



# Shortest Paths

CSE 332 25Sp  
Lecture 17

# Announcements

	Monday	Tuesday	Wed	Thursday	Friday
This Week					<b>TODAY</b> → Ex 6 (sorting, gs) due Ex 8 ( <u>Dijkstra</u> , gs) out
Next Week	<u>Ex 7 (DFS, prog)</u> due Ex 9 (reductions, gs) out				Ex 8 due Ex 10 (parallel, prog)

MT grades  
→ Solutions on Ed  
regards given tomorrow  
post early next

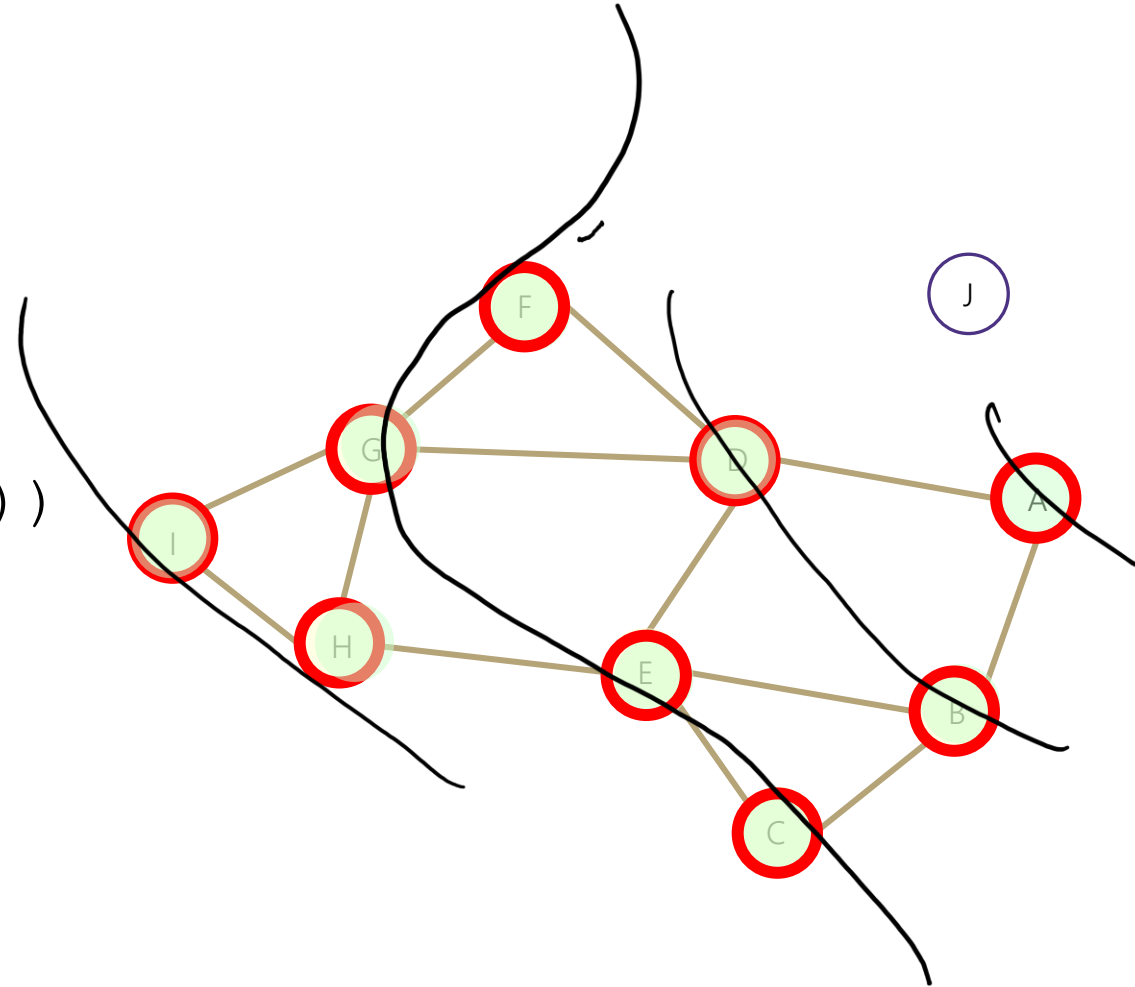
# Breadth First Search

```
search(graph)
  toVisit.enqueue(first vertex)
  mark first vertex as visited
  while(toVisit is not empty)
    current = toVisit.dequeue()
    for (V : current.neighbors())
      if (v is not visited)
        toVisit.enqueue(v)
        mark v as visited
    finished.add(current)
```

Current node: I

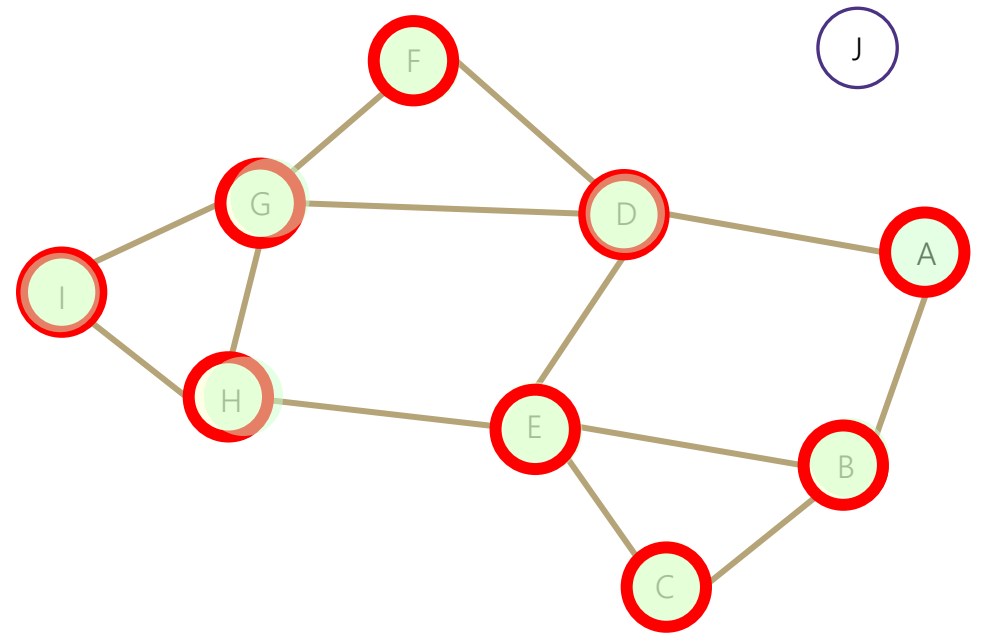
Queue: B D E C F G H I

Finished: A B D E C F G H I



# Breadth First Search

```
search(graph)
  toVisit.enqueue(first vertex)
  mark first vertex as visited
  while(toVisit is not empty)
    current = toVisit.dequeue()
    for (V : current.neighbors())
      if (v is not visited)
        toVisit.enqueue(v)
        mark v as visited
    finished.add(current)
```



What's the running time of this algorithm?

We visit each vertex at most twice, and each edge at most once:  $O(|V| + |E|)$

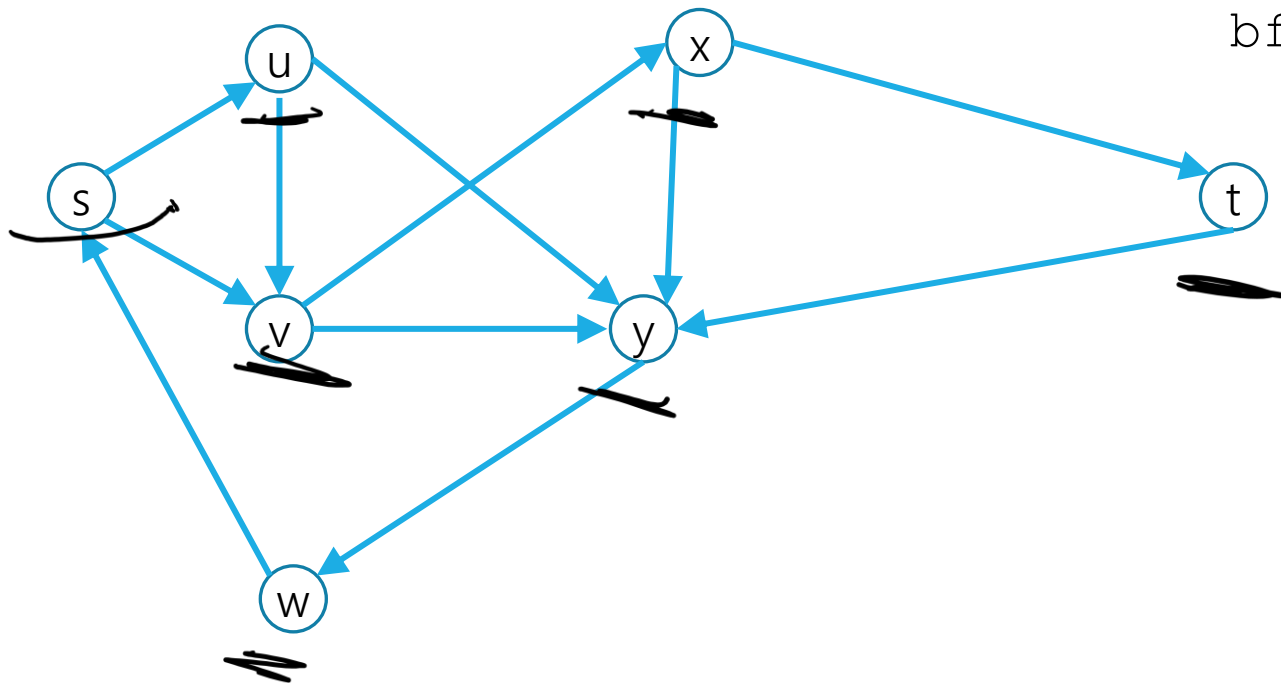
# Extra Practice

Run Breadth First Search on this graph starting from s.

