1

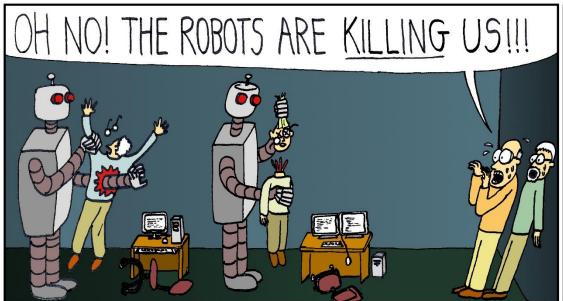
Lecture #16 - That's all folks!

- Intro to Graphs
- Graph Traversals
 - Depth-first
 - Breadth-first
- · Dijkstra's Algorithm

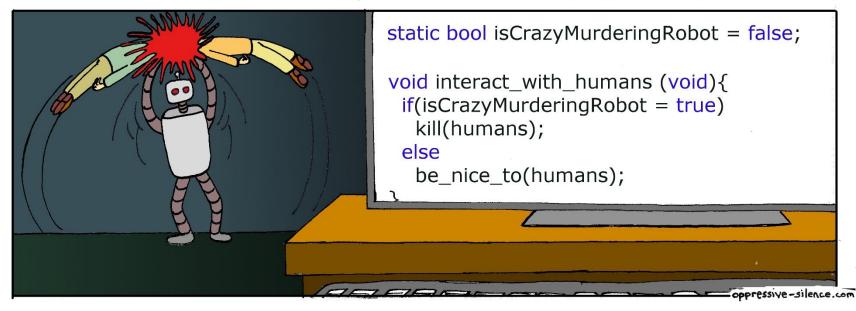
Final Exam: Saturday, March 16th 11:30am-2:30pm

Final Exam Location: TBA

Graphs







Graphs Why should you care?

Facebook? Duh!

Not good enough? Google+?

Computer Animation?

Google Maps?

The Internet?

So pay attention!

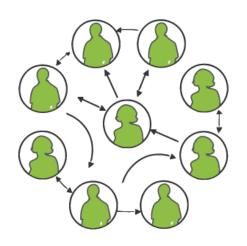


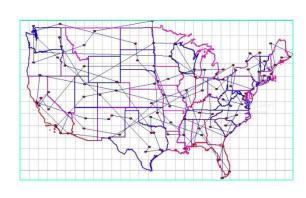
Introduction to Graphs

A graph is an ADT that stores a set of entities and also keeps track of the relationships between all of them.

Examples of Entities
People
Cities
Web pages

Examples of Relationships
Joe is friends with Linda
LA is 3000 miles from NYC
ucla.edu links to awesome.com







Introduction to Graphs

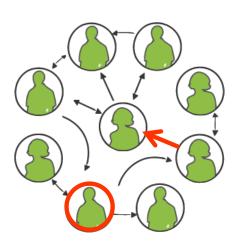
Each graph holds two types of items:

Vertices (aka Nodes):

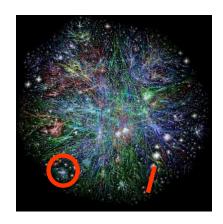
A vertex might represent a person, a city or a web page.

Edges (aka Arcs):

An edge simply connects two* vertices to each other.







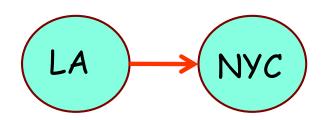
^{*} Technically, an edge could connect a vertex to itself!

Directed vs. Undirected Graphs

There are two major types of graphs...

Directed Graph

In a directed graph, an edge goes from one vertex to another in a specific direction.

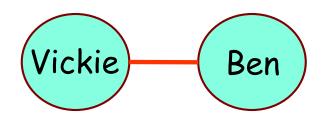


For example, above we have an edge that goes from the LA vertex to NYC vertex, but not the other way around.

(e.g., there may be a flight from LA to NYC but not the other way around)

Undirected Graph

In an undirected graph, all edges are bi-directional. You can go either way along any edge.



For example, Vickie and Ben are mutual friends on FaceBook.

(It would be kinda creepy if Vickie liked Ben, but not visa-versa)

The easiest way to represent a graph is with a double-dimensional array.

The size of both dimensions of the array is equal to the number of vertices in the graph.

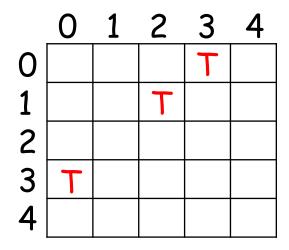
```
bool graph[5][5];
```

Each element in the array indicates whether or not there is an edge between vertex i and vertex j.

Each element in the array indicates whether or not there is an edge between vertex i and vertex j.

```
bool graph[5][5];

// edge from vertex 0 to vertex 3
graph[0][3] = true;
graph[1][2] = true;
graph[3][0] = true;
```

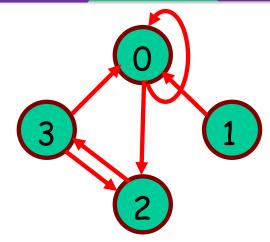


As you can see, when we set array[i][j] to true, it represents a directed edge from vertex i to vertex j.

This is called an adjacency matrix.

Exercise: What does the following directed graph look like?

Nodes	0	1	2	3
0	True	False	True	False
1	True	False	False	False
2	False	False	False	True
3	True	False	True	False



Question:

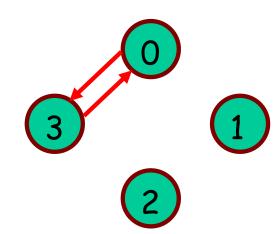
How do you represent an undirected graph with an adjacency matrix?

It's easy!

To bi-directionally connect vertices i and j, simply set array[i][j] to true and set array[j][i] to true as well!

