

# Heuristic Search

Computer Science cpsc322, Lecture 7

*(Textbook Chpt 3.6)*

Sept, 20, 2013



# Course Announcements

**Marks for Assignment0:** will be posted on Connect next week

**Assignment1:** will also be posted on Mon/Tue

If you are confused on basic search algorithm, different search strategies..... Check **learning goals** at the end of lectures. Work on the **Practice Exercises** and **Please do come to office hours**

*Giuseppe* : Fri 2-3, my office CICS R 105

*Kamyar Ardekani* Mon 2-3, X150 (Learning Center)

*Tatsuro Oya* Thur 11-12, X150 (Learning Center)

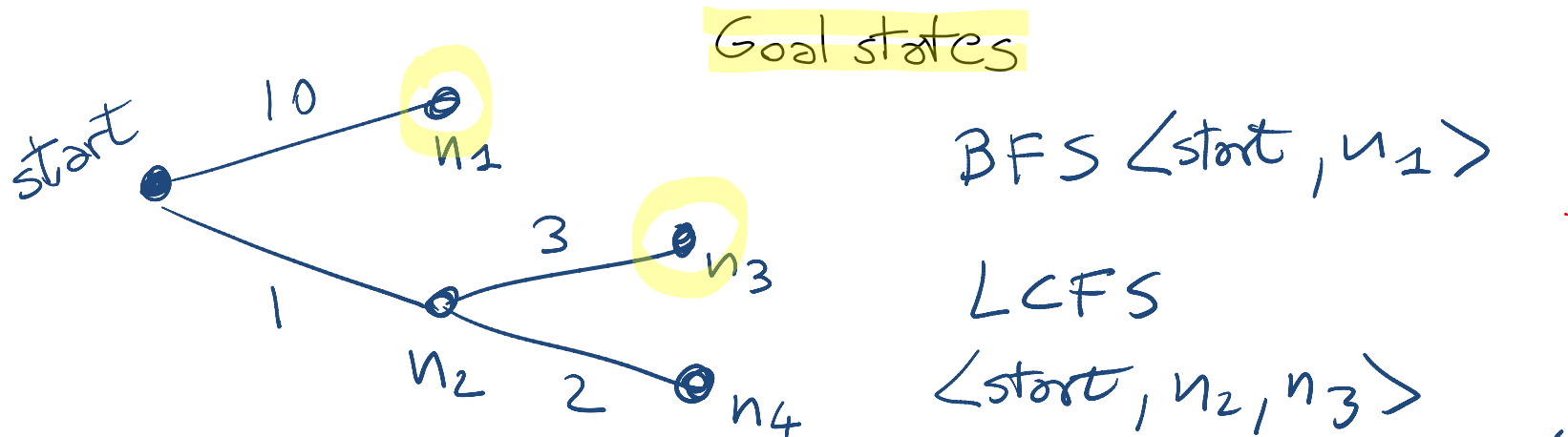
*Xin Ru (Nancy) Wang* Tue 2-3, X150 (Learning Center)

# Lecture Overview

- **Recap**
  - **Search with Costs**
  - **Summary Uninformed Search**
- Heuristic Search

# Recap: Search with Costs

- Sometimes there are **costs** associated with arcs.
  - The cost of a path is the sum of the costs of its arcs.



- **Optimal solution:** not the one that minimizes the *number of links*, but the one that minimizes *cost*
- **Lowest-Cost-First Search:** expand paths from the frontier in order of their costs.

# Recap Uninformed Search

	Complete	Optimal	Time	Space
DFS	N	N	$O(b^m)$	$O(mb)$
	Y if no cycles and finite search space			
BFS	Y	Y	$O(b^m)$	$O(b^m)$
IDS	Y	Y	$O(b^m)$	<u><math>O(mb)</math></u>
LCFS	Y	Y	$O(b^m)$	$O(b^m)$
	Costs > 0	Costs <u><math>\geq 0</math></u>		

# Recap Uninformed Search

- Why are all these strategies called uninformed?

Because they do not consider any information about **the states (end nodes)** to decide which path to expand first on the frontier

eg

$(\langle n_0, n_2, n_3 \rangle 12)$ ,  $(\langle n_0, n_3 \rangle 8)$ ,  $(\langle n_0, n_1, n_4 \rangle 13)$

In other words, they are general they do not take into account the **specific nature of the problem.**

# Lecture Overview

- **Recap**
  - Search with Costs
  - Summary Uninformed Search
- **Heuristic Search**

# Beyond uninformed search....

iclicker.

What information we could use to better select paths from the frontier?