What order are vertices placed on the queue?

When processing a vertex insert neighbors in alphabetical order.

In a directed graph, BFS only follows an edge in the direction it points.



```
bfs(graph)
  toVisit.enqueue(first vertex)
  mark first vertex as visited
  while(toVisit is not empty)
    current = toVisit.dequeue()
    for (V : current.outneighbors())
      if (v is not visited)
        toVisit.enqueue(v)
        mark v as visited
    finished.add(current)
```

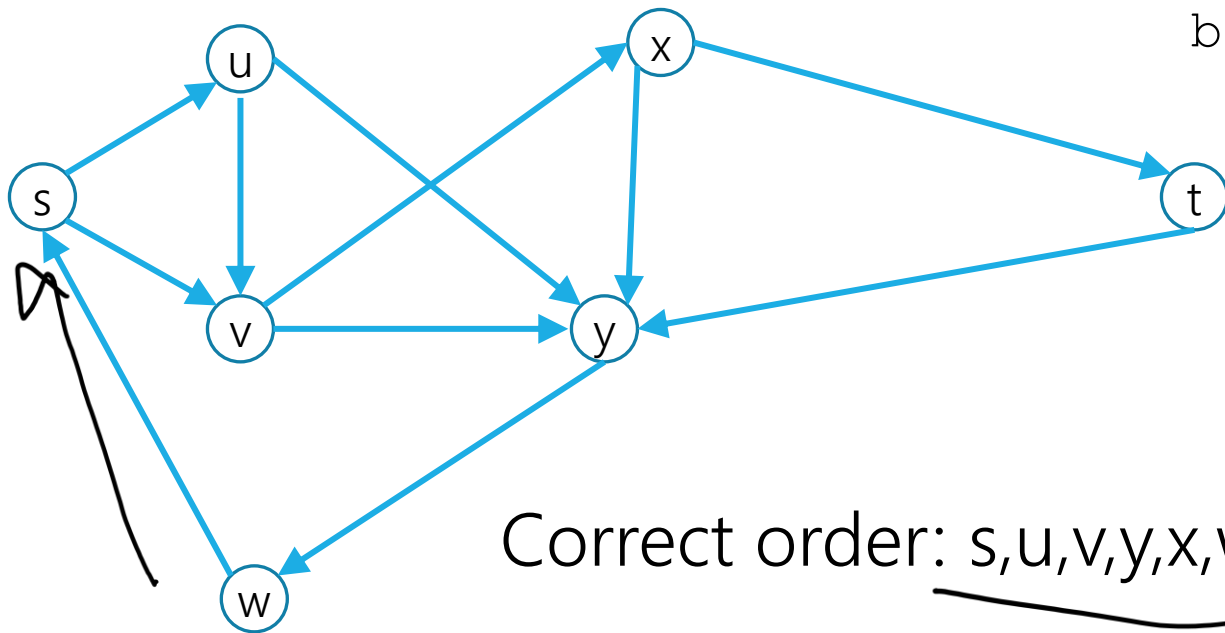
# Extra Practice

Run Breadth First Search on this graph starting from s.

What order are vertices placed on the queue?

When processing a vertex insert neighbors in alphabetical order.

In a directed graph, BFS only follows an edge in the direction it points.



Correct order: s, u, v, y, x, w, t

```
bfs(graph)
```

```
  toVisit.enqueue(first vertex)
```

```
  mark first vertex as visited
```

```
  while(toVisit is not empty)
```

```
    current = toVisit.dequeue()
```

```
    for (V : current.outneighbors())
```

```
      if (v is not visited)
```

```
        toVisit.enqueue(v)
```

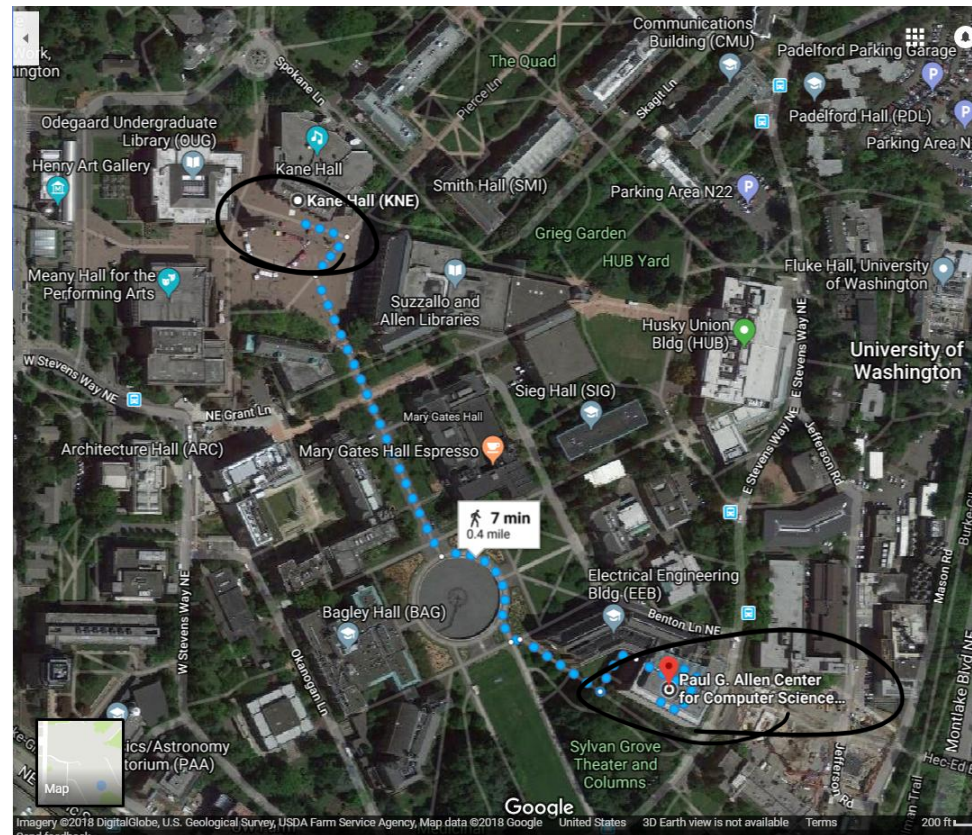
```
        mark v as visited
```

```
  finished.add(current)
```

