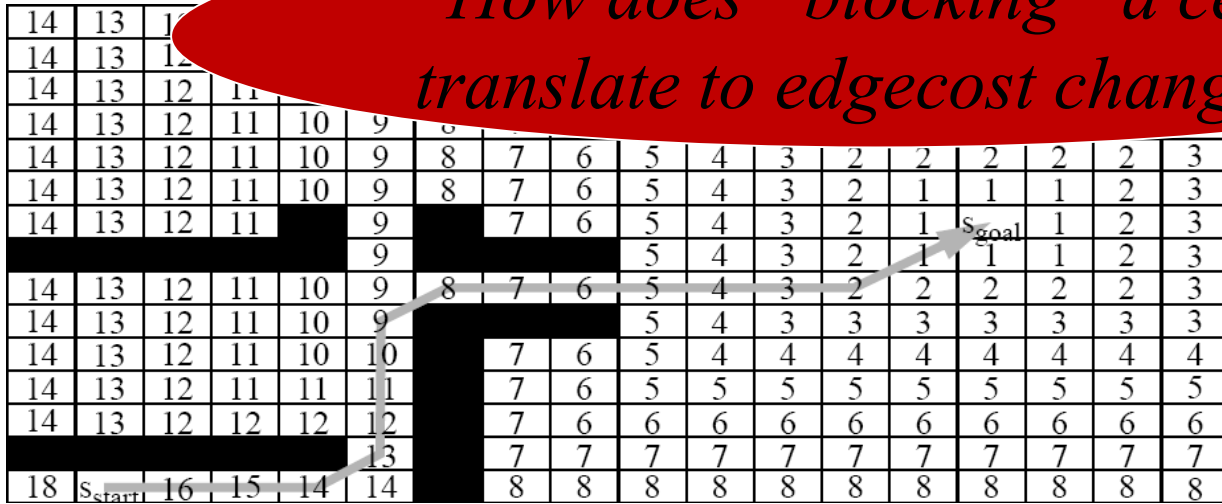
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15	14	13	13	13	13		7	6	5	4	4	4	4	4	4	4	4
15	14	14	14	14	14		7	6	5	5	5	5	5	5	5	5	5
15	15	15	15	15	15		7	6	6	6	6	6	6	6	6	6	6
					16		7	7	7	7	7	7	7	7	7	7	7
21	20	19	18	17	17		8	8	8	8	8	8	8	8	8	8	8



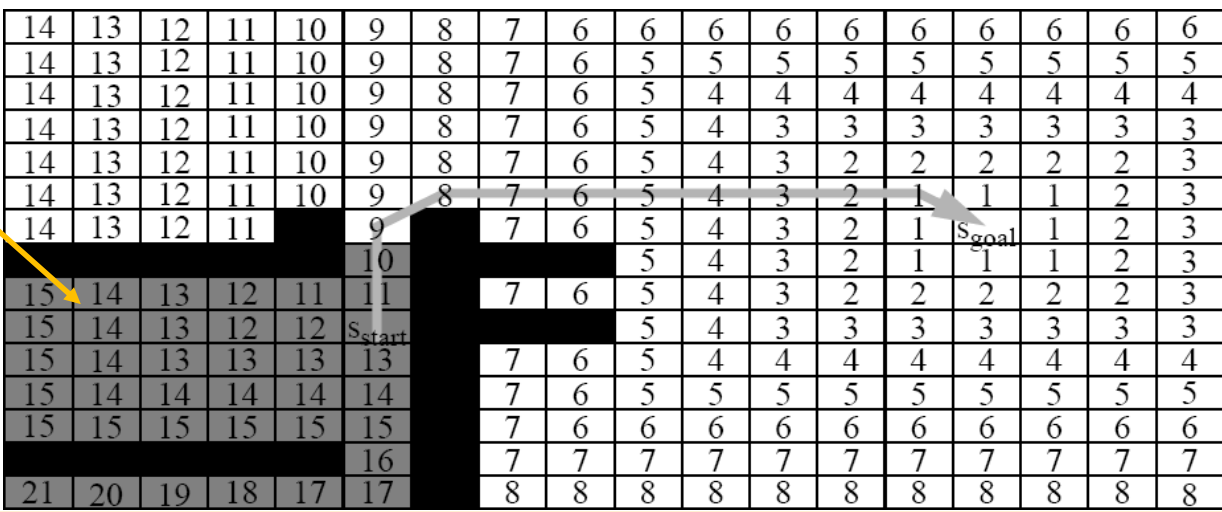
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*How does "blocking" a cell translate to edgecost changes?*