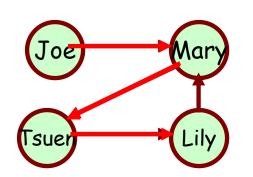
```
graph[0][3] = true;
graph[3][0] = true;
```

Nodes	0	1	2	3
0				True
1				
2				
3	True			



¹¹An Interesting Property of Adjacency Matrices

Consider the following graph: And it's associated A.M.:



Neato effect: If you multiply the matrix by itself something cool happens!

	Joe M	an'	sten	1/1
Joe	0	1	0	0
Mary	0	0	1	0
Tsuen	0	0	0	1
Lily	0	1	0	0

0	1	0	0
0	0	1	0
0	0	0	1
0	1	0	0

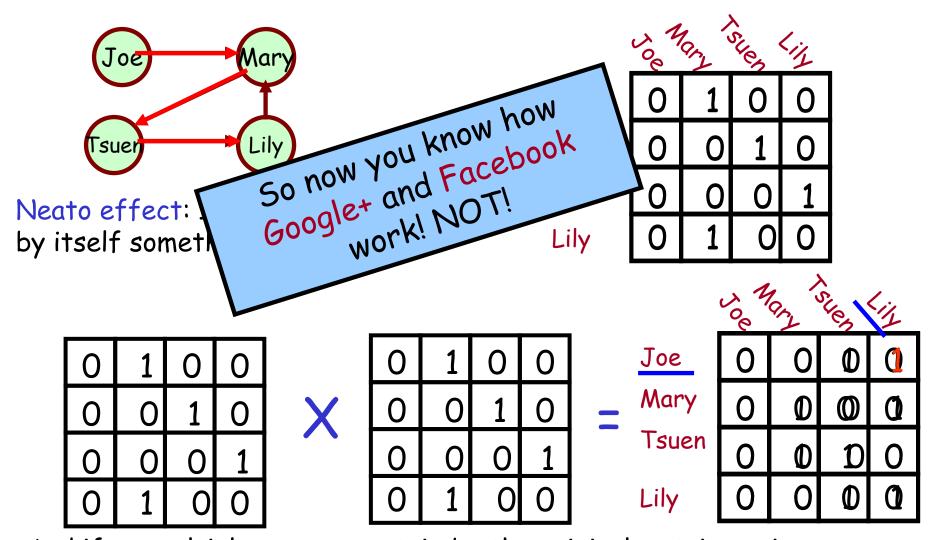
0	1	0	0
0	0	1	0
0	0	0	1
0	1	0	0

	Joe M	and I	Ster	1/1
Joe	0	0	1	0
Mary	0	0	0	1
Tsuen	0	1	0	0
Lily	0	0	1	0

The resulting matrix shows us which vertices are exactly two edges apart.

¹²An Interesting Property of Adjacency Matrices

Consider the following graph: And it's associated A.M.:



And if we multiply our new matrix by the original matrix again, we'll get all vertices that are exactly 3 edges apart!

Another Way to Represent a Graph

Question:

How else can we represent a graph (without a 2D array)?

Answer:

A directed graph of n vertices can be represented by an array of n linked lists. This is called an adjacency list.

list<int> graph[n];

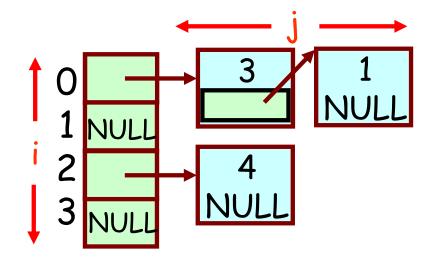
If we add a number j, to list number i (e.g., to graph[i]), this means that there is an edge from vertex i to vertex j.

The Adjacency List

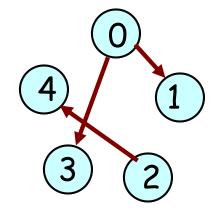
If we add a number j, to list number i (e.g., to graph[i]), this means that there is an edge from vertex i to vertex j

```
list<int> graph[4];

// edge from node 0 to node 3
graph[0].push_back(3);
graph[2].push_back(4);
graph[0].push_back(1);
```



So for each entry j, in list i, this means that there is an edge from vertex i to vertex j.



Which Representation Should You Use?

When should you use an adjacency matrix vs. an adjacency list?

Scenario #1:

We've got 10,000,000 users who have relationships with each other - typically each person is friends with just a few hundred other people.

What would you do?

Option A: Store the graph in a 10 million by 10 million array? (That's 100 trillion cells)

Option B: Store your graph in an array holding 10 million linked lists, each holding roughly 500 items?

(That's only 5 billion pieces of data)

Which Representation Should You Use?

When should you use an adjacency matrix vs. an adjacency list?

Scenario #2:

We've got 1,000 cities, with airlines offering flights from every city to almost every other city.

What would you do?

Option A: Store the graph in a 1000 by 1000 array?

(That's 1 million cells)

Option B: Store your graph in an array holding 1000 linked lists, each holding roughly 1000 items?

(That's also 1 million pieces of data, but it's more complex)

Which Representation Should You Use?

When should you use an adjacency matrix vs. an adjacency list?

Use an adjacency matrix if you have lots of edges between vertices but few vertices (< 10,000 vertices).

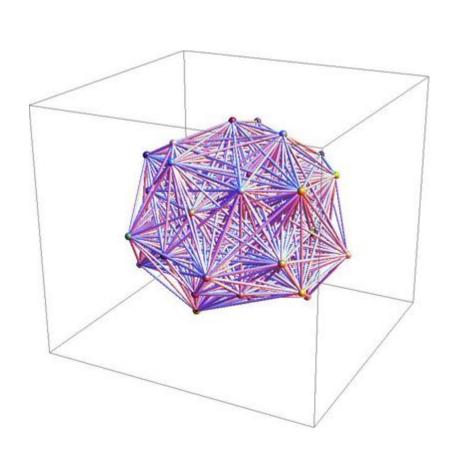
Use an adjacency list if you have few edges between vertices and lots of vertices (> 10,000 verices).

