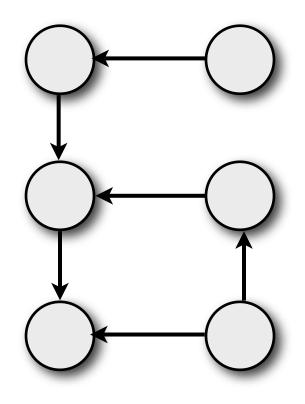
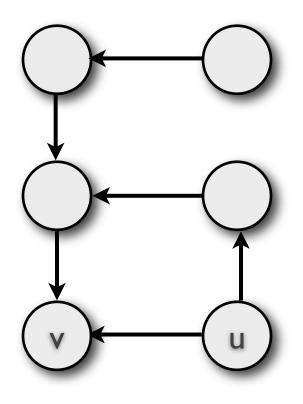
## Graph

#### Ooi Wei Tsang School of Computing, NUS

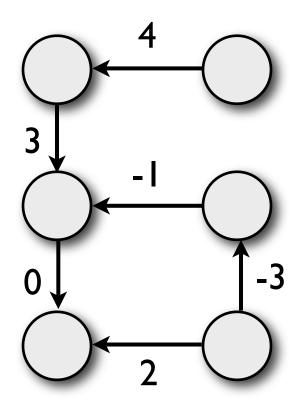
1



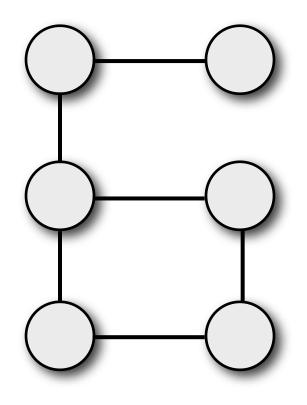
#### A graph consists of edges and vertices.



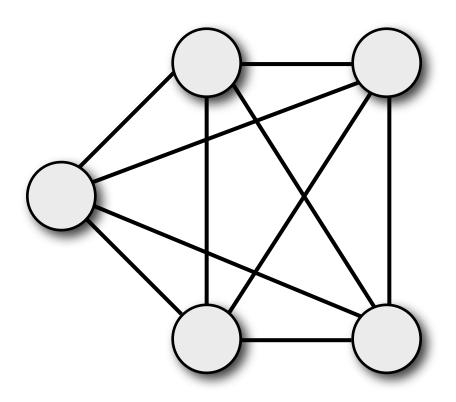
A vertex u is a neighbor of v, if there is an edge from v to u. We say u is adjacent to v. The number of neighbors of a vertex is called degree.



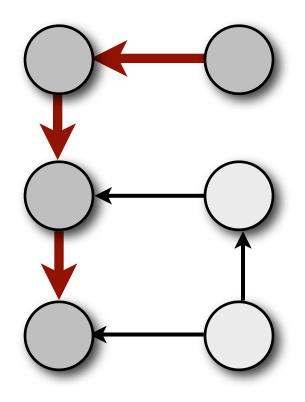
A weighted graph has a value associated with its edges.



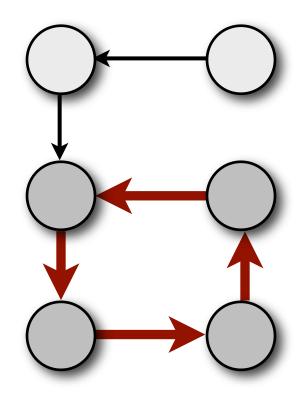
Direction of edges does not matter in a undirected graph.



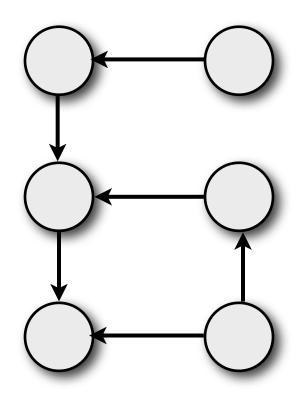
#### In a complete graph, every vertex is connected to every other vertices.



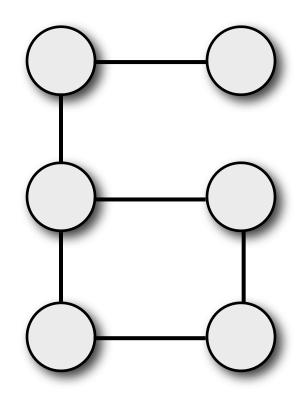
## A path consists of a sequence of vertices adjacent to each other.



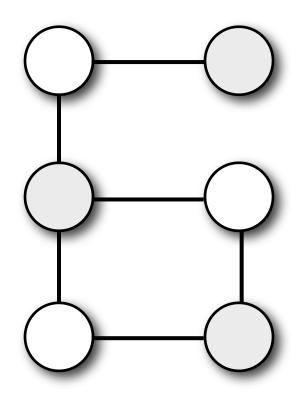
### A cycle is a path that starts and ends with the same vertex.



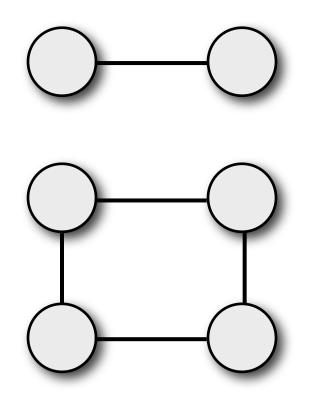
A graph is acyclic if it contains no cycle. It is cyclic otherwise.



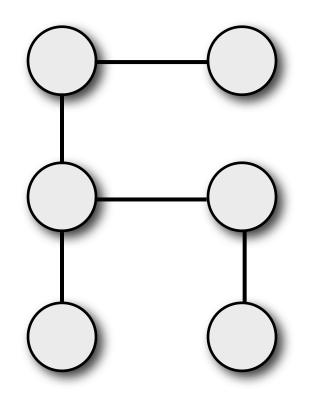
#### A undirected graph is **connected** if there is a path between any two vertices.



A undirected graph is bipartite if we can partition the vertices into two sets and there are no edges between two vertices of the same set.



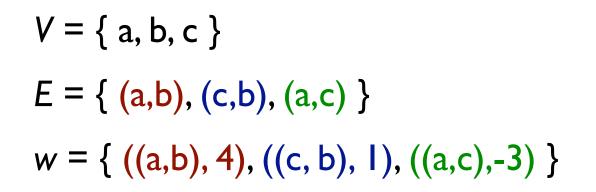
A unconnected graph consists of two connected components.

