16-350 Planning Techniques for Robotics

Search Algorithms: A* Search

Maxim Likhachev
Robotics Institute
Carnegie Mellon University

Uninformed A* Search

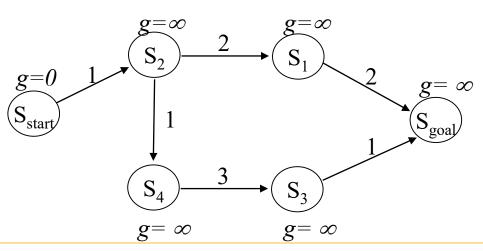
Computes g*-values for relevant (not all) states

Main function

```
g(s_{start}) = 0; all other g-values are infinite; OPEN = \{s_{start}\}; ComputePath(); publish solution; //compute least-cost path using g-values
```

ComputePath function

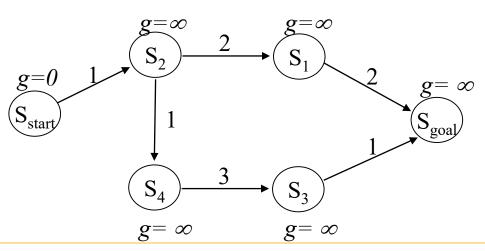
while(s_{goal} is not expanded and $OPEN \neq 0$) remove s with the smallest g(s) from OPEN; expand s;



Uninformed A* Search

Computes g*-values for relevant (not all) states

```
while(s_{goal} is not expanded and OPEN \neq 0)
remove s with the smallest g(s) from OPEN;
insert s into CLOSED;
for every successor s of s such that s not in CLOSED
if g(s') > g(s) + c(s,s')
g(s') = g(s) + c(s,s');
insert s into OPEN;
```



Uninformed A* Search

Computes g*-values for relevant (not all) states

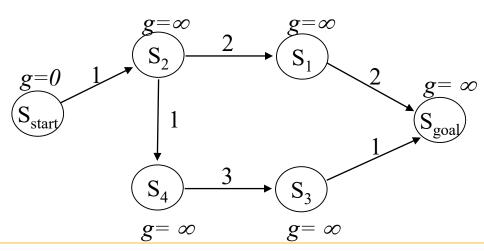
ComputePath function

while(s_{goal} is not expanded and $OPEN \neq 0$)
remove s with the smallest g(s) from OPEN;
insert s into CLOSED;
for every successor s of s such that s not in CLOSED

if
$$g(s') > g(s) + c(s,s')$$

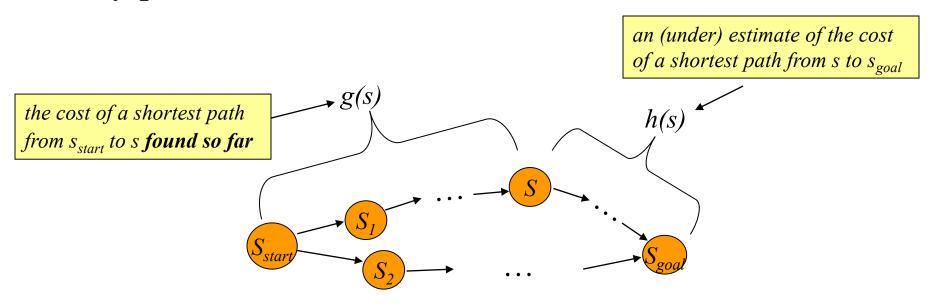
 $g(s') = g(s) + c(s,s')$;
insert s' into OPEN;

