

# **Heuristic search wrap-up, and local search**

# Announcements

- Assignment 0 released
- Survey released (closing tonight)

# A\* Search

- Best First Search with  $f(s) = g(s) + h(s)$ , where:
    - $g(s)$  = cost of best path found so far to  $s$
    - $h(s)$  = **admissible** heuristic function
1. If GOAL?(initial-state) then return **initial-state**
  2. INSERT(initial-node, FRINGE)
  3. Repeat:
  4. If empty(FRINGE) then return **failure**
  5.  $s \leftarrow \text{REMOVE}(\text{FRINGE})$
  6. If GOAL?( $s$ ) then return **s** and/or path
  7. For every state  $s'$  in SUCC( $s$ ):
  8. INSERT( $s'$ , FRINGE)

# Reminder: Admissible Heuristic

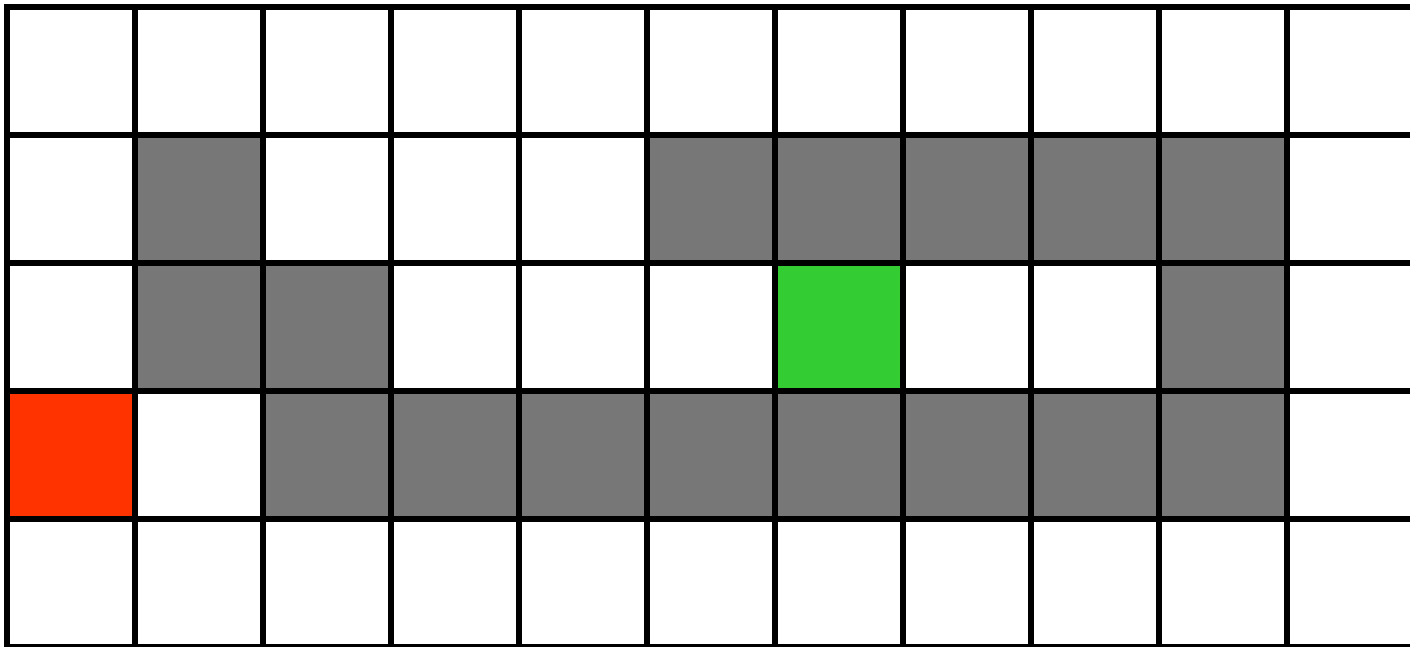
- An heuristic  $h(s)$  is **admissible** if for any state  $s$ ,

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

where  $h^*(s)$  is the optimal cost from  $s$  to a goal.

- In other words, an admissible heuristic **never overestimates the cost to the goal**
  - We'll need to design  $h(s)$  so that it's always less than  $h^*(s)$ , even though we don't know  $h^*(s)$ !