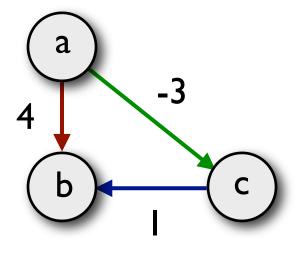


A connected, undirected, acyclic graph is called a tree.

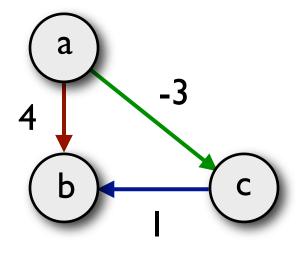
# A weighted graph G = (V, E, w), where

- V is the set of vertices
- E is the set of edges
- *w* is the weight function





# adj(v) : set of vertices adjacent to vertex v adj(a) = {b, c} adj(b) = { } adj(c) = {b}



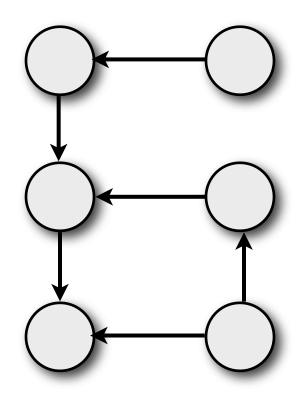
## **Review Questions**

• How many edges are there in a undirected complete graph with N vertices?

