

# Incremental Path Planning

Continuous Planning and Dynamic A\*

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## Outline

- Optimal Path Planning in Partially Known Environments.
- Continuous Optimal Path Planning
  - Dynamic A\*
  - Incremental A\* (LRTA\*) [Appendix]

## [Zellinsky, 92]

1. Generate global path plan from initial map.
2. Repeat until goal reached or failure:
  - Execute next step in current global path plan
  - Update map based on sensors.
  - If map changed generate new global path from map.

## Compute Optimal Path



J	M	N	O
E	I	L	G
B	D	H	K
S	A	C	F

## Begin Executing Optimal Path

h=4 J	h=3 M	h=2 N	h=1 O
h=3 E	h=2 I	h=1 L	h=0 G
h=4 B	h=3 D	h=2 H	h=1 K
h=5 S	h=4 A	h=3 C	h=2 F

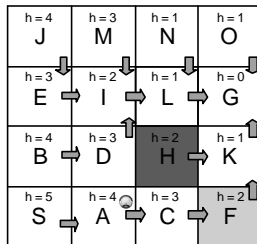
- Robot moves along backpointers towards goal.
- Uses sensors to detect discrepancies along way.

## Obstacle Encountered!

h=4 J	h=3 M	h=1 N	h=1 O
h=3 E	h=2 I	h=1 L	h=0 G
h=4 B	h=3 D	h=2 H	h=1 K
h=5 S	h=4 A	h=3 C	h=2 F

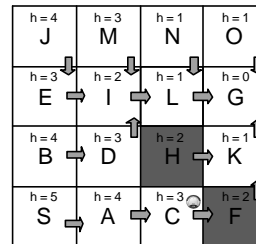
- At state A, robot discovers edge from D to H is blocked (cost 5,000 units).
- Update map and reinvok planner.

## Continue Path Execution



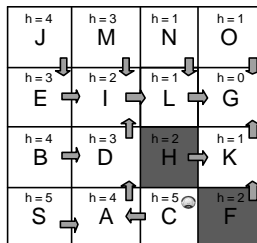
- A's previous path is still optimal.
- Continue moving robot along back pointers.

## Second Obstacle, Replan!



- At C robot discovers blocked edges from C to F and H (cost 5,000 units).
- Update map and reinvoked planner.

## Path Execution Achieves Goal



- Follow back pointers to goal.
- No further discrepancies detected; goal achieved!

## Outline

- Optimal Path Planning in Partially Known Environments.
- Continuous Optimal Path Planning
  - Dynamic A\*
  - Incremental A\* (LRTA\*) [Appendix]

## What is Continuous Optimal Path Planning?

- Supports search as a repetitive online process.
- Exploits similarities between a series of searches to solve much faster than solving each search starting from scratch.
- Reuses the identical parts of the previous search tree, while updating differences.
- Solutions guaranteed to be optimal.
- On the first search, behaves like traditional algorithms.
  - D\* behaves exactly like Dijkstra's.
  - Incremental A\* A\* behaves exactly like A\*.

## Dynamic A\* (aka D\*) [Stenz, 94]

1. Generate global path plan from initial map.
  2. Repeat until Goal reached, or failure.
    - Execute next step of current global path plan.
    - Update map based on sensor information.
    - Incrementally update global path plan from map changes.
- 1 to 3 orders of magnitude speedup relative to a non-incremental path planner.

## Map and Path Concepts

- $c(X, Y)$  :  
Cost to move from Y to X.  
 $c(X, Y)$  is undefined if move disallowed.
- $Neighbors(X)$  :  
Any Y such that  $c(X, Y)$  or  $c(Y, X)$  is defined.
- $o(G, X)$  :  
True optimal path cost to Goal from X.
- $h(G, X)$  :  
Estimate of optimal path cost to goal from X.
- $b(X) = Y$  : *backpointer* from X to Y.  
Y is the first state on path from X to G.

