

Search: Advanced Topics

Computer Science cpssc322, Lecture 9

(Textbook Chpt 3.6)

Sept, 24, 2010



Lecture Overview

$$f = \underbrace{c + h}$$

- **Recap A***
- Branch & Bound
- A* tricks
- Other Pruning

Branch-and-Bound Search

- What is the biggest advantage of A*?

uses heuristics

- What is the biggest problem with A*?

space

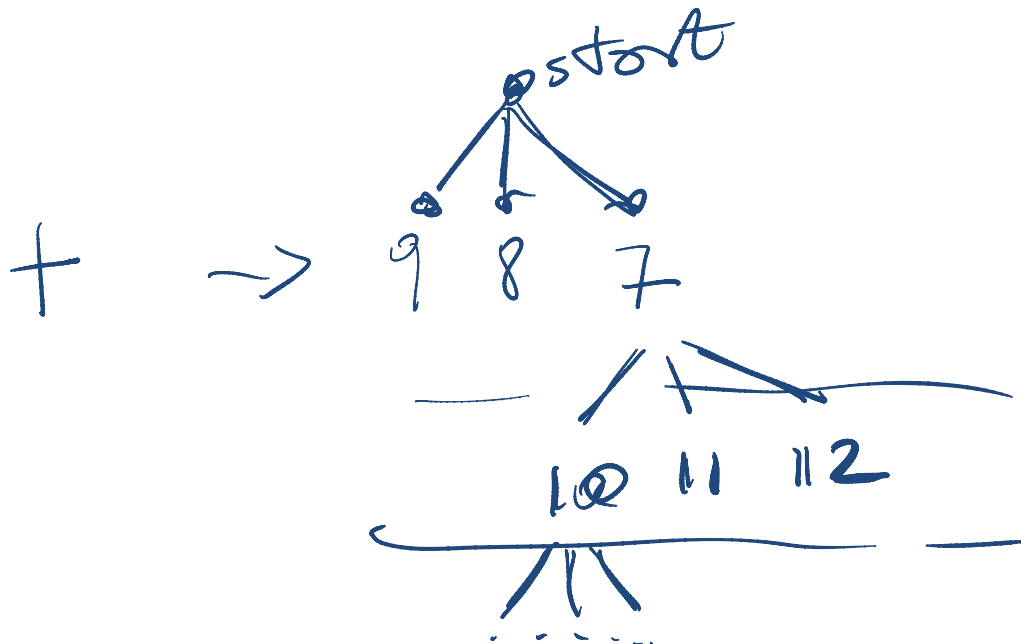
- Possible Solution:

DFS + h

Branch-and-Bound Search Algorithm

- Follow exactly the same search path as **depth-first search**
 - treat the frontier as a stack: expand the most-recently added path first
 - the order in which neighbors are expanded can be governed by some arbitrary node-ordering heuristic ←