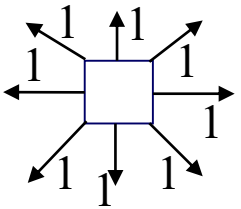
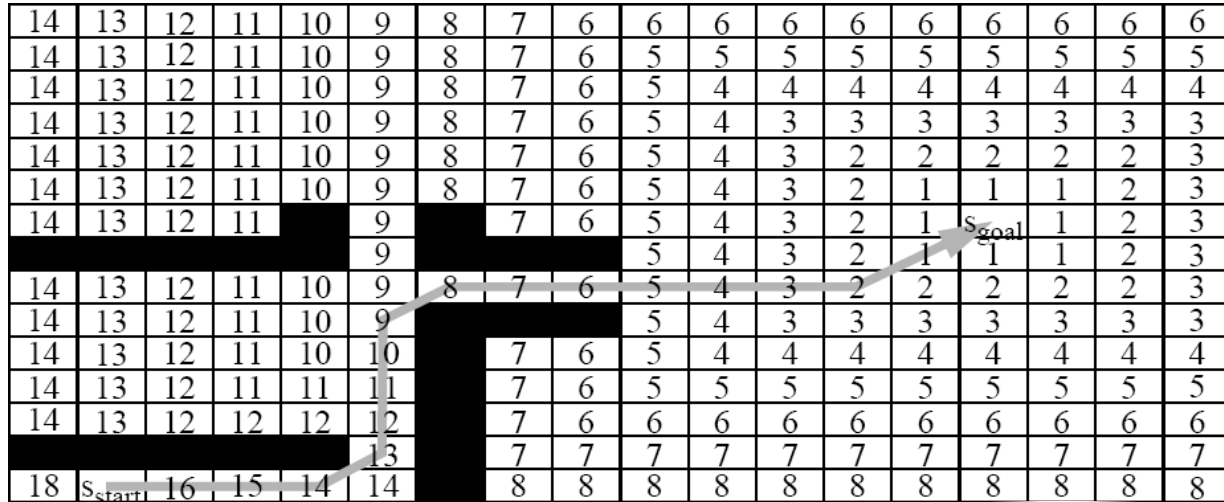
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*States with changed g-values*

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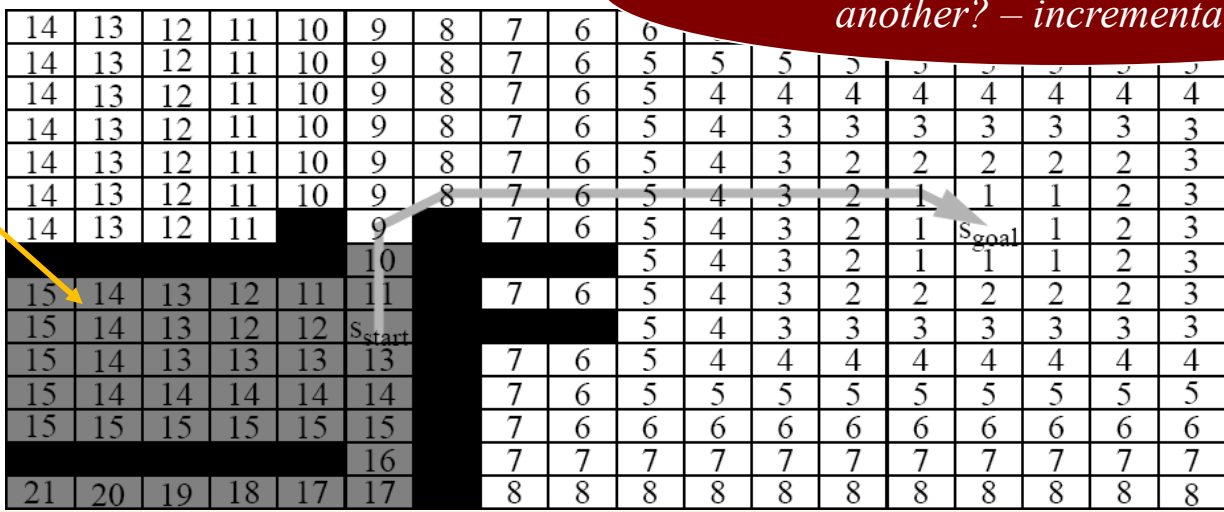
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States with changed g-values

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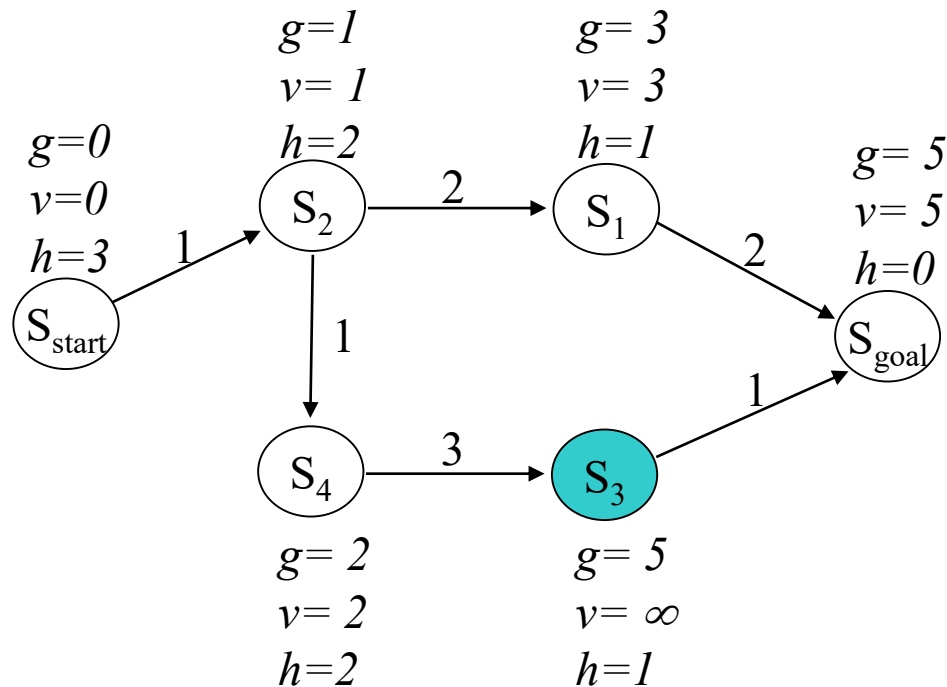
Can we reuse these g-values from one search to another? – incremental A\*