## D\* Search Concepts

- *State tag*  $t(X)$  :
  - *NEW* : has no estimate h.
  - *OPEN* : estimate needs to be propagated.
  - *CLOSED* : estimate propagated.
- *OPEN list* :  
States with estimates to be propagated to other states.
  - States on list tagged OPEN
  - Sorted by key function k (defined below).

## D\* Fundamental Search Concepts

- $k(G, X)$  : key function  
Minimum of
  - $h(G, X)$  before modification, and
  - all values assumed by  $h(G, X)$  since X was placed on the OPEN list.
- *Lowered state* :  $k(G, X) = \text{current } h(G, X)$ ,  
□ Propagate decrease to descendants and other nodes.
- *Raised state* :  $k(G, X) < \text{current } h(G, X)$ ,  
□ Propagate increase to descendants and other nodes.  
□ Try to find alternate shorter paths.

## Running D\* First Time on Graph

Initially

- Mark G Open and Queue it
- Mark all other states New
- Run Process\_States on queue until path found or empty.

When edge cost  $c(X, Y)$  changes

- If X is marked Closed, then
  - Update  $h(X)$
  - Mark X open and queue with key  $h(X)$ .

## Use D\* to Compute Initial Path



J NEW	M NEW	N NEW	O NEW
E NEW	I NEW	L NEW	G NEW
B NEW	D NEW	H NEW	K NEW
S NEW	A NEW	C NEW	F NEW

States initially tagged NEW (no cost determined yet).

## Use D\* to Compute Initial Path

J NEW	M NEW	N NEW	O NEW
E NEW	I NEW	L NEW	G OPEN
B NEW	D NEW	H NEW	K NEW
S NEW	A NEW	C NEW	F NEW

OPEN List	
1	(0, G)

```

8: if kold = h(X) then
9: for each neighbor Y of X:
10: if t(Y) = NEW or
11:    (b(Y) = X and h(Y) ? h(X) + c(X, Y)) or
12:    (b(Y) ? X and h(Y) > h(X) + c(X, Y)) then
13:    b(Y) = X; Insert(Y, h(X) + c(X, Y))
    
```

- Add Goal node to the OPEN list.
- Process OPEN list until the robot's current state is CLOSED.

## Process\_State: New or Lowered State

- Remove from Open list, state X with lowest k
- If X is a new/lowered state, its path cost is optimal!  
Then propagate to each neighbor Y
  - If Y is New, give it an initial path cost and propagate.
  - If Y is a descendant of X, propagate any change.
  - Else, if X can lower Y's path cost, Then do so and propagate.

## Use D\* to Compute Initial Path

J NEW	M NEW	N NEW	O NEW
E NEW	I NEW	L NEW	G OPEN
B NEW	D NEW	H NEW	K NEW
S NEW	A NEW	C NEW	F NEW

OPEN List	
1	(0, G)

8: if kold = h(X) then  
 9: for each neighbor Y of X:  
 10: if t(Y) = NEW or  
 11: (b(Y) = X and h(Y) ? h(X) + c(X,Y)) or  
 12: (b(Y) ? X and h(Y) > h(X) + c(X,Y)) then  
 13: b(Y) = X; Insert(Y, h(X) + c(X,Y))

- Add new neighbors of G onto the OPEN list
- Create backpointers to G.

## Use D\* to Compute Initial Path

J NEW	M NEW	N NEW	O OPEN
E NEW	I NEW	L OPEN	G CLOSED
B NEW	D NEW	H NEW	K OPEN
S NEW	A NEW	C NEW	F NEW

OPEN List	
1	(0, G)
2	(1, K) (1, L) (1, O)

8: if kold = h(X) then  
 9: for each neighbor Y of X:  
 10: if t(Y) = NEW or  
 11: (b(Y) = X and h(Y) ? h(X) + c(X,Y)) or  
 12: (b(Y) ? X and h(Y) > h(X) + c(X,Y)) then  
 13: b(Y) = X; Insert(Y, h(X) + c(X,Y))

- Add new neighbors of G onto the OPEN list
- Create backpointers to G.