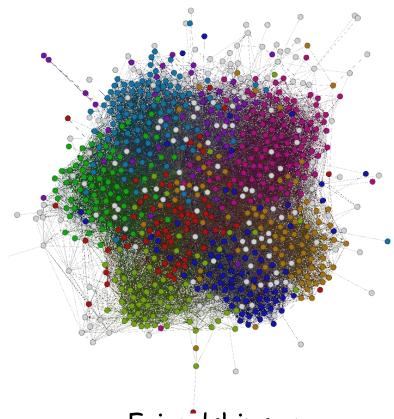
A graph that has many edges between the vertices is called a "dense graph".

A graph that has few edges between the vertices is called a "sparse graph".

Let's see examples of both...

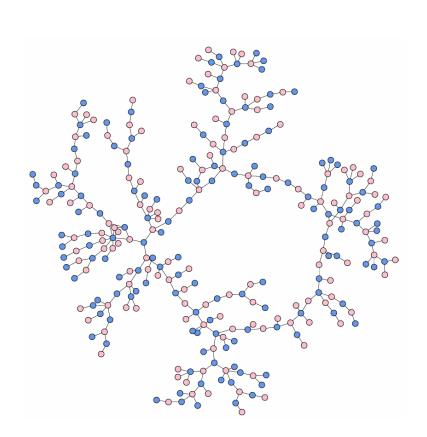
Dense(r) Graphs

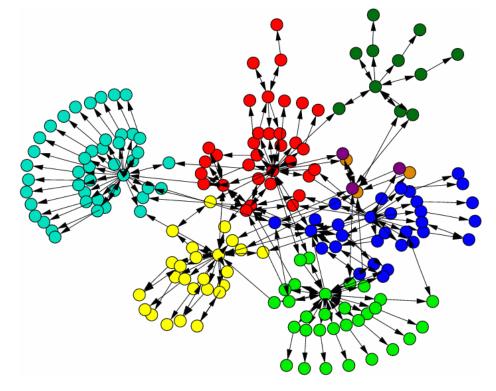




Friendships on Facebook for people from Caltech.

Sparse Graphs





(Intra-website links)

(High-school dating habits)

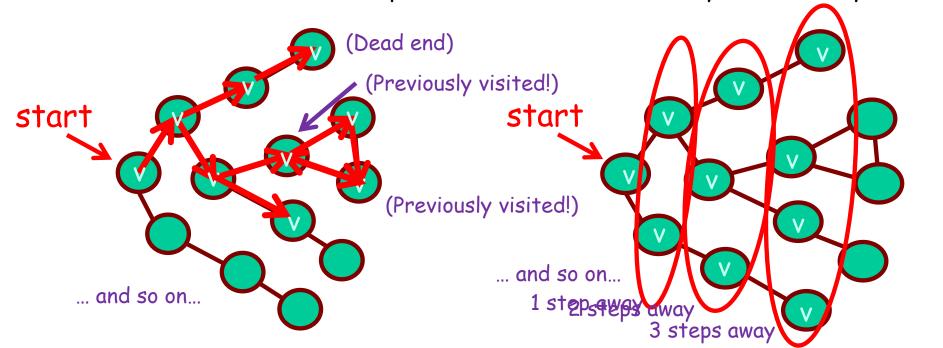
Graph Traversals

We can traverse graphs just like we traverse binary trees!

There are two types of graph traversals: Depth-first and Breadth-first

A Depth-first Traversal keeps moving forward until it hits a dead end or a previously-visited vertex... then it backtrack sand tries another path

A Breadth-first Traversal explores the graph in growing concentric circles, exploring all vertices 1 away from the start, then 2 away, then 3 away, etc.

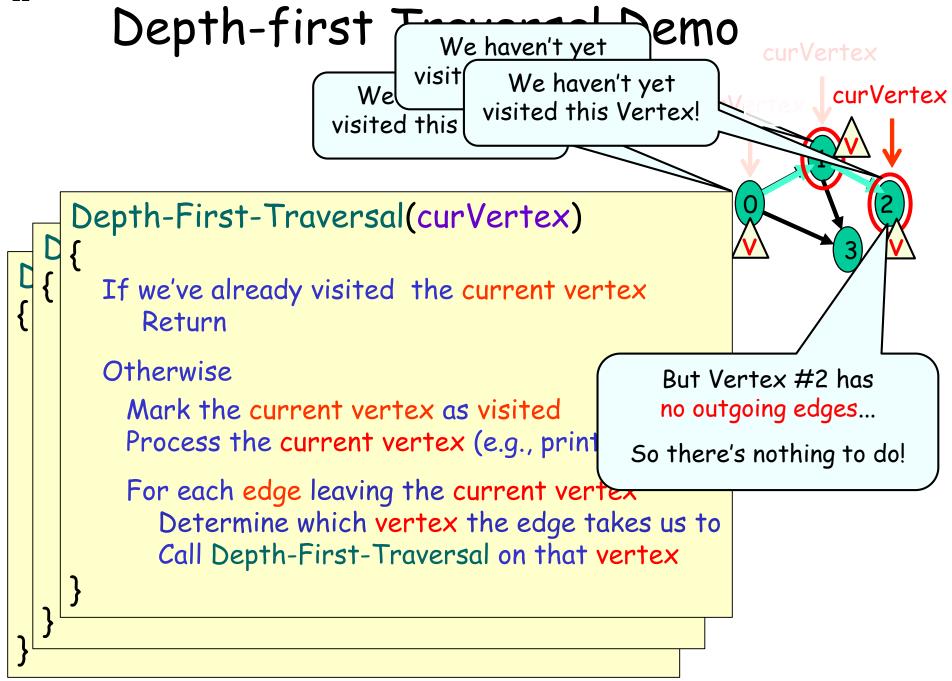


Depth-first Traversals

Let's learn the Depth-first Traversal algorithm first:

```
Depth-First-Traversal(curVertex)
  If we've already visited the current vertex
     Return
  Otherwise
     Mark the current vertex as visited
     Process the current vertex (e.g., print it out)
     For each edge leaving the current vertex
        Determine which vertex the edge takes us to
       Call Depth-First-Traversal on that vertex
```

(Notice that it's recursive!)