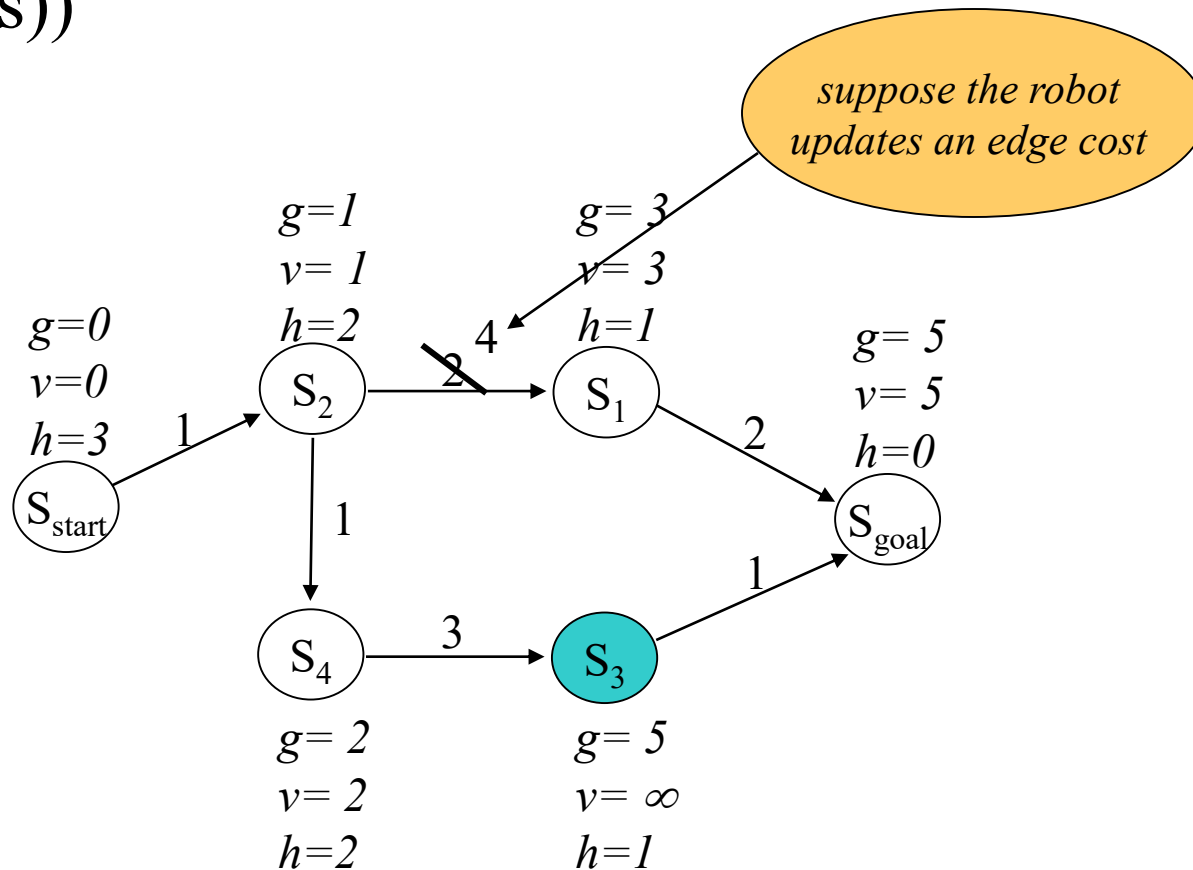
# A\* with Reuse of State Values

- So far, ComputePathwithReuse() could only deal with states whose  $v(s) \geq g(s)$  (overconsistent or consistent)
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