<u>clarification:</u> updates g(s') if s' is already in OPEN



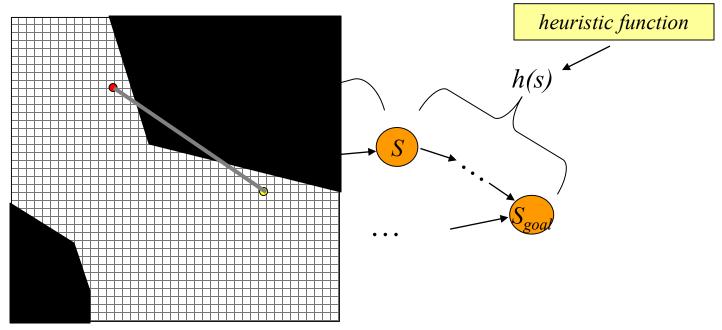
• Computes optimal g-values for relevant states

at any point of time:



Computes optimal g-values for relevant states

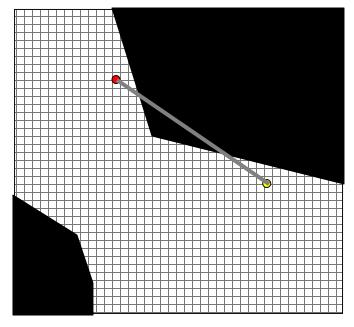
at any point of time:



one popular heuristic function – Euclidean distance

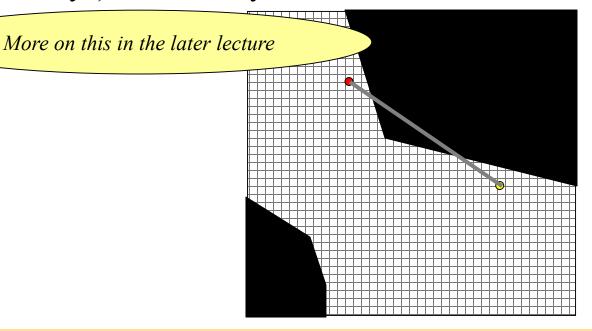
 $minimal\ cost\ from\ s\ to\ s_{goal}$

- Heuristic function must be:
 - admissible: for every state s, $h(s) \le c *(s, s_{goal})$
 - consistent (satisfy triangle inequality): $h(s_{goal}, s_{goal}) = 0 \text{ and for every } s \neq s_{goal}, h(s) \leq c(s, succ(s)) + h(succ(s))$
 - admissibility <u>provably</u> follows from consistency and often (<u>not</u> <u>always</u>) consistency follows from admissibility



 $minimal\ cost\ from\ s\ to\ s_{goal}$

- Heuristic function must be:
 - admissible: for every state s, $h(s) \le c *(s, s_{goal})$
 - consistent (satisfy triangle inequality). Why triangle inequality? $h(s_{goal}, s_{goal}) = 0$ and for every $s \neq s_{goal}$, $h(s) \leq c(s, succ(s)) + h(succ(s))$
 - admissibility <u>provably</u> follows from consistency and often (<u>not always</u>) consistency follows from admissibility



A*: Uninformed vs. Informed Search

- A*: expands states in the order of f = g + h values
- Uninformed A*: expands states in the order of g values

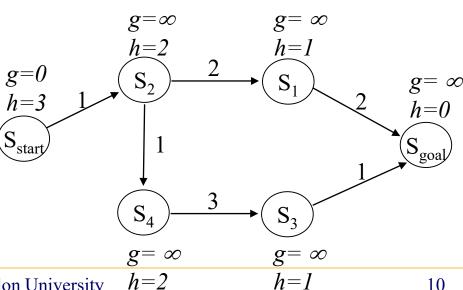
Computes optimal g-values for relevant states

Main function

```
g(s_{start}) = 0; all other g-values are infinite; OPEN = \{s_{start}\};
ComputePath();
publish solution;
```