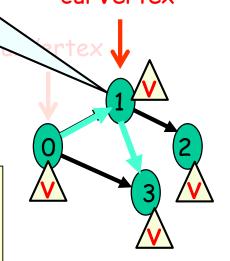
Depth-first Traversal Demo We haven't yet curV visited this Vertex! Depth-First-Traversal(curVertex) If we've already visited the current vertex Return rVertex Alas, Vertex #3 has Otherwise no outgoing edges... sed vertex 0! Mark the current vertex sed vertex 1! So there's nothing to do! Process the current verte sed vertex 2! rocessed vertex 3! For each edge leaving the current vertex Determine which vertex the edge takes us to Call Depth-First-Traversal on that vertex

Depth-firs

Vertex #1 has no MORE outgoing edges...

So there's nothing to do!

mo sun/ant



Depth-First-Traversal(curVertex)

If we've already visited the current vertex Return

Otherwise

Mark the current vertex as visited Process the current vertex (e.g., print it out)

For each edge leaving the current vertex

Determine which vertex the edge takes us to

Call Depth-First-Traversal on that vertex

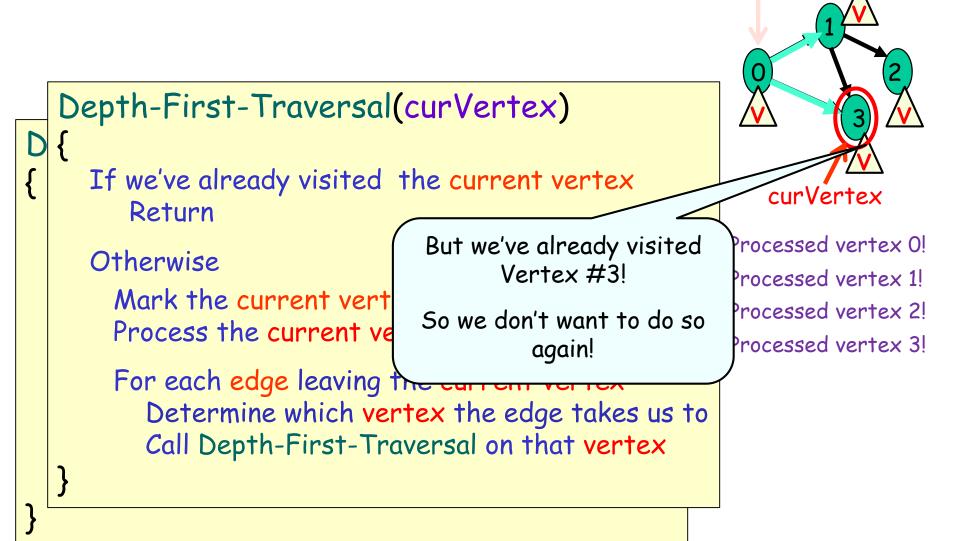
Processed vertex 0!

Processed vertex 1!

Processed vertex 2!

Processed vertex 3!

Depth-first Traversal Demo



Depth-first Travance Demo

no MORE outgoing edges...

So there's nothing to do!

curVertex

Depth-First-Traversal(curVertex)

If we've already visited the current vertex Return

Otherwise

Mark the current vertex as visited Process the current vertex (e.g., print it out)

For each edge leaving the current vertex

Determine which vertex the edge takes us to

Call Depth-First-Traversal on that vertex

Processed vertex 0!

Processed vertex 1!

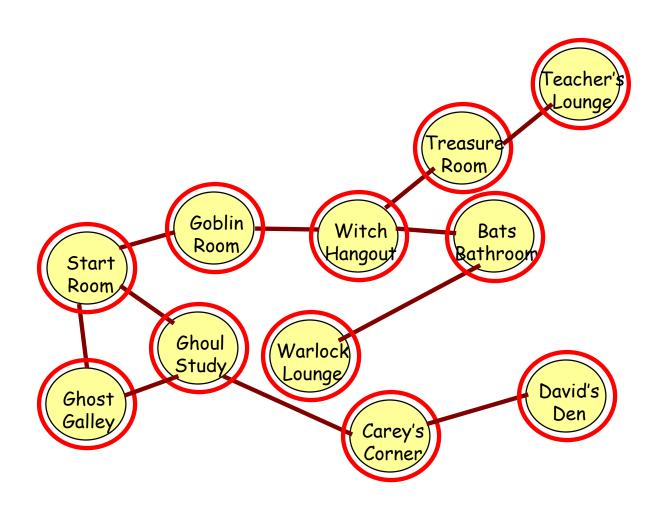
Processed vertex 2!

Processed vertex 3!

And we're done!

Depth-first Traversal Challenge

What does a Depth-first Traversal look like on this graph?



Implementing Depth-first Traversal w/Stack!

You can also implement your Depth-first Traversal with a stack if you like! (What's not to like???)

```
Depth-First-Search-With-Stack(start_room)

Push start_room on the stack

While the stack is not empty

Pop the top item off the stack and put it in variable c

If c hasn't been visited yet

Drop a breadcrumb (we've visited the current room)

For each door d leaving the room

If the room r behind door d hasn't been visited

Push r onto the stack.
```

Basically, the stack allows you to simulate recursion...

Or does the recursion allow you to simulate a stack?

Hmmmmmmm!