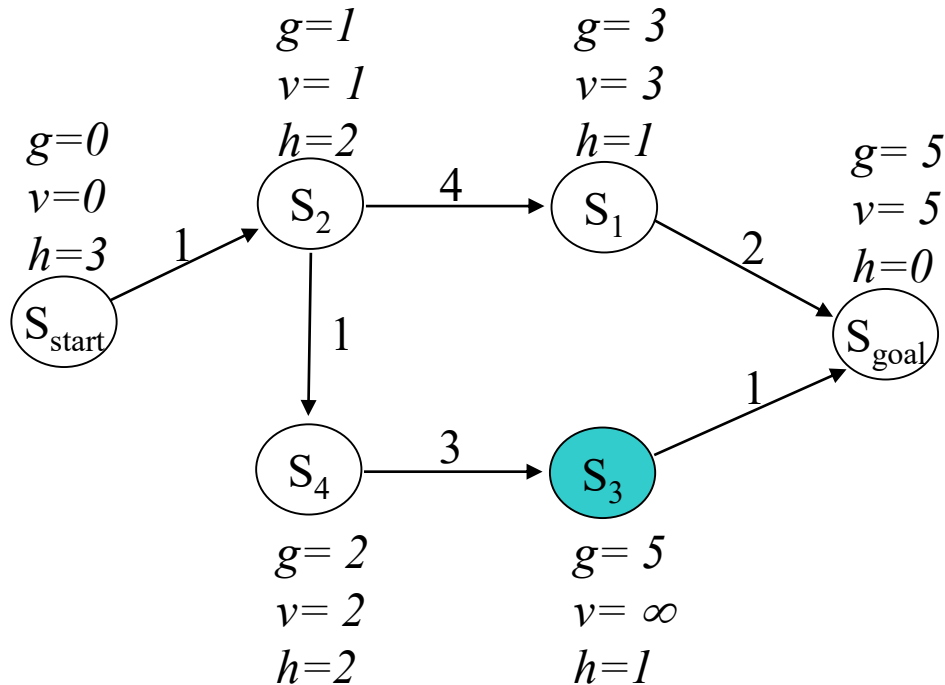


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*ComputePathwithReuse invariant:*  
 $g(s') = \min_{s'' \in \text{pred}(s')} v(s'') + c(s'', s')$

↓  
*need to update  $g(s_1)$*



