

Heuristic Search: BestFS and A*

Computer Science cpsc322, Lecture 8

(Textbook Chpt 3.6)

January, 20, 20010



Course Announcements

Posted on WebCT

- Second Practice Exercise (uninformed Search)
- Assignment 1

DEPARTMENT OF COMPUTER SCIENCE



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Distinguished Lecture Series 2008 - 2009

Speaker: Michael Littman Rutgers University

Date: Thursday, January 22, 2009 Time: 3:30 - 4:50pm

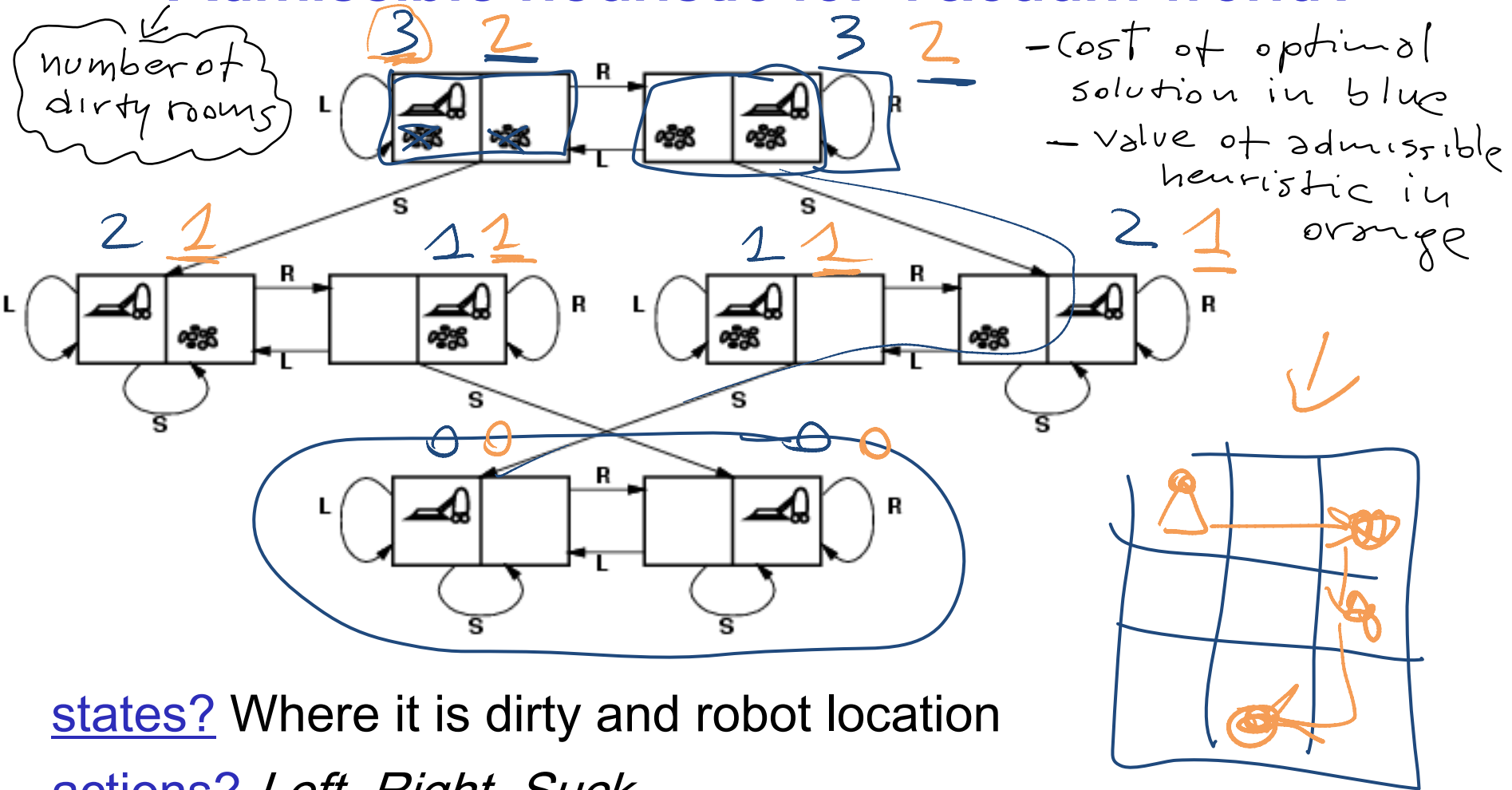
Venue: Hugh Dempster Pavilion Room 310

Title: Efficiently Learning to Behave Efficiently

# Lecture Overview

- **Recap Heuristic Function**
- Best First Search
- $A^*$

# Admissible heuristic for Vacuum world?



# Lecture Overview

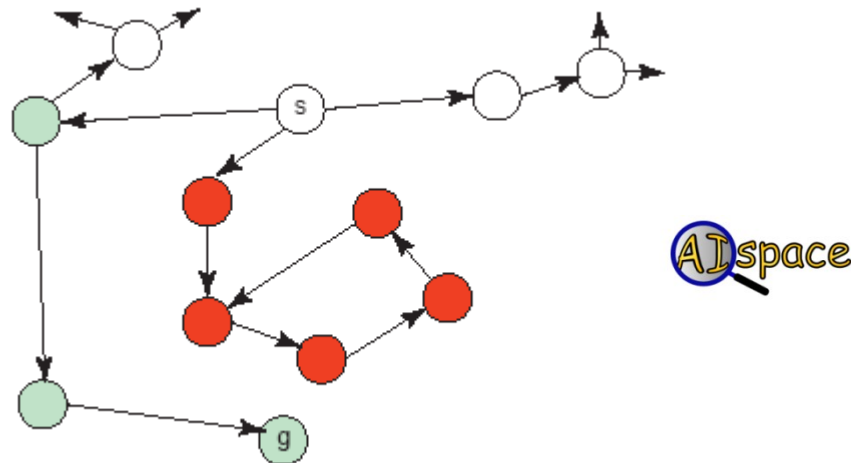
- Recap Heuristic Function
- **Best First Search**
- $A^*$




# Best-First Search

- **Idea:** select the path whose end is closest to a goal according to the heuristic function.
- Best-First search selects a path on the frontier with minimal  $h$ -value (for the end node).
- It treats the frontier as a priority queue ordered by  $h$ .  
(similar to ?) LCFS  $\rightarrow$  Cost