## Use D\* to Compute Initial Path

J NEW	M NEW	N NEW	O OPEN
E NEW	I NEW	L OPEN	G CLOSED
B NEW	D NEW	H OPEN	K CLOSED
S NEW	A NEW	C NEW	F OPEN

OPEN List	
1	(0, G)
2	(1, K) (1, L) (1, O)
3	(1, L) (1, O) (2, F) (2, H)

8: if kold = h(X) then  
 9: for each neighbor Y of X:  
 10: if t(Y) = NEW or  
 11: (b(Y) = X and h(Y) ? h(X) + c(X,Y)) or  
 12: (b(Y) ? X and h(Y) > h(X) + c(X,Y)) then  
 13: b(Y) = X; Insert(Y, h(X) + c(X,Y))

- Add new neighbors of K onto the OPEN list
- Create backpointers.

## Use D\* to Compute Initial Path

J NEW	M NEW	N OPEN	O CLOSED
E NEW	I OPEN	L OPEN	G CLOSED
B NEW	D NEW	H OPEN	K CLOSED
S NEW	A NEW	C NEW	F OPEN

OPEN List	
1	(0, G)
2	(1, K) (1, L) (1, O)
3	(1, L) (1, O) (2, F) (2, H)
4	(2, F) (2, H) (2, I) (2, N)

8: if kold = h(X) then  
 9: for each neighbor Y of X:  
 10: if t(Y) = NEW or  
 11: (b(Y) = X and h(Y) ? h(X) + c(X,Y)) or  
 12: (b(Y) ? X and h(Y) > h(X) + c(X,Y)) then  
 13: b(Y) = X; Insert(Y, h(X) + c(X,Y))

- Add new neighbors of L, then O on to the OPEN list
- Create backpointers.

## Use D\* to Compute Initial Path

J NEW	M NEW	N OPEN	O CLOSED
E NEW	I OPEN	L CLOSED	G CLOSED
B NEW	D NEW	H OPEN	K CLOSED
S NEW	A NEW	C OPEN	F CLOSED

OPEN List	
1	(0, G)
2	(1, K) (1, L) (1, O)
3	(1, L) (1, O) (2, F) (2, H)
4	(2, F) (2, H) (2, I) (2, N)
5	(2, H) (2, I) (2, N) (3, C)

- Continue until current state S is closed.

## Use D\* to Compute Initial Path

J NEW	M NEW	N OPEN h=2	O CLOSED h=1
E NEW	I OPEN h=2	L CLOSED h=1	G CLOSED h=0
B NEW	D OPEN h=3	H CLOSED h=2	K CLOSED h=1
S NEW	A NEW	C OPEN h=3	F CLOSED h=2

OPEN List	
1	(0,G)
2	(1,K) (1,L) (1,O)
3	(1,L) (1,O) (2,F) (2,H)
4	(2,F) (2,H) (2,I) (2,N)
5	(2,H) (2,I) (2,N) (3,C)
6	(2,I) (2,N) (3,C) (3,D)

Continue until current state S is closed.

## Use D\* to Compute Initial Path

J NEW	M OPEN h=3	N OPEN h=2	O CLOSED h=1
E OPEN h=3	I CLOSED h=2	L CLOSED h=1	G CLOSED h=0
B NEW	D OPEN h=3	H CLOSED h=2	K CLOSED h=1
S NEW	A NEW	C OPEN h=3	F CLOSED h=2

OPEN List	
1	(0,G)
2	(1,K) (1,L) (1,O)
3	(1,L) (1,O) (2,F) (2,H)
4	(2,F) (2,H) (2,I) (2,N)
5	(2,H) (2,I) (2,N) (3,C)
6	(2,I) (2,N) (3,C) (3,D)
7	(2,N) (3,C) (3,D) (3,E) (3,M)

Continue until current state S is closed.