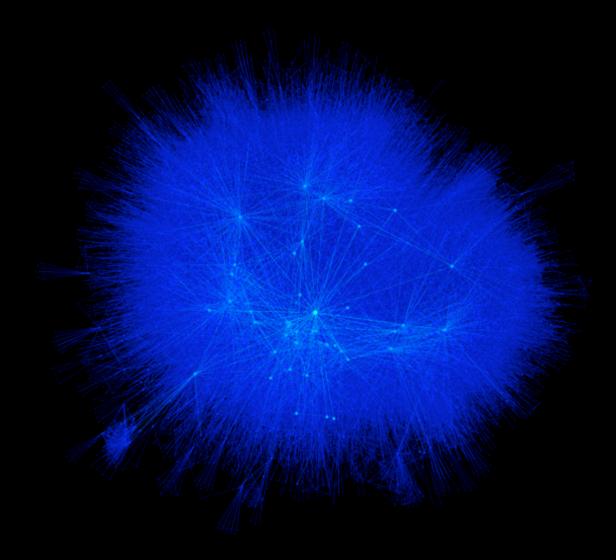
#### **Review Questions**

 $\sum |adj(u)| = ?$  $u \in V$ 

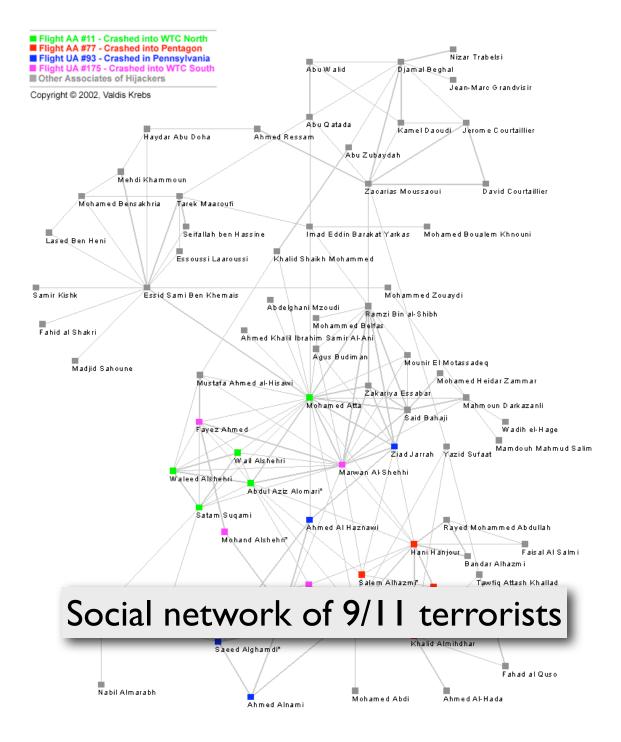
#### **Example Applications**

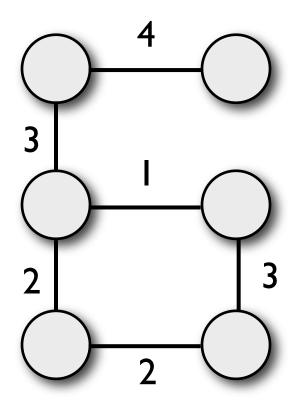


#### Representing a social network. (u,v) in E if u knows v.

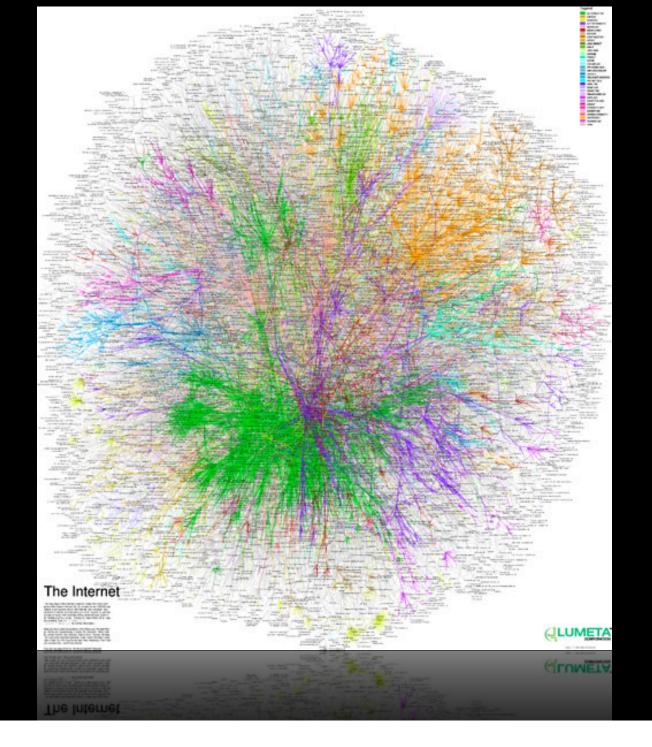


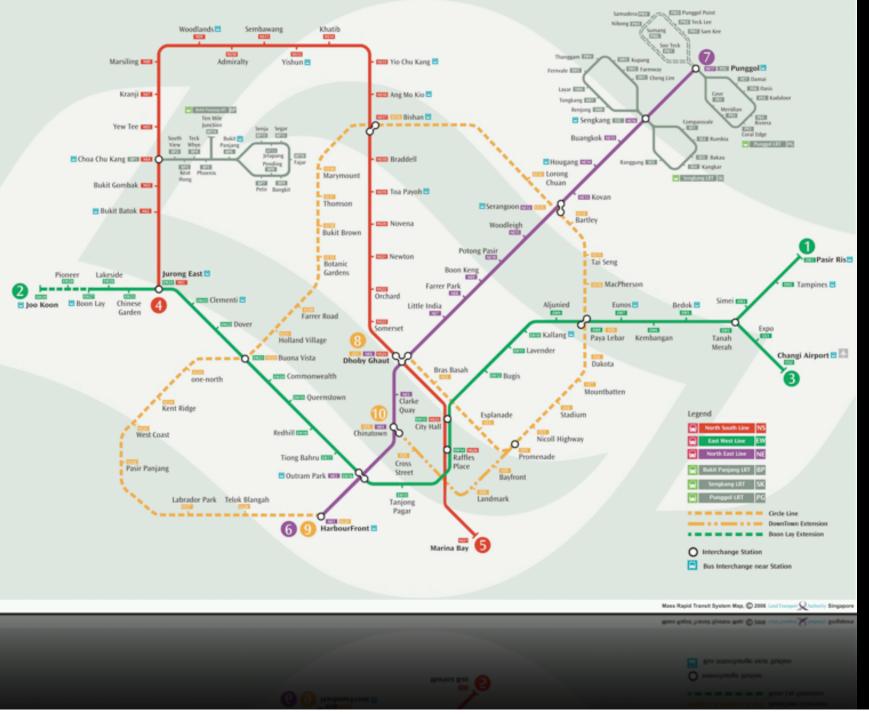
Jeffrey Heer's Social Network from Friendster (47471 people, 432430 edges)





Representing places and routes. (u,v) exists if there is a direct route from u to v. Weight w(u,v) is the distance or cost. We are often interested in finding the cheapest path between between two places.

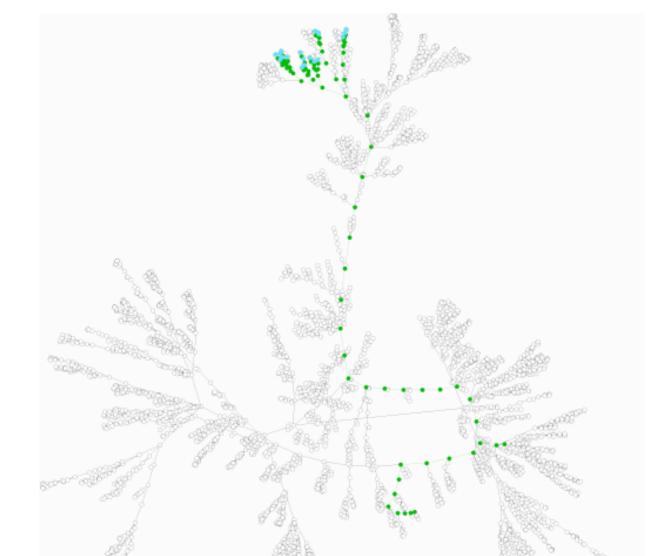




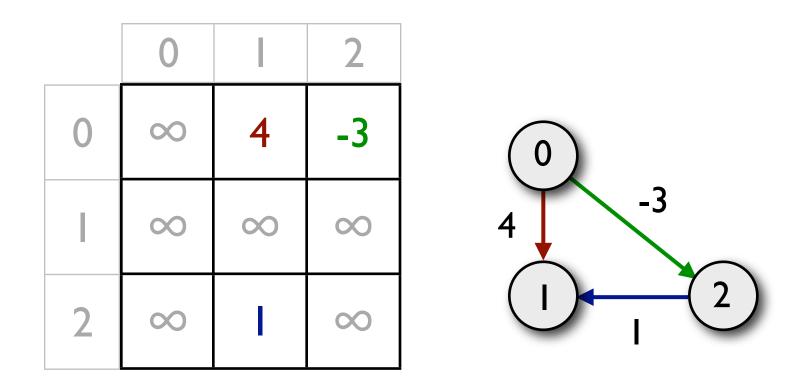
#### MRT ROUTE MAP



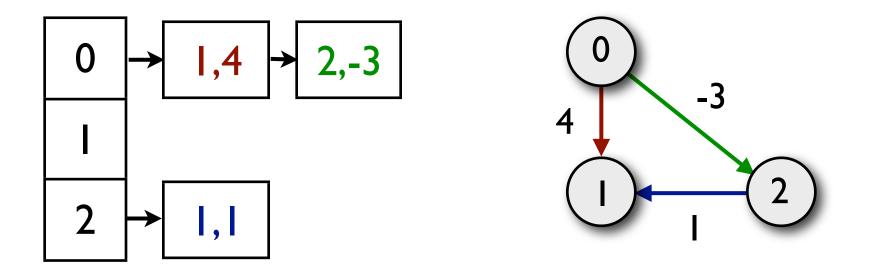




#### Implementation



**Adjacency Matrix**: Use a 2D array. Store w(u,v) in a[u] [v] if edge (u,v) exists. Store an invalid value otherwise.



## **Adjacency List**: Use an array of link list. a[u] stores adj(u) and the associated weight.

How long does it take to delete an edge for
(a) adjacency matrix ?
(b) adjacency list ?

 How long does it take to go through all neighbors of a vertex v for

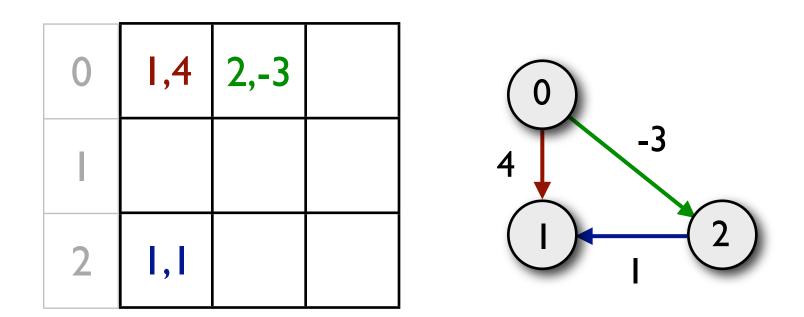
(a) adjacency matrix ?

(b) adjacency list ?

• How much space is needed to store a graph of size N if we are using

(a) adjacency matrix ?

(b) adjacency list ?



## Adjacency List in Matrix: Use a 2D array. Each row is an array-representation of the adjacency list.

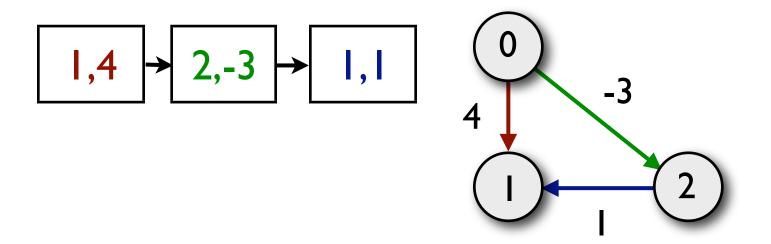
Avoid using pointers in competitive programming.

