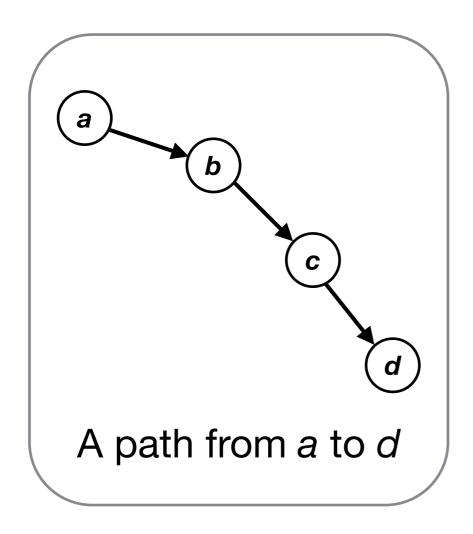
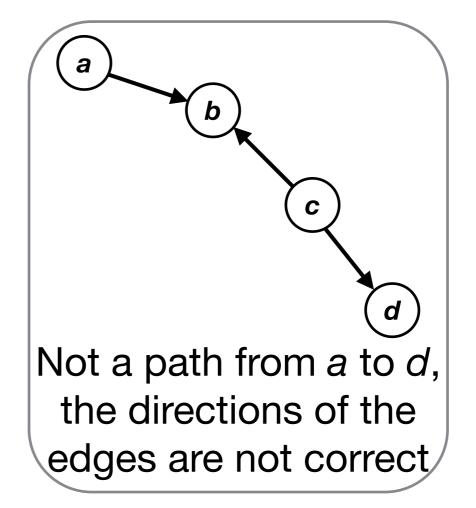
COMP3251 Lecture 8: Breadth-First Search (Chapter 4.1 and 4.2)

Some Definitions

A path is a sequence of edges in which the end vertex of an edge equals the start vertex of the following edge.

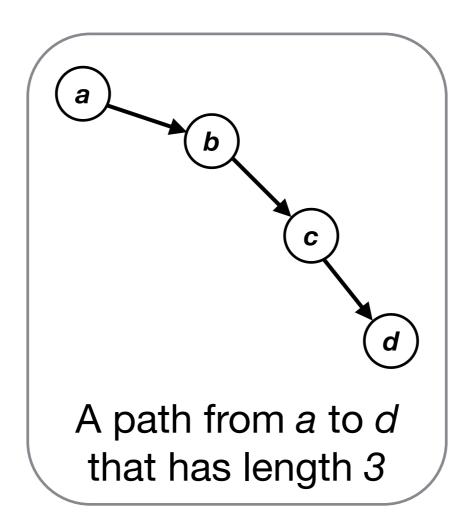




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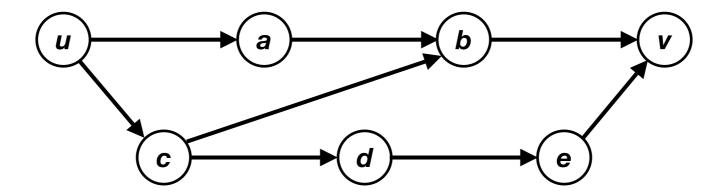


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The distance d(u, v) from u to v is the length of the shortest path (the one with the smallest length) from u to v.



There are three paths from u to v, and the shortest ones have length three. Thus, d(u,v) = 3.

Single-Source Shortest Paths Problem

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Assumption (for simplicity): All vertices are reachable from s. **Notations:**

- For any vertex v, let dist(v) = d(s, v).
- For any $k \ge 0$, let L_k be the set of vertices v with dist(v) = k.
- Let $adj(L_k)$ be the set of vertices that are adjacent to some vertices in L_k , i.e.,

$$adj(L_k) = \{ u : (v, u) \in E \text{ for some } v \in L_k \}$$

• In essence, our problem is to find L_k for all $k \ge 0$.