## Use D\* to Compute Initial Path

J NEW	M OPEN h=3	N CLOSED h=2	O CLOSED h=1
E OPEN h=3	I CLOSED h=2	L CLOSED h=1	G CLOSED h=0
B NEW	D OPEN h=3	H CLOSED h=2	K CLOSED h=1
S NEW	A NEW	C OPEN h=3	F CLOSED h=2

OPEN List	
1	(0,G)
2	(1,K) (1,L) (1,O)
3	(1,L) (1,O) (2,F) (2,H)
4	(2,F) (2,H) (2,I) (2,N)
5	(2,H) (2,I) (2,N) (3,C)
6	(2,I) (2,N) (3,C) (3,D)
7	(2,N) (3,C) (3,D) (3,E) (3,M)
8	(3,C) (3,D) (3,E) (3,M)

Continue until current state S is closed.

## Use D\* to Compute Initial Path

J NEW	M OPEN h=3	N CLOSED h=2	O CLOSED h=1
E OPEN h=3	I CLOSED h=2	L CLOSED h=1	G CLOSED h=0
B NEW	D OPEN h=3	H CLOSED h=2	K CLOSED h=1
S NEW	A OPEN h=4	C CLOSED h=3	F CLOSED h=2

OPEN List	
1	(0,G)
2	(1,K) (1,L) (1,O)
3	(1,L) (1,O) (2,F) (2,H)
4	(2,F) (2,H) (2,I) (2,N)
5	(2,H) (2,I) (2,N) (3,C)
6	(2,I) (2,N) (3,C) (3,D)
7	(2,N) (3,C) (3,D) (3,E) (3,M)
8	(3,C) (3,D) (3,E) (3,M)
9	(3,D) (3,E) (3,M) (4,A)

Continue until current state S is closed.

## Use D\* to Compute Initial Path

J NEW	M OPEN h=3	N CLOSED h=2	O CLOSED h=1
E OPEN h=3	I CLOSED h=2	L CLOSED h=1	G CLOSED h=0
B OPEN h=4	D CLOSED h=3	H CLOSED h=2	K CLOSED h=1
S NEW	A OPEN h=4	C CLOSED h=3	F CLOSED h=2

OPEN List	
1	(0,G)
2	(1,K) (1,L) (1,O)
3	(1,L) (1,O) (2,F) (2,H)
4	(2,F) (2,H) (2,I) (2,N)
5	(2,H) (2,I) (2,N) (3,C)
6	(2,I) (2,N) (3,C) (3,D)
7	(2,N) (3,C) (3,D) (3,E) (3,M)
8	(3,C) (3,D) (3,E) (3,M)
9	(3,D) (3,E) (3,M) (4,A)
10	(3,E) (3,M) (4,A) (4,B)

Continue until current state S is closed..

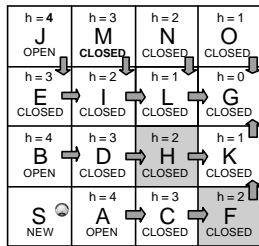
## Use D\* to Compute Initial Path

J OPEN h=4	M OPEN h=3	N CLOSED h=2	O CLOSED h=1
E CLOSED h=3	I CLOSED h=2	L CLOSED h=1	G CLOSED h=0
B OPEN h=4	D CLOSED h=3	H CLOSED h=2	K CLOSED h=1
S NEW	A OPEN h=4	C CLOSED h=3	F CLOSED h=2

OPEN List	
10	(3,E) (3,M) (4,A) (4,B)
11	(3,M) (4,A) (4,B) (4,J)
12	
13	
14	

Continue until current state S is closed.

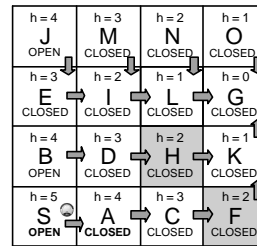
## Use D\* to Compute Initial Path



OPEN List	
10	(3,E) (3,M) (4,A) (4,B)
11	(3,M) (4,A) (4,B) (4,J)
12	(3,M) (4,A) (4,B)
13	
14	

- Continue until current state S is closed.