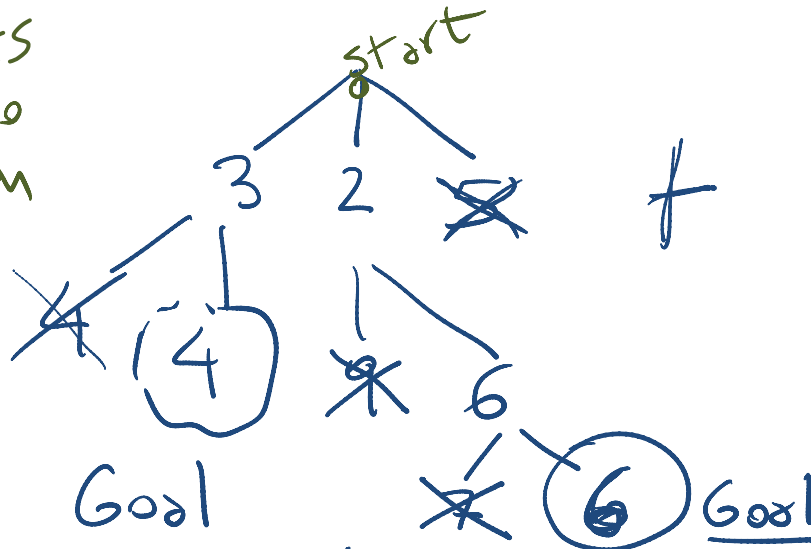
we can use
 $f = c + h$



Branch-and-Bound Search Algorithm

- Keep track of a lower bound and upper bound on solution cost at each path
 - lower bound: $LB(p) = f(p) = cost(p) + h(p)$
 - upper bound: $UB = \text{cost of the best solution found so far.}$
 - ✓ if no solution has been found yet, set the upper bound to ∞ .
- When a path p is selected for expansion:
 - if $LB(p) \geq UB$, remove p from frontier without expanding it (pruning)
 - else expand p , adding all of its neighbors to the frontier

The numbers
correspond to
f for the path
from start
to that node



$UB = \infty$
 ↑
 6
 4
 same for
 all paths
 at any
 given time

Branch-and-Bound Analysis

- Complete ?

yes

no

It depends

- Optimal ?

yes

no

It depends

- Space complexity?

$O(b^m)$

$O(m^b)$

$O(bm)$

$O(b+m)$

- Time complexity?

Branch-and-Bound Analysis

- **Completeness**: no, for the same reasons that DFS isn't complete
 - however, for many problems of interest there are no infinite paths and no cycles
 - hence, for many problems B&B is complete
- **Time complexity**: $O(b^m)$
- **Space complexity**: $O(bm)$
 - Branch & Bound has the same space complexity as DFS
 - this is a big improvement over A^*!
- **Optimality**: yes.....

Lecture Overview

- Recap A^*
- Branch & Bound
- A^* tricks
- Pruning Cycles and Repeated States

Other A^* Enhancements

The main problem with A^* is that it uses exponential space. Branch and bound was one way around this problem. Are there others?

- *Iterative Deepening A^** \swarrow *IDA **
.....
- Memory-bounded A^*

(Heuristic) Iterative Deepening – IDA*

B & B can still get stuck in infinite (extremely long) paths

- Search depth-first, but to a fixed depth / *bound*
 - if you don't find a solution, increase the depth tolerance and try again
 - depth is measured in..... *f*
start node $f(\text{start}) = h(\text{start})$
- Then update with the *lowest f* that passed the previous bound

Analysis of Iterative Deepening A* (IDA*)

- Complete and optimal:

yes

no

It depends

- Space complexity:

$O(b^m)$

$O(m^b)$

$O(bm)$

$O(b+m)$

- Time complexity:

$O(b^m)$

$O(m^b)$

$O(bm)$

$O(b+m)$

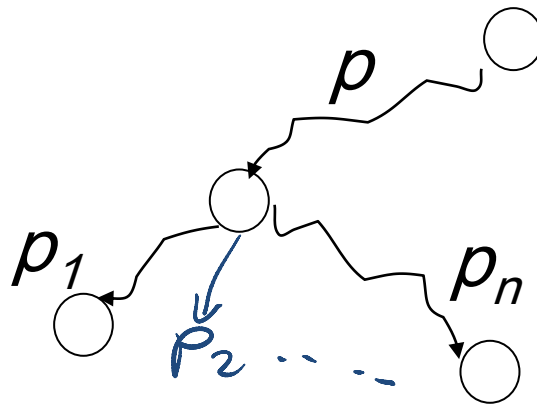
(Heuristic) Iterative Deepening – IDA*

- Counter-intuitively, the asymptotic complexity is not changed, even though we visit paths multiple times (go back to slides on uninformed ID)

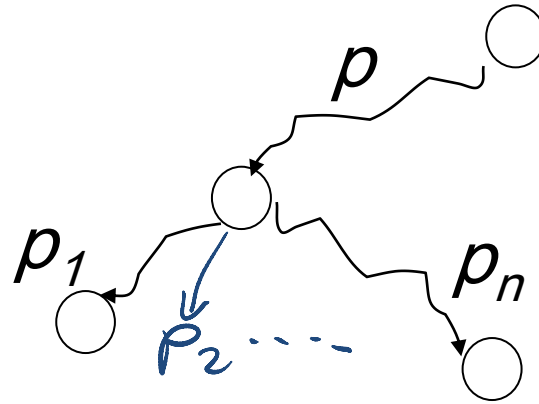
$$\left(\frac{b}{b-1}\right)^2$$

Memory-bounded A^*

- Iterative deepening A^* and B & B use a tiny amount of memory
- what if we've got more memory to use?
- keep as much of the fringe in memory as we can
- if we have to delete something:
 - delete the worst paths (with *highest* *f*)
 - ``back them up" to a common ancestor



