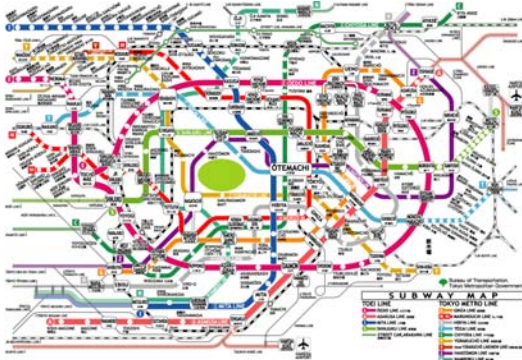


## More Graphs

Lecture 21  
CS211 – Fall 2005



## Announcements

- Upcoming talk
  - "The Many Careers of a Computer Scientist"
    - Or how a Computer Science degree empowers you to do much more than code
  - Dan Huttenlocher, Professor in the Department of Computer Science and Johnson Graduate School of Management
  - 5:00 PM, Wednesday, November 9th
  - Upton Lounge
  - FREE PIZZA!
- ACSU (Association of Computer Science Undergraduates)

## Prelim 2 Reminder

- Prelim 2
  - Tuesday, Nov 15, 7:30-9pm
  - One week from today!
  - Topics: all material through Nov 1
  - Does *not* include
    - Graphs
    - GUIs in Java
- Note that this week's Section meetings are last before the exam
- Exam conflicts
  - Email Kelly Patwell (ASAP)
- Prelim 2 Review Session
  - Sunday, Nov 13, 1:30-3:00pm, Kimball B11
  - See *Exams* on course website for more information
  - Individual appointments are available if you cannot attend the review session (email *one* TA to arrange appointment)
- Old exams are available for review on the course website

## Implementing Digraphs

- Adjacency Matrix
 

$g[u][v]$  is true iff there is an edge from  $u$  to  $v$

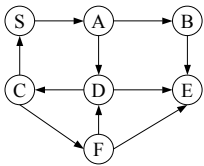
	0	1	2	3
0		T		T
1			T	
2	T			
3				
- Adjacency List
 

The list for  $u$  contains  $v$  iff there is an edge from  $u$  to  $v$

```

0 → [1, 3]
1 → [2]
2 → [0]
3 → []
            
```

## Shortest Paths for Unweighted Graphs



```

bfsDistance(s):
// s is the start vertex
// dist[v] is length of s-to-v path
// Initially dist[v] = ∞ for all v
dist[s] = 0;
Q.insert(s);

while (Q.nonempty()) {
    v = Q.get();
    for (each w adjacent to v) {
        if (dist[w] == ∞) {
            dist[w] = dist[v]+1;
            Q.insert(w);
        }
    }
}
    
```

## Analysis for bfsDistance

- How many times can a vertex be placed in the queue?
  - How much time for the for-loop?
    - Depends on representation
      - Adjacency Matrix:  $O(n)$
      - Adjacency List:  $O(m_v)$
  - Time:
    - $O(n^2)$  for adj matrix
    - $O(m+n)$  for adj list
- ```

bfsDistance(s):
// s is the start vertex
// dist[v] is length of s-to-v path
// Initially dist[v] = ∞ for all v
dist[s] = 0;
Q.insert(s);

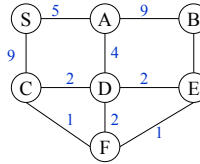
while (Q.nonempty()) {
    v = Q.get();
    for (each w adjacent to v) {
        if (dist[w] == ∞) {
            dist[w] = dist[v]+1;
            Q.insert(w);
        }
    }
}
    
```

## If There are Edge Costs?

- Idea #1
  - Add false nodes so that all edge costs are 1
  - But what if edge costs are large?
  - What if the costs aren't integers?
- Idea #2
  - Nothing "interesting" happens at the false nodes
    - Can't we just jump ahead to the next "real" node
  - Rule: always do the closest (real) node first
  - Use the array `dist[]` to
    - Report answers
    - Keep track of what to do next

## Dijkstra's Algorithm

- Intuition
  - Edges are threads, vertices are beads
  - Pick up at `s`; mark each node as it leave the table
- Note: Negative edge-costs are *not allowed*