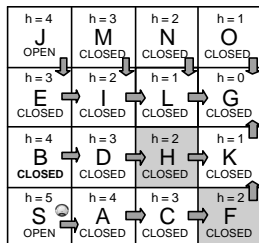
## Use D\* to Compute Initial Path



OPEN List	
10	(3,E) (3,M) (4,A) (4,B)
11	(3,M) (4,A) (4,B) (4,J)
12	(3,M) (4,A) (4,B)
13	(4,B) (4,J) (5,S)
14	
15	

- Continue until current state S is closed.

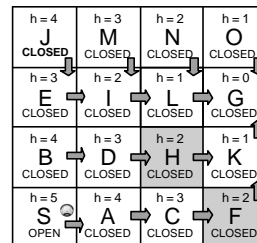
## Use D\* to Compute Initial Path



OPEN List	
10	(3,E) (3,M) (4,A) (4,B)
11	(3,M) (4,A) (4,B) (4,J)
12	(3,M) (4,A) (4,B)
13	(4,B) (4,J) (5,S)
14	(4,J) (5,S)
15	

- Continue until current state S is closed.

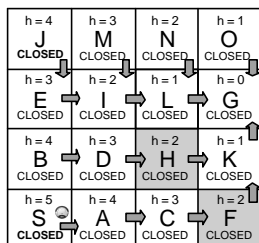
## Use D\* to Compute Initial Path



OPEN List	
10	(3,E) (3,M) (4,A) (4,B)
11	(3,M) (4,A) (4,B) (4,J)
12	(3,M) (4,A) (4,B)
13	(4,B) (4,J) (5,S)
14	(4,J) (5,S)
15	(5,S)

- Continue until current state S is closed.

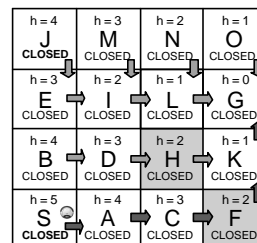
## D\* Completed Initial Path



OPEN List	
10	(3,E) (3,M) (4,A) (4,B)
11	(3,M) (4,A) (4,B) (4,J)
12	(3,M) (4,A) (4,B)
13	(4,B) (4,J) (5,S)
14	(4,J) (5,S)
15	(5,S)
16	NULL

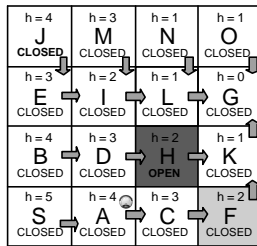
- Done: Current state S is closed, and Open list is empty.

## Begin Executing Optimal Path



- Robot moves along backpointers towards goal
- Uses sensors to detect discrepancies along way.

## Obstacle Encountered!



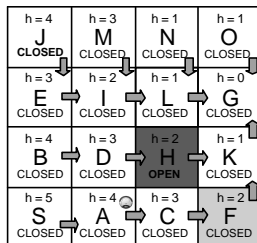
- At state A, robot discovers edge D to H is blocked off (cost 5,000 units).
- Update map and reinvoked D\*

## Running D\* After Edge Cost Change

When edge cost  $c(X,Y)$  changes

- If X is marked Closed, then
  - Update  $h(X)$
  - Mark X open and queue, key is new  $h(X)$ .
- Run Process\_State on queue
  - until path to current state is shown optimal,
  - or queue Open List is empty.

## D\* Update From First Obstacle

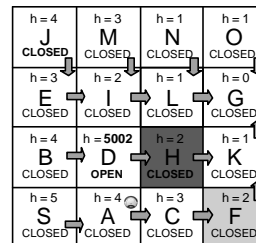


OPEN List	
1	(2,H)
2	
3	
4	

Function: Modify-Cost(X,Y,eval)  
 1:  $c(X,Y) = \text{eval}$   
 2: if  $h(X) = \text{CLOSED}$  then Insert( $X, h(X)$ )  
 3: return Get-Kmin()

