# Graph Algorithms

## What's on today

- Adjacency matrix and list
- BFS and DFS

### Graphs

- You've see it already Trees are technically graphs
- large number of practical problems we can model as graph problems
  - network design
  - flow design
  - planning
  - scheduling
  - route finding
- You will see it again in later subjects

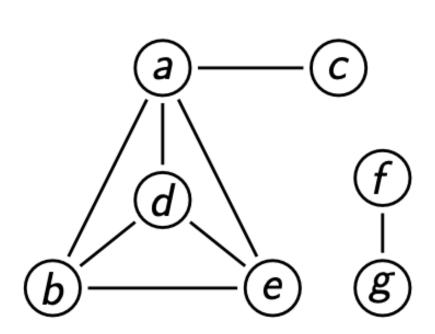
## Graph, mathematically

#### Jargon warning

- Connected vs unconnected
- Directed vs undirected
- Node/vertex, edges, degrees
- Path(length)/Simple path/Cy cle
- Cyclic/Acyclic/DAG
- Dense/Sparse

### Adjacency Matrix and List

Which one to use?



	a	b	С	d	e	f	g
а	0	1	1	1	1	0	0
b	1	0	0	1 1	1	0	0
С	1	0	0	0	0	0	0
d	1	1	0	0	1	0	0
e	1	1	0	1	0	0	0
f	0	0	0	0 0 1 0	0	0	1
g	0	0	0	0	0	1	0

The adjacency matrix for the graph.

$$\begin{array}{c|cccc}
a & \rightarrow b \rightarrow c \rightarrow d \rightarrow e \\
b & \rightarrow a \rightarrow d \rightarrow e \\
c & \rightarrow a \\
d & \rightarrow a \rightarrow b \rightarrow e \\
e & \rightarrow a \rightarrow b \rightarrow d \\
f & \rightarrow g \\
g & \rightarrow f
\end{array}$$

The adjacency list representation.

(Assuming lists are kept in sorted order.)

## Searching, on Graphs

#### Or, Graph traversal

- Exhaustive search
- Mark as we go
- Breadth-first search (BFS)
  - exploring all neighbouring nodes before we go deeper
- Depth-first search (DFS)
  - Go all the way until we hit a dead end

#### DFS

- Based on back tracking
- Stack
- Time complexity?
  - Adj list vs matrix?

This works both for directed and undirected graphs.

#### DFS

- Time complexity?
  - Using an adjacency matrix, we need to consider adj[v, w] for each w in V.
  - Here the complexity of graph traversal is  $\Theta(|V|^2)$ .
  - Using adjacency lists, for each v, we traverse the list adj[v].
  - In this case, the complexity of traversal is  $\Theta(|V| + |E|)$ .

```
function DFS(\langle V, E \rangle)

mark each node in V with 0

count ← 0

for each v in V do

if v is marked 0 then

DFSEXPLORE(v)

function DFSEXPLORE(v)

count ← count + 1

mark v with count

for each edge (v, w) do

v is v is neighbour

if v is marked with 0 then

DFSEXPLORE(v)
```

This works both for directed and undirected graphs.

#### DFS

#### Applications

- Connected
- Cyclic

#### BFS

- Queue
- No backtrack needed
- Same complexity as DFS

```
function BFS(\langle V, E \rangle)
    mark each node in V with 0
    count \leftarrow 0, init(queue)

    ▷ create an empty queue

    for each v in V do
       if v is marked 0 then
           count \leftarrow count + 1
           mark v with count
           inject(queue, v)

▷ queue containing just v

           while queue is non-empty do
               u \leftarrow eject(queue)
                                                          ⊳ dequeues u
               for each edge (u, w) adjacent to u do
                   if w is marked with 0 then
                       count \leftarrow count + 1
                       mark w with count
                       inject(queue, w)
                                                         ▷ enqueues w
```

### Topology sort

#### MORE SORTING!!!

- We use DFS:
  - Perform DFS and note the order in which nodes are popped off the stack.
  - List the nodes in the reverse of that order
- Why is topology sort useful?
  - Dependency Resolution
  - Task Scheduling
  - Course Prerequisites
  - Build Systems

## Graph, Greedy and DP

It's all coming back together:))

- Shortest path problem (TSP), again
- Assume graph is connected
- Now think how you can implement it
- Time complexity?

# Floyd's algorithm

```
// Let dist be a 2D array of size V x V (where V is the number of vertices in the graph)
// Initialize the dist matrix with the given weights of the edges
// If there is no edge between vertex i and j, set dist[i][j] to infinity
// Set dist[i][i] to 0 for all vertices i

for k from 0 to V-1 do // Iterates over each vertex as an intermediate vertex.
    for i from 0 to V-1 do // Iterates over each starting vertex.
    for j from 0 to V-1 do // Iterates over each ending vertex.
    if dist[i][j] > dist[i][k] + dist[k][j] then
        dist[i][j] = dist[i][k] + dist[k][j]
```

// The dist array now contains the shortest paths between all pairs of vertices

- O(V<sup>3</sup>) time
- DP: using the shortest paths between intermediate vertices to update the shortest paths between all pairs of vertices.

#### Dijkstra

#### Dike-struh

```
function DIJKSTRA(\langle V, E \rangle, v_0)
    for each v \in V do
        dist[v] \leftarrow \infty
        prev[v] \leftarrow nil
    dist[v_0] \leftarrow 0
    Q \leftarrow \text{InitPriorityQueue}(V)
                                                   > priorities are distances
    while Q is non-empty do
        u \leftarrow \text{EJECTMIN}(Q)
        for each (u, w) \in E do
             if u in Q and dist[u] + weight(u, w) < dist[w] then
                 dist[w] \leftarrow dist[u] + weight(u, w)
                 prev[w] \leftarrow u
                 UPDATE(Q, w, dist[w]) \triangleright rearranges priority queue
```

### Summary

- Graphs are super useful
- A lot more cool algorithms Algorithms and Complexity (COMP90038)
- A lot of resource on line
- More on the proof heavy side Graph Theory (MAST30011)
- If you are really really really keen on math heavy graph related topics Group Theory and Linear Algebra (MAST20022)