MBA*: Compute New $h(p)$

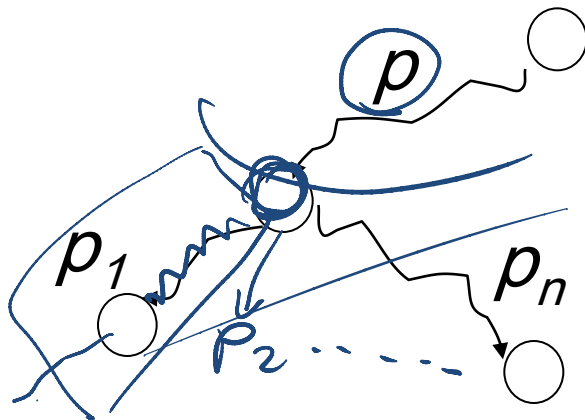


$$\text{New } h(p) = \min \left(\max_i [(\text{cost}(p_i) - \text{cost}(p)) + h(p_i)], \text{Old } h(p) \right)$$

$$\text{New } h(p) = \max \left(\min_i [(\text{cost}(p_i) - \text{cost}(p)) + h(p_i)], \text{Old } h(p) \right)$$

Memory-bounded A^*

- Iterative deepening A^* and B & B use a tiny amount of memory
- what if we've got more memory to use?
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- if we have to delete something:
 - delete the worst paths (with *highest f*)
 - "back them up" to a common ancestor