- Assign cost of 5,000 for D to H
- Propagate changes starting at H

## D\* Update From First Obstacle



OPEN List	
1	(2,H)
2	(3,D)
3	
4	

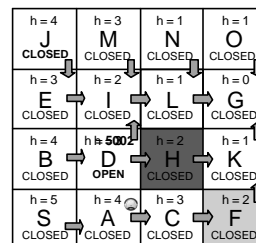
8: if kold =  $h(X)$  then  
 9: for each neighbor Y of X:  
 10: if  $h(Y) = \text{NEW}$  or  
 11:  $(b(Y) = X \text{ and } h(Y) > h(X) + c(X,Y))$  or  
 12:  $(b(Y) \neq X \text{ and } h(Y) > h(X) + c(X,Y))$  then  
 13:  $b(Y) = X; h(Y) = h(X) + c(X,Y)$

- Raise cost of H's descendant D, and propagate.

## Process\_State: Raised State

- If X is a raise state its cost might be suboptimal.
  - Try reducing cost of X using an *optimal neighbor* Y.
    - $h(Y) = h(X)$  before it was raised]
  - Propagate X's cost to each neighbor Y
    - If Y is New, Then give it an initial path cost and propagate.
    - If Y is a descendant of X, Then propagate ANY change .
    - If X can lower Y's path cost,
      - Postpone: Queue X to propagate when optimal (reach current  $h(X)$ )
    - If Y can lower X's path cost, and Y is suboptimal,
      - Postpone: Queue Y to propagate when optimal (reach current  $h(Y)$ ).
- Postponement avoids creating cycles.

## D\* Update From First Obstacle

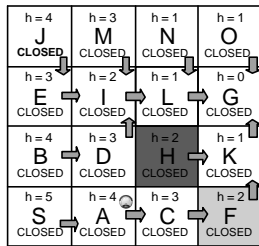


OPEN List	
1	(2,H)
2	(3,D)
3	
4	

4: if kold <  $h(X)$  then  
 5: for each neighbor Y of X:  
 6: if  $h(Y) = \text{kold}$  and  $h(X) > h(Y) + C(Y,X)$  then  
 7:  $b(X) = Y; h(X) = h(Y) + c(Y,X)$

- D may not be optimal, check neighbors for better path.
- Transitioning to I is better, and I's path is optimal, so update D.

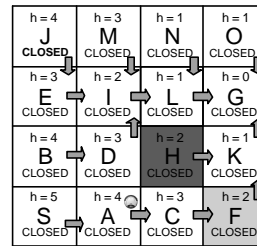
## D\* Update From First Obstacle



OPEN List	
1	
2	(3,D)
3	NULL
4	

- All neighbors of D have consistent h-values.
- No further propagation needed.

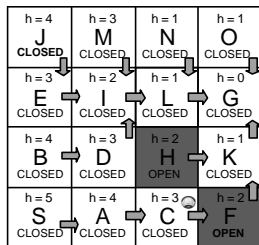
## Continue Path Execution



OPEN List	
1	
2	(3,D)
3	NULL
4	

- A's path optimal.
- Continue moving robot along backpointers.

## Second Obstacle!

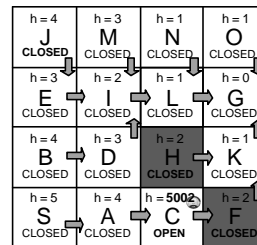


OPEN List	
1	(2,F) (2,H)
2	
3	
4	

Function: ModifyCost(X,Y,eval)  
 1:  $c(X,Y) = \text{eval}$   
 2: if  $t(X) = \text{CLOSED}$  then  $\text{Insert}(X, h(X))$   
 3: return  $\text{Get-Kmin}()$

- At C robot discovers blocked edges C to F and H (cost 5,000 units).
- Update map and reinvoke D\* until H (current position optimal).