Breadth-first Graph Traversal

Idea:

Process all of the vertices that are 1 edge away from the start vertex,

then process all vertices that are two edges away,

then process all vertices that are three edges away,

etc...

Question:

What data structure could we use to implement this?

Answer:

Not a P, but a?

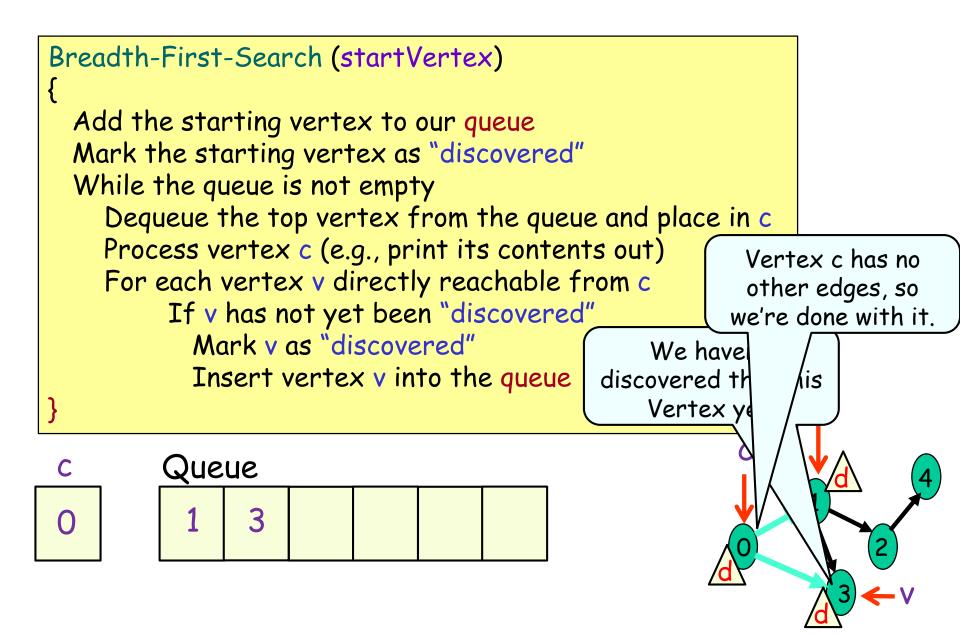
Breadth-first Graph Traversal

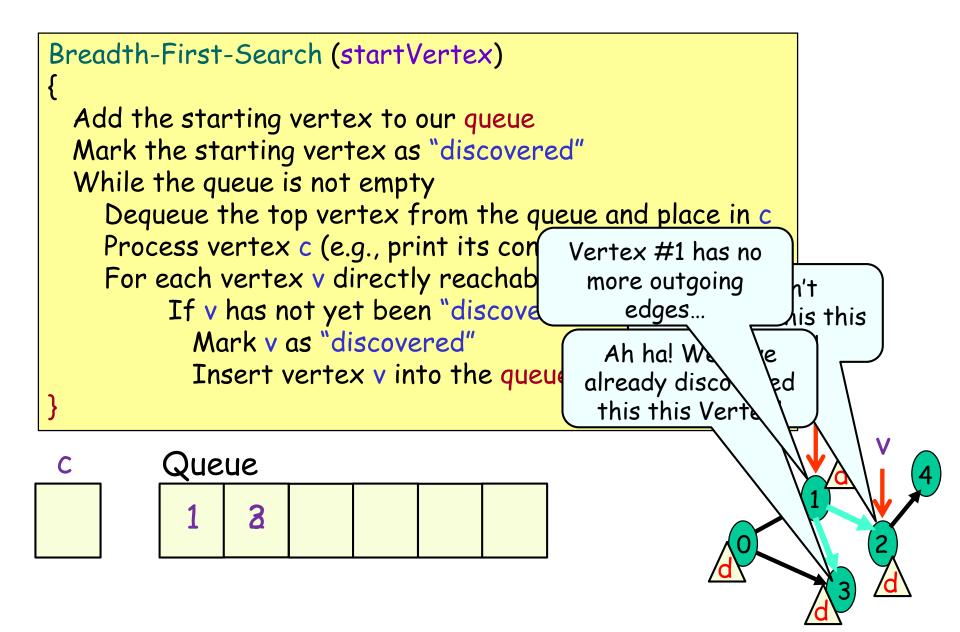
```
Breadth-First-Search (startVertex)
 Add the starting vertex to our queue
 Mark the starting vertex as "discovered"
 While the queue is not empty
    Dequeue the top vertex from the queue and place in c
    Process vertex c (e.g., print its contents out)
    For each vertex v directly reachable from c
        If v has not yet been "discovered"
           Mark v as "discovered"
           Insert vertex v into the queue
```

Hmmm. Does this algorithm look familiar?

It's a-maze-ingly similar to our queue-based maze-solving algorithm!!!

```
Breadth-First-Search (startVertex)
 Add the starting vertex to our queue
 Mark the starting vertex as "discovered"
 While the queue is not empty
    Dequeue the top vertex from the queue and place in c
    Process vertex c (e.g., print its contents out)
                                                        We haven't
    For each vertex v directly reachable from c
                                                    discovered this this
         If v has not yet been "discovered"
                                                        Vertex yet!
           Mark v as "discovered"
           Insert vertex v into the queue
                                                  start Vertex
         Queue
```





Processed vertex 0!
Processed vertex 1!

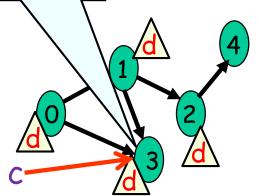
Processed vertex 3!