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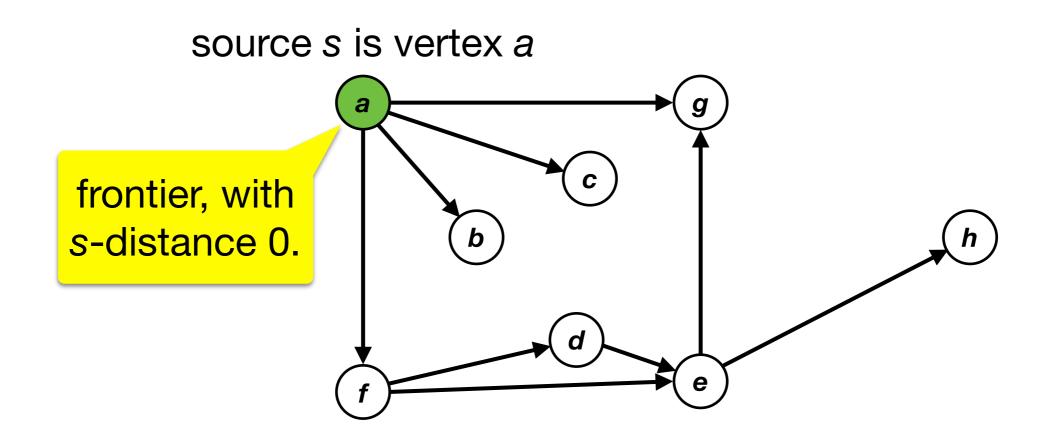
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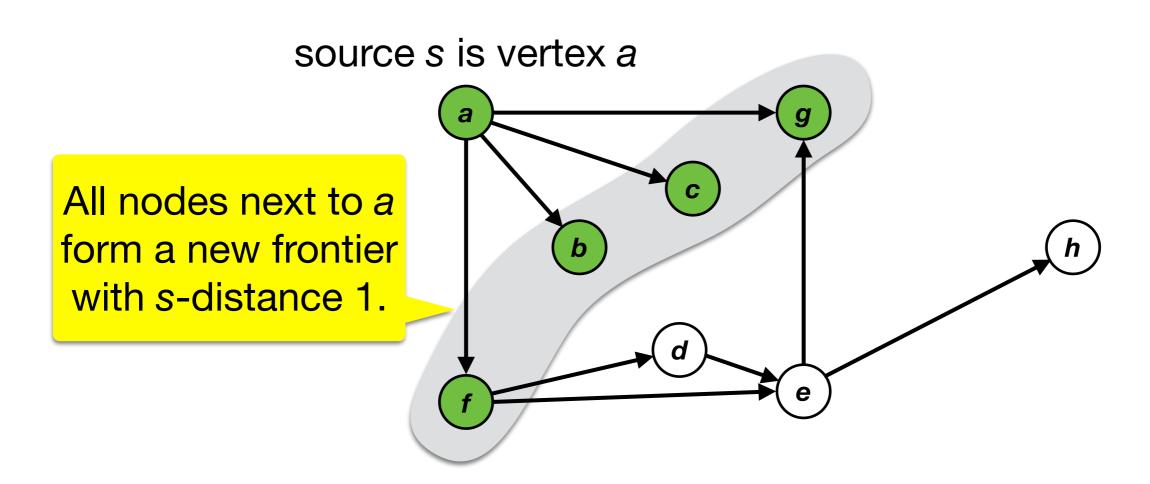
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- **Observation:** Suppose we have determined $L_0, L_1, ..., L_k$, but not $L_{k+1}, L_{k+2}, ...$ Then, for any v in $adj(L_k)$,
 - if visited(v) = true, then v is in L_i for some $i \le k$;
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 - if visited(v) = true, then v is in L_i for some $i \le k$;
 - if *visited*(v) = *false*, then v is in L_{k+1} .
- Hence, given L_0 , we can find L_1 by picking vertices v in $adj(L_0)$ with visited(v) = false; and then similarly find L_2 , L_3 ...

Breadth-First Search implements the idea directly. Breadth-first means to expand the frontier between discovered and undiscovered vertices uniformly across the breadth of the frontier (just like water-front), i.e., the algorithm discovers all vertices in L_k before discovering any vertices L_{k+1} .