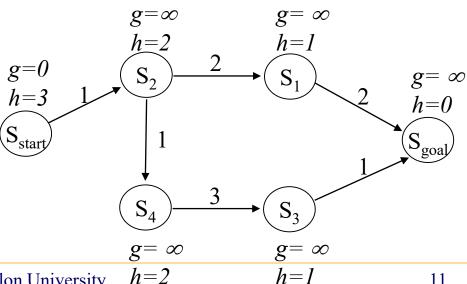
ComputePath function

while(s_{goal} is not expanded and $OPEN \neq 0$) remove s with the smallest [f(s) = g(s) + h(s)] from OPEN; expand s;



Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
     if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
       insert s' into OPEN;
```

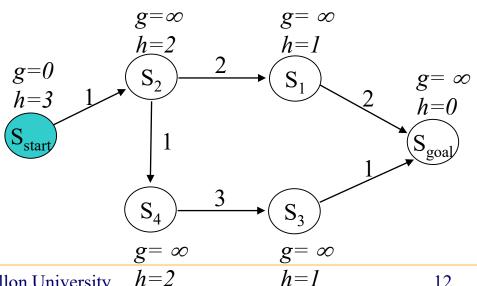


Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
    if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
      insert s' into OPEN;
```

$$CLOSED = \{\}$$

 $OPEN = \{s_{start}\}$
 $next \ state \ to \ expand: \ s_{start}$



Computes optimal g-values for relevant states

ComputePath function

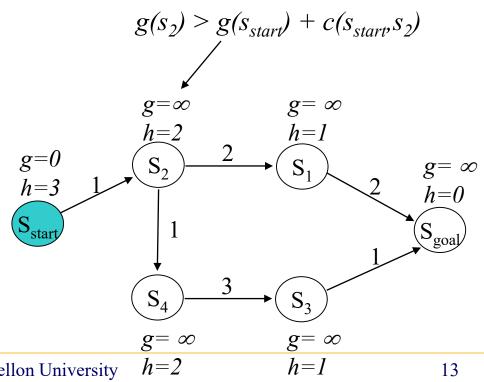
while(s_{goal} is not expanded and $OPEN \neq 0$) remove s with the smallest [f(s) = g(s) + h(s)] from *OPEN*; insert s into CLOSED; for every successor s' of s such that s'not in CLOSED

if
$$g(s') > g(s) + c(s,s')$$

 $g(s') = g(s) + c(s,s')$;
insert s' into OPEN;

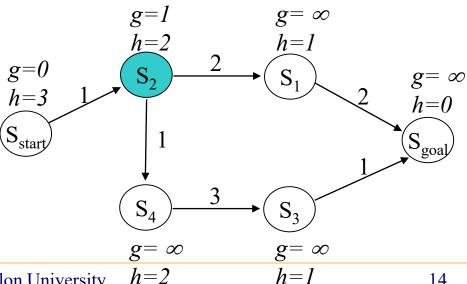
$$CLOSED = \{\}$$

 $OPEN = \{s_{start}\}$
 $next \ state \ to \ expand: \ s_{start}$



Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
     if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
       insert s' into OPEN;
```

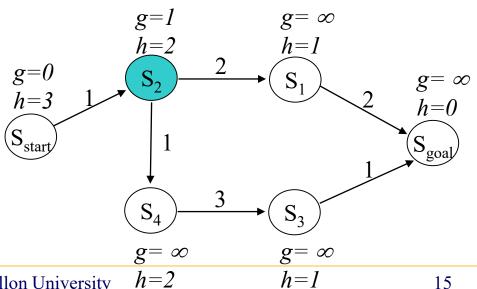


Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
     if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
       insert s' into OPEN;
```