$\max(\min$

$$h(p) = \left[\text{cost}(p_i) - \text{cost}(p) \right] + h(p_i)$$

original $h(p)$

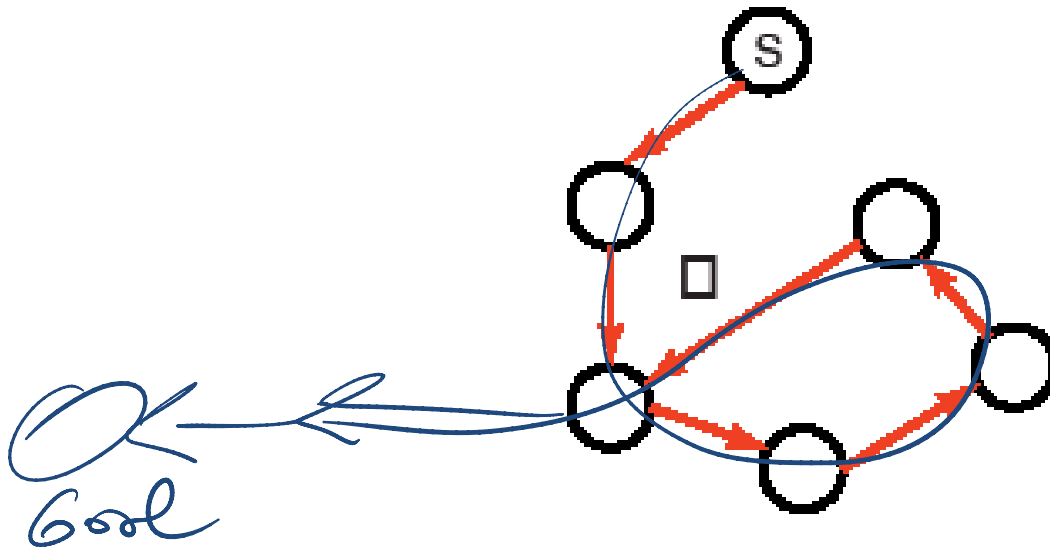
min max

max min

Lecture Overview

- Recap A^*
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Cycle Checking



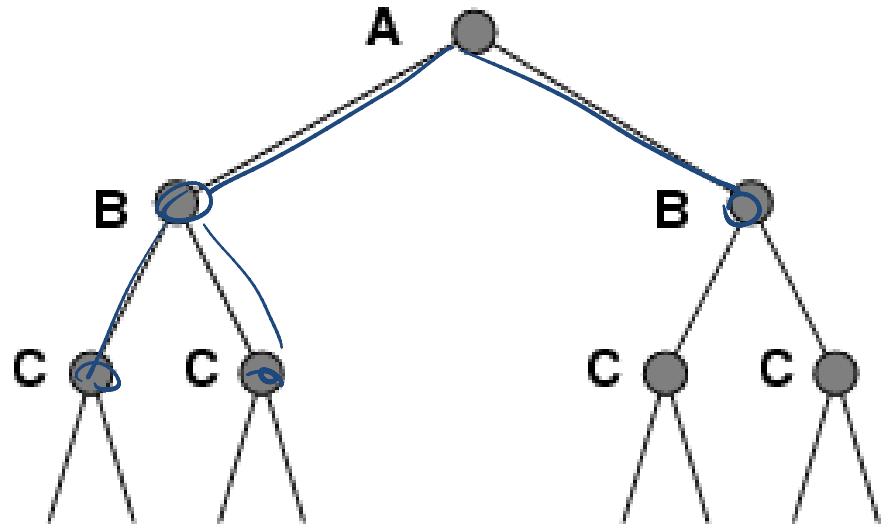
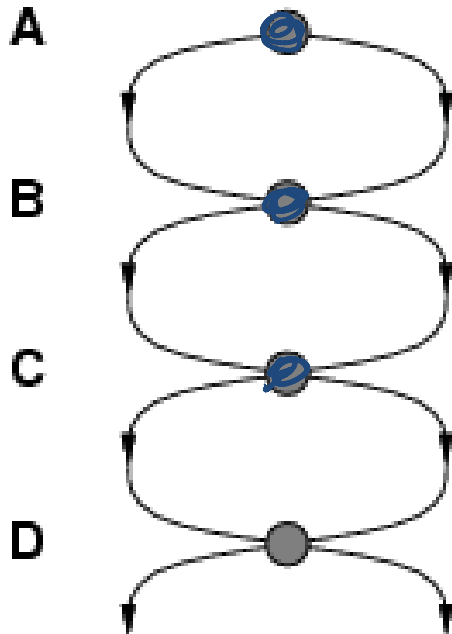
You can prune a path that ends in a node already on the path. This pruning cannot remove an optimal solution.

- The time is *linear* in path length.

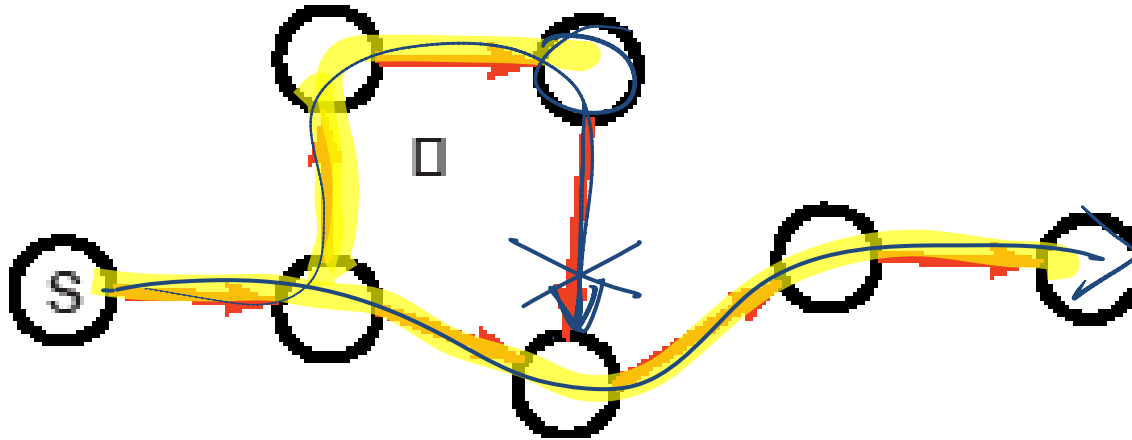


Repeated States / Multiple Paths

Failure to detect repeated states can turn a linear problem into an exponential one!