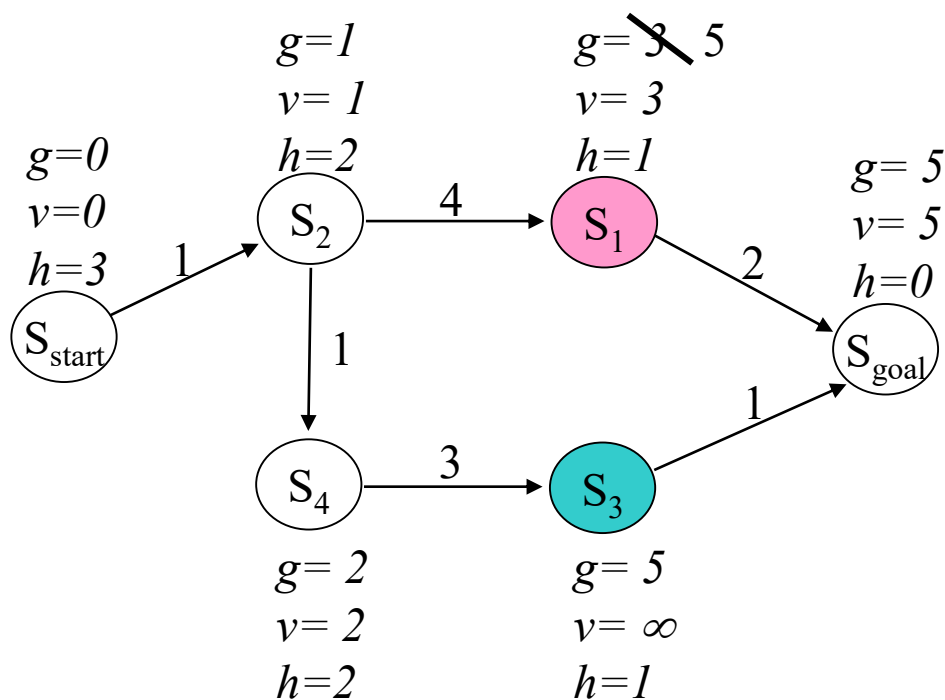
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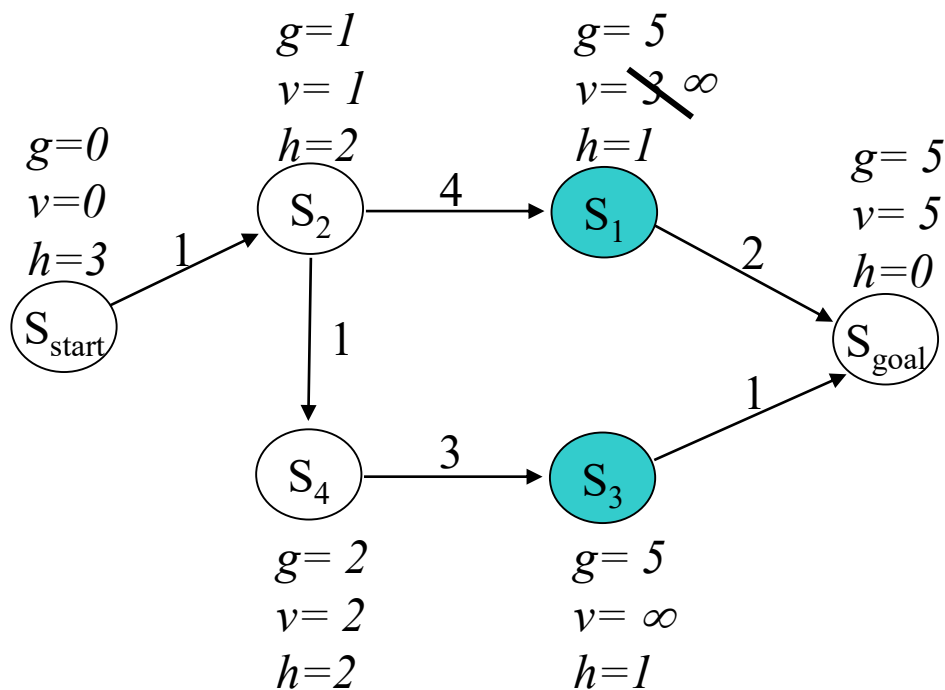
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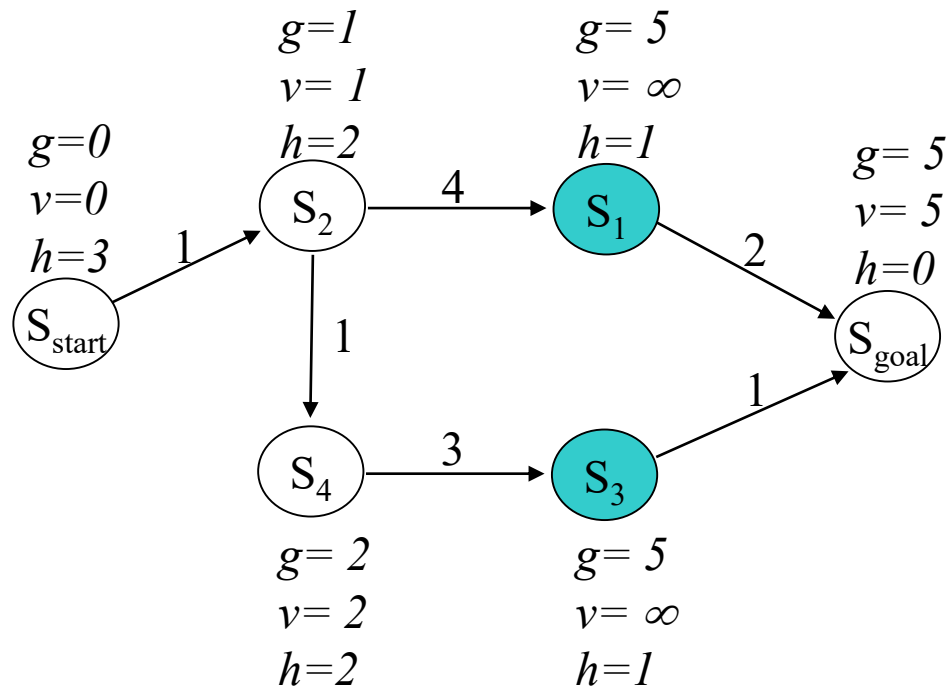