$$CLOSED = \{s_{start}\}$$

 $OPEN = \{s_2\}$
 $next \ state \ to \ expand: \ s_2$

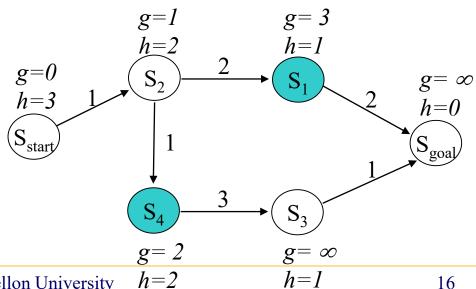


Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
     if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
       insert s' into OPEN;
```

$$CLOSED = \{s_{start}, s_2\}$$

 $OPEN = \{s_1, s_4\}$
 $next \ state \ to \ expand: \ s_1$

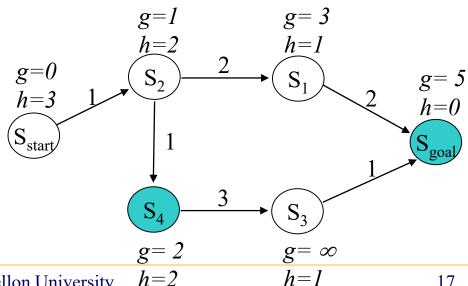


Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
    if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
      insert s' into OPEN;
```

$$CLOSED = \{s_{start}, s_2, s_1\}$$

 $OPEN = \{s_4, s_{goal}\}$
 $next \ state \ to \ expand: \ s_4$

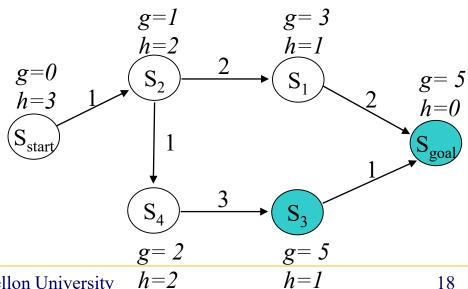


Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
    if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
      insert s' into OPEN;
```

$$CLOSED = \{s_{start}, s_2, s_1, s_4\}$$

 $OPEN = \{s_3, s_{goal}\}$
 $next\ state\ to\ expand:\ s_{goal}$

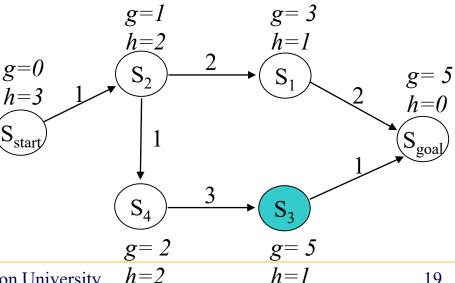


Computes optimal g-values for relevant states

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
    if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
      insert s' into OPEN;
```

$$CLOSED = \{s_{start}, s_2, s_1, s_4, s_{goal}\}$$

 $OPEN = \{s_3\}$
 $done$



Computes optimal g-values for relevant states

ComputePath function

```
while (s_{goal}) is not expanded and OPEN \neq 0)
remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
insert s into CLOSED;
for every successor s of s such that s not in CLOSED
if g(s') > g(s) + c(s,s')
g(s') = g(s) + c(s,s');
insert s into OPEN;
```

for every expanded state g(s) is optimal

for every other state g(s) is an upper bound g=0 h=2 S_2 h=3 S_3 S_4 S_4 S_4 S_4 S_5 S_6 S_8 S_8

h=1

h=2

Computes optimal g-values for relevant states

ComputePath function

```
while(s_{goal} is not expanded and OPEN \neq 0)
 remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
 insert s into CLOSED;
 for every successor s' of s such that s'not in CLOSED
    if g(s') > g(s) + c(s,s')
      g(s') = g(s) + c(s,s');
      insert s' into OPEN;
```

for every expanded state g(s) is optimal for every other state g(s) is an upper bound we can now compute a least-cost path