Most of the time, graph are static (no insert/ delete after initialization).

Maximum size is often given.

typedef struct neighbor {
 int id;
 int weight;
} neighbor;

// N is max num of vertices; neighbor graph[N][N]; int num\_of\_vertices;

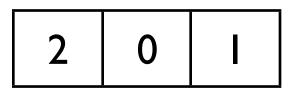


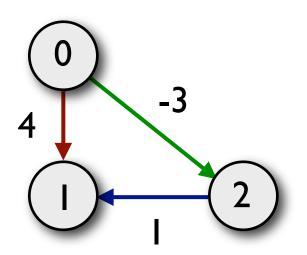
#### **Edge List**: Use a linked list of edges.





Degree





#### **Edge List**: Use a array of edges.

typedef struct edge {
 int from;
 int to;
 int weight;
} edge;

edge graph[MAX\_NUM\_OF\_EDGES];
int num\_of\_edges;

Pick the simplest implementation that meets the requirements.

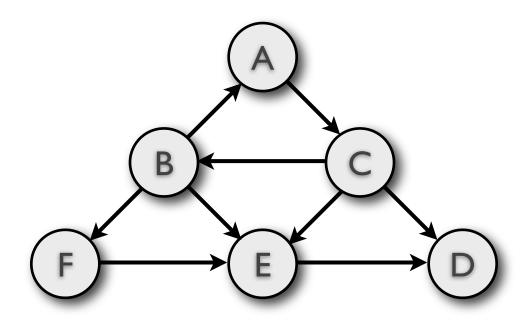
## Graph Traversal

How to systematically visit the whole graph?

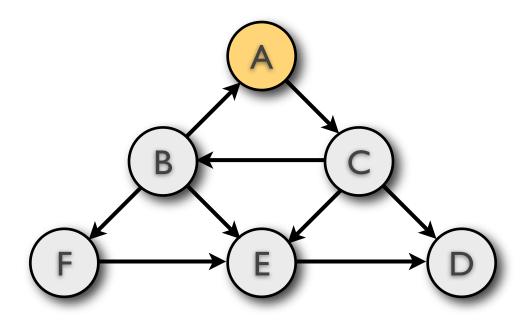
# Breadth-First Search

or BFS

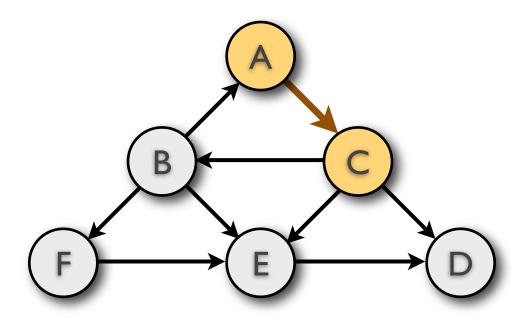
- **Basic idea**: pick a source and visit the vertices in increasing distance from the source
  - visit all vertices one hop away
  - visit all vertices two hops away etc.
- Note: A vertex u is k-hop away from the v if the shortest path from u to v consists of k edges.



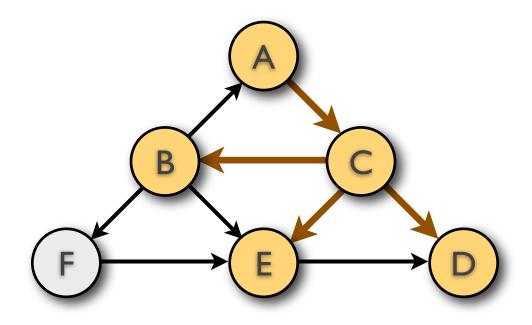
Example: F is 3-hop away from A. E is 2-hop away from A.



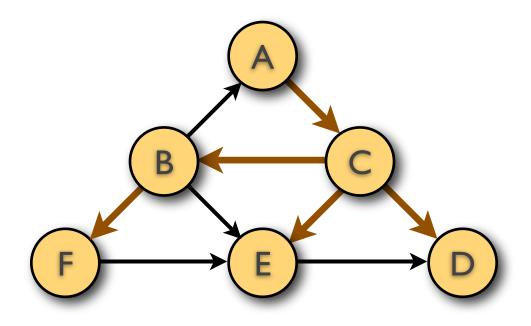
Let A be the source. We first visit the source. I colored visited vertices yellow.



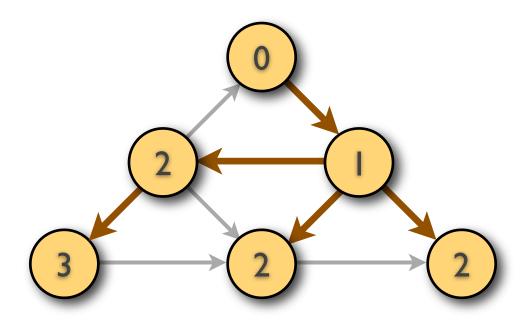
Next, visit the vertices that are one-hop away.



Next, visit the vertices that are two hops away. (i.e, all unvisited vertices that are neighbors of one-hop neighbor of A.



Edges that lead to undiscovered node during traversal are colored brown.

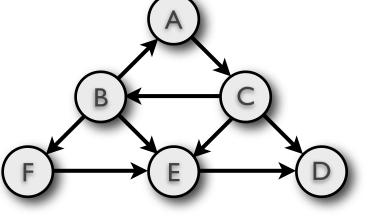


These edges form the breadth-first tree. Level of vertices in the tree is the hop distance from source.

- An implementation needs to keep track of vertices we have discovered.
- To visit the vertices in increasing order of hop distance, we need to visit the nodes the order we discover them (FIFO).

Q = new Queue enqueue source into Q

while Q is not empty v = dequeue from Qmark v as visited for each neighbor u of v if u is not visited and not already in Q enqueue u into Q

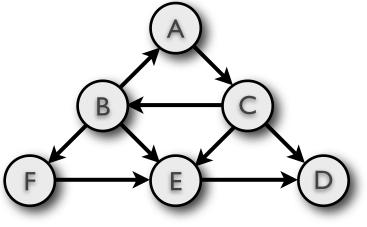


### **Review Questions**

• Suppose we want to keep track of breadthfirst tree by marking the edges in the tree as brown. How should we change the algorithm?