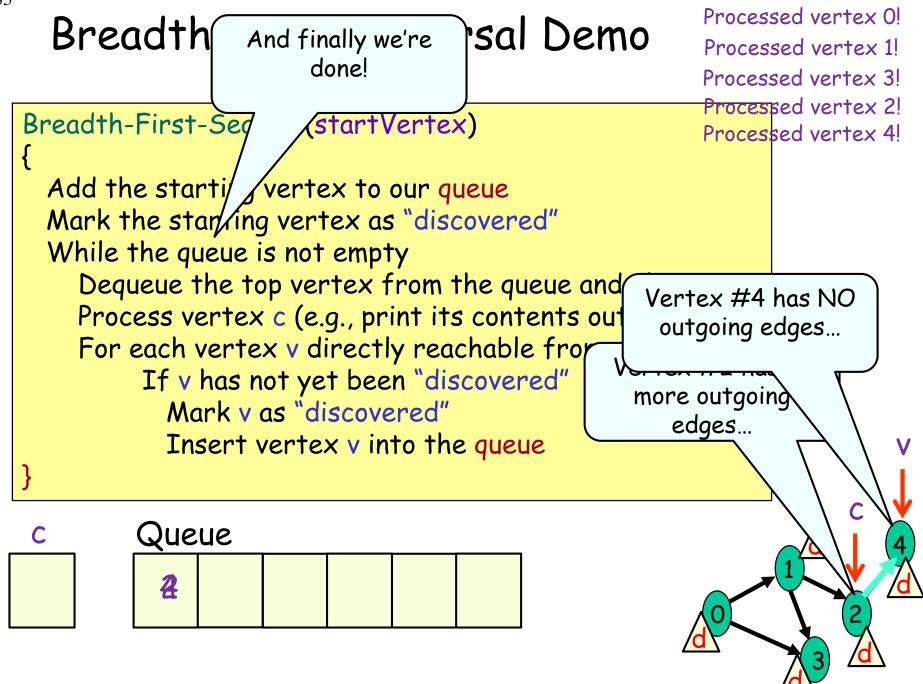
```
Breadth-First-Search (startVertex)
 Add the starting vertex to our queue
 Mark the starting vertex as "discovered"
 While the queue is not empty
    Dequeue the top vertex from the queue and place in c
    Process vertex c (e.g., print its contents out)
    For each vertex v directly reachable from c
         If v has not yet been "discovered"
           Mark v as "discovered"
                                          Vertex #3 has NO
           Insert vertex v into the queu
                                           outgoing edges at
                                           all! So we're done.
```

С

Queue

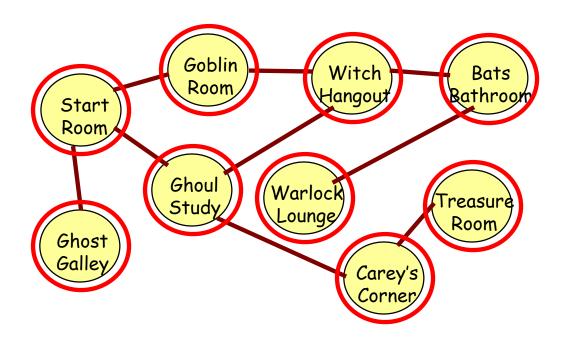
3 2





Breadth-first Traversal Challenge

What does a Breadth-first Traversal look like on this graph?

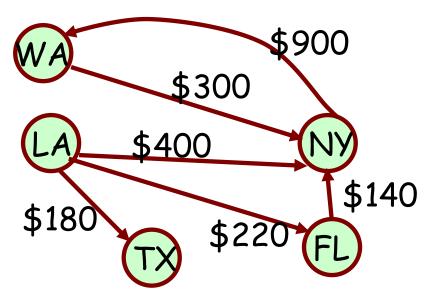


Graphs With Weighted Edges

What does it mean for a graph to have weighted edges?

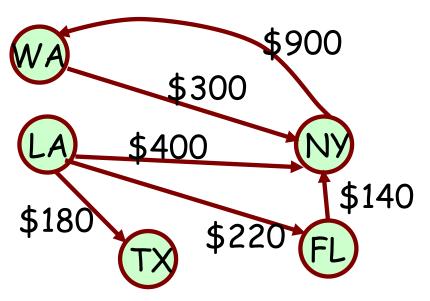
Definition: Each edge connecting vertex u with vertex v has a weight or cost associated with it.

Question: Why would we want to have weighted edges?



Graphs With Weighted Edges

Definition: The weight of a path from vertex u to vertex v is the sum of the weights of the edges between the two vertices.



Question: What's the cost of traveling from LA to NY to WA?

Question: What's the shortest path from LA to WA?

Definition: The shortest path between two vertices is the path with the lowest total cost of edges between the two vertices. (The shortest path is a set of vertices)

Finding the Shortest Path

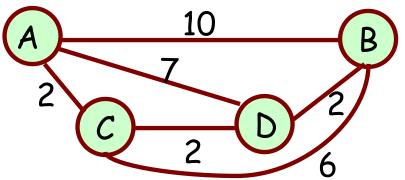
Question: How can we find the shortest path between any two nodes in a graph?

Answer: Dijkstra's Algorithm (the dorm guy?)

Dijkstra's Algorithm:

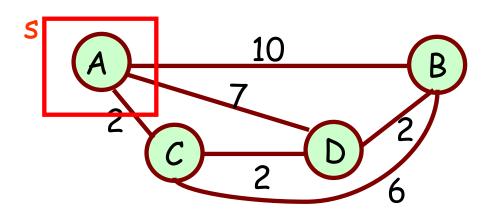
the length of

This algorithm determines the shortest path (i.e. set of vertices) from a start vertex s to all other vertices in the graph.



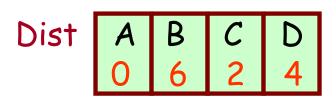
So Dijkstra(A) would give us a value of 6 for A to B, a value of 2 for A to C, and 4 for A to D.

Input: A graph G, and a starting vertex s



G must not have any negative edge values.

Output: An array called Dist of optimal distances from s to every other node in the graph.