 S_2 S_4 g=2g=5h=2h=1

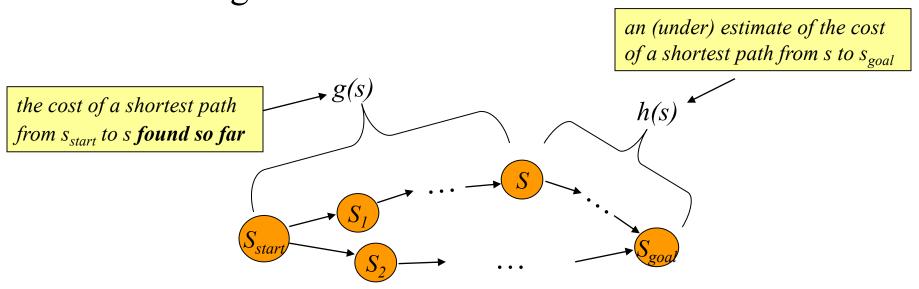
h=1

g=0

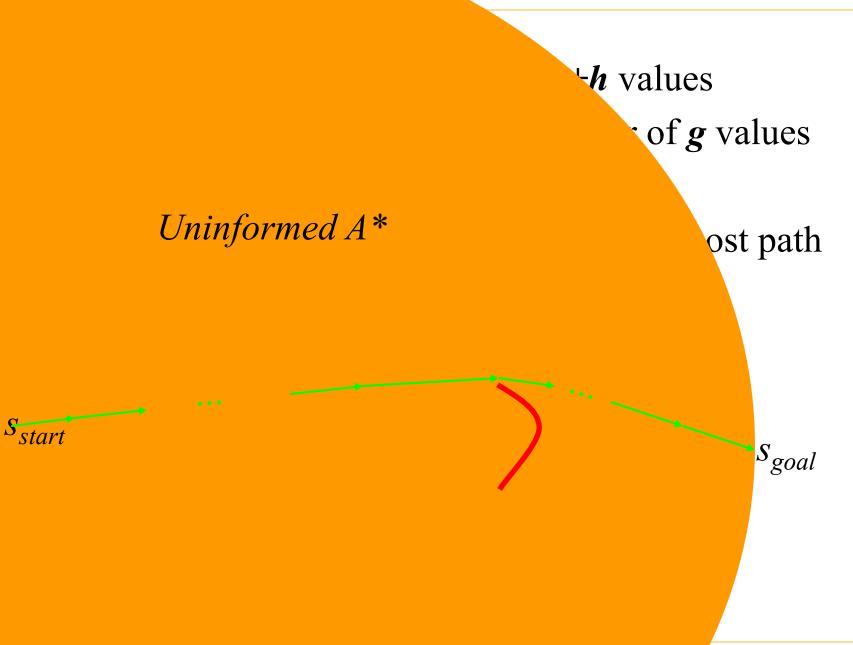
h=3

A*: Uninformed vs. Informed Search

- A*: expands states in the order of f = g + h values
- Uninformed A^* : expands states in the order of g values
- Intuitively: f(s) estimate of the cost of a least cost path from start to goal via state s



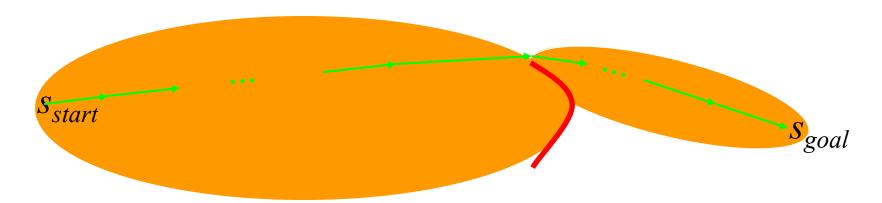
Informed Search



A*: Uninformed vs. Informed Search

- A*: expands states in the order of f = g + h values
- Uninformed A*: expands states in the order of g values

• Intuitively: f(s) – estimate of the cost of a least cost path from start to goal via state s