Q = **new** Queue enqueue source into Q

while Q is not empty v = dequeue from Qmark v as visited for each neighbor u of v if u is not visited and not already in Q mark (v,u) as brown enqueue u into Q



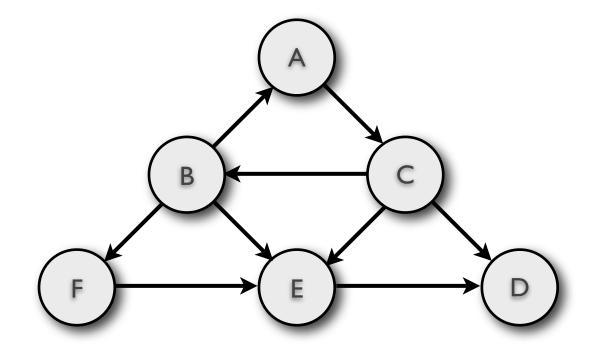
### **Review Questions**

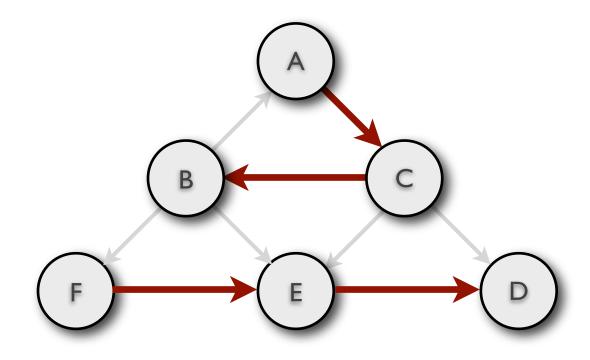
 Suppose we want to keep track of hop distance from the source. How should we change the algorithm? Q = **new** Queue enqueue source into Qlevel[source] = 0 while Q is not empty v = dequeue from Qmark v as visited for each neighbor u of v if u is not visited and not already in Q level[u] = level[v] + l

enqueue u into Q

### **Review Questions**

• Can we always visit every vertex using the previous algorithm?



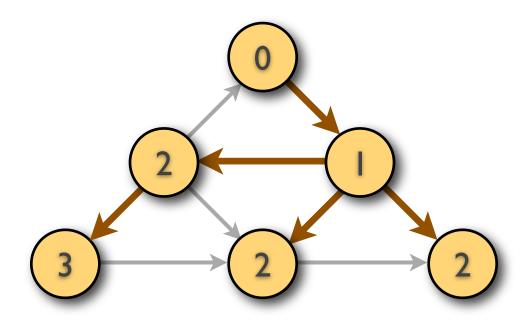


If we pick F as the source, then we can't visit A, B, and C, and need to visit them through another source.

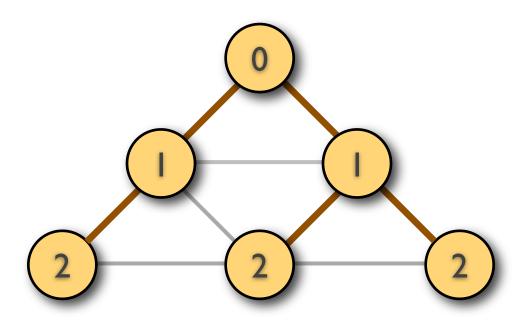
Mark all vertices as unvisited

for each vertex v
if v is not visited
 use v as source and run BFS

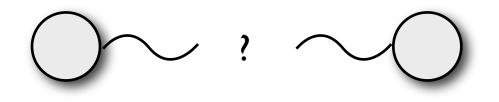
### Applications of BFS



On an unweighted graph, the breadth-first tree tells us the shortest path from source to all the other vertices.



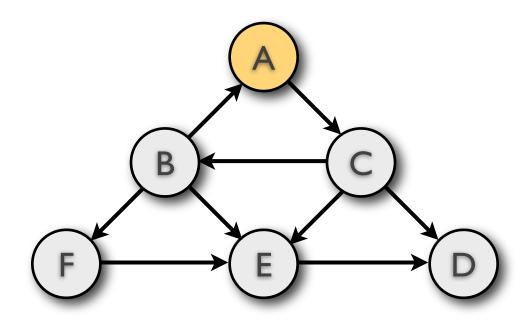
The algorithm works for undirected graph too.



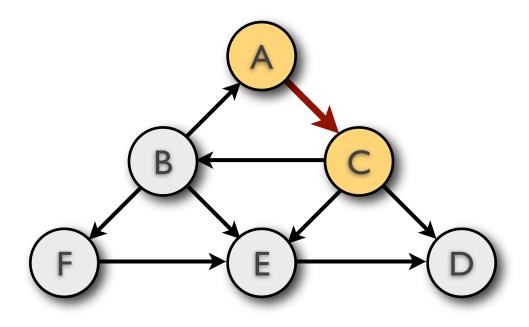
#### We can check if two vertices are connected using BFS.

### Depth-First Search or, DFS

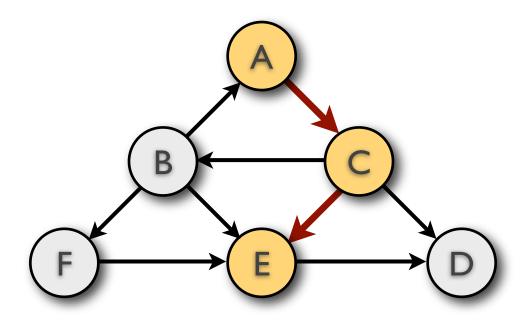
- **Basic idea**: Starting from a source, repeatedly visit a neighbor of the current vertex until we hit a dead-end (no unvisited neighbors), then backtrack.
- After we visit a vertex v, we visit all vertices reachable from v.



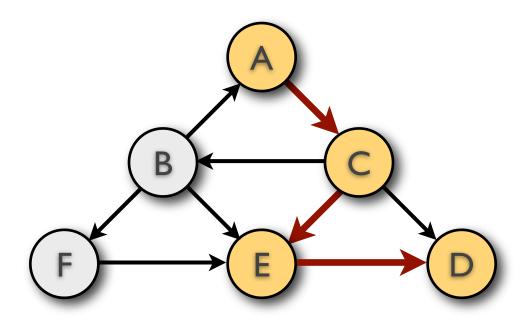
Let A be the source.



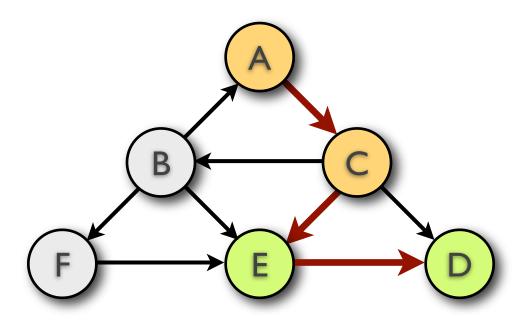
Visit a neighbor of A (say, C).