# Robot Navigation



# Robot Navigation

$f(s) = h(s)$ , with  $h(s)$  = Manhattan distance to the goal  
(not A\*)

8	7	6	5	4	3	2	3	4	5	6
7		5	4	3						5
6			3	2	1	0	1	2		4
7	6									5
8	7	6	5	4	3	2	3	4	5	6

# Robot Navigation

$f(s) = h(s)$ , with  $h(s)$  = Manhattan distance to the goal  
(not A\*)

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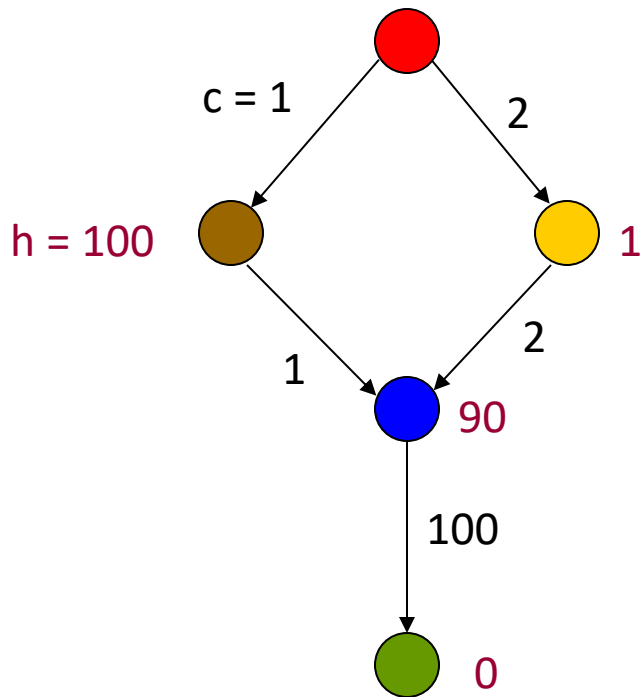
# Robot Navigation

$f(s) = g(s) + h(s)$ , with  $h(s)$  = Manhattan distance to goal ( $A^*$ )

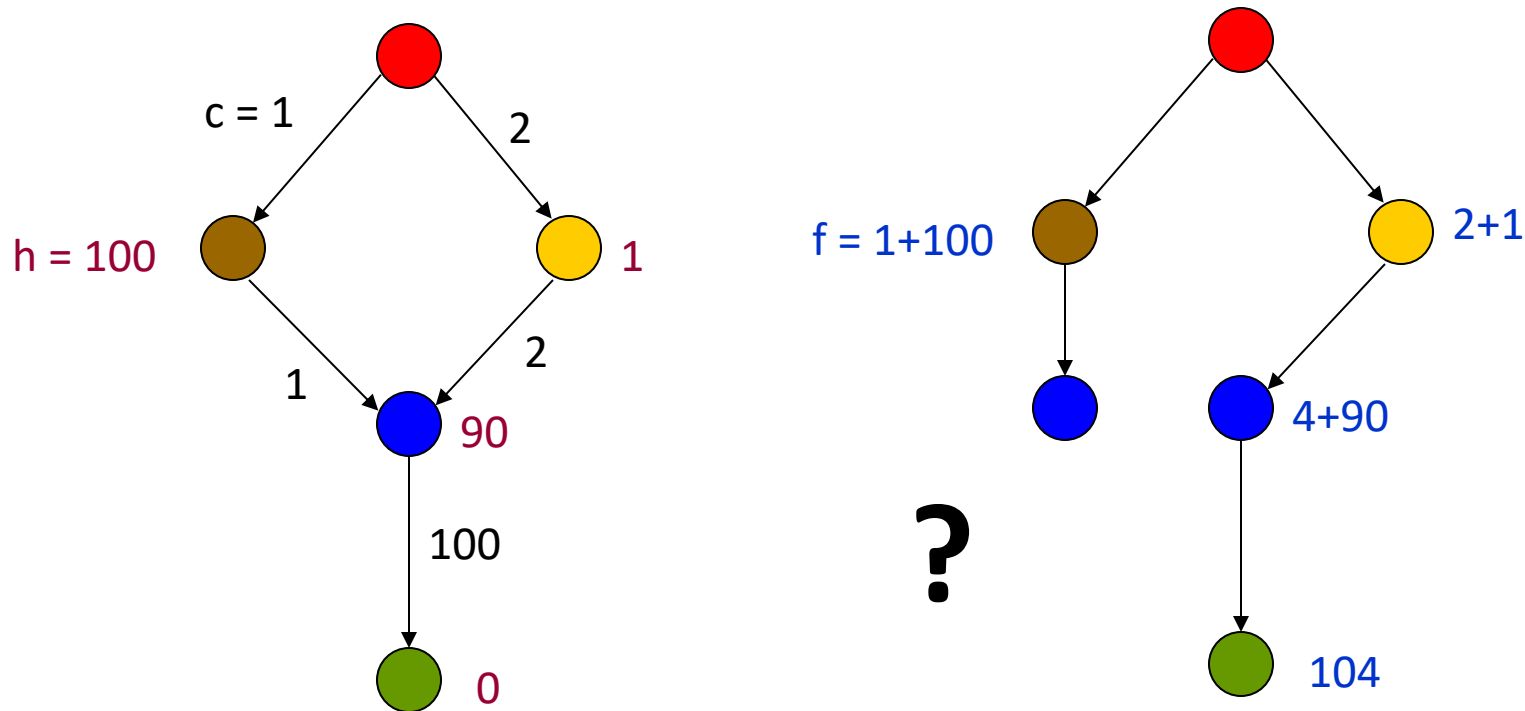
8+3	7+4	6+5	5+6	4+7	3+8	2+9	3+10	4	5	6
7+2		5+6	4+7	3+8						5
6+1			3	2+9	1+10	0+11	1	2		4
7+0	6+1									5
8+1	7+2	6+3	5+4	4+5	3+6	2+7	3+8	4	5	6