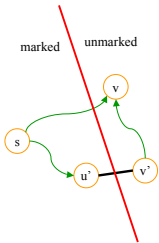


```

dijkstra(s):
  dist[s] = 0;
  while (some vertices are unmarked) {
    v = unmarked vertex with
      smallest dist;
    Mark v;
    for (each w adj to v) {
      dist[w] = min
        ( dist[w], dist[v] + c(v,w) );
    }
  }
    
```

## Proof for Dijkstra's Algorithm

- Claim: When vertex `v` is marked, `dist[v]` is the length of the shortest path from `s` to `v`
- Proof
  - Suppose there is a shorter path `P` from `s` to `v`
  - Consider the first edge of `P` that links a marked vertex to an unmarked vertex
    - Such an edge must exist because we know `s` is marked and `v` is not
    - Call this edge  $(u', v')$
  - Note that the length of the path from `s` to `u'` to `v'` is less than the length of `P`
    - Thus `v'` would be chosen in the algorithm instead of `v`
    - Contradiction!



## Dijkstra's Algorithm using Adj Matrix

- While-loop is done `n` times
- Within the loop
  - Choosing `v` takes  $O(n)$  time
    - Could do this faster using PQ, but no reason to
  - For-loop takes  $O(n)$  time
- Total time =  $O(n^2)$

```

dijkstra(s):
  dist[s] = 0;
  while (some vertices are unmarked) {
    v = unmarked vertex with
      smallest dist;
    Mark v;
    for (each w adj to v) {
      dist[w] = min
        ( dist[w], dist[v] + c(v,w) );
    }
  }
    
```

## Dijkstra's Algorithm using Adj List

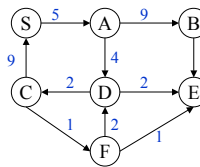
- Looks like we need a PQ
  - Problem: priorities are updated as algorithm runs
  - Can insert pair  $(v, dist[v])$  in PQ whenever `dist[v]` is updated
  - At most `m` things in PQ
- Time  $O(n + m \log m)$
- Using a more complicated PQ (e.g., Pairing Heap), time can be brought down to  $O(m + n \log n)$

```

dijkstra(s):
  dist[s] = 0;
  while (some vertices are unmarked) {
    v = unmarked vertex with
      smallest dist;
    Mark v;
    for (each w adj to v) {
      dist[w] = min
        ( dist[w], dist[v] + c(v,w) );
    }
  }
    
```

## Dijkstra's Algorithm for Digraphs

- Algorithm works on both undirected and directed graphs without modification
- As before: Negative edge-costs are *not allowed*



```

dijkstra(s):
  dist[s] = 0;
  while (some vertices are unmarked) {
    v = unmarked vertex with
      smallest dist;
    Mark v;
    for (each w adj to v) {
      dist[w] = min
        ( dist[w], dist[v] + c(v,w) );
    }
  }
    
```

## Greedy Algorithms

- Dijkstra's Algorithm is an example of a *Greedy Algorithm*
- The Greedy Strategy is an algorithm design technique
  - Like Divide & Conquer
- The Greedy Strategy is used to solve optimization problems
  - The goal is to find the best solution
- Works when the problem has the *greedy-choice property*
  - A global optimum can be reached by making locally optimum choices
- Problem: Given an amount of money, find the smallest number of coins to make that amount
- Solution: Use a Greedy Algorithm
  - Give as many large coins as you can
- This greedy strategy produces the optimum number of coins for the US coin system
- Different money system  $\Rightarrow$  greedy strategy may fail
  - For example: suppose the US introduces a 4¢ coin

## Minimum Spanning Trees

### Definition

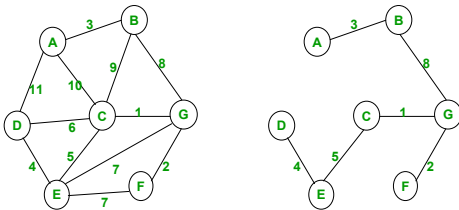
A *spanning tree* of an undirected graph  $G$  is a *tree* whose nodes are the vertices of  $G$  and whose edges are a subset of the edges of  $G$