## D\* Update From Second Obstacle

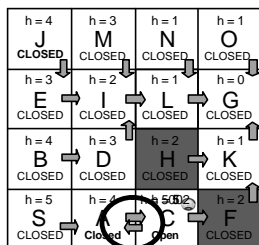


OPEN List	
1	(2,F) (2,H)
2	(3,C)
3	
4	

8: if  $\text{kold} = h(X)$  then  
 9: for each neighbor Y of X:  
 10: if  $t(Y) = \text{NEW}$  or  
 11:  $(b(Y) = X \text{ and } h(Y) ? h(X) + c(X,Y))$  or  
 12:  $(b(Y) ? X \text{ and } h(Y) > h(X) + c(X,Y))$  then  
 13:  $b(Y) = X; \text{Insert}(Y, h(X) + c(X,Y))$

- Processing F raises descendant C's cost, and propagates.
- Processing H does nothing.

## D\* Update From Second Obstacle

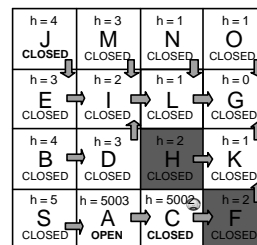


OPEN List	
1	(2,F) (2,H)
2	(3,C)
3	
4	

4: if  $\text{kold} < h(X)$  then  
 5: for each neighbor Y of X:  
 6: if  $h(Y) = \text{kold}$  and  $h(X) > h(Y) + c(Y,X)$  then  
 7:  $b(X) = Y; h(X) = h(Y) + c(Y,X)$

- C may be suboptimal, check neighbors; → Better path through A!
- However, A may be suboptimal, and updating would create a loop!

## D\* Update From Second Obstacle



OPEN List	
1	(2,F) (2,H)
2	(3,C)
3	(4,A)
4	

15: for each neighbor Y of X:  
 16: if  $t(Y) = \text{NEW}$  or  
 17:  $(b(Y) = X \text{ and } h(Y) ? h(X) + c(X,Y))$  then  
 18:  $b(Y) = X; \text{Insert}(Y, h(X) + c(X,Y))$

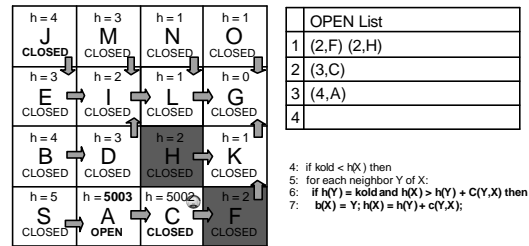
- Don't change C's path to A (yet).
- Instead, propagate increase to A.



## Process\_State: Raised State

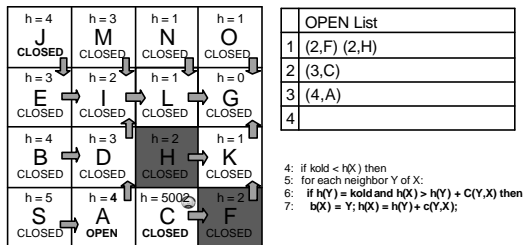
- If X is a raise state its cost might be suboptimal.
  - Try reducing cost of X using an *optimal neighbor* Y.
    - $h(Y) = \lfloor h(X) \rfloor$  before it was raised]
  - propagate X's cost to each neighbor Y
    - If Y is New, Then give it an initial path cost and propagate.
    - If Y is a descendant of X, Then propagate ANY change .
    - If X can lower Y's path cost,
      - Postpone: Queue X to propagate when optimal (reach current  $h(X)$ )
    - If Y can lower X's path cost, and Y is suboptimal,
      - Postpone: Queue Y to propagate when optimal (reach current  $h(Y)$ ).
- Postponement avoids creating cycles.

## D\* Update From Second Obstacle



- A may not be optimal, check neighbors for better path.
- Transitioning to D is better, and D's path is optimal, so update A.

## D\* Update From Second Obstacle

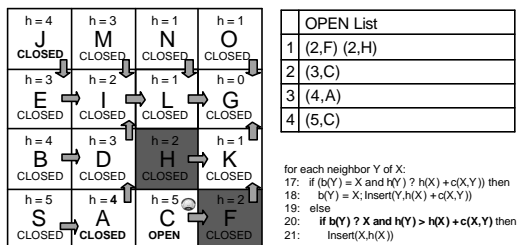


- A may not be optimal, check neighbors for better path.
- Transitioning to D is better, and D's path is optimal, so update A.