# What to do with revisited states?



# What to do with revisited states?



If we discard this new node, A\* expands the goal next, returning a non-optimal solution

1. Is  $A^*$  complete?
2. Is  $A^*$  optimal?
3. What is the running time of  $A^*$ ?
4. What are the memory requirements of  $A^*$ ?

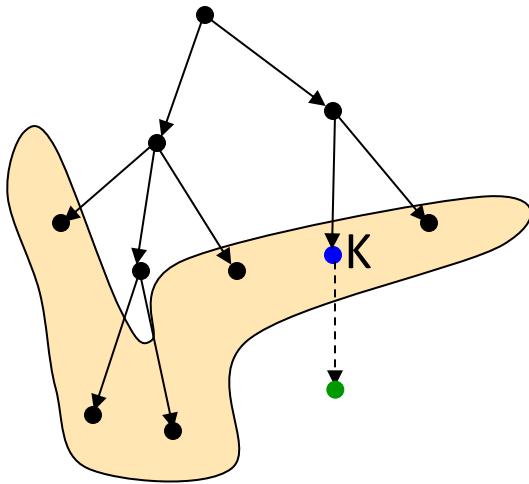
# $A^*$ is complete and optimal if:

- Duplicate states are revisited, and
- $h$  is admissible.
- Proof sketch:
  - First show that if a solution exists,  $A^*$  terminates and finds a solution.
  - Then show that whenever  $A^*$  expands a node, the path to that node is optimal.

$A^*$  is **complete** and **optimal** if  $h$  is admissible.

Proof sketch:

- If a solution exists,  $A^*$  terminates and returns a solution

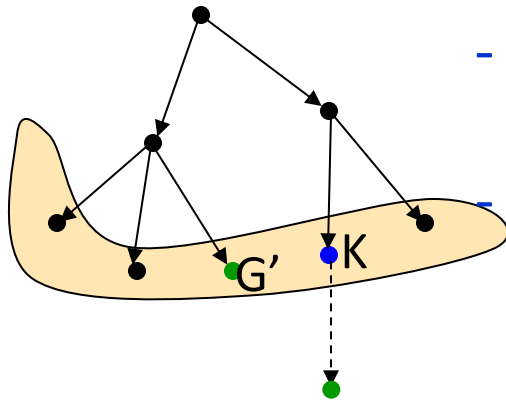


- As long as  $A^*$  hasn't terminated, a node  $K$  on the fringe lies on a solution path
- Since each node expansion increases the length of one path,  $K$  will eventually be selected for expansion, unless a solution is found along another path