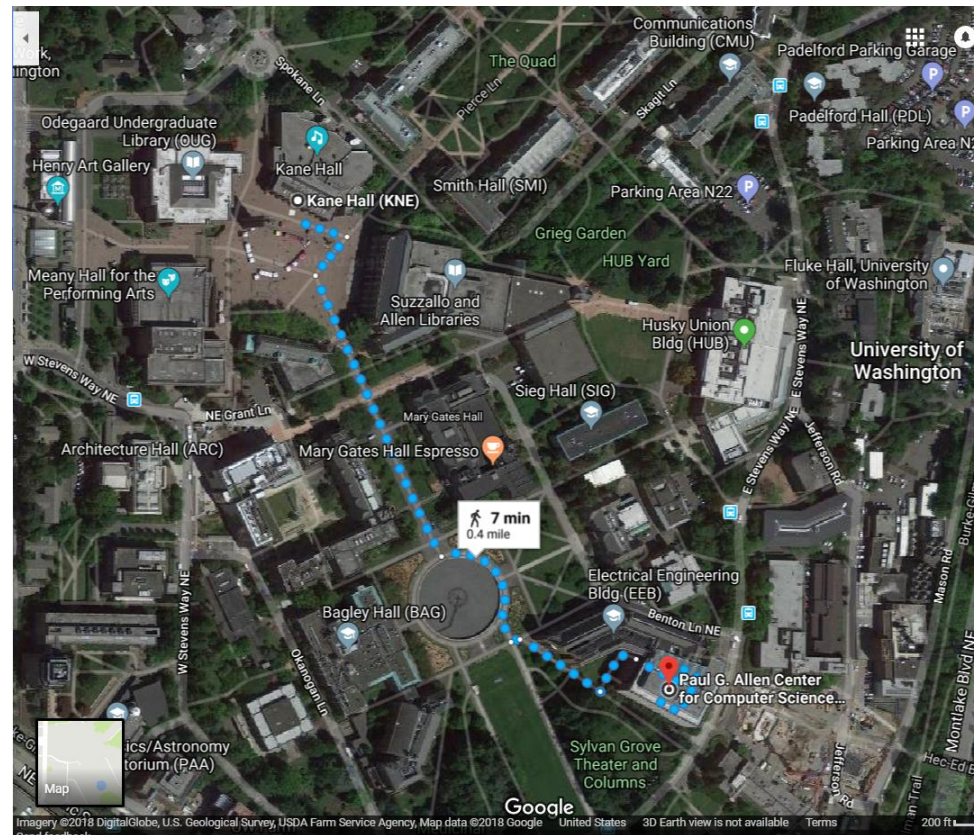


# Shortest Paths

How does Google Maps figure out this is the fastest way to get from Kane Hall to CSE?

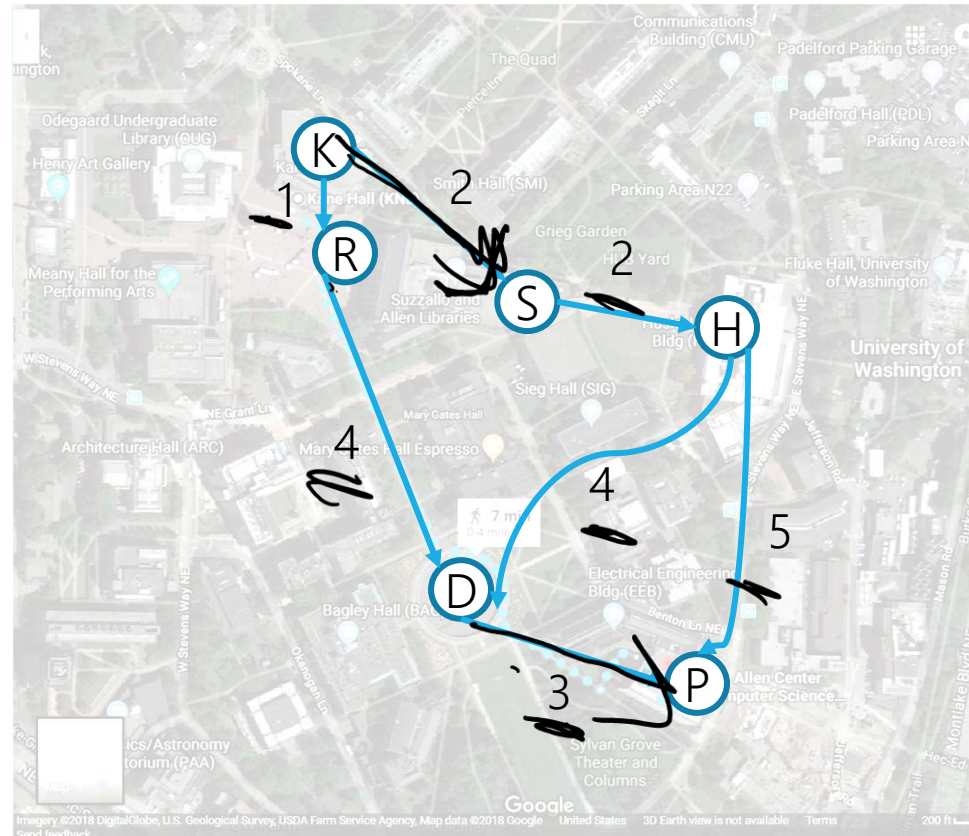


How do we represent a map as a graph? What are the vertices and edges?





# Representing Maps as Graphs

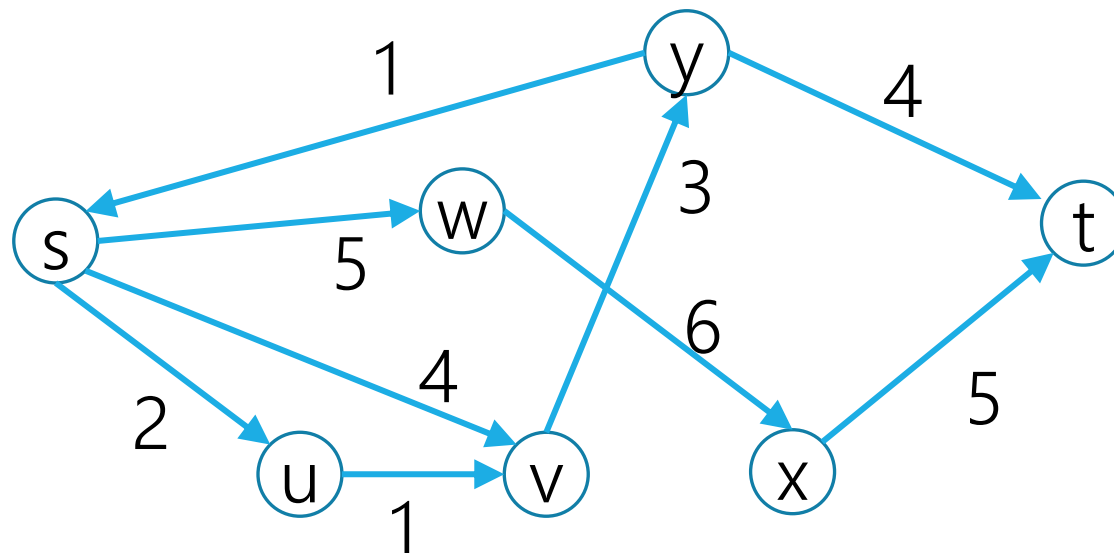


# Shortest Paths

The length of a path is the sum of the edge weights on that path.

## Shortest Path Problem

Given a directed graph and vertices  $s$  and  $t$   
Find: the shortest path from  $s$  to  $t$ .

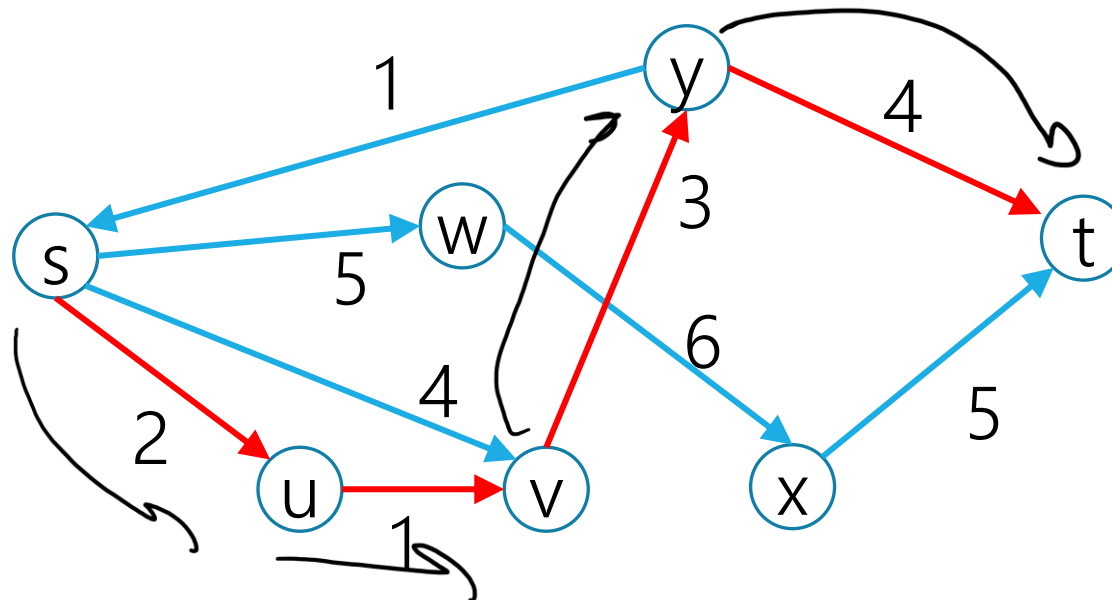


# Shortest Paths

The **length** of a path is the sum of the edge weights on that path.