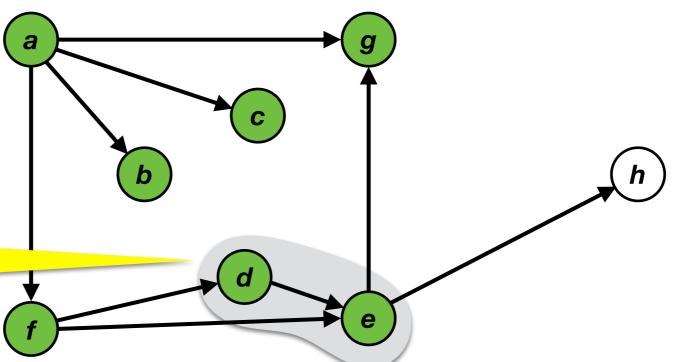
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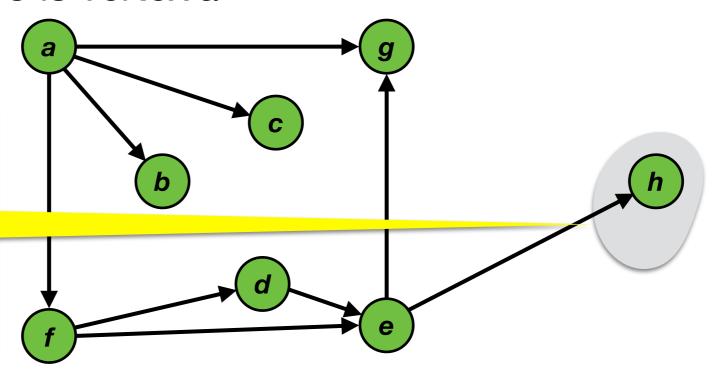
All undiscovered vertices next to a s-distance 1 vertex form a new frontier with s-distance 2.



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All undiscovered vertices next to a s-distance 2 vertex form a new frontier with s-distance 3.



```
BFS(s):
  Set Discovered[s] = true and Discovered[v] = false for all other v
  Initialize L[0] to consist of the single element s
  Set the layer counter i = 0
  Set the current BFS tree T = \emptyset
  While L[i] is not empty
    Initialize an empty list L[i+1]
    For each node u \in L[i]
      Consider each edge (u, v) incident to u
      If Discovered[v] = false then
        Set Discovered[v] = true
        Add edge (u, v) to the tree T
        Add v to the list L[i+1]
      Endif
    Endfor
    Increment the layer counter i by one
  Endwhile
```

Time Complexity of BFS

- 1) Every vertex will be put in some L[i] exactly once and be checked once. This takes O(|V|) step.
- 2) When we explore a vertex, we explore all its adjacency edges once. This takes O(|E|) steps.

In sum, the time complexity of BFS is O(|V| + |E|).

Implementation in the Textbook

```
1) initialize dist(s) = 0 and dist(u) = \infty for all other u \in V.
2) initialize queue Q = [s] (a queue containing just s).
3) while Q is not empty:
4)
     u = \mathbf{eject}(Q).
5)
     for all edges (u, v) \in E:
6)
         if dist(v) = \infty:
            inject(Q, v).
7)
8)
            dist(v) = dist(u) + 1.
```

Note: The two implementations are essentially the same.

Exercise: Give an implementation of DFS similar to the above, using a stack instead of a queue.

Retrieving the Shortest Path

In our discussion, we only focused on how to determine dist(u).

Can we also retrieve the shortest path?

- This is easy!
 - For each vertex v, let the algorithm remember *prev[v]*, the vertex immediately precedes v in shortest path.
 - To do that, each time that the algorithm discovers a new vertex *v* through an edge (*u*, *v*), let *prev*[*v*] = *u*.

DFS vs. BFS

Why two different search algorithms?

	DFS	BFS
Detecting cycles		X 1
Topological ordering		X 2
Finding CCs	✓	✓
Finding SCCs	✓	X 3
Shortest path problem	X 4	✓

- 1. When BFS encounters a non-tree edge, it is not easy to check whether it is a back edge.
- 2. The "post-ordering" numbers in BFS are not meaningful.
- 3. Same as above.
- 4. DFS focuses on going deep instead of using the shortest path.