- A. an estimate of the distance from the last node on the path to the goal
- B. an estimate of the distance from the start state to the goal
- C. an estimate of the cost of the path
- D. None of the above



# Heuristic Search

Uninformed/Blind search algorithms do not take into account the goal until they are at a goal node.

Often there is extra knowledge that can be used to guide the search: an *estimate* of the distance from node  $n$  to a goal node.

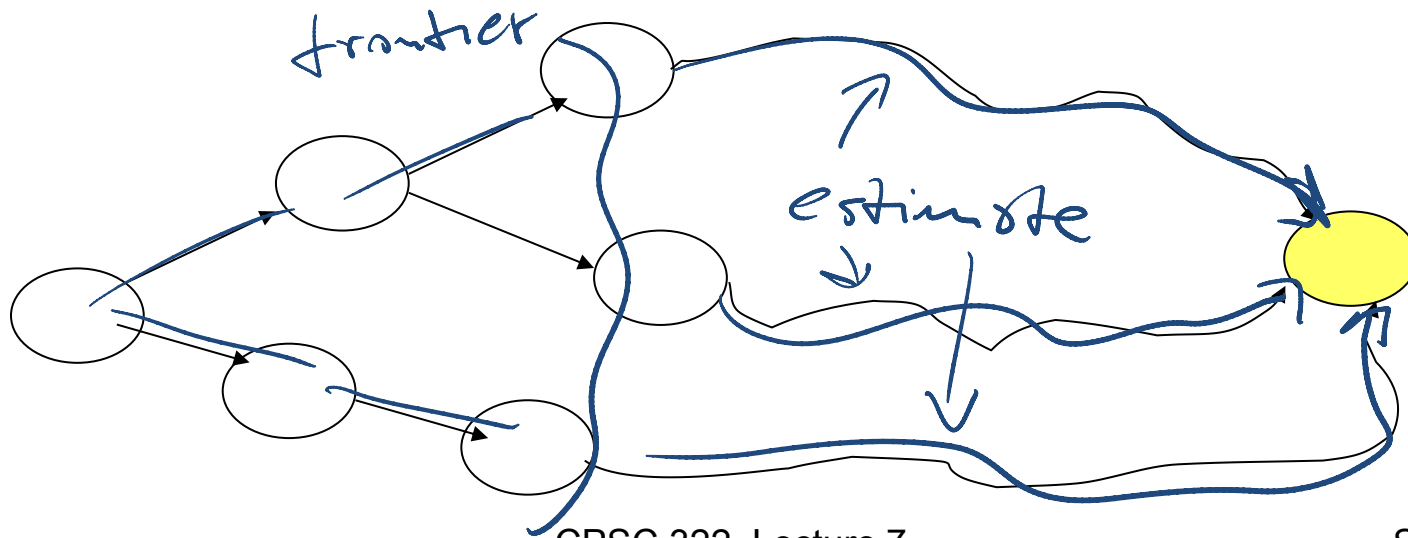
This is called a *heuristic*

# More formally

Definition (search heuristic)

A search heuristic  $h(n)$  is an estimate of the cost of the shortest path from node  $n$  to a goal node.

- $h$  can be extended to paths:  $h(\langle n_0, \dots, n_k \rangle) = h(n_k)$
- For now think of  $h(n)$  as only using readily obtainable information (that is easy to compute) about a node.



# More formally (cont.)

## Definition (admissible heuristic)

A search heuristic  $h(n)$  is **admissible** if it is never an overestimate of the cost from  $n$  to a goal. ←

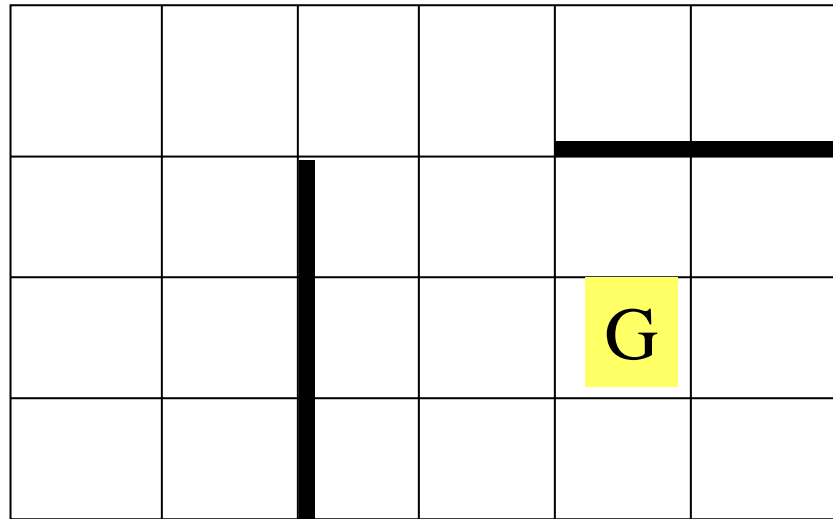
- There is never a path from  $n$  to a goal that has path cost less than  $h(n)$ .
- another way of saying this:  $h(n)$  is a **lower bound** on the cost of getting from  $n$  to the nearest goal.