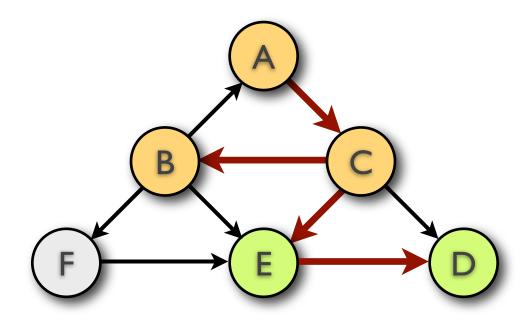
#### Visit a neighbor of C (say, E).



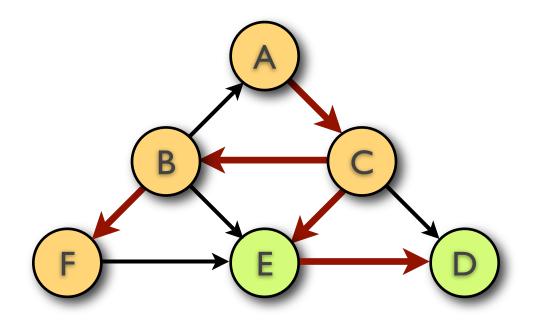
#### Visit a neighbor of E (say, D).



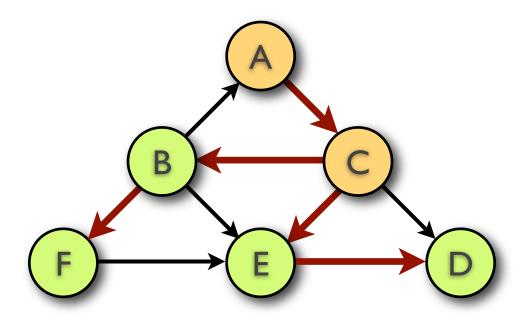
D has no neighbor. Back to E. E has no unvisited neighbor. Back to C.



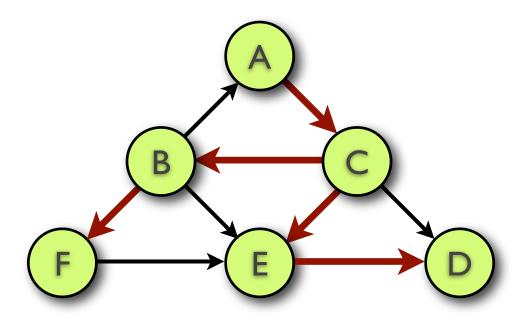
Visit B.



Visit F.



F has no unvisited neighbor. Back to B. B has no unvisited neighbor. Back to C.



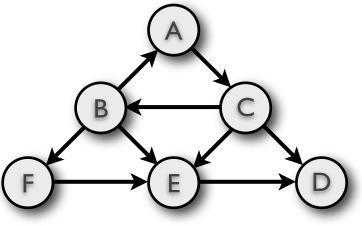
C has no unvisited neighbor. Back to A. A, the source, has no unvisited neighbor. Done!

- An implementation needs to keep track of vertices we have discovered.
- When backtrack, we need to go back to the last vertex we visited. (LIFO).

S = **new** Stack push *source* onto S

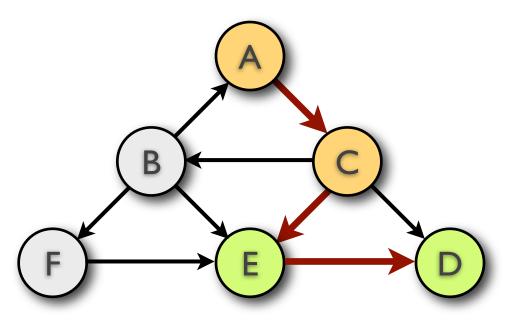
while S is not empty
v = top of S
if v has a unvisited neighbor u
mark u as visited
push u onto S
else

pop v from S



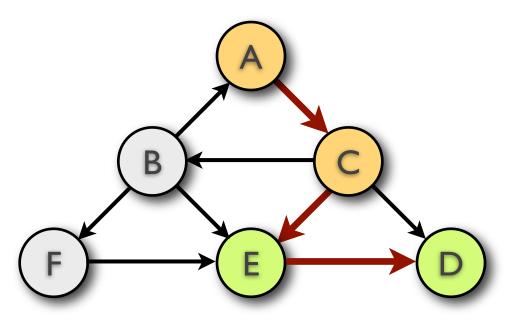
Mark all vertices as unvisited

for each vertex v
if v is not visited
 use v as source and run DFS



D has no neighbor. Back to E. E has no unvisited neighbor. Back to C.

What is the color of a vertex:
(a) before it is inserted into the stack ?
(b) while it is inside the stack ?
(c) after it is pop from the stack ?

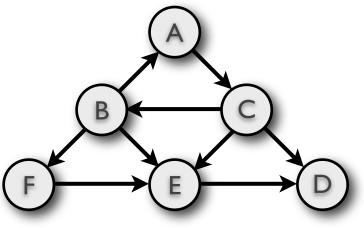


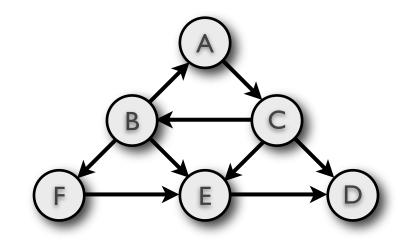
D has no neighbor. Back to E. E has no unvisited neighbor. Back to C.

A vertex can be in three states: unvisited, visiting, visited.

S = **new** Stack push *source* onto S

while S is not empty v = top of Sif v has a unvisited neighbor u mark *u* as "visiting" push *u* onto S else pop v from S mark *u* as "visited"



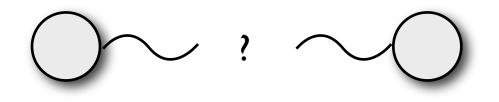


proc DFS(u):
// recursive version of DFS
mark u as "visiting"
for each unvisited neighbor v of u
 DFS(v)
mark u as "visited"

### **Review Questions**

- True/False? : There is always a path from the vertices in the stack to the vertex at the top of the stack.
- (Alternatively: There is always a path from a vertex marked "visiting" to the current vertex.)