## Shortest Path Problem

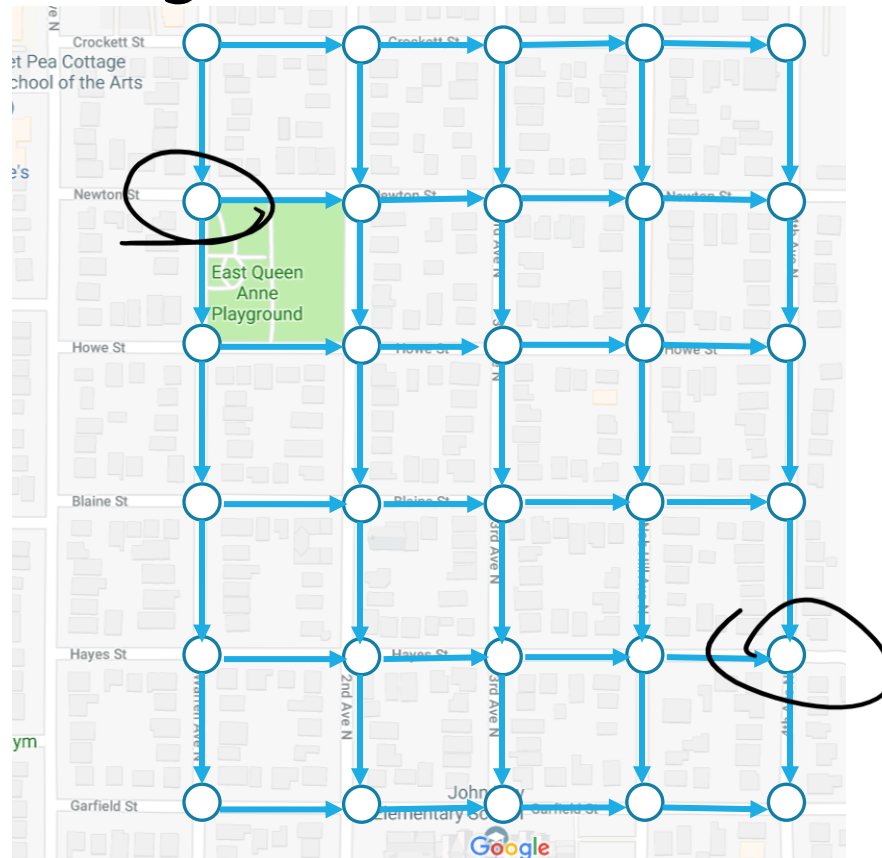
Given a directed graph and vertices  $s$  and  $t$   
Find: the shortest path from  $s$  to  $t$ .



# Unweighted Graphs

Let's start with a simpler version: the edges are all the same weight

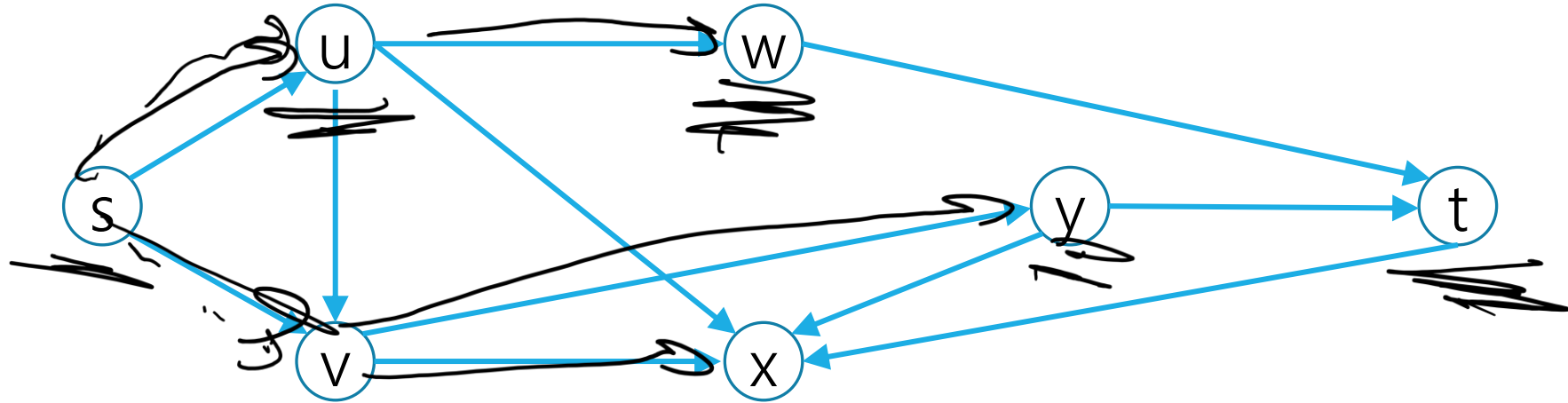
If the graph is **unweighted**, how do we find a shortest paths?





# Unweighted Graphs

If the graph is unweighted, how do we find a shortest paths?



What's the shortest path from s to ~~s~~?

Well....we're already there.

What's the shortest path from s to u or v?

Just go on the edge from s

From s to w, x, or y?

Can't get there directly from s, for length 2 path, have to go through u or v.

# Unweighted Graphs: Key Idea

To find the set of vertices at distance  $k$ , just find the set of vertices at distance  $k-1$ , and check for outgoing edge to an undiscovered vertex.

Do we already know an algorithm that does something like that?

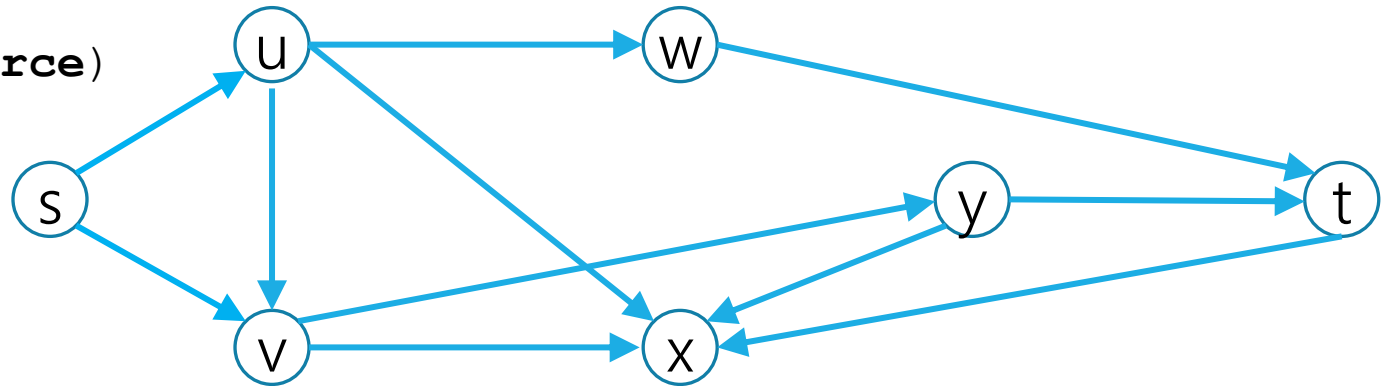
Yes! BFS!

```
bfsShortestPaths(graph G, vertex source)  
    toVisit.enqueue(source)  
    → source.dist = 0  
    mark source as visited  
    while(toVisit is not empty){  
        current = toVisit.dequeue()  
        for (v : current.outNeighbors()){  
            if (v is not yet visited){  
                [ v.distance = current.distance + 1  
                  v.predecessor = current ]  
                toVisit.enqueue(v)  
                mark v as visited  
            }  
        }  
    }  
}
```

# Unweighted Graphs

Use BFS to find shortest paths in this graph.