- This is a **greedy** approach: it always takes the path which appears locally best

# Analysis of Best-First Search

- **Complete** no: a low heuristic value can mean that a cycle gets followed forever.



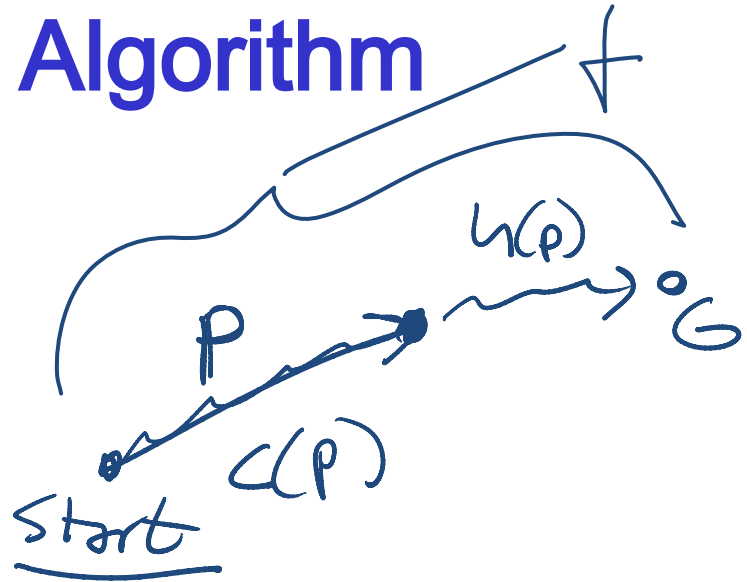
- **Optimal**: no (why not?) 
- **Time complexity** is  $O(b^m)$  
- **Space complexity** is  $O(b^m)$  

# Lecture Overview

- Recap Heuristic Function
- Best First Search
- A\* Search Strategy



# A\* Search Algorithm



- $A^*$  is a mix of:
  - lowest-cost-first and
  - best-first search
- $A^*$  treats the frontier as a priority queue ordered by  $\underline{f(p) = C(p) + h(p)}$
- It always selects the node on the frontier with the lowest estimated total distance.

# Analysis of $A^*$

Let's assume that arc costs are strictly positive.

- Time complexity is  $O(b^m)$ 
  - the heuristic could be completely uninformative and the edge costs could all be the same, meaning that  $A^*$  does the same thing as BFS
- Space complexity is  $O(b^m)$  like BFS,  $A^*$  maintains a frontier which grows with the size of the tree
- Completeness: yes.
- Optimality: yes.

