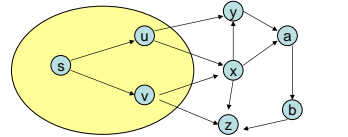


CSE 421 Algorithms

Autumn 2015
Lecture 10
Minimum Spanning Trees

Dijkstra's Algorithm Implementation and Runtime

```
S = {}; d[s] = 0; d[v] = infinity for v != s
While S != V
  Choose v in V-S with minimum d[v]
  Add v to S
  For each w in the neighborhood of v
    d[w] = min(d[w], d[v] + c(v, w))
```

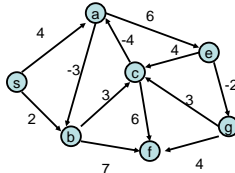


HEAP OPERATIONS
n Extract Mins
m Heap Updates

Edge costs are assumed to be non-negative

Shortest Paths

- Negative Cost Edges
 - Dijkstra's algorithm assumes positive cost edges
 - For some applications, negative cost edges make sense
 - Shortest path not well defined if a graph has a negative cost cycle

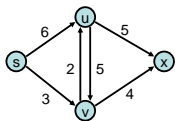


Negative Cost Edge Preview

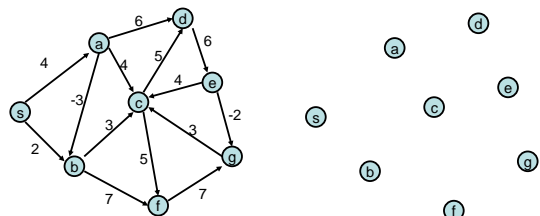
- Topological Sort can be used for solving the shortest path problem in directed acyclic graphs
- Bellman-Ford algorithm finds shortest paths in a graph with negative cost edges (or reports the existence of a negative cost cycle).

Bottleneck Shortest Path

- Define the bottleneck distance for a path to be the maximum cost edge along the path

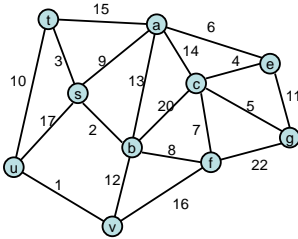


Compute the bottleneck shortest paths



Greedy Algorithm 3 Reverse-Delete Algorithm

- Delete the most expensive edge that does not disconnect the graph



Construct the MST with the reverse-delete algorithm
Label the edges in order of removal

Dijkstra's Algorithm for Minimum Spanning Trees

$S = \{s\}$; $d[s] = 0$; $d[v] = \text{infinity}$ for $v \