Multiple-Path Pruning

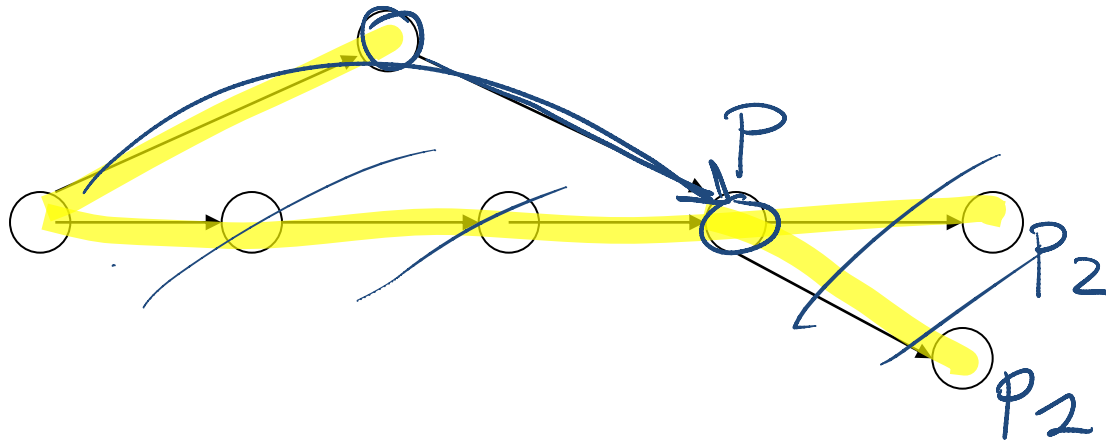


- You can prune a path to node n that you have already found a path to
- (if the new path is longer – more costly).

Multiple-Path Pruning & Optimal Solutions

Problem: what if a subsequent path to n is shorter than the first path to n ?

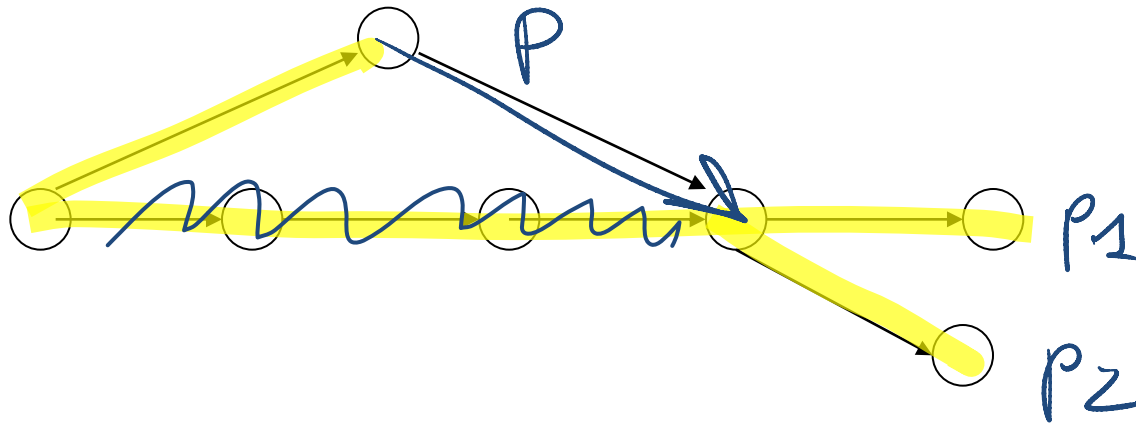
- You can remove all paths from the frontier that use the longer path. (as these can't be optimal)



Multiple-Path Pruning & Optimal Solutions

Problem: what if a subsequent path to n is shorter than the first path to n ?

- You can change the initial segment of the paths on the frontier to use the shorter path.



Learning Goals for today's class

- Define/read/write/trace/debug different search algorithms
 - With / Without cost
 - Informed / Uninformed
- Pruning cycles and Repeated States

Next class

- Dynamic Programming
- Recap Search
- Start Constraint Satisfaction Problems (CSP)
- Chp 4.

- Start working on assignment-1 !