# Optimality of $A^*$

If  $A^*$  returns a solution, that solution is guaranteed to be optimal, as long as

When

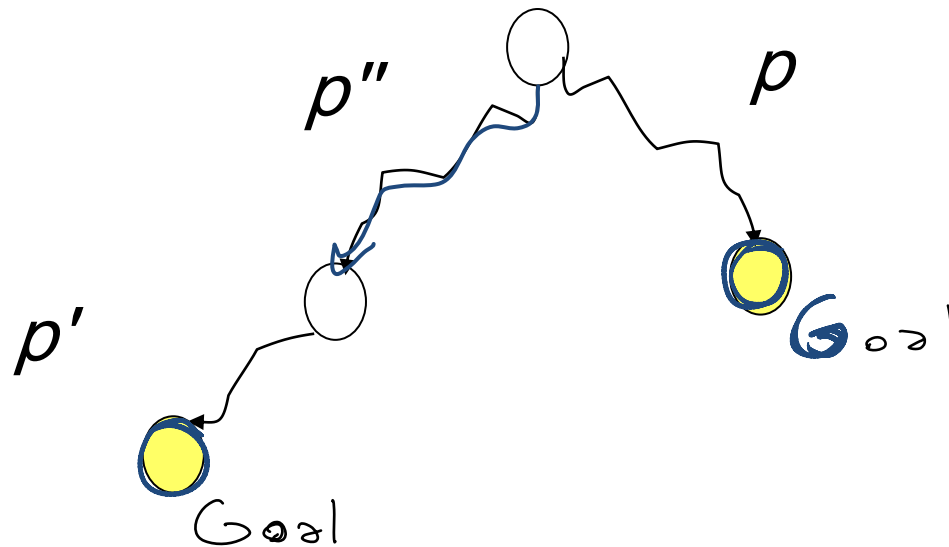
- the branching factor is finite
  - arc costs are strictly positive
  - $h(n)$  is an underestimate of the length of the shortest path from  $n$  to a goal node, and is non-negative
- admissible* ↙

## Theorem

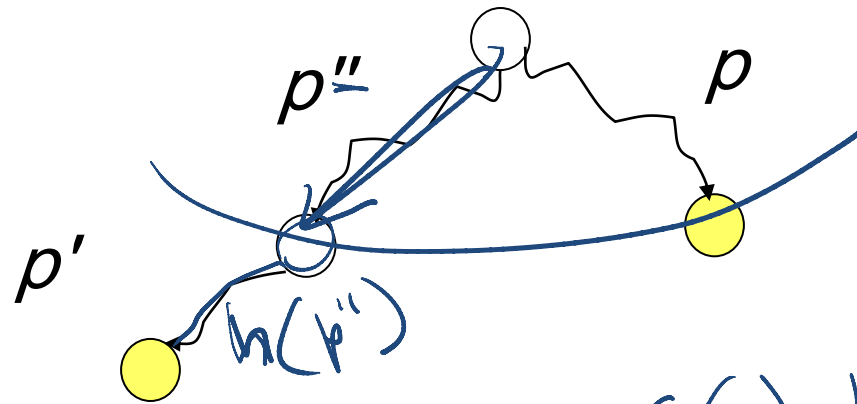
If  $A^*$  selects a path  $p$ ,  $p$  is the shortest (i.e., lowest-cost) path.

# Why is $A^*$ optimal? $f = c + h$

- Assume for contradiction that some other path  $p'$  is actually the shortest path to a goal  $\text{cost}(p') < \text{cost}(p)$
- Consider the moment when  $p$  is chosen from the frontier. Some part of path  $p'$  will also be on the frontier; let's call this partial path  $p''$ .



# Why is $A^*$ optimal? (cont')



- Because  $p$  was expanded before  $p''$ ,  $f(p) \leq f(p'')$
- Because  $p$  is a goal,  $h(p) = 0$ . Thus  $c(p) \leq c(p'') + h(p'')$
- Because  $h$  is admissible,  $cost(p'') + h(p'') \leq cost(p')$  for any path  $p'$  to a goal that extends  $p''$
- Thus  $cost(p) \leq cost(p')$  for any other path  $p'$  to a goal.