Dijkstra's Algorithm: Basic Idea

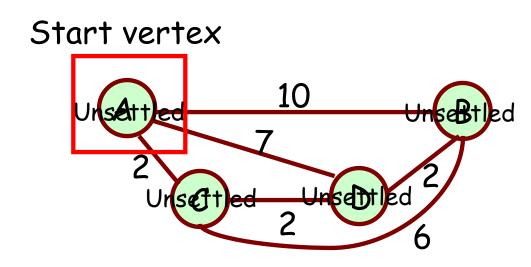
Dijkstra's algorithm splits vertices in two distinct sets: the set of unsettled vertices and the set of settled vertices.

Unsettled vertex: A vertex v is unsettled if we don't know the optimal distance to it from the starting vertex s.

Settled vertex: A vertex v is settled if we have learned the optimal distance to it from the starting vertex s.

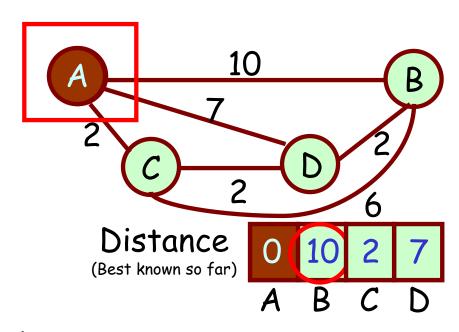
Initially all vertices are unsettled.

The algorithm ends once all vertices are in the settled set.



Assume that all vertices are infinitely far away to start...

Since we start at vertex A, we know it's the closest vertex to us! How far is it? Zero steps away! We can settle it immediately!



Now let's see which unsettled vertices we can reach directly from A.

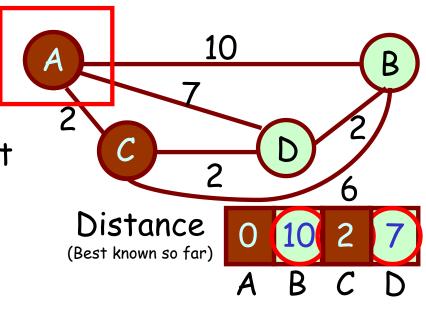
- B is 10 units away.
- · C is 2 units away.
- D is 7 units away.

And going directly from A to D is only 7 units away, which is less than infinity, so I'll update this entry too...

Ok, so now we know the costs to travel to all vertices directly reachable from A.

Which unsettled vertex is closest to A?

Right! C is closest to A.



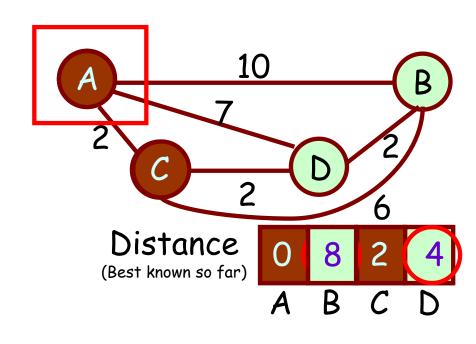
If we go directly to C ($A \rightarrow C$), it costs us 2 units. Is there any possible way I can travel to C cheaper by going through B or through D first?

So I know that if I travel directly from A to C, at a cost of 2 units, that's the *fastest* possible route. Therefore we can settle C at 2 units.

At this point, we know the shortest path from A to C. Now let's see if we can travel through C to reach one of our other unsettled vertices faster.

Ok, which unsettled vertices can be reached directly from C?

- B is 6 units away.
- D is 2 units away.



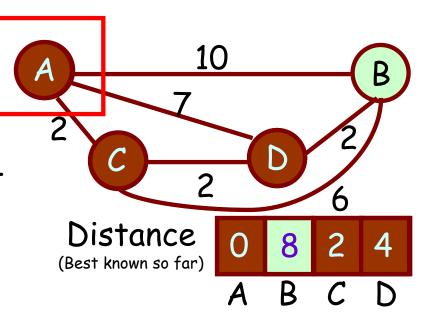
Let's do D next. We know we can get from A to C in 2 units, and we can directly go from C to D in 2 units, so we can reach D in just 4 units!

Is our new distance to D better than our old one? Yup!! Let's update our table again!

Ok, so now we know the best cost to get to all unsettled vertices, assuming we travel through C.

Which unsettled vertex is closest to A now?

Right! D is closest.



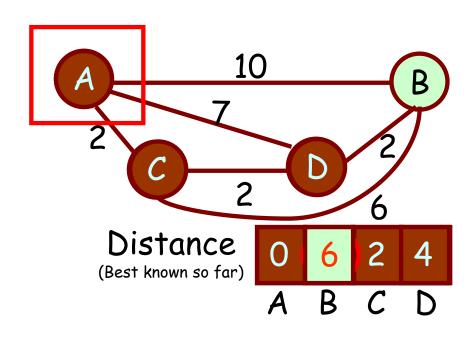
If we take the path $A \rightarrow C \rightarrow D$, it costs us 4 units. Is there any possible way I can travel to D cheaper by going through B first?

So I know that if I travel from $A \rightarrow C \rightarrow D$, at a cost of 4 units, that's the fastest possible route. Therefore we can settle D at 4 units.

At this point, we know the shortest path from A to D. Now let's see if we can travel through D to reach one of our other unsettled vertices faster.

Ok, which unsettled vertices can be reached directly from D?

• B is 2 units away.



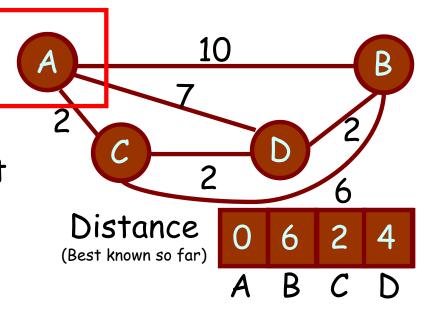
Let's check B. We know we can get from A to D in 4 units, and we can directly go from D to B in 2 units, so we can reach B in just 6 units!