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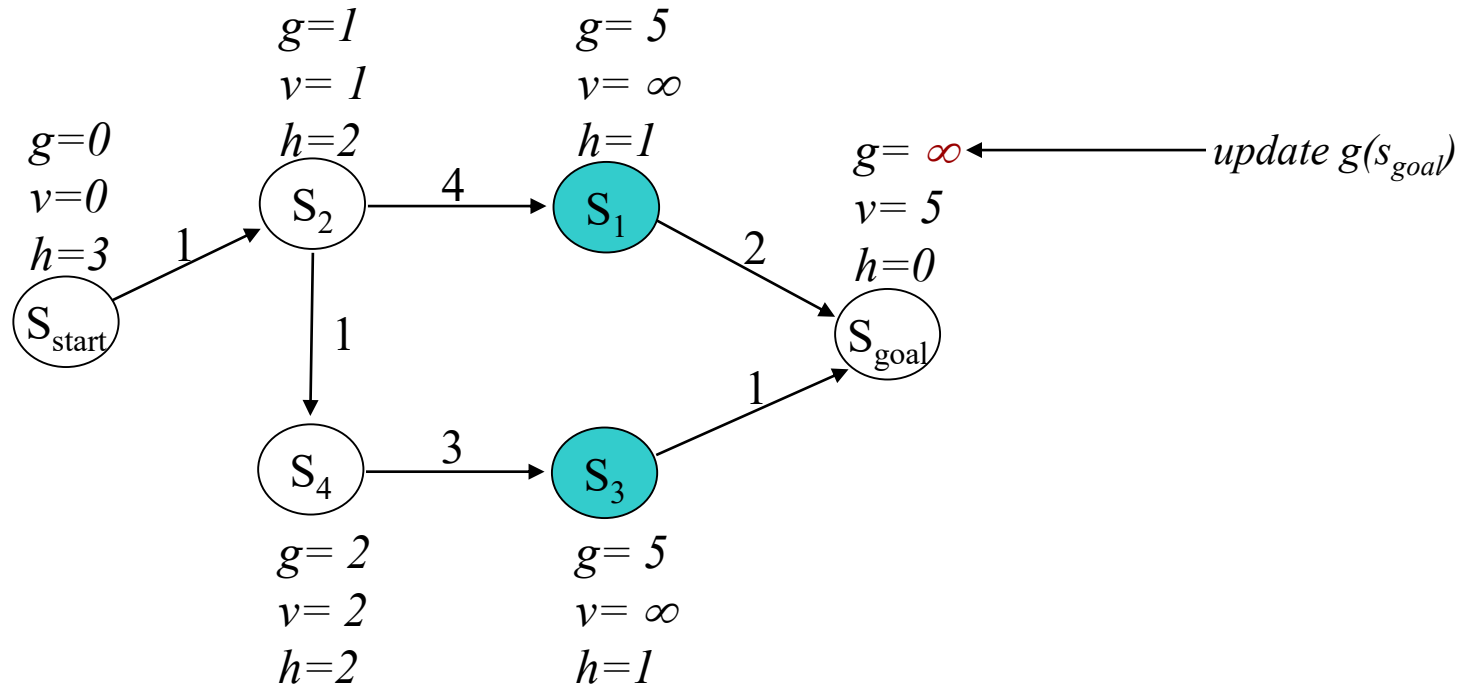
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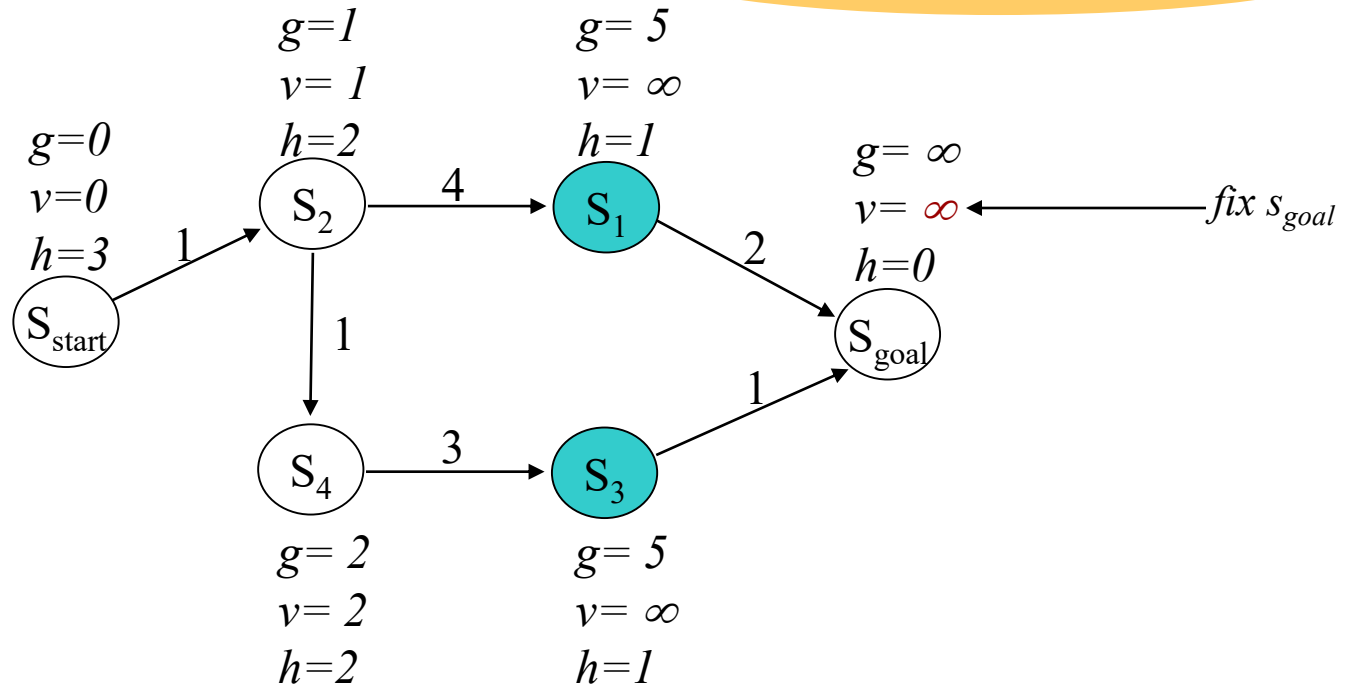