**A\*** is **complete** and **optimal** if **h** is admissible.

Proof sketch:

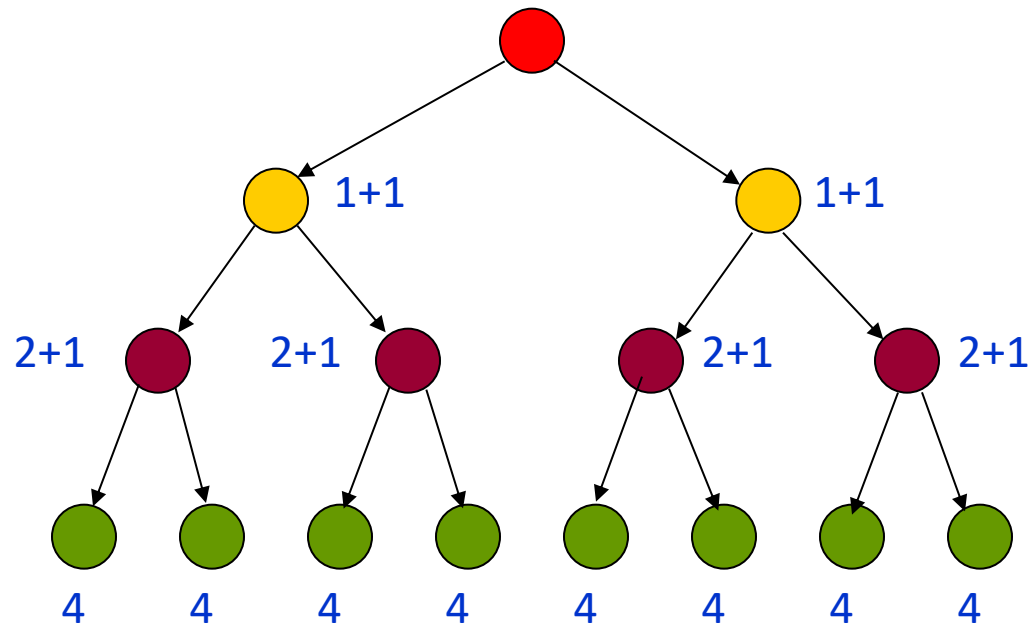
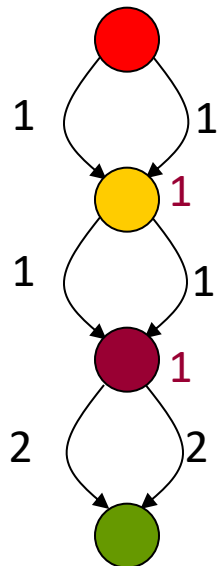
- Whenever A\* chooses to expand a goal node, the path to this node is optimal



- $C^*$  = cost of the optimal solution path
- $G'$ : non-optimal goal node in the fringe  
 $f(G') = g(G') + h(G') = g(G') > C^*$
- A node K in the fringe lies on an optimal path:  
 $f(K) = g(K) + h(K) \leq C^*$
- So,  $G'$  will not be selected for expansion

# Complexity of A\*

- A\* expands all nodes with  $f(s) < C^*$ 
  - May also expand non-goal states with  $f(s) = C^*$
  - May be an exponential number of nodes unless the heuristic is sufficiently *accurate*

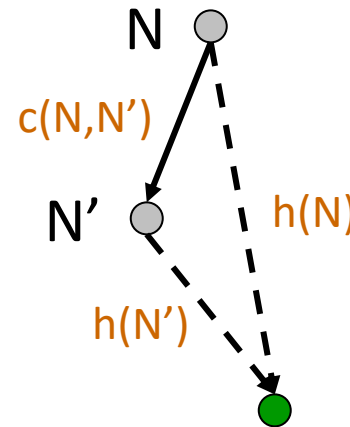


# Consistent Heuristic

- An **admissible** heuristic  $h$  is **consistent** (or **monotone**) if for each node  $N$  and each child  $N'$  of  $N$ :