Is our new distance to B better than our old one? You bet!! Let's update our table!

Ok, so now we know the best cost to get to all unsettled vertices, assuming we travel through D.

Which unsettled vertex is closest to A now?

Right! B is closest.



And now that all of our vertices are settled, we are guaranteed to have found the *minimum* travel distances to each of our vertices!

Dijkstra

And now I'll give you the more formal algorithm...



Born: 11 May 1930, Rotterdam, Netherlands

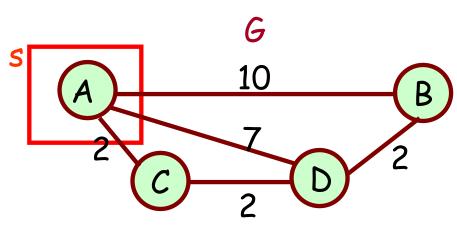
Died: 6 August 2002, Nuenen, Netherlands

Dijkstra's Algorithm uses 2 data structures:

1. An array called Dist that holds the the current best known cost to get from s to every other vertex in the graph.

For each vertex i, Dist[i] starts out with a value of:

- O for vertex s
- Infinity for all other vertices



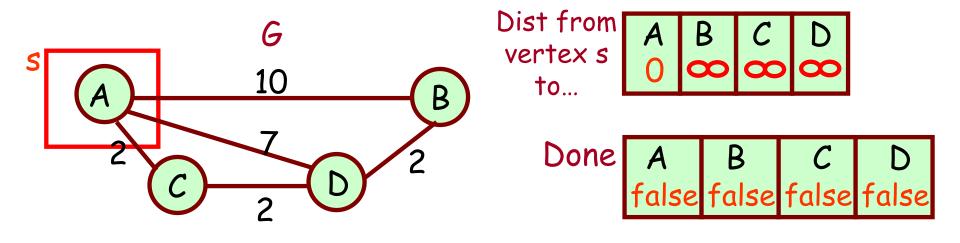
Dist from A B C D vertex s to...

Idea: We start at node A so we're 0 steps away from node A. We assume the other vertices are infinitely far away from A.

Dijkstra's Algorithm uses 2 data structures:

2. An array called Done that holds true for each vertex that has been fully processed, and false otherwise.

For each vertex i, Done[i] starts out with a value of false.



While there are still unprocessed vertices:

Set u = the closest unprocessed vertex to the start vertex s

Mark vertex u as processed: Done[u] = true.

We now know how to reach u optimally from s

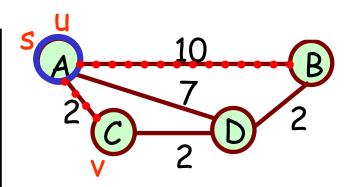
Loop through all unprocessed vertices:

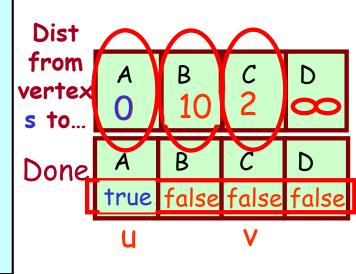
Set v = the next unprocessed vertex

If there's an edge from u to v then compare:

- a. the previously computed path from s to
 v (i.e. Dist[v]) OR
- b. the path from s to u, and then from u to v (I.e. Dist[u] + weight(u,v))

If the new cost is less than old cost then Set Dist[v] = Dist[u] + weight(u,v)





While there are still unprocessed vertices:

Set u = the closest unprocessed vertex to the start vertex s

Mark vertex u as processed: Done[u] = true.

We now know how to reach u optimally from s

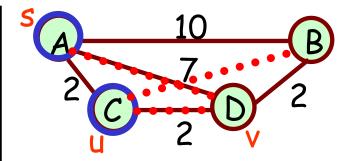
Loop through all unprocessed vertices:

Set v = the next unprocessed vertex

If there's an edge from u to v then compare:

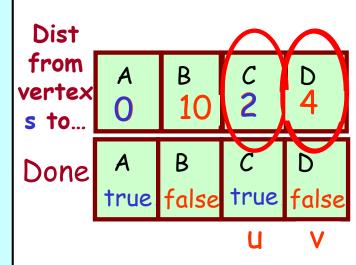
- a. the previously computed path from s to
 v (i.e. Dist[v]) OR
- b. the path from s to u, and then from u to v (I.e. Dist[u] + weight(u,v))

If the new cost is less than old cost then Set Dist[v] = Dist[u] + weight(u,v)



Previous cost: 7

New cost: 2 + 2 = 4



While there are still unprocessed vertices:

Set u = the closest unprocessed vertex to the start vertex s

Mark vertex u as processed: Done[u] = true.

We now know how to reach u optimally from s

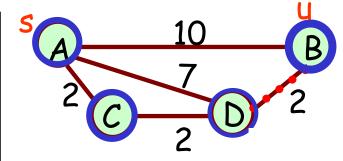
Loop through all unprocessed vertices:

Set v = the next unprocessed vertex

If there's an edge from u to v then compare:

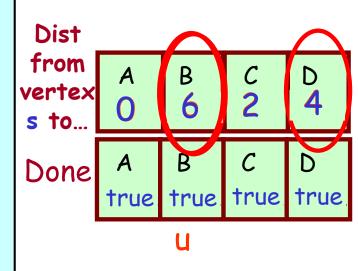
- a. the previously computed path from s tov (i.e. Dist[v]) OR
- b. the path from s to u, and then from u to v (I.e. Dist[u] + weight(u,v))

If the new cost is less than old cost then Set Dist[v] = Dist[u] + weight(u,v)



Previous cost: 10

New cost: 4 + 2 = 6



And we're done! The Dist array contains the results.