#### Applications of DFS



#### We can check if two vertices are connected using DFS.



#### We can check if a graph is acyclic/cyclic using DFS.



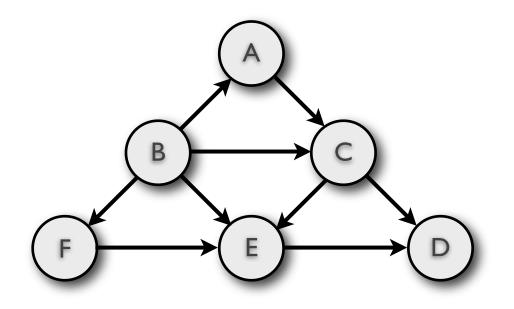
There is a cycle iff we found an edge from current vertex to a visiting vertex (called backward edge)

#### **proc** DFS(u):

mark u as "visiting" for each neighbor v of u if v is marked as "visiting" we found a cycle! else if v is marked as "unvisited" DFS(v) mark u as "visited"

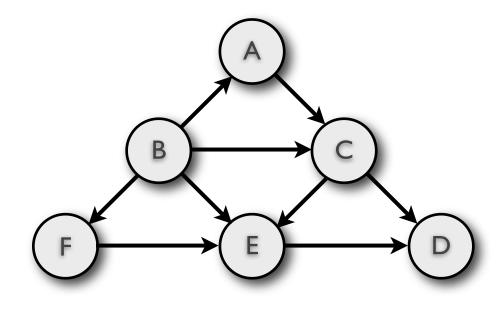
## **Topological Sort**

**Goal**: Given a directed acyclic graph, order the vertices such that if there is a path from *u* to *v*, then *u* appears before *v* in the output. **Goal**: Given a directed acyclic graph, order the vertices such that if there is a path from *u* to *v*, then *u* appears before *v* in the output.

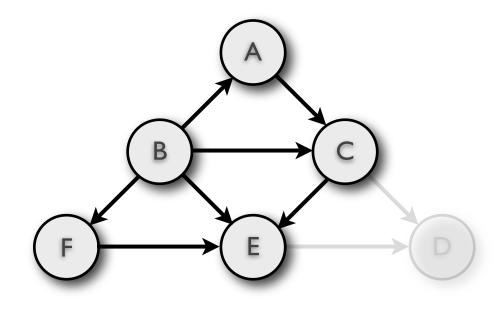


BACFED? BCAFED? BFACED?

### **Idea**: The first vertex marked "visited" can appear last in the topological order.



Now, we remove that vertex from consideration, and repeat -- the next vertex marked as visited can appear last in the topological sort order.



**proc** DFS(u):

# for each unvisited neighbor v of u DFS(v) push u onto a stack

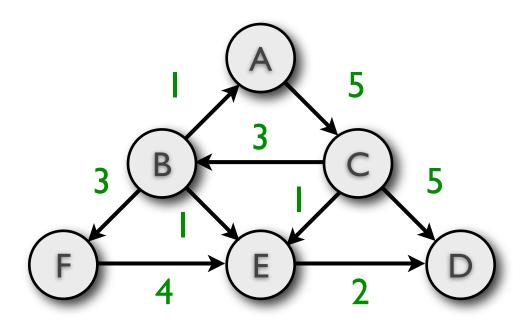
To output in topological sort order, pop from stack and print after completing DFS.

#### Dijkstra's Algorithm

### Single-Source Shortest Path

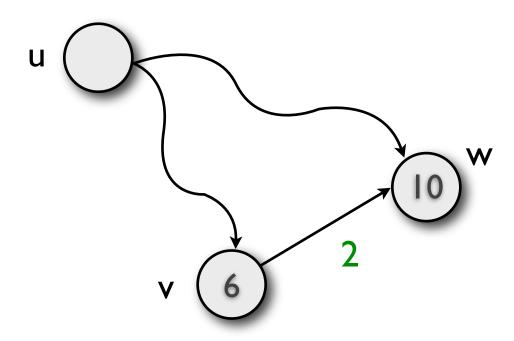
• **Problem**: Given a weighted graph G and a vertex v in G, find the shortest (or least cost) path from v to all other vertices.

• Restrict ourselves to **positive** weight.



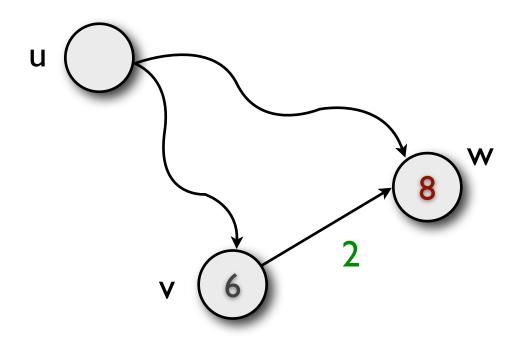
Shortest Path from A to D = A-C-E-D (Cost = 8)

- Must keep track of smallest distance so far.
- If we found a new, shorter path, update the distance.



Let d[v] be the current known shortest distance from u to v.

d[v] = 6, d[w] = 10

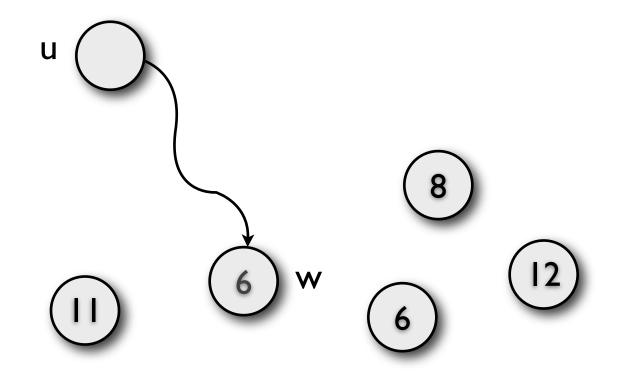


We just found a shorter path from u to w. Update d[w] = d[v] + cost(v,w). We call this step relax(v,w). **proc** relax (v,w):

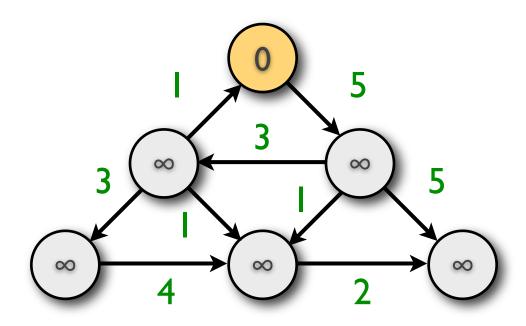
```
Let d = d[v] + cost(v,w)

if d[w] > d

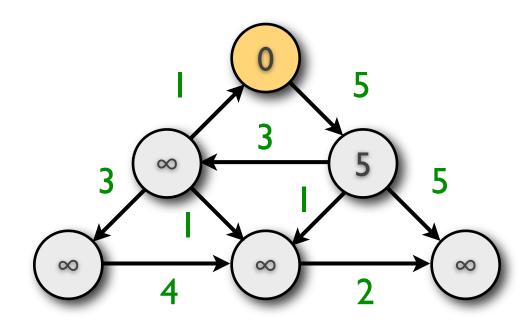
d[w] = d
```