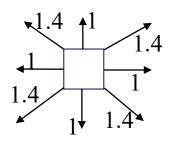


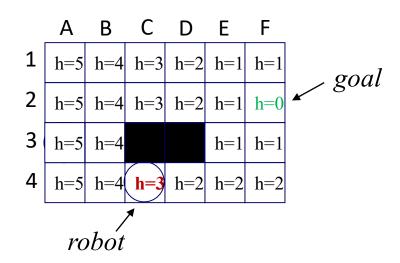
A* with Heuristics=Euclidean Distance

• Example on a Grid-based Graph:

$$h(cell < x,y>) = max(|x-x_{goal}|,|y-y_{goal}|)$$

8-connected grid





Theorem 1. For every expanded state s, it is guaranteed that g(s)=g*(s)

Sketch of proof by induction:

- assume all previously expanded states have optimal g-values
- next state to expand is s: f(s) = g(s) + h(s) min among states in OPEN
- assume g(s) is suboptimal (we will prove that it is impossible by contradiction)
- then there must be at least one state s' on an optimal path from start to s such that it is in OPEN but wasn't expanded
- $g(s') + h(s') \ge g(s) + h(s)$
- but g(s') + c*(s',s) < g(s) =>
- g(s') + c*(s',s) + h(s) < g(s) + h(s) => (from consistency of h-values)
- g(s') + h(s') < g(s) + h(s) => CONTRADICTION
- thus it must be the case that g(s) is optimal

Theorem 2. Once the search terminates, it is guaranteed that $g(s_{goal}) = g*(s_{goal})$

Sketch of proof:



Sketch of proof:

- every backtracking step from state s moves to a predecessor state s' that continues to be on a least-cost path (because all predecessors u not on a least-cost path will have have $g(u)+\cos t(u,s)$ that are strictly larger than $g(s')+\cos t(s',s)$)

Theorem 4 (complexity). No state is expanded more than once by A*

Sketch of proof:



Theorem 5. Given a graph and a heuristic function, **A* performs a minimal number of expansions to find a provably optimal path** (provided goal state is always expanded first among the states with the same f-values in OPEN)

Implementation Details of A* Search

Computes optimal g-values for relevant states

ComputePath function

```
while (s_{goal}) is not expanded and OPEN \neq 0)
remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
insert s into CLOSED;
for every successor s of s such that s not in CLOSED
if g(s') > g(s) + c(s,s')
```

if
$$g(s') > g(s) + c(s,s')$$

 $g(s') = g(s) + c(s,s')$;
insert s' into OPEN;

How to implement OPEN?

How to implement CLOSED?

Implementation Details of A* Search

Computes optimal g-values for relevant states

ComputePath function

```
while (s_{goal} \text{ is not expanded and } OPEN \neq 0)
remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
insert s into CLOSED;
for every successor s of s such that s not in CLOSED
```

if
$$g(s') > g(s) + c(s,s')$$

 $g(s') = g(s) + c(s,s')$;
insert s' into *OPEN*;

How to implement OPEN?

Typically, a priority queue built using a binary heap

How to implement CLOSED?

Typically, each state has a Boolean flag indicating if it was already closed

A* Search with Backpointers

• After search terminates, least-cost path is given by backtracking backpointers from s_{goal} to s_{start}

Main function

```
g(s_{start}) = 0; all other g-values are infinite; OPEN = \{s_{start}\}; set all backpointers bp to NULL;
ComputePath();
publish solution; //backtrack least-cost path using backpointers bp
ComputePath function
while(s_{goal} is not expanded and OPEN \neq 0)
remove s with the smallest [f(s) = g(s) + h(s)] from OPEN;
insert s into CLOSED;
for every successor s of s such that s not in CLOSED
```

if
$$g(s') > g(s) + c(s,s')$$

 $g(s') = g(s) + c(s,s')$; $bp(s') = s$;
insert s' into *OPEN*;

What You Should Know...

- Operation of A*
- Understand why A* returns an optimal solution (e.g., understand the sketch of proof)
- Theoretical properties of A*
- Properties of heuristics (e.g., admissibility, consistency)