### Definition

A *Minimum Spanning Tree (MST)* for a weighted graph  $G$  is the spanning tree of least cost (sum of edge-weights)

- Alternately, an MST can be defined as the least-cost set of edges so that all the vertices are connected
  - This has to be a tree... Why?
- A greedy strategy works for this problem
  - Add vertices one at a time
  - Always add the one that is closest to the current tree
  - This is called *Prim's Algorithm*

## An Example Graph and Its MST



## Prim's Algorithm

- $s$  is the start vertex
- $c(i,j)$  is the cost from  $i$  to  $j$
- Initially, vertices are unmarked
- $\text{dist}[v]$  is length of smallest tree-to- $v$  edge
- Initially,  $\text{dist}[v] = \infty$ , for all  $v$

```

prim(s):
  dist[s] = 0;
  while (some vertices are unmarked) {
    v = unmarked vertex with
        smallest dist;
    Mark v;
    for (each w adj to v) {
      dist[w] = min[ dist[w], c(v,w) ];
    }
  }
    
```

- Runtime analysis
  - $O(v^2)$  for adj matrix
    - While-loop is executed  $v$  times
    - For-loop takes  $O(v)$  time
  - $O(e + v \log v)$  for adj list
    - Use a PQ
    - Regular PQ produces time  $O(v + e \log e)$
    - Can improve to  $O(e + v \log v)$  by using fancier heap

## Similar Code Structures

- ```

while (some vertices are unmarked) {
  v = best of unmarked vertices;
  Mark v;
  for (each w adj to v)
    Update w;
}
    
```
- **bfsDistance**
    - best: next in queue
    - update:  $\text{dist}[w] = \text{dist}[v] + 1$
  - **dijkstra**
    - best: next in PQ
    - update:  $\text{dist}[w] = \min [ \text{dist}[w], \text{dist}[v] + \text{cost}(v,w) ]$
  - **prim**
    - best: next in PQ
    - update:  $\text{dist}[w] = \min [ \text{dist}[w], \text{cost}(v,w) ]$

## Remembering Your Choices

- How can you remember which choices were made?
    - Whenever  $\text{dist}[w]$  is updated we can remember the current  $v$  by using  $\text{parent}[w] = v$ ;
    - Can use the parent info to construct the *bfs tree*, the *shortest path tree*, or the *minimum spanning tree*
- ```

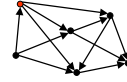
while (some vertices are unmarked) {
  v = best of unmarked vertices;
  Mark v;
  for (each w adj to v)
    Update w;
    if (w changed) parent[w] = v;
}
    
```

## Depth-First Search

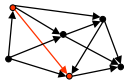
- Follow edges depth-first starting from an arbitrary vertex  $s$ , using a *Stack* to remember where you came from
- When you encounter a vertex previously visited, or there are no outgoing edges, retreat and try another path
- Eventually visit all vertices reachable from  $s$
- If there are still unvisited vertices, repeat

Easy to see this takes  $O(m)$  time

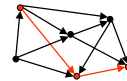
## Depth-First Search



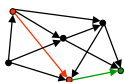
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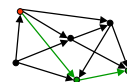
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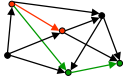
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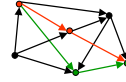
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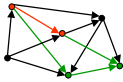
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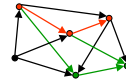
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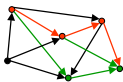
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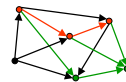
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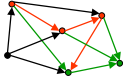
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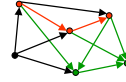
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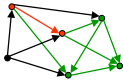
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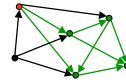
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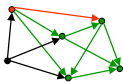
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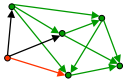
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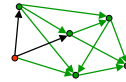
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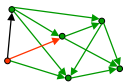
### Depth-First Search



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## Depth-First Search



## Depth-First Search



## Depth-First Search



## DFS Notes

- Same as BFS, except we use a Stack instead of a Queue to determine which edge to explore next
- Can also implement DFS recursively
  - The Stack is represented *implicitly* in the Stack Frames created by the recursive calls