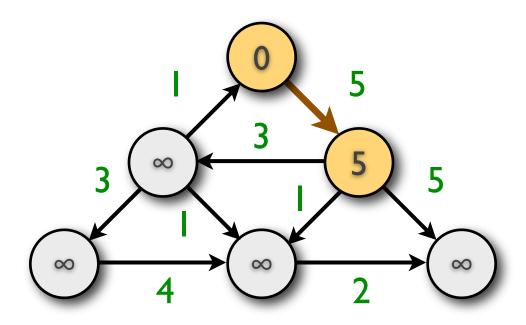
If d[w] is the smallest among the "remaining" vertices, then d[w] is the smallest possible (can't be relaxed further)



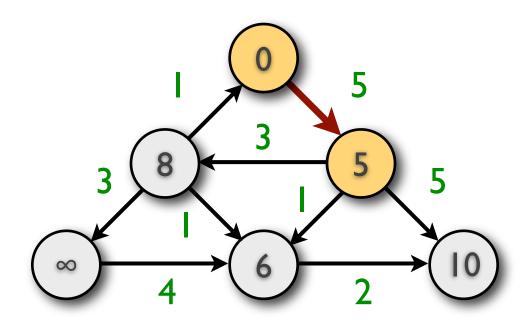
At the beginning, we know d[A]. But the rest is unknown and is set to infinity.



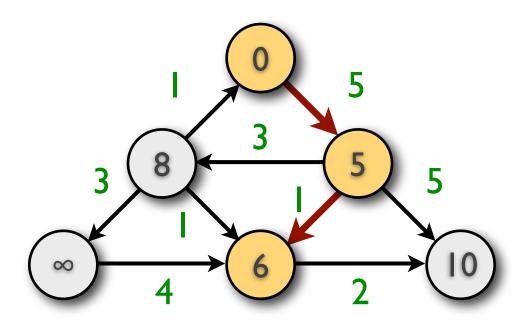
Relax all neighbors of A.



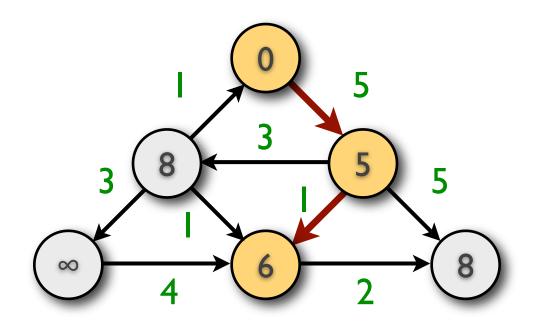
Pick a white vertex with smallest d[]. Color it yellow.



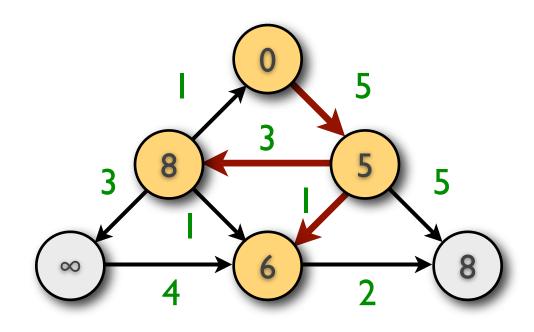
#### Relax all neighbors of this vertex.

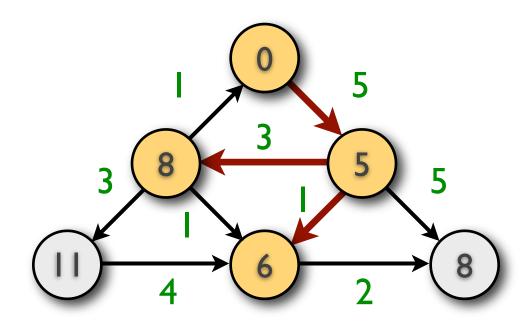


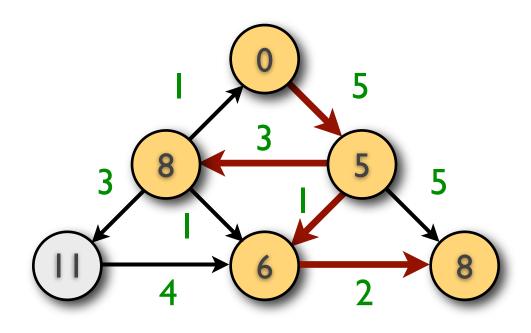
Repeat: pick a white vertex with smallest d[].

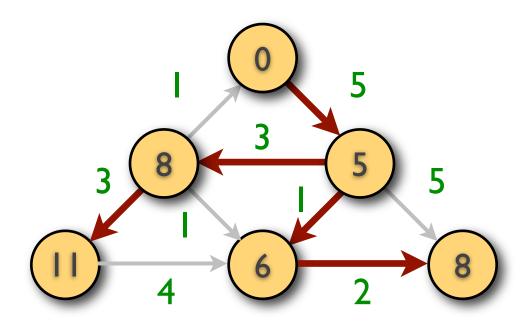


Relax its neighbors









#### Everyone is yellow. Done!

proc Dijkstra(s):

for each vertex v in G d[w] = infinity color[w] = white

d[s] = 0

while there exists a white vertex

let u be a white vertex with smallest d
color[u] = yellow
for each neighbor v of u
relax(u,v)

### Array Implementation

```
while there exists a white vertex
min = infinity
for each vertex v
if color[v] is white and d[v] < min
min = d[v]
u = v</pre>
```

```
color[u] = yellow
for each neighbor v of u
```

### Priority Queue Implementation

while there exists a white vertex

```
u = q.getMin()
color[u] = yellow
for each neighbor v of u
relax(u,v)
```

### Priority Queue Implementation

**proc** relax (v,w):

```
Let d = d[v] + cost(v,w)

if d[w] > d

d[w] = d

q.decreaseCost(w, d)
```

# Summary: Graph

- Basic terms
- Representations
- Applications
- BFS
  - find shortest path in unweighted path
  - finding connected component

# Summary: Graph

#### • DFS

- finding connected component
- check for cycles
- topological sort
- Dijkstra algorithm
  - finding shortest path from a single source in a weighted graph with positive weights.