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*no more underconsistent states!*

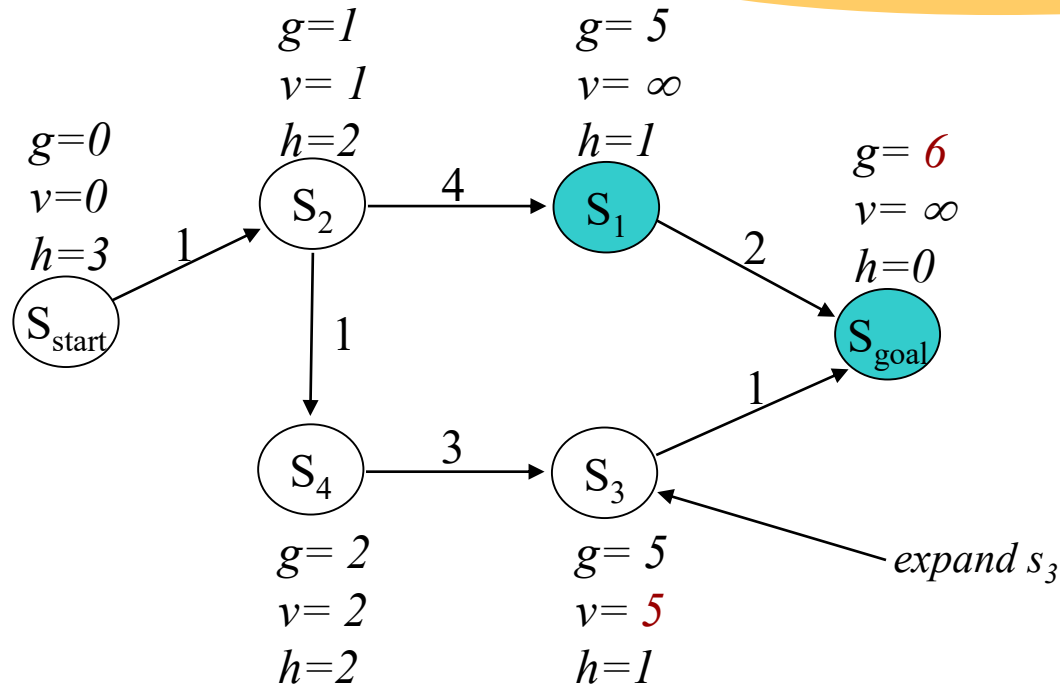


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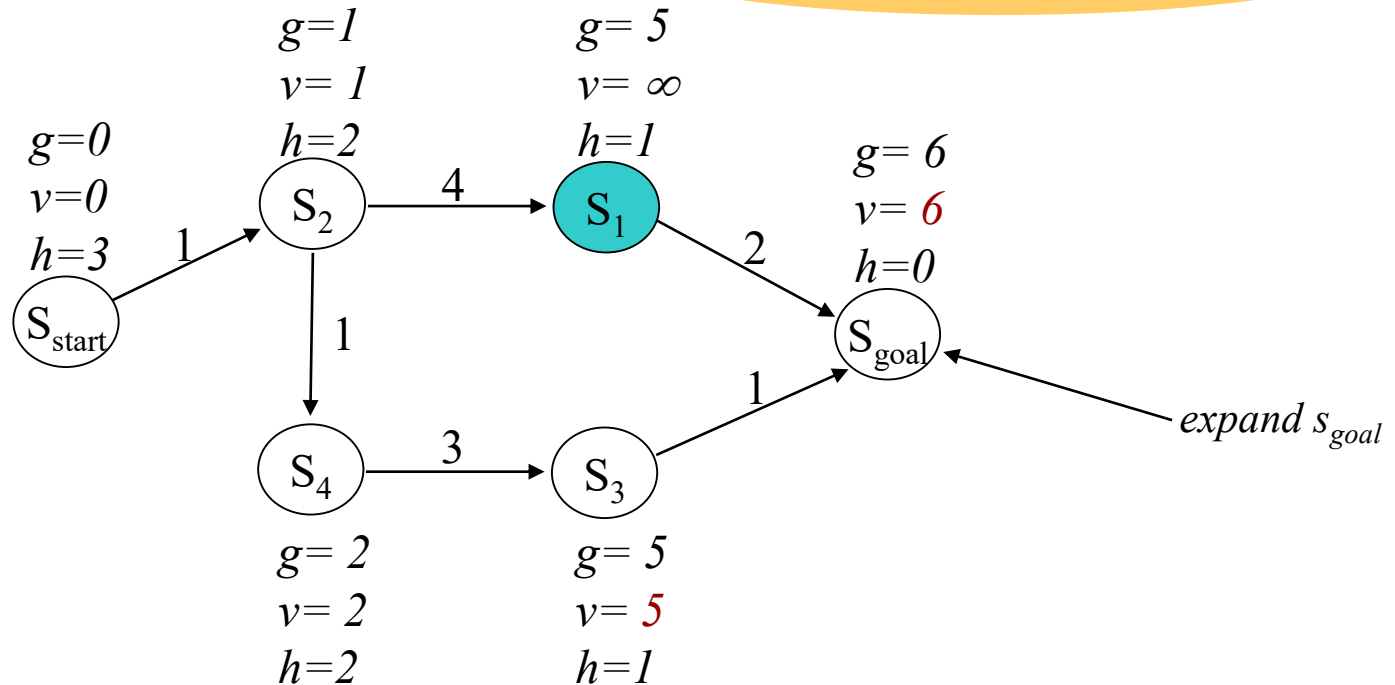