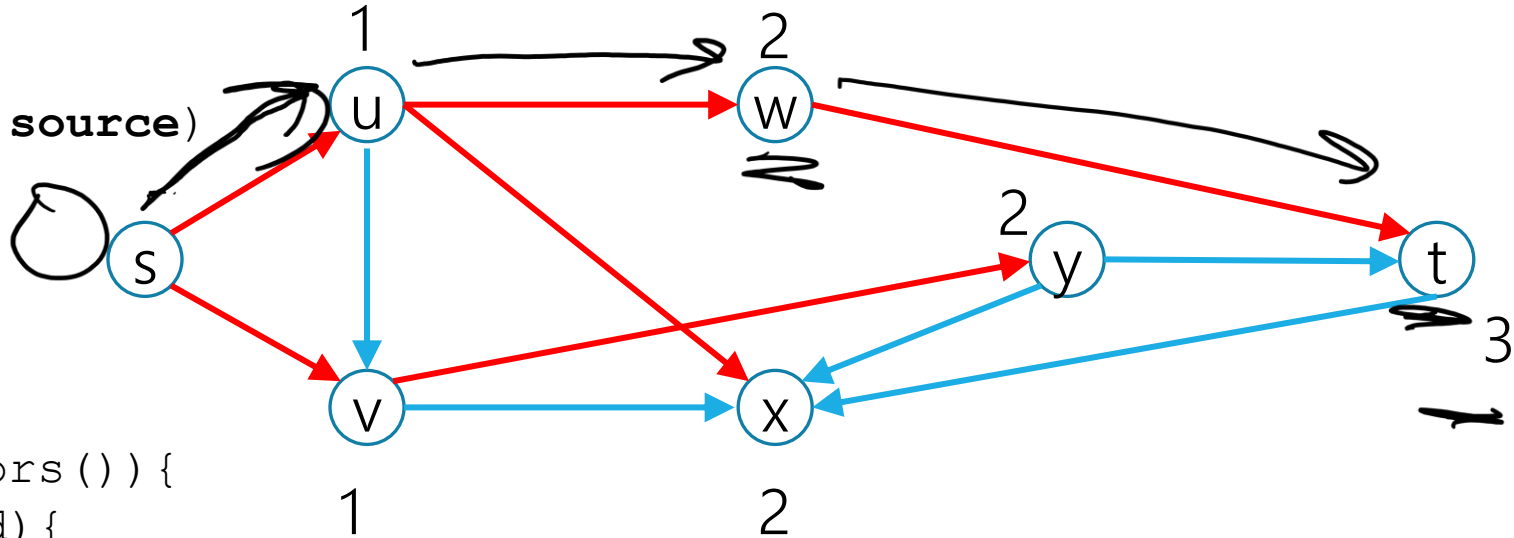
```
bfsShortestPaths(graph G, vertex source)  
  toVisit.enqueue(source)  
  source.dist = 0  
  mark source as visited  
  while(toVisit is not empty){  
    current = toVisit.dequeue()  
    for (v : current.outNeighbors()){  
      if (v is not yet visited){  
        v.distance = current.distance + 1  
        v.predecessor = current  
        toVisit.enqueue(v)  
        mark v as visited  
      }  
    }  
  }  
}
```



# Unweighted Graphs

Use BFS to find shortest paths in this graph.

```
bfsShortestPaths(graph G, vertex source)
  toVisit.enqueue(source)
  source.dist = 0
  mark source as visited
  while(toVisit is not empty){
    current = toVisit.dequeue()
    for (v : current.outNeighbors()){
      if (v is not yet visited){
        v.distance = current.distance + 1
        v.predecessor = current
        toVisit.enqueue(v)
        mark v as visited
      }
    }
  }
}
```





# What about the target vertex?

## Shortest Path Problem

Given a directed graph and vertices  $s$  and  $t$   
Find: the shortest path from  $s$  to  $t$ .

BFS didn't mention a target vertex...

It actually finds the shortest path from  $s$  to every other vertex.

If you know your target, you can stop the algorithm early, when the target is removed from the queue.

# Weighted Graphs

Each edge should represent the “time” or “distance” from one vertex to another.

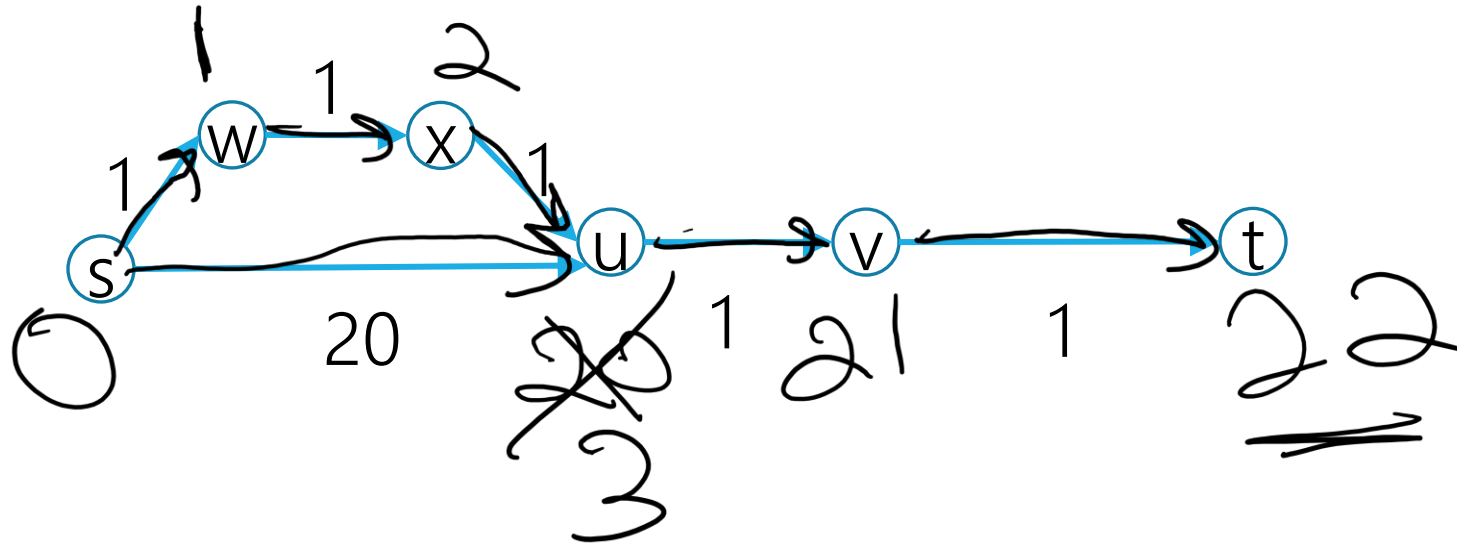
Sometimes those aren’t uniform, so we put a weight on each edge to record that number.

The length (or “**weight**” or “**cost**”) of a path in a weighted graph is the sum of the weights along that path.

# Weighted Graphs: Take 1

BFS works if the graph is unweighted.

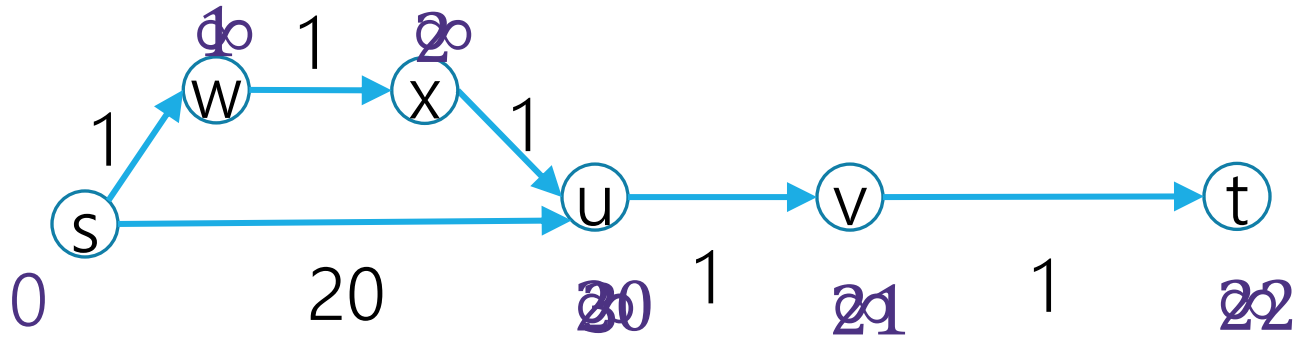
Maybe it just works for weighted graphs too?



# Weighted Graphs: Take 1

BFS works if the graph is unweighted.

Maybe it just works for weighted graphs too?



What went wrong?

When we found a shorter path from s to u, we needed to update the distance to v but BFS doesn't do that.

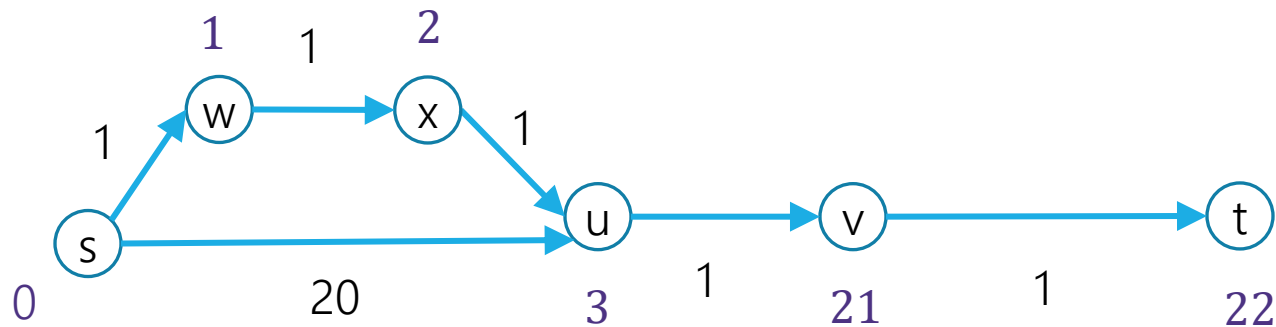


# Weighted Graphs: Take 2

So we can't just do a reduction.