$$h(N) \leq c(N, N') + h(N') \quad (\text{triangle inequality})$$

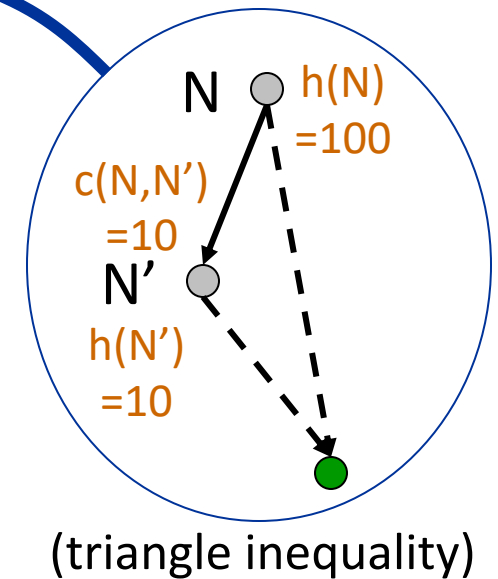
$$\text{Or: } h(N) - h(N') \leq c(N, N')$$





# Consistency Violation

If  $h$  says that  $N$  is 100 units from the goal, then moving from  $N$  along an edge costing 10 units should **not** lead to a node  $N'$  that  $h$  estimates to be 10 units away from the goal



Should satisfy:  $h(N) - h(N') \leq c(N, N')$

# 8-Puzzle

5		8
4	2	1
7	3	6

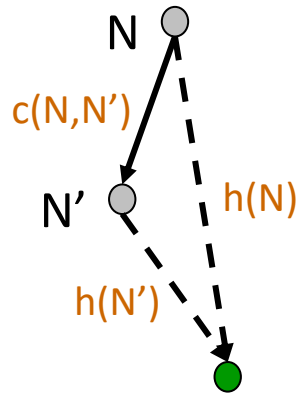
STATE(N)

1	2	3
4	5	6
7	8	

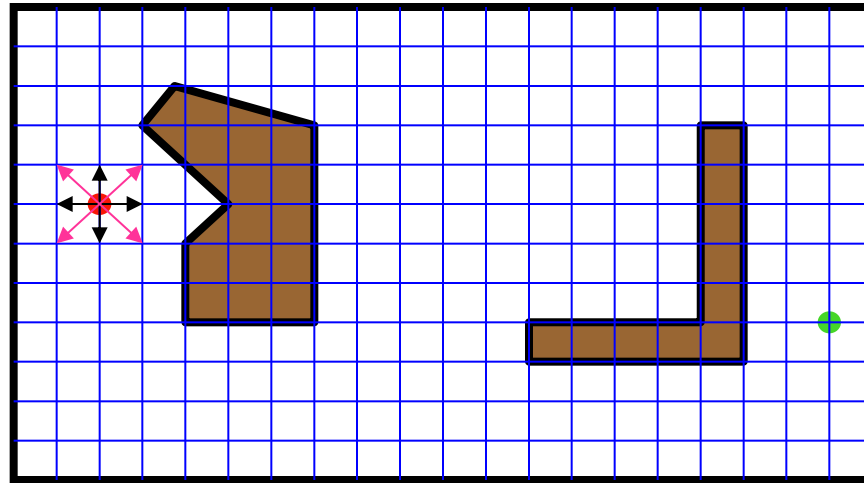
goal

- $h_1(N)$  = number of misplaced tiles
- $h_2(N)$  = sum of the (Manhattan) distances of every tile to its goal position

# Robot Navigation



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



Cost of one horizontal/vertical step = 1  
 Cost of one diagonal step =  $\sqrt{2}$

