

1 Strongly Connected Components

The issue with our algorithm is that we recompute the postorder for every SCC we need to find. However, we don't need to do this; rather, after removing some strongly connected components to get G' , the largest postorder number of vertices in G' is still in a sink component of G' .

1.1 Better Algorithm

```
SCCs(G)
  Run DFS( $G^R$ ), record postorders
  Mark all vertices as unvisited
  For v in V in reverse postorder
    If v not in a component yet
      explore(v) on G-components found,
      marking new component
```

So, really, this is just 2 DFSs, so the runtime is $O(|V| + |E|)$.

2 Paths in Graphs

DFS and `explore` allow us to determine *if* it is possible to get from one vertex to another, and using the DFS tree, you can also find a path. However, this is often not an efficient path.

2.1 Goal

Given a graph G with two vertices s and t in the same connected component, find the *best* path from s to t . What do we mean by the best?

- Least expensive.
- Best scenery.
- Shortest.

For now, we want the fewest edges.

2.2 Observation

If there is a path from s to v with length at most d , then there is some w adjacent to v with a length at most $\leq (d - 1)$ for a path from s to w .

2.3 Algorithm Idea

For each d , create a list of all vertices at distance d from s .

- For $d = 0$, this is just $\{s\}$.
- For larger d , we want all new vertices adjacent to vertices at distance $d - 1$.

```
1 ShortestPaths(G, s)
2   Initialize Array A
3   A[0] = {s}
4   dist(s) = 0
5   For d = 0 to n
6     For u in A[d]
7       For (u, v) in E
8         if dist(v) undefined
```

```

9           dist(v) = d + 1
10          add v to A[d + 1]

```

How can we improve this?

- What if `dist(v)` undefined at end? We can set the distances of all vertices to undefined.
- The algorithm goes through `A[0]`, `A[1]`, in order. We can just use a queue.

```

1  ShortestPaths(G, s)
2      Initialize Queue Q
+   Q.enqueue(s)
4   dist(s) = 0
+   While Q not empty
+   u = front(Q)
7       For (u, v) in E
8           if dist(v) = infinity
9               dist(v) = dist(u) + 1
10              Q.enqueue(v)

```

- What if we want to keep track of the paths?

```

1  ShortestPaths(G, s)
2      Initialize Queue Q
3      Q.enqueue(s)
4      dist(s) = 0
5      While Q not empty
6      u = front(Q)
7          For (u, v) in E
8              if dist(v) = infinity
9                  dist(v) = dist(u) + 1
10                 Q.enqueue(v)
+                 v.prev = u

```

2.4 Breadth First Search

In our last change above, we note that we simply have BFS.

```

BFS(G, s)
  For v in V, dist(v) = infinity
  Initialize Queue Q
  Q.enqueue(s)
  dist(s) = 0
  While Q is not empty
    u = front(Q)
    For (u, v) in E
      If dist(v) = infinity
        dist(v) = dist(u) + 1
        Q.enqueue(v)
        v.prev = u

```

The total runtime is $O(|V| + |E|)$.

2.5 DFS vs. BFS

- Similarities:
 - The way both algorithms process vertices is the same (`visited` vs. `dist < infinity`).
 - For each vertex, process all unprocessed neighbors.
- Differences:
 - DFS uses a stack to store vertices waiting to be processed.
 - BFS uses a queue.
- Big Effect:
 - DFS goes depth-first: very long path. Get a very “skinny” tree.
 - BFS is breadth first: visits all side paths. Get a very shallow tree since we process all of the neighbors.

2.6 Edge Length

The number of edges in a path is not always the right measure of distance. Sometimes, taking several shorter steps is preferable to taking a few longer ones.

We can assign each edge (u, v) a non-negative length $\ell(u, v)$. The length of a path is the sum of the lengths of its edges.

2.7 Problem: Shortest Path

Coming soon!