Instead figure out why BFS worked in the unweighted case, try to make the same thing happen in the weighted case.

How did we avoid this problem:



# Weighted Graphs: Take 2

In BFS When we used a vertex  $u$  to update shortest paths we already knew the exact shortest path to  $u$ .

So we never ran into the update problem

If we process the vertices in order of distance from  $s$ , we have a chance.

# Weighted Graphs: Take 2

Goal: Process the vertices in order of distance from  $s$

Idea:

[ Have a set of vertices that are "known"  
- (we know at least one path from  $s$  to them). ]

[ Record an estimated distance  
- (the best way we know to get to each vertex). ]

If we process only the vertex closest in estimated distance, we won't ever find a shorter path to a processed vertex.

- This statement is the key to proving correctness.
- It's nice if you want to practice induction/understand the algorithm better.

# Dijkstra's Algorithm

Dijkstra(Graph G, Vertex source)

initialize distances to  $\infty$

mark source as distance 0

mark all vertices unprocessed

while(there are unprocessed vertices){

let u be the closest unprocessed vertex

foreach(edge (u,v) leaving u){

if(~~u.dist+weight(u,v) < v.dist~~) {

v.dist = u.dist+weight(u,v)

v.predecessor = u

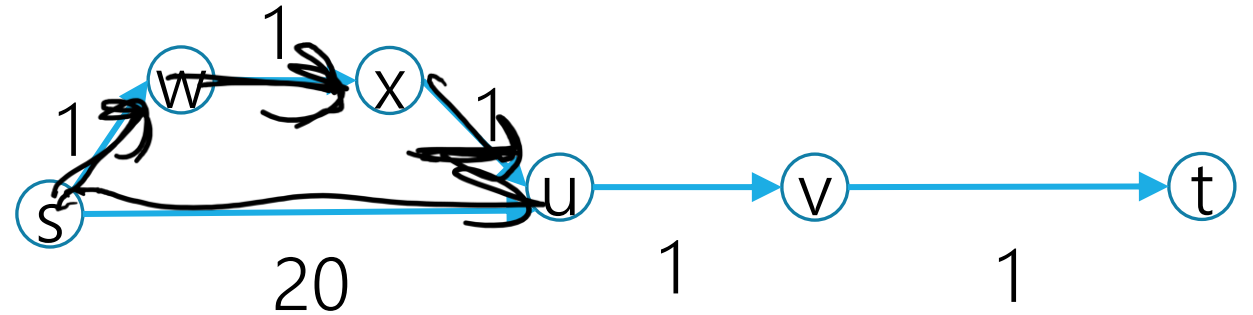
}

}

mark u as processed

}

Vertex	Distance	Predecessor	Processed
s	0	—	✓
w	1	s	✓
x	2	w	✓
u	<del>2</del> 3	<del>s</del> x	✓
v	4	u	✓
t	5	v	✓

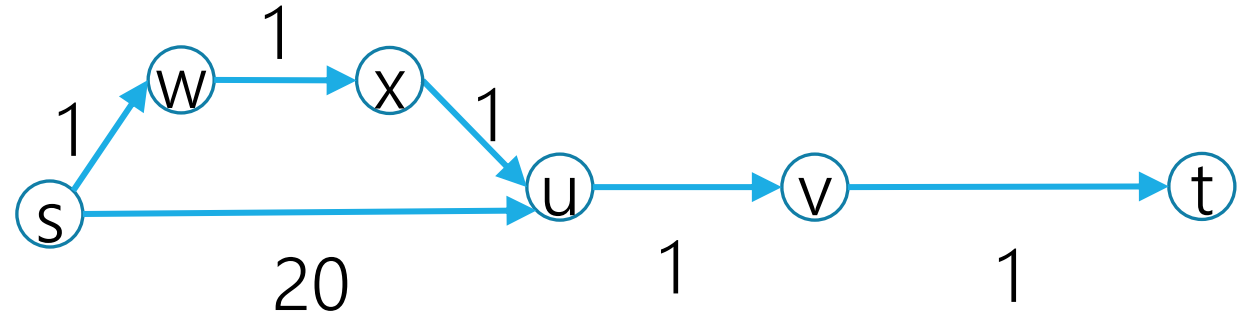




# Dijkstra's Algorithm

```
Dijkstra(Graph G, Vertex source)
  initialize distances to  $\infty$ 
  mark source as distance 0
  mark all vertices unprocessed
  while(there are unprocessed vertices){
    let u be the closest unprocessed vertex
    foreach(edge (u,v) leaving u){
      if( $u.\text{dist} + \text{weight}(u,v) < v.\text{dist}$ ){
         $v.\text{dist} = u.\text{dist} + \text{weight}(u,v)$ 
         $v.\text{predecessor} = u$ 
      }
    }
    mark u as processed
  }
```

Vertex	Distance	Predecessor	Processed
s	0	--	Yes
w	1	s	Yes
x	2	w	Yes
u	<del>2</del> 3	<del>s</del> x	Yes
v	4	u	Yes
t	5	v	Yes



# Implementation Details

One of those lines of pseudocode was a little sketchy

```
> let u be the closest unprocessed vertex
```



What ADT have we talked about that might work here?

Minimum Priority Queues!

# Making Minimum Priority Queues Work

They won't quite work "out of the box".

We need the "updatePriority" method, which means we need the KeyToIndex dictionary (like Ex 2).

- We'll ignore the updates to the dictionary, but it'll be in that method.

## Min Priority Queue ADT

### state

Set of comparable values

- Ordered by "priority"

### behavior

**peek()** – find the element with the smallest priority

**insert(value)** – add new element to collection

**removeMin()** – returns and removes element with the smallest priority

**DecreaseKey(e, p)** – decreases priority of element e down to p.

# Running Time Analysis

$$\underline{\underline{O(n \log n + m \log n)}}$$

Dijkstra(Graph G, Vertex source)

initialize distances to  $\infty$ , source.dist to 0

mark all vertices unprocessed

initialize MPQ as a Min Priority Queue

add source at priority 0

while(MPQ is not empty){

**u = MPQ.removeMin()**

foreach(edge (u,v) leaving u){

if(u.dist+weight(u,v) < v.dist){

if(v.dist ==  $\infty$ ) //if v not in MPQ

**MPQ.insert(v, u.dist+weight(u,v))**

else

**MPQ.decreaseKey(v, u.dist+weight(u,v))**

v.dist = u.dist+weight(u,v)

v.predecessor = u

}

}

mark u as processed

}

$$O(\log n)$$

$$O(\log n)$$

$$O(\log n)$$

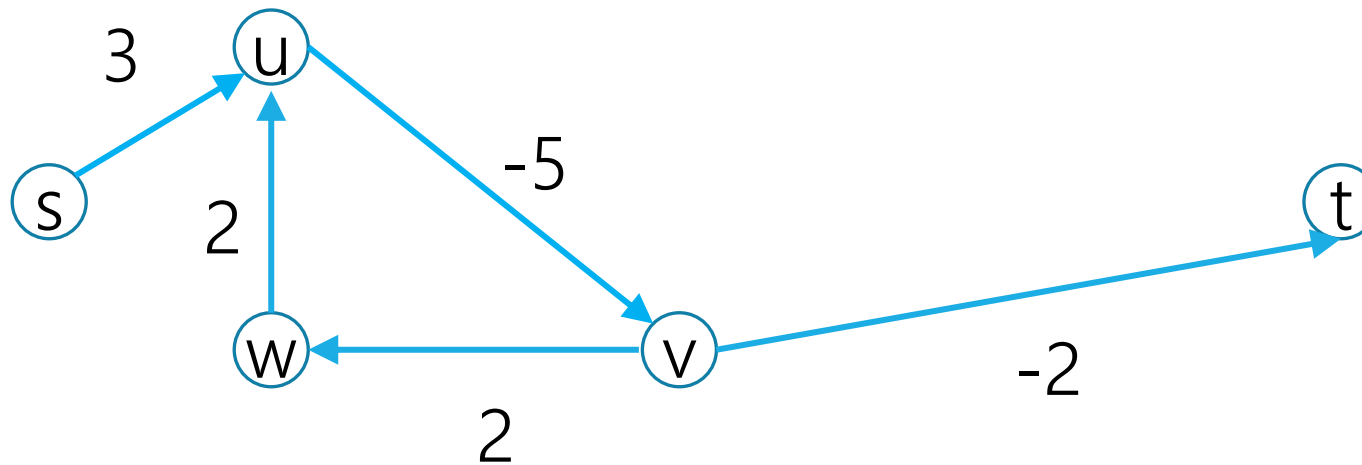


# Negative Edge Weights



# Negative Edge Weights

What's the shortest way to get from s to t?



s, u,v,w, u,v,w, u,v,w, ...