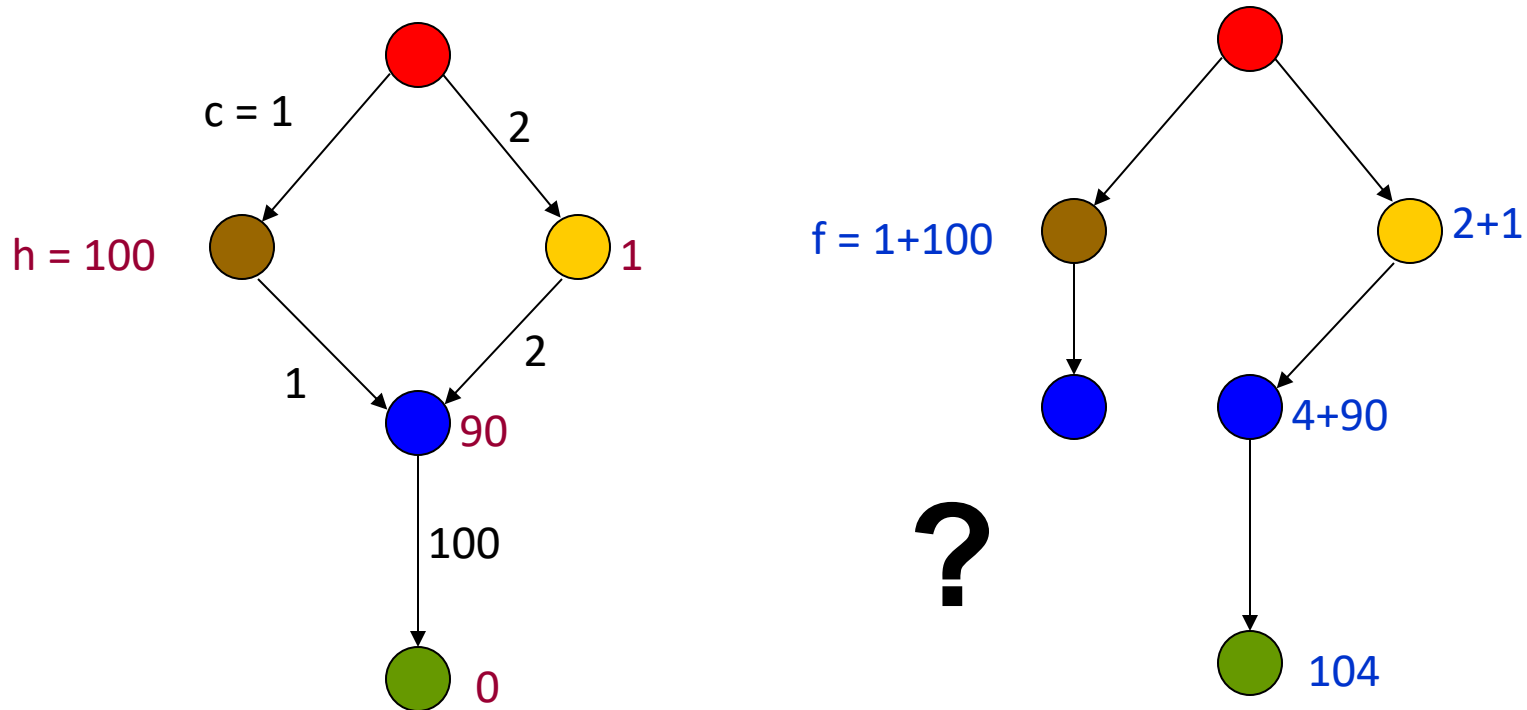
$$h_1(N) = \sqrt{(x_N - x_g)^2 + (y_N - y_g)^2}$$

$$h_2(N) = |x_N - x_g| + |y_N - y_g|$$

# Admissibility and Consistency

- A consistent heuristic is also admissible
- An admissible heuristic may not be consistent, but many admissible heuristics are consistent

# Revisiting - Is $h(\cdot)$ consistent?



# Updated Algorithm

1. If GOAL?(initial-state) then return initial-state
2. INSERT(initial-node, FRINGE)
3. Repeat:
4. If empty(FRINGE) then return failure
5.  $s \leftarrow \text{REMOVE}(\text{FRINGE})$
6. INSERT( $s$ , CLOSED)
7. If GOAL?( $s$ ) then return  $s$  and/or path
8. For every state  $s'$  in SUCC( $s$ ):
9. If  $s'$  in CLOSED, discard  $s'$
10. If  $s'$  in FRINGE with larger  $s'$ , remove from FRINGE
11. If  $s'$  not in FRINGE, INSERT( $s'$ , FRINGE)

# A\* is optimal if

- h is **admissible** (but not necessarily consistent)

- Revisited states not discarded

1. If GOAL?(initial-state) then return **initial-state**
2. INSERT(initial-node, FRINGE)
3. Repeat:
4. If empty(FRINGE) then return **failure**
5.  $s \leftarrow \text{REMOVE}(\text{FRINGE})$
6. If GOAL?(**s**) then return **s** and/or path
7. For every state **s'** in SUCC(**s**):
8.     INSERT(**s'**, FRINGE)

