

CS 5100

Search Algorithms

Assessing an algorithm

- Completeness
- Optimality
- time complexity
- Space complexity

Breadth First Search (FIFO)

- yes complete
- not optimal
- $O(b^d)$ time
- $O(b^d)$ space

b : # nodes branched out from one node
 d : depth of the shallowest goal node

Depth First Search (LIFO)

- not complete (infinite # of states)
- not optimal
- $O(b^d)$ time
- we could write DFS in a way to get $O(m)$

m : max depth of the tree

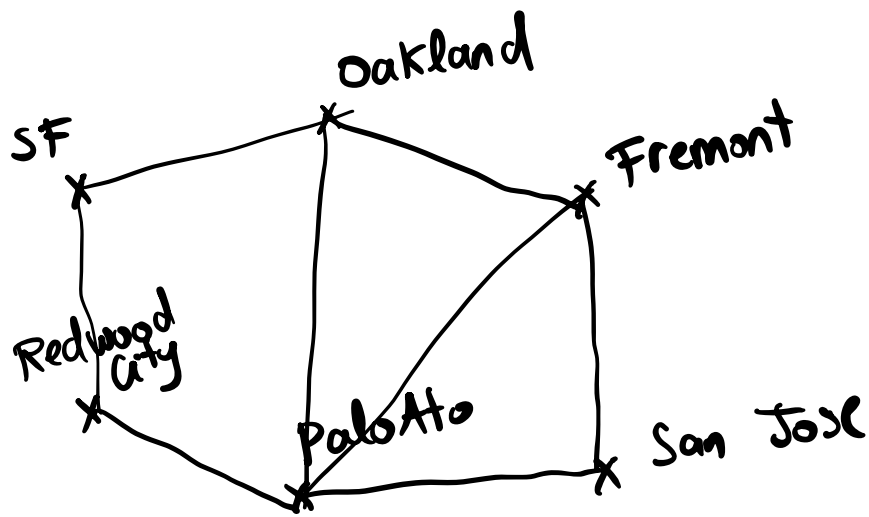
we could limit the depth of the tree we reach. What depth limit l to choose?

If we choose $l < d \Rightarrow$ not complete

If we choose $l \geq d \Rightarrow$ complete

Depth limited search

resolve completeness but not optimality



Knowledge of
the problem
can help

6 cities
 $l = 5$

(could also choose
a smaller value
 $l = 3$ for instance)
the smallest value we
can choose is known as
the diameter

What if I don't have any useful information to help me
select l ? Iterative deepening search

We can try different values of l , we iterate from
0, then 1 then 2 until a goal node is found

$l = d$ (depth of the
shallowest
goal node)

How much time complexity will it cost me?

$l=0$

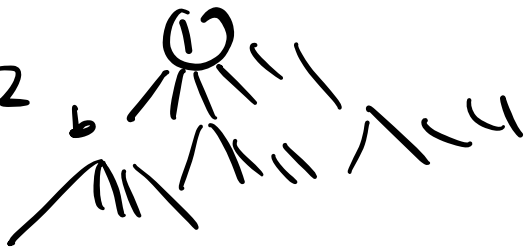
$l=1$



1 node

$1+b$ nodes

$l=2$



$1+b+b^2$ nodes

$1+b+b^2+b^3$ nodes

$l=3$

\vdots

$l=d$

$1+b+b^2+b^3+\dots+b^d$

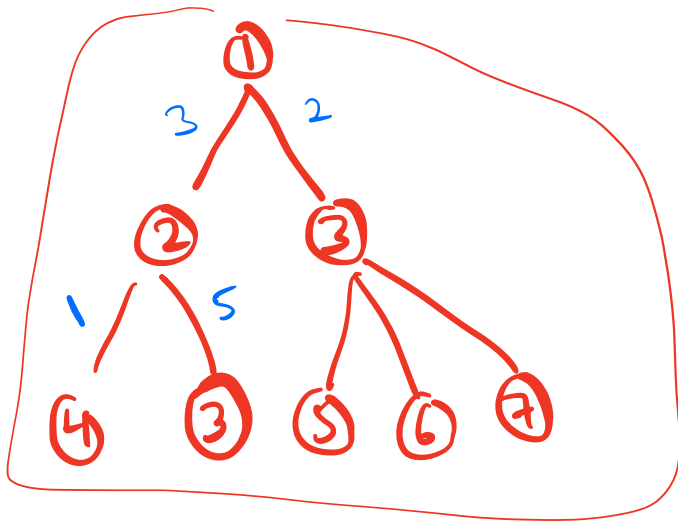
$$\begin{aligned} &1 \\ + &1+b \\ + &1+b+b^2 \\ + &1+b+b^2+b^3 \\ + &\vdots \\ + &1+b+b^2+b^3+\dots+b^d \end{aligned}$$

$$\begin{aligned} &= 1(d+1) + bd \\ &+ b^2(d-1) \\ &+ \dots + b^{d-1} \cdot 2 \\ &+ b^d \cdot 1 \end{aligned}$$