There is no shortest way. You can always go around u,v,w once more.

If there's a **negative weight cycle** shortest paths are **undefined**.

Undefined means "there is no correct answer" (or " $-\infty$  is the closest thing to a correct answer")

# Negative Edge Weights

If there are negative edge weights, but no negative weight cycle, shortest paths are still defined.

For today we'll assume all of the weights are positive

- For GoogleMaps that definitely makes sense.
- Sometimes negative weights make sense.
- Dijkstra's algorithm doesn't work for those graphs**
- There are other algorithms that do work (ask Robbie later)



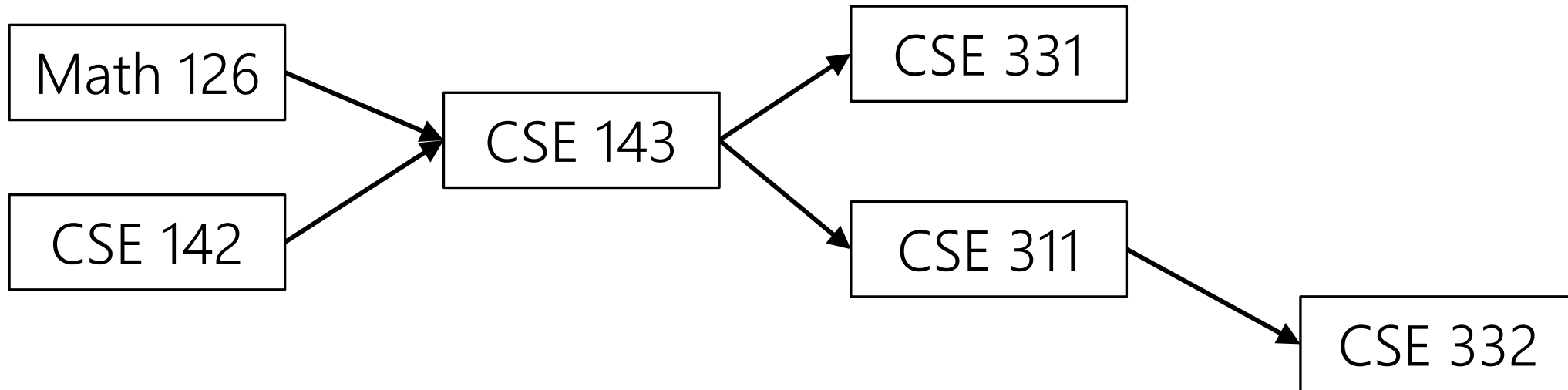


## One more Graph Algorithm

---

# Ordering Dependencies

Today's next problem: Given a bunch of courses with prerequisites, find an order to take the courses in.



# Ordering Dependencies

Given a directed graph  $G$ , where we have an edge from  $u$  to  $v$  if  $u$  must happen before  $v$ .

Can we find an order that **respects dependencies**?

## Topological Sort (aka Topological Ordering)

**Given:** a directed graph  $G$

**Find:** an ordering of the vertices so all edges go from left to right.

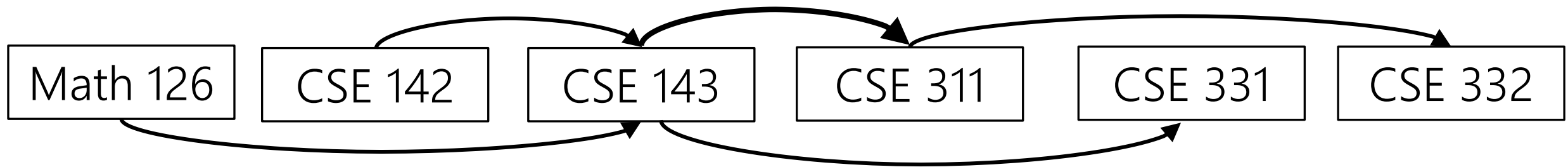
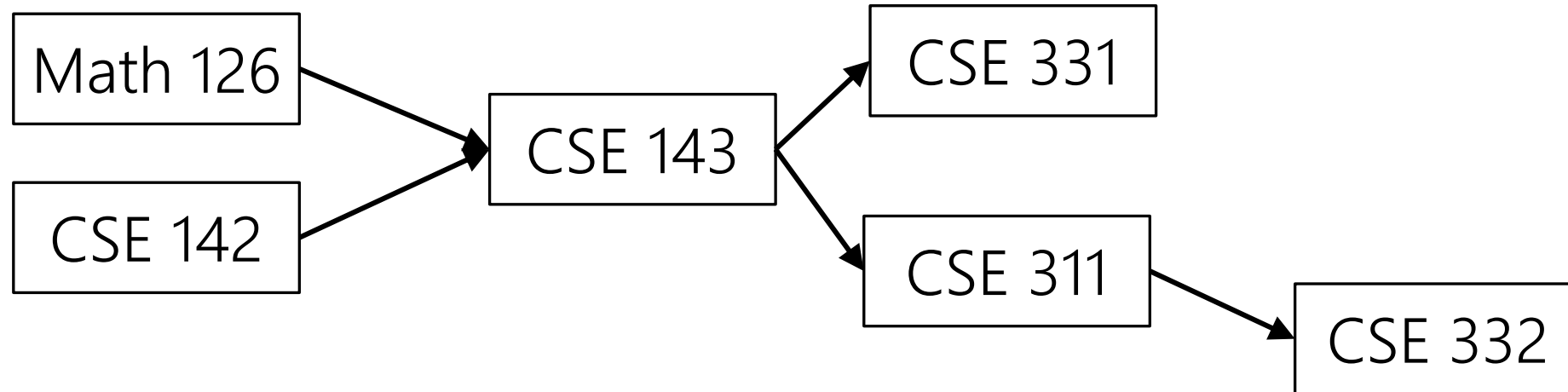
Uses:

Compiling multiple files

Graduating.

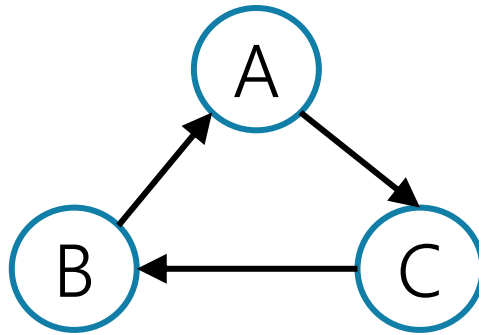
# Topological Ordering

A course prerequisite chart and a possible topological ordering.



# Can we always order a graph?

Can you topologically order this graph?



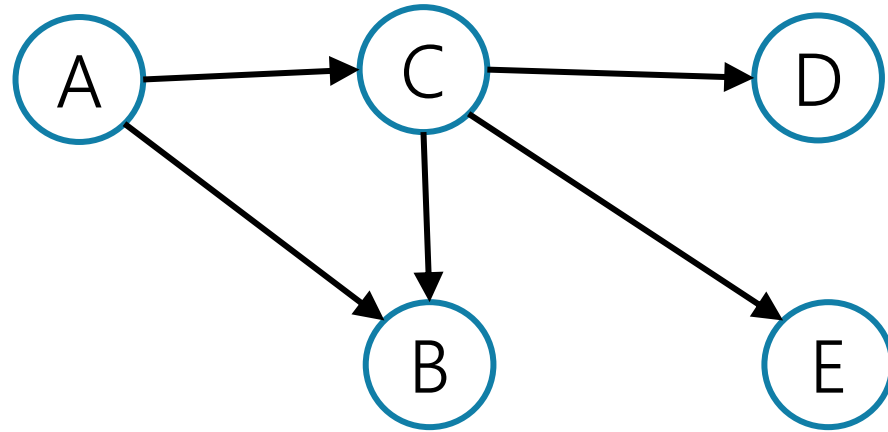
## Directed Acyclic Graph (DAG)

A directed graph without any cycles.

A graph has a topological ordering if and only if it is a DAG.