# Example Admissible Heuristic Functions

**Search problem:** robot has to find a route from start location to goal location on a grid (discrete space with obstacles)

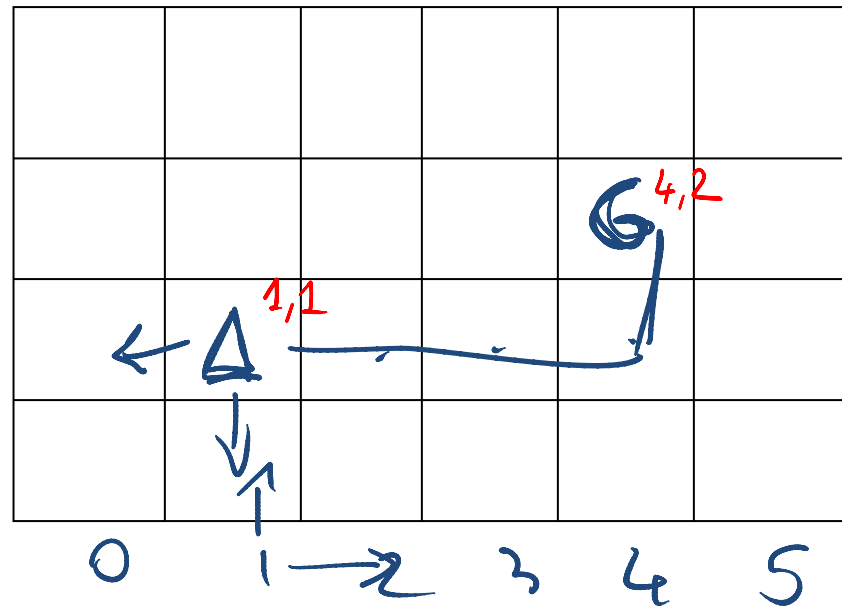
**Final cost** (quality of the solution) is the number of steps



# Example Admissible Heuristic Functions

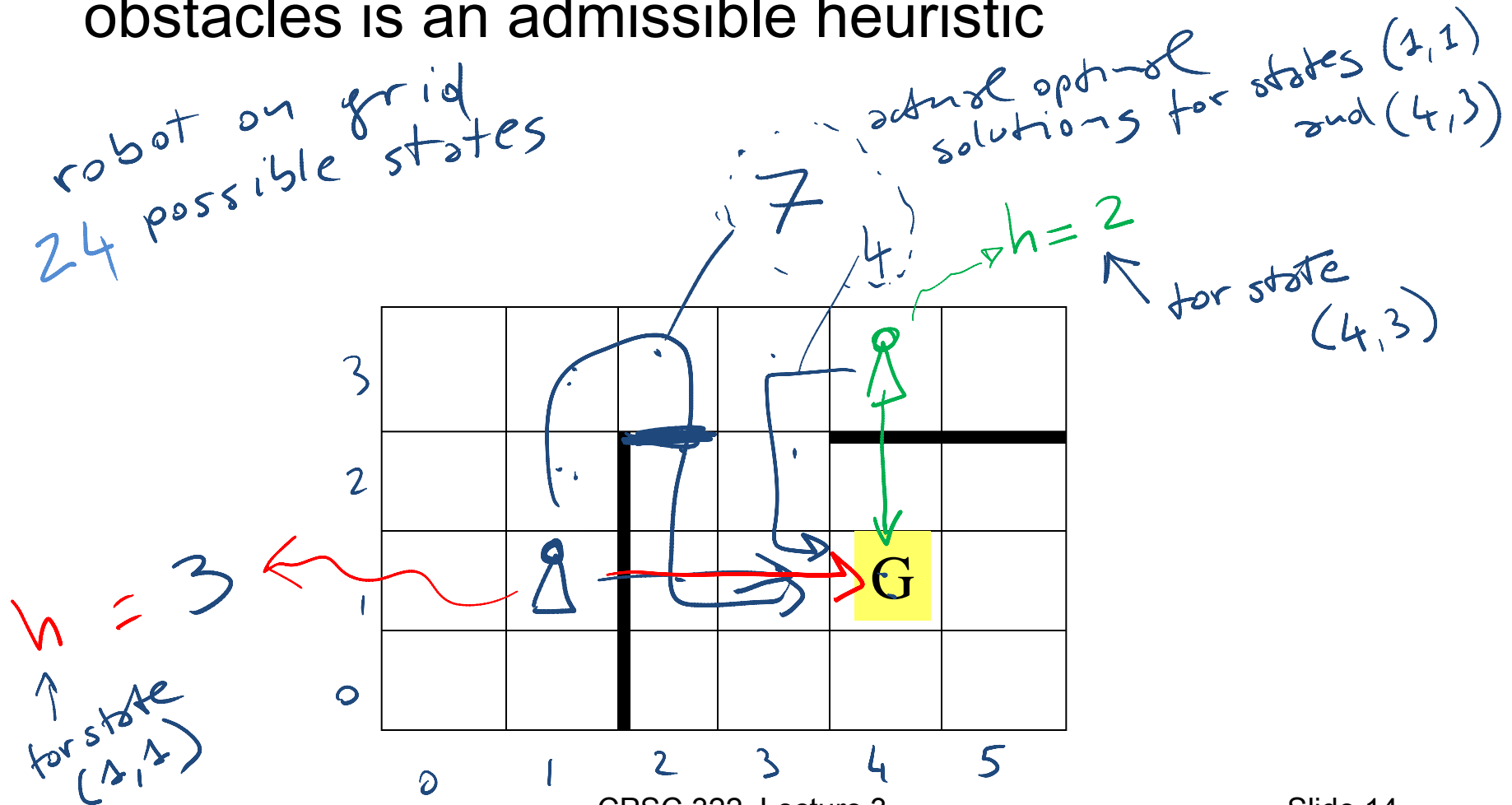
If no obstacles, cost of optimal solution is...