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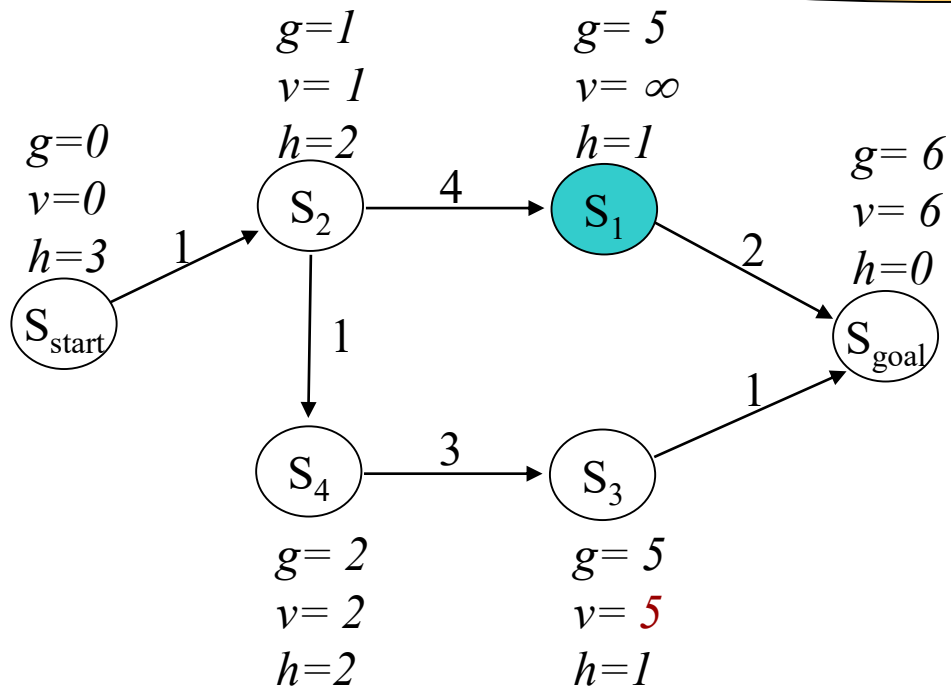


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after *ComputePathwithReuse* terminates:  
all  $g$ -values of states are equal to final A\*  $g$ -values

we can backtrack an optimal path  
(start at  $s_{goal}$ , proceed to pred that minimizes  $g+c$ )

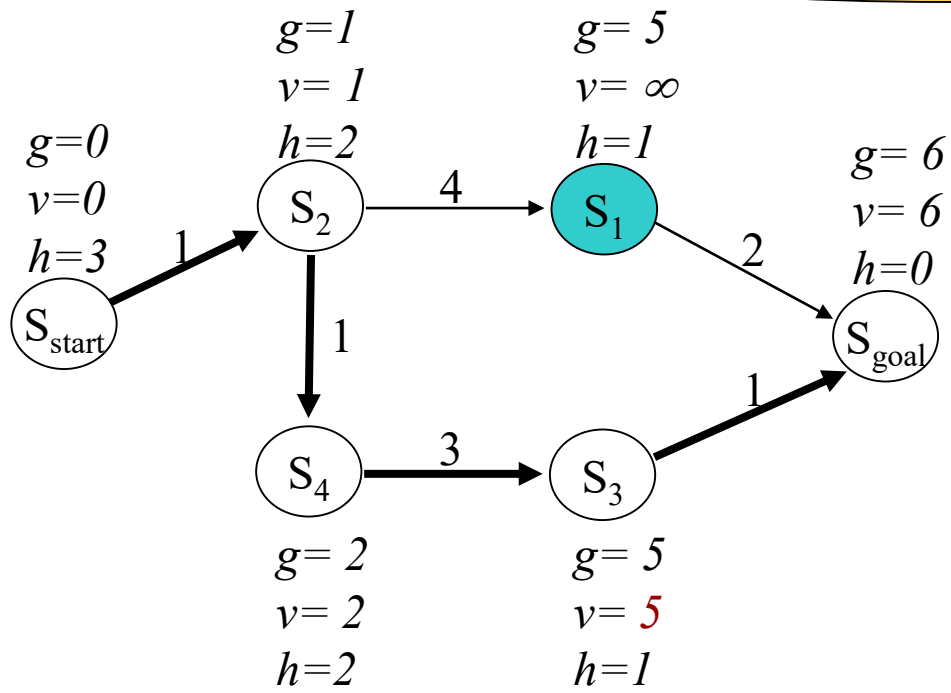


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# D\* Lite

- Optimal re-planning algorithm
- Simpler and with nicer theoretical properties version of D\*

until goal is reached

    ComputePathwithReuse();    *//modified to fix underconsistent states*

    publish optimal path;

    follow the path until map is updated with new sensor information;

    update the corresponding edge costs;

    set  $s_{\text{start}}$  to the current state of the agent;



# Anytime Incremental Heuristic Search

- Anytime D\*:
  - decrease  $\epsilon$  and update edge costs at the same time
  - re-compute a path by reusing previous state-values

set  $\epsilon$  to large value;

until goal is reached

    ComputePathwithReuse();    //modified to fix underconsistent states

    publish  $\epsilon$ -suboptimal path;

    follow the path until map is updated with new sensor information;

    update the corresponding edge costs;

    set  $s_{\text{start}}$  to the current state of the agent;

    if significant changes were observed

        increase  $\epsilon$  or replan from scratch;

    else

        decrease  $\epsilon$ ;

*What for?*

# What You Should Know...

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- How to handle changes to Robot Pose Only or Goal Only
- What is Freespace Assumption
- What is D\*/D\* Lite and the general principles behind it (don't need to know the exact algorithm)