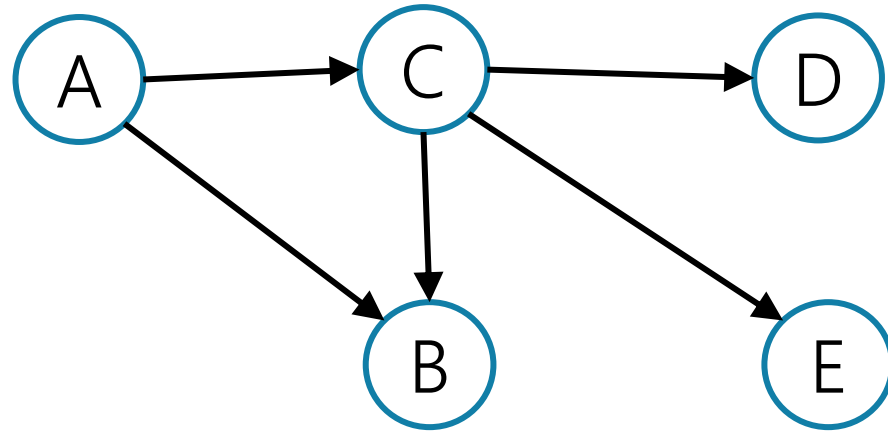
# Ordering a DAG

Does this graph have a topological ordering? If so find one.



# Ordering a DAG

Does this graph have a topological ordering? If so find one.



If a vertex doesn't have any edges going into it, we can add it to the ordering.

More generally, if the only incoming edges are from vertices already in the ordering, it's safe to add.



# How Do We Find a Topological Ordering?

```
TopologicalSort(Graph G, Vertex source)
    count how many incoming edges each vertex has
    Collection toProcess = new Collection()
    foreach(Vertex v in G){
        if(v.edgesRemaining == 0)
            toProcess.insert(v)
    }
    topOrder = new List()
    while(toProcess is not empty){
        u = toProcess.remove()
        topOrder.insert(u)
        foreach(edge (u,v) leaving u){
            v.edgesRemaining--
            if(v.edgesRemaining == 0)
                toProcess.insert(v)
        }
    }
}
```

# What's the running time?

```
TopologicalSort(Graph G, Vertex source)
    count how many incoming edges each vertex has
    Collection toProcess = new Collection()
    foreach(Vertex v in G){
        if(v.edgesRemaining == 0)
            toProcess.insert(v)
    }
    topOrder = new List()
    while(toProcess is not empty){
        u = toProcess.remove()
        topOrder.insert(u)
        foreach(edge (u,v) leaving u){
            v.edgesRemaining--
            if(v.edgesRemaining == 0)
                toProcess.insert(v)
        }
    }
}
```

Running Time:  $O(|V| + |E|)$

# Finding a Topological Ordering

Instead of counting incoming edges, you can actually modify DFS to find you one (think about why).

But the “count incoming edges” is a bit easier to understand (for me 😊 )



## More Graph Applications

---

# Another Application of Shortest Paths

Shortest path algorithms are obviously useful for GoogleMaps.

The wonderful thing about graphs is they can encode **arbitrary** relationships among objects.

I don't care if you remember all the details.

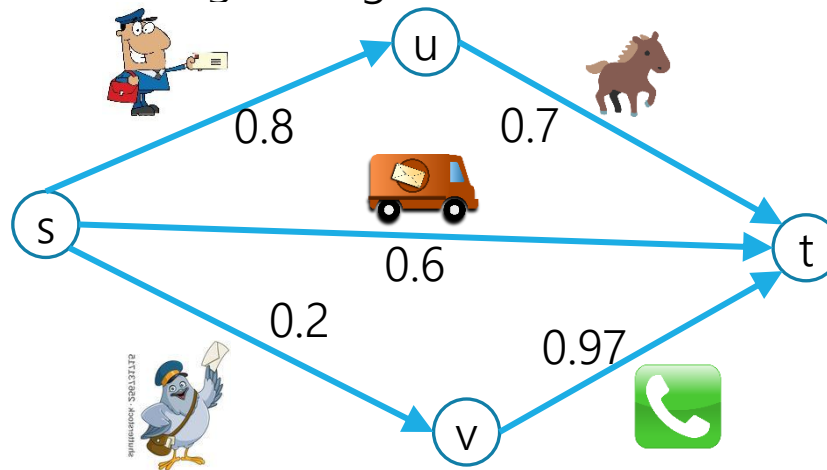
I just want you to see that these algorithms have non-obvious applications.

# Another Application of Shortest Paths

I have a message I need to get from point s to point t.

But the connections are unreliable.

What path should I send the message along so it has the best chance of arriving?



## Maximum Probability Path

**Given:** a directed graph  $G$ , where each edge weight is the probability of successfully transmitting a message across that edge

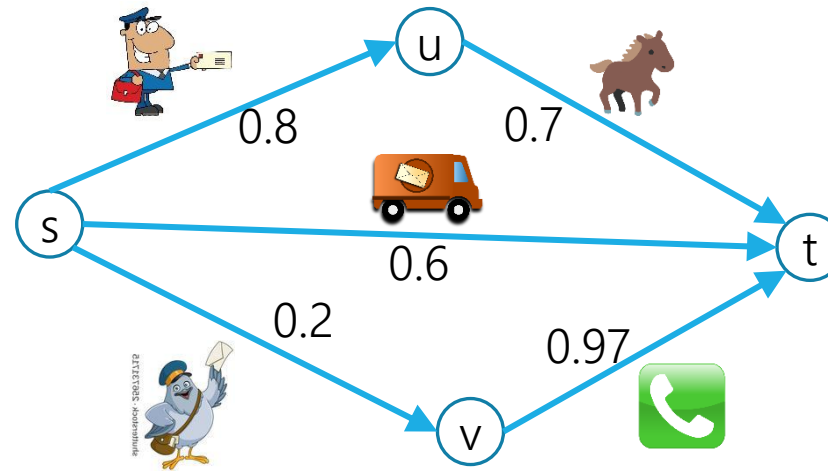
**Find:** the path from  $s$  to  $t$  with maximum probability of message transmission

# Another Application of Shortest Paths

Let each edge's weight be the probability a message is sent successfully across the edge.

What's the probability we get our message all the way across a path?

- It's the product of the edge weights.



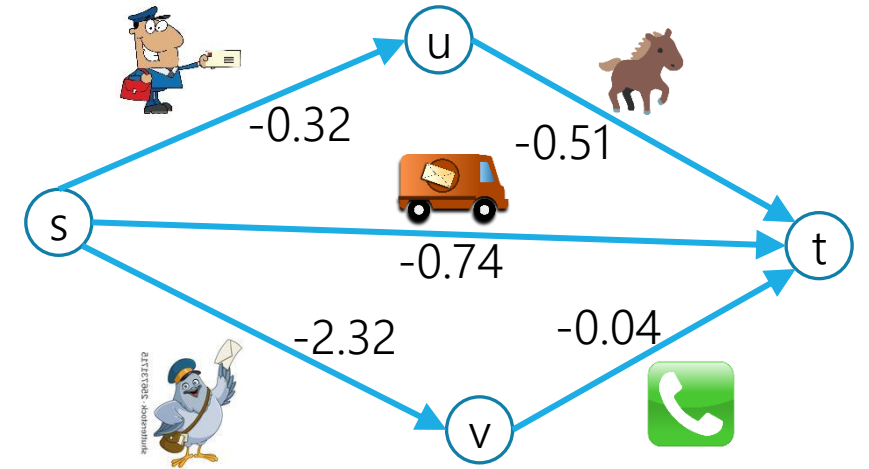
We only know how to handle sums of edge weights.

Is there a way to turn products into sums?

$$\log(ab) = \log a + \log b$$



# Another Application of Shortest Paths



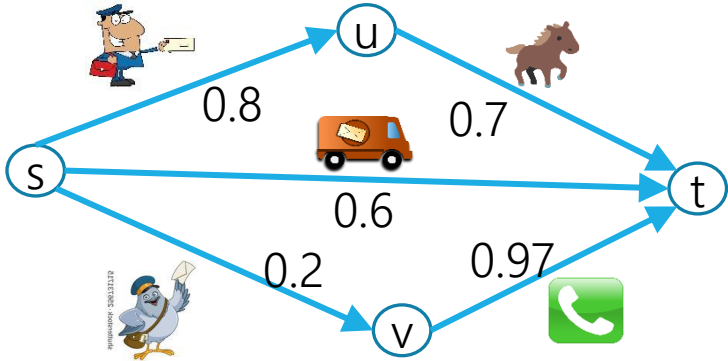
We've still got two problems.

1. When we take logs, our edge weights become negative.
2. We want the *maximum* probability of success, but that's the longest path not the shortest one.

Multiplying all edge weights by negative one fixes both problems at once!

We **reduced** the maximum probability path problem to a shortest path problem by taking  $-\log()$  of each edge weight.

# Maximum Probability Path Reduction



Transform Input

Weighted Shortest Paths

Transform Output