- h is **consistent**

- (Many) revisited states discarded:

1. If GOAL?(initial-state) then return **initial-state**
2. INSERT(initial-node, FRINGE)
3. Repeat:
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10.    If **s'** in FRINGE with larger **s'**, remove from FRINGE
11.    If **s'** not in FRINGE, INSERT(**s'**, FRINGE)

# Heuristic Accuracy

- If  $h_1$  and  $h_2$  are consistent heuristics such that  $h_1(N) \leq h_2(N)$  for all nodes  $N$ , then  $h_2$  is more **accurate** (or **informative**) than  $h_1$ 
  - The more accurate  $h$  is, the less work  $A^*$  has to do!

5		8
4	2	1
7	3	6

STATE(N)

1	2	3
4	5	6
7	8	

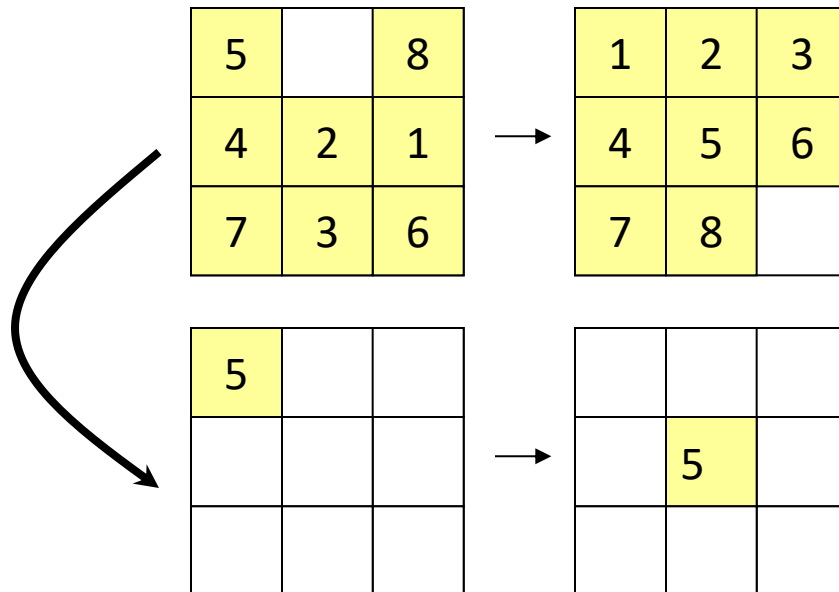
Goal state

- $h_1(N)$  = number of misplaced tiles
- $h_2(N)$  = sum of distances of every tile to its goal position
- Which is more accurate?



# How to create good heuristics?

- One approach: solve “relaxed” problems that ignore some constraints
  - E.g. ignore interactions among parts of the problem
  - In the 8-puzzle, the sum of the distances of each tile to its goal position ( $h_2$ ) corresponds to solving 8 simple problems:



$d_i$  is length of shortest path to move tile  $i$  to its goal position, ignoring the other tiles, and  
 $h_2 = \sum_{i=1,\dots,8} d_i$

# Can we do better?

- For example, we could consider two more complex relaxed problems:

$d_{1234}$  = length of shortest  
path  
to goal  
ignoring

5		8
4	2	1



1	2	3
4	5	6

→ Several order-of-magnitude speedups  
for the 15- and 24-puzzle (see R&N)

4	2	1
	3	



1	2	3
4		

5		8
7		6



	5	6
7	8	

- $h = d_{1234} + d_{5678}$  [disjoint pattern heuristic]
- These distances can be pre-computed and stored  
[Each requires generating a tree of 3,024 states (breadth-first search)]

# Experimental Results

(see R&N for details)

- Random 8-puzzles with:
  - $h_1$  = number of misplaced tiles
  - $h_2$  = sum of distances of tiles to their goal positions
- Average “effective branching factors” based on actual # of nodes expanded:

d	IDS	$A_1^*$	$A_2^*$
2	2.45	1.79	1.79
6	2.73	1.34	1.30
12	2.78 (3,644,035)	1.42 (227)	1.24 (73)
16	--	1.45	1.25
20	--	1.47	1.27
24	--	1.48 (39,135)	1.26 (1,641)

# Next class

- Local search
- Start with adversarial search and game playing