## Process\_State: New or Lowered State

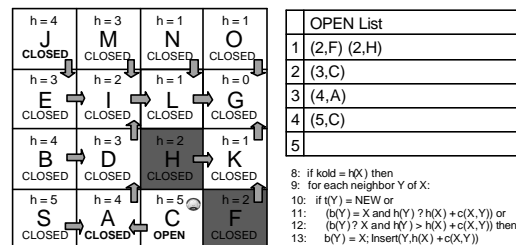
- Remove from Open list , state X with lowest k
- If X is a new/lowered state its path cost is optimal, Then propagate to each neighbor Y
  - If Y is New, give it an initial path cost and propagate.
  - If Y is a descendant of X, propagate any change.
  - Else, if X can lower Y's path cost, Then do so and propagate.

## D\* Update From Second Obstacle



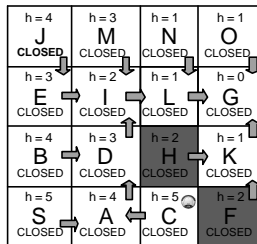
- A can improve neighbor C, so queue C.

## D\* Update From Second Obstacle



- C lowered to optimal; no neighbors affected.
- Current state reached, so Process\_State terminates.

## Complete Path Execution



- Follow back pointers to Goal.
- No further discrepancies detected; goal achieved!

## D\* Pseudo Code

```

Function: Process-State()
1: X ← Min-State()
2: if X=NULL then return -1
3: kold ← Get-Kmin(); Delete(X)
4: if kold < h(X) then
5:   for each neighbor Y of X:
6:     if h(Y) = kold and h(X) > h(Y) + C(Y,X) then
7:       b(X) = Y; h(X) = h(Y) + c(Y,X)
8:   if kold = h(X) then
9:     for each neighbor Y of X:
10:      if t(Y) = NEW or
11:         (b(Y) = X and h(Y) > h(X) + c(X,Y)) or
12:         (b(Y) ≠ X and h(Y) > h(X) + c(X,Y)) then
13:        b(Y) = X; Insert(Y, h(X) + c(X,Y))
14:   else
15:     for each neighbor Y of X:
16:       if t(Y) = NEW or
17:          (b(Y) = X and h(Y) > h(X) + c(X,Y)) then
18:        b(Y) = X; Insert(Y, h(X) + c(X,Y))
19:   else
20:     if b(Y) ≠ X and h(X) > h(Y) + c(X,Y) then
21:       Insert(X, h(X))
22:   else
23:     if b(Y) ≠ X and h(X) > h(Y) + c(X,Y) and
24:        t(Y) = CLOSED and h(Y) > kold then
25:       Insert(Y, h(Y))
26: return Get-Kmin()

Function: Modify-Cost(X,Y,eval)
1: c(X,Y) = eval
2: if t(X) = CLOSED
   then Insert(X, h(X))
3: return Get-Kmin()

Function: Insert(X, h_new)
1: if t(X) = NEW
   then k(x) = h_new
2: else if t(X) = OPEN
   then k(X) = min(k(X), h_new)
3: else
   k(X) = min(h(X), h_new)
4: h(X) = h_new
5: t(X) = OPEN
    
```

## Recap: Continuous Optimal Planning

- Generate global path plan from initial map.
  - Repeat until Goal reached, or failure.
    - Execute next step of current global path plan.
    - Update map based on sensor information.
    - Incrementally update global path plan from map changes.
- 1 to 3 orders of magnitude speedup relative to a non-incremental path planner.

## Recap: Dynamic A\*

- Supports search as a repetitive online process.
- Exploits similarities between a series of searches to solve much faster than from scratch.
- Reuses the identical parts of the previous search tree, while updating differences.
- Solutions guaranteed to be optimal.
- On the first search, behaves like traditional Dijkstra.