

TERMINOLOGY:

- A **node** in a graph represents state, parent, action, and path-cost.
- **Expanding** a node involved **generating** children & a goal test.
- The **fringe/frontier** is the set of reachable nodes yet to be expanded.
- We define a **strategy** as the criteria for choosing the next node to expand.

To decide on a search strategy, we need some criteria to evaluate them.
We have four major criteria:

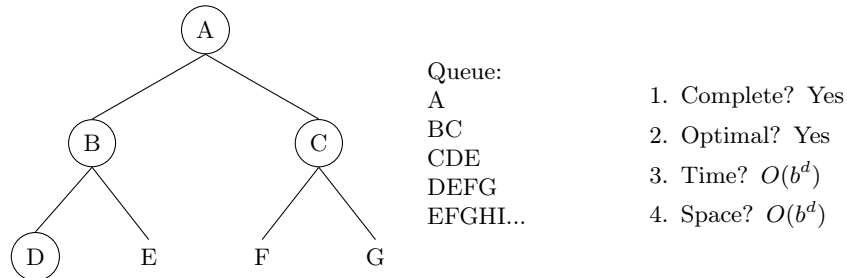
1. **completeness**: does it find a solution (if one exists)?
2. **optimality**: does it always find the best solution?
3. **time complexity**: number of worst case nodes expanded
4. **space complexity**: maximum number of nodes stored in memory

What do we use to measure these?

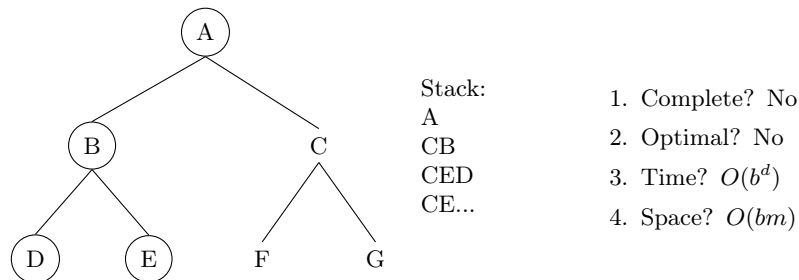
- b – max branching factor
- d – depth of least cost solution
- m – max depth of trees

Let's consider our two basic search algorithms:

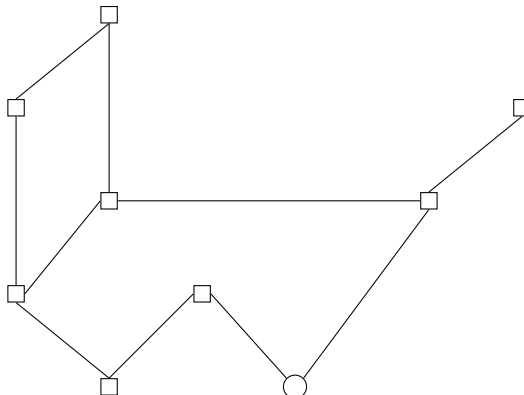
Breadth-First Search



Depth-First Search



We can't seem to improve the time complexity, but can we improve by blending BFS & DFS?
Consider a navigation problem:



We can see there are only 9 locations, so we can limit our path length!
We can use this to define a new algorithm:

Limited-Depth Search

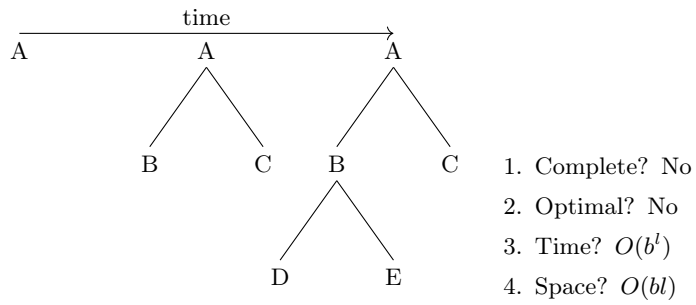
w/ length cutoff L , using priority queue (early = high priority)

1. complete? No
2. optimal? No
3. time? $O(b^L)$
4. space? $O(bL)$

We can extend this iteratively, increasing the length each round, for:

Iterative Deepening

where L denotes the iterative length.



But shouldn't we expect the time complexity to increase from limited-depth search?

$$t = b^0 \left(\frac{b}{b-1} \right) + b^1 \left(\frac{b}{b-1} \right) + \dots + b^d \left(\frac{b}{b-1} \right)$$

$$\begin{aligned} \text{Constant} &= (b = 2) \rightarrow 2 \\ &= (b = 3) \rightarrow 1.5 \\ &= (b = 10) \rightarrow 1.1 \end{aligned}$$

Therefore the constant approaches 1!