$$\begin{array}{l} \text{Goal state} \\ \hline x_G \quad y_G \\ \hline \text{Current state} \\ \hline x_c \quad c \\ \hline |x_G - x_c| + |y_G - y_c| \\ \hline \text{In example} \\ \hline |4 - 1| + |2 - 1| \\ \hline = 4 \\ \hline \text{Manhattan distance} \end{array}$$



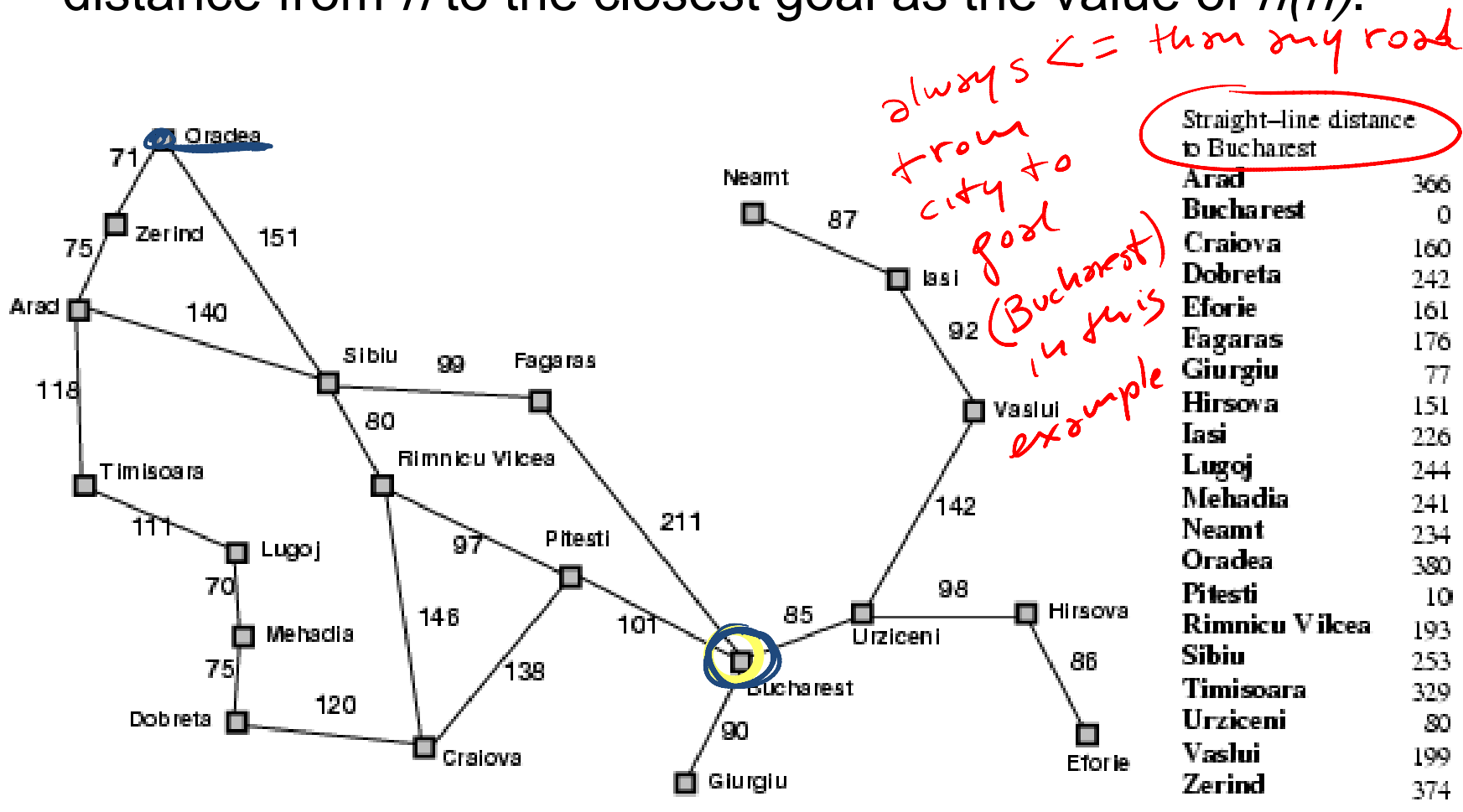
# Example Admissible Heuristic Functions

If there are obstacles, the optimal solution without obstacles is an admissible heuristic

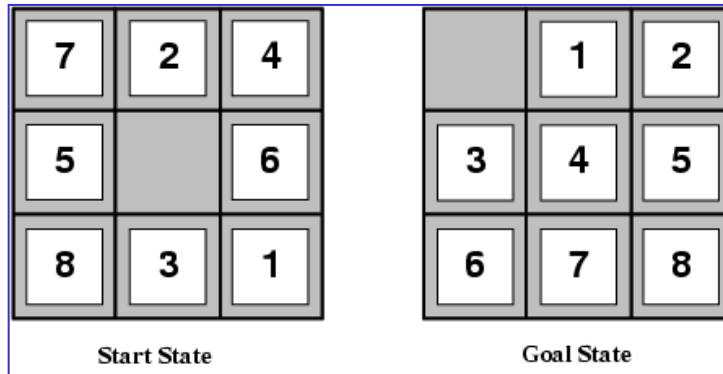


# Example Admissible Heuristic Functions

- Similarly, If the nodes are **points on a Euclidean plane** and the cost is the distance, we can use the straight-line distance from  $n$  to the closest goal as the value of  $h(n)$ .



# Heuristic Function for 8-puzzle



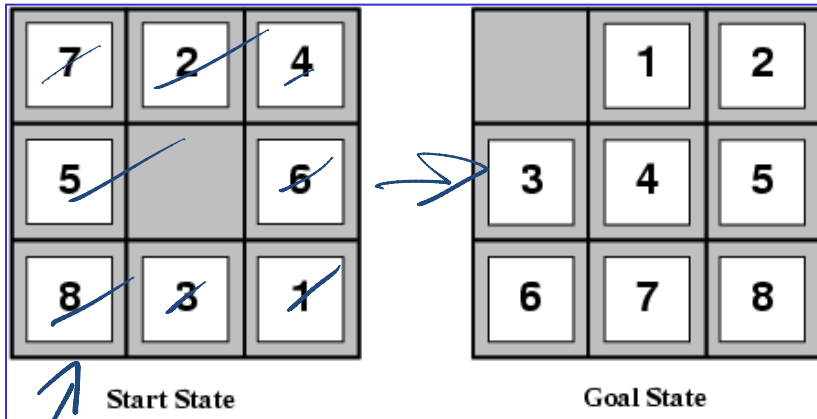
A reasonable heuristics for the 8-puzzle is?

- A. Number of misplaced tiles plus number of correctly place tiles
- B. Number of misplaced tiles
- C. Number of correctly placed tiles
- D. None of the above

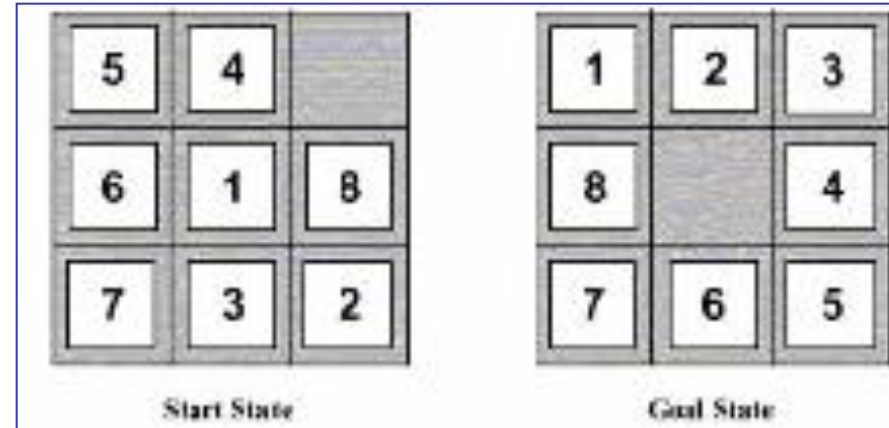


# Example Heuristic Functions <sup>(1)</sup>

- In the 8-puzzle, we can use the number of misplaced tiles



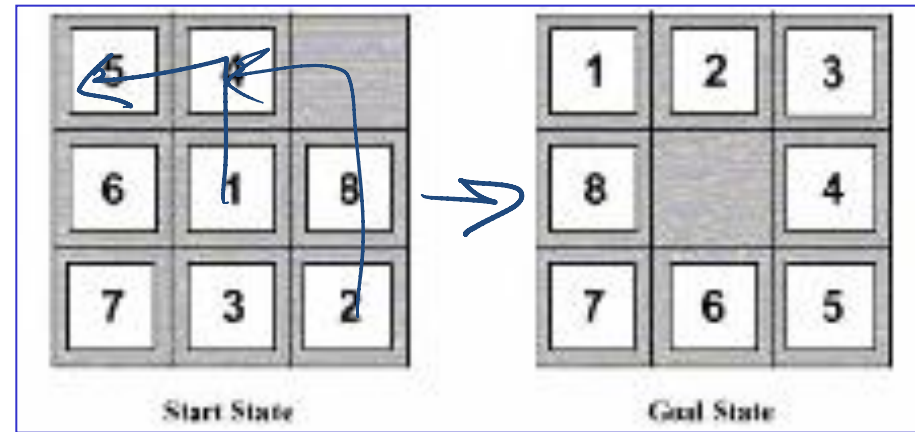
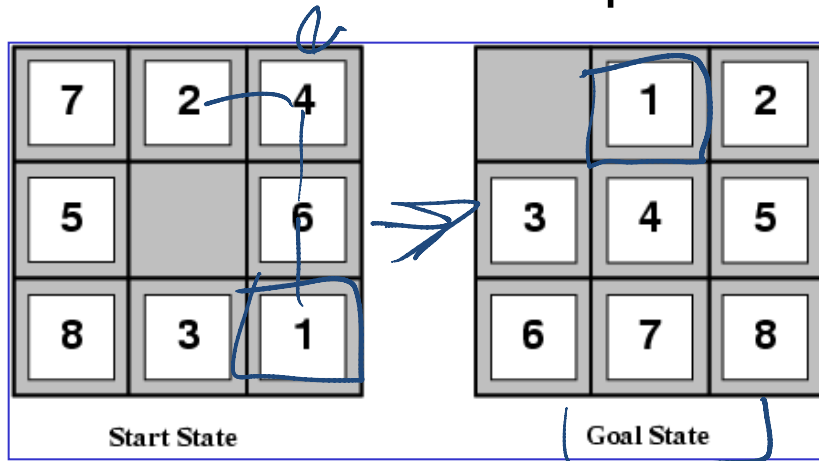
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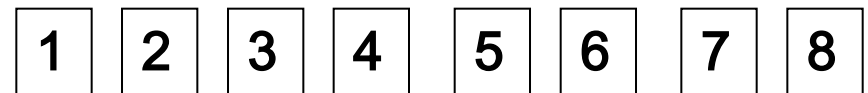
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# Example Heuristic Functions (2)

- Another one we can use the number of moves between each tile's current position and its position in the solution



tiles



$\{3, 1, 2, 2, 2, 3, 3, 2\}$       $\{2, 3, \dots, \dots\}$

$= 18$

# How to Construct a Heuristic

You identify **relaxed version of the problem**:

- where one or more constraints have been dropped
- problem with fewer restrictions on the actions ←

**Robot**: the agent **can move through walls** ←

**Driver**: the agent **can move straight** ←

**8puzzle**: (1) tiles **can move anywhere** ←

(2) tiles can move to **any adjacent square** ←

**Result**: The cost of an optimal solution to the relaxed problem is an admissible heuristic for the original problem (because it is always weakly less costly to solve a less constrained problem!)

# How to Construct a Heuristic (cont.)

You should identify constraints which, when dropped, make the problem extremely easy to solve

- this is important because heuristics are not useful if they're as hard to solve as the original problem!

This was the case in our examples

**Robot:** *allowing* the agent to move through walls. Optimal solution to this relaxed problem is Manhattan distance

**Driver:** *allowing* the agent to move straight. Optimal solution to this relaxed problem is straight-line distance

**8puzzle:** (1) tiles **can move anywhere** Optimal solution to this relaxed problem is number of misplaced tiles

(2) tiles can move to **any adjacent square**....

# Another approach to construct heuristics

Solution cost for a subproblem

1 2 3 4

Original Problem

	1	3
8	2	5
7	6	4

Current node

1	2	3
8		4
7	6	5

Goal node

*simpler!*

SubProblem

	1	3
@	2	@
@	@	4

1	2	3
@		4
@	@	@

Good

# Heuristics: Dominance

If  $h_2(n) \geq h_1(n)$  for every state  $n$  (both admissible)  
then  $h_2$  **dominates**  $h_1$

*Which one* is better for search ?



*A.  $h_1$*

*B.  $h_2$*

*C. It depends*

# Heuristics: Dominance

8puzzle: (1) tiles can move anywhere

(2) tiles can move to any adjacent square

(Original problem: tiles can move to an adjacent square if it is empty)

*Iterative deepening (not using any heuristic)*

search costs for the 8-puzzle (average number of paths expanded):

*→ depth of solution*

$d=12$  IDS = 3,644,035 paths  
 $A^*(h_1)$  = 227 paths  
 $A^*(h_2)$  = 73 paths

$d=24$  IDS = too many paths  
 $A^*(h_1)$  = 39,135 paths  
 $A^*(h_2)$  = 1,641 paths



*why*

	$h_1$	$h_2$
If tile in correct position	0	0
If tile 1 move from correct position	1	1
otherwise	1	>1

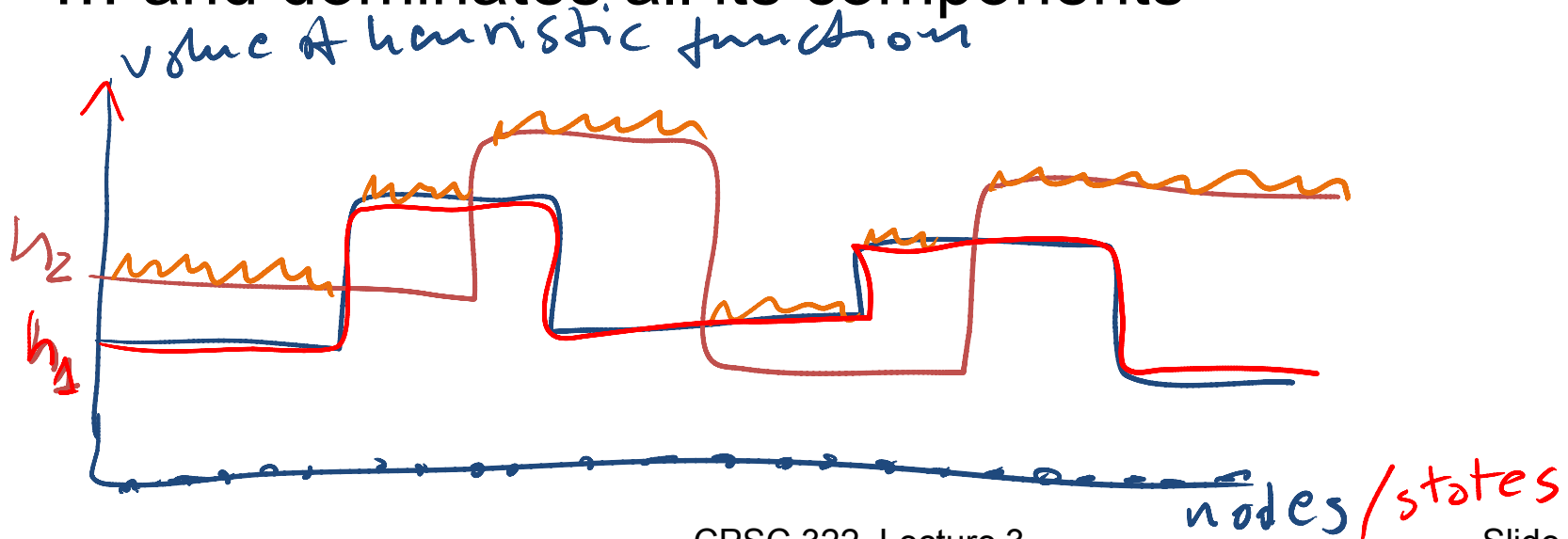
# Combining Heuristics

How to combine heuristics when there is no dominance?