$O(b^d)$

it will cost a bit more than the regular DFS but it still has the same performance in terms of big O notation (asymptotically same as DFS)

Yay for completeness



Goal node is 3

$l=2$

we still have the issue of optimality

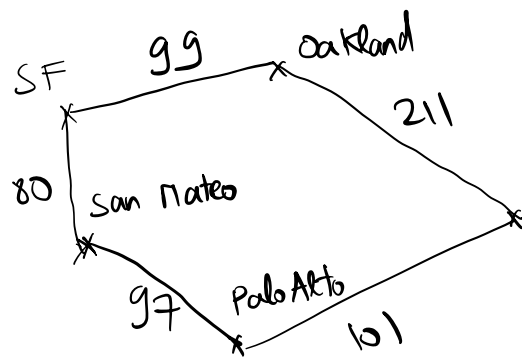
Let's explore a search algorithm that can find the optimal solution

Uniform Cost Search

the node chosen to be expanded will have the lowest path cost

→ ~~*~~ goal test is applied when it is selected for expansion (rather than when it is generated)

* we'll add another test in case a better path is found to a node that is currently in frontier



function UNIFORM-COST-SEARCH(*problem*) **returns** a solution, or failure

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→ node ← a node with STATE = problem.INITIAL-STATE, PATH-COST = 0
→ frontier ← a priority queue ordered by PATH-COST, with node as the only element
→ explored ← an empty set
loop do
  → if EMPTY?(frontier) then return failure
  → node ← POP(frontier) /* chooses the lowest-cost node in frontier */
  → if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
  → add node.STATE to explored
  → for each action in problem.ACTIONS(node.STATE) do
    → child ← CHILD-NODE(problem, node, action)
    → if child.STATE is not in explored or frontier then
      frontier ← INSERT(child, frontier)
    else if child.STATE is in frontier with higher PATH-COST then
      replace that frontier node with child
  
```

Goal SF to San Jose

node = SF

frontier = { SF }

explored = { }

loop

node popped = SF (not goal)

explored = { SF }

loop actions

child = San Mateo⁸⁰
frontier = { San Mateo }

child = Oakland

frontier = { San Mateo⁸⁰, Oakland⁹⁹ }

node popped = San Mateo (not goal)

explored = { SF, San Mateo }

loop actions

child = Palo Alto⁹⁹
frontier = { Oakland, Palo Alto⁸⁰⁺⁹⁹⁼¹⁷⁷ }

node popped = Oakland (not goal)

explored = { SF, San Mateo, Oakland }

loop actions

child = San Jose¹⁷⁷
frontier = { Palo Alto, San Jose⁹⁹⁺¹⁷⁷⁼²⁷⁶ }

node popped = Palo Alto (not goal)
explored = {SF, San Mateo, Oakland, Palo Alto}

loop actions

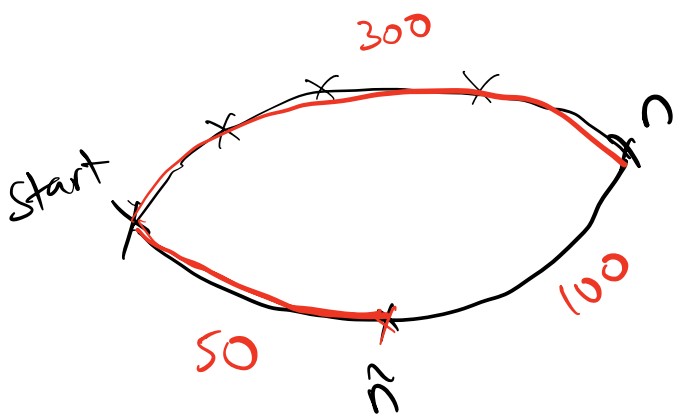
child = San Jose (already in frontier but with a higher path cost \Rightarrow will be replaced)
Frontier = {San Jose} $177 + 101 = 278$

node popped = San Jose (Yes, it is the goal node)

return San Jose with path cost of 278

Why is it the optimal path?

* Whenever UCS selects some node n for expansion, the optimal path to that node has been found



Assume path in red was not the optimal one

there should have been another node \hat{n} on the optimal path from start to n with \hat{n} having lower cost than n (requiring it to get selected first)

* Assume that the step costs are non-negative (paths never get less costly as nodes are added)

\Rightarrow UCS expands nodes in order of their optimal path cost.

* Cost to the optimal solution is C^*

* You had edges with costs ϵ (a very small positive quantity)

$$O(b^{\lceil \frac{C^*}{\epsilon} \rceil})$$

could become very costly especially if

C^* is really large and ϵ is really small

We will try to speed up the process by using some approximations / heuristics in the algorithm.

\Rightarrow informed search strategies