This contradicts our assumption that  $p'$  is the shortest path.  
that  $cost(p') < cost(p)$

# Optimal efficiency of $A^*$

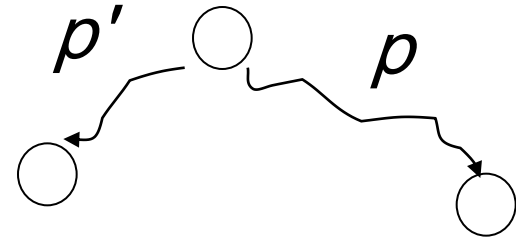
next class

- In fact, we can prove something even stronger about  $A^*$ : in a sense (given the particular heuristic that is available) **no search algorithm could do better!**
- **Optimal Efficiency:** Among all optimal algorithms that **start from the same start node** and **use the same heuristic  $h$** ,  $A^*$  **expands the minimal number of paths.**

# Why is $A^*$ optimally efficient?

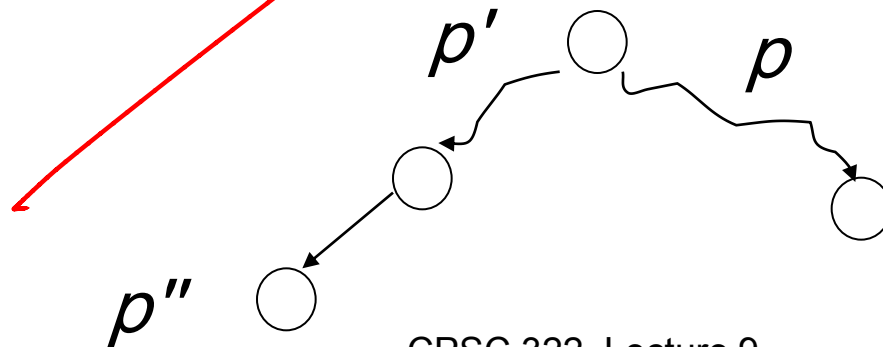
**Theorem:**  $A^*$  is optimally efficient.

- Let  $f^*$  be the cost of the shortest path to a goal.
- Consider any algorithm  $A'$ 
  - the same start node as  $A^*$ ,
  - uses the same heuristic
  - fails to expand some path  $p'$  expanded by  $A^*$ , for which  $f(p') < f^*$ .
- Assume that  $A'$  is optimal.



# Why is $A^*$ optimally efficient? (cont')

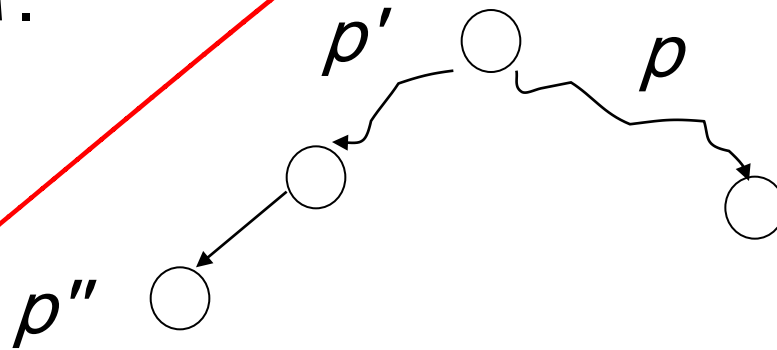
- Consider a different search problem
  - identical to the original
  - on which  $h$  returns the same estimate for each path
  - **except that**  $p'$  has a child path  $p''$  which is a goal node, and the true cost of the path to  $p''$  is  $f(p')$ .
  - that is, the edge from  $p'$  to  $p''$  has a cost of  $h(p')$ : the heuristic is exactly right about the cost of getting from  $p'$  to a goal.





# Why is $A^*$ optimally efficient? (cont')

- $A'$  would behave identically on this new problem.
  - The only difference between the new problem and the original problem is beyond path  $p'$ , which  $A'$  does not expand.
- Cost of the path to  $p''$  is lower than cost of the path found by  $A'$ .



- This violates our assumption that  $A'$  is optimal.

# Learning Goals for today's class

- Define/read/write/trace/debug different search algorithms

- With / Without cost

- Informed / Uninformed

they use  $h$   
Best First  $h$   
 $A^* \min c + h$

- Formally prove  $A^*$  optimality.

- Define optimally efficient and formally prove that  $A^*$  is optimally efficient

to be done

# Next class

Finish Search (finish Chpt 3)

- Branch-and-Bound
- A\* enhancements
- Non-heuristic Pruning
- Backward Search
- Dynamic Programming