If  $h_1(n)$  is admissible and  $h_2(n)$  is also admissible then

$h(n) = \max(h_1, h_2)$  is also admissible

... and dominates all its components





# Combining Heuristics: Example

In 8-puzzle, solution cost for the 1,2,3,4 subproblem is substantially more accurate than Manhattan distance in some cases

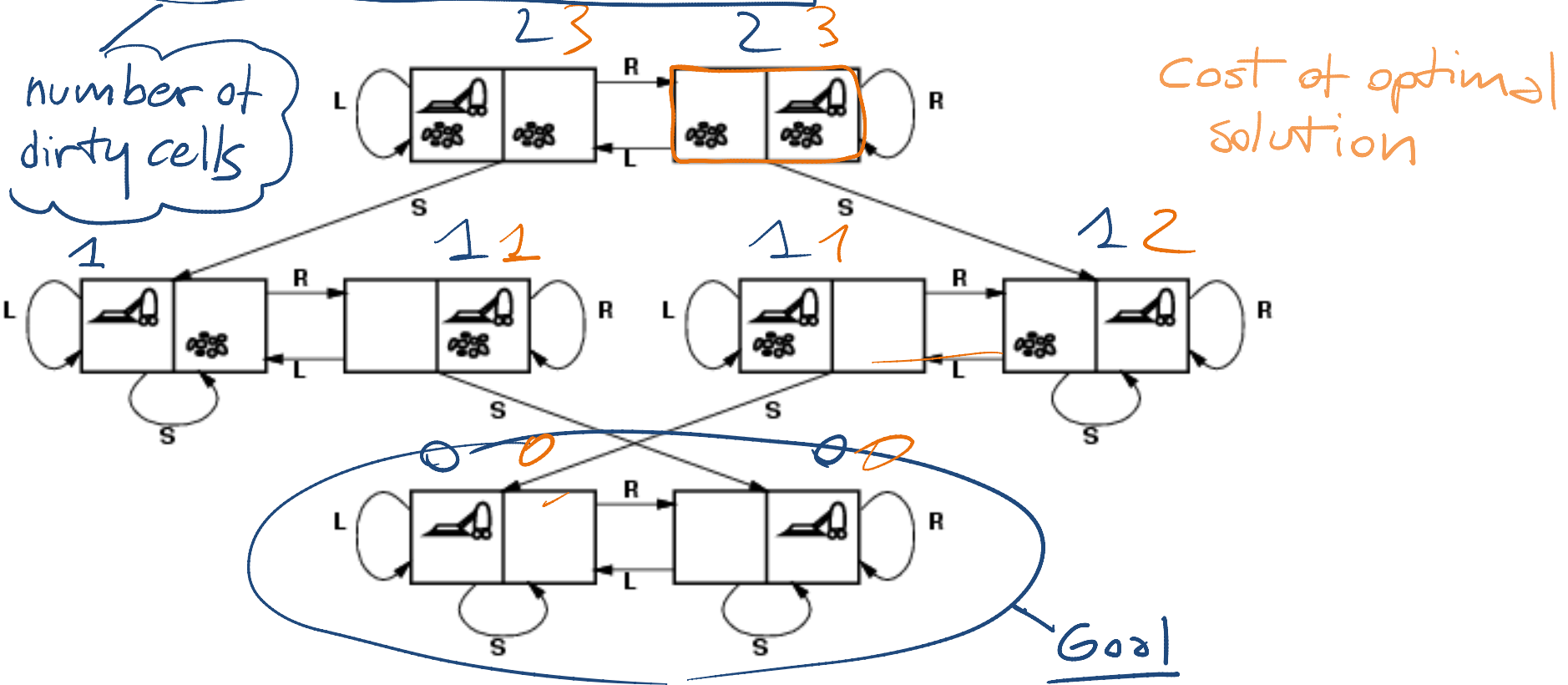
So.....

of each tile  
from its position  
in goal

sum of

max ( )  
better heuristic

# Admissible heuristic for Vacuum world?



states? Where it is dirty and robot location

actions? *Left, Right, Suck*

Possible goal test? no dirt at all locations

# Learning Goals for today's class

- Construct admissible heuristics for a given problem.
  - Verify Heuristic Dominance.
  - Combine admissible heuristics
- 
- From previous classes
- Define/read/write/trace/debug different search algorithms
- With / Without cost
  - Informed / Uninformed

# Next Class

- Best-First Search
- Combining LCFS and BFS: A\* (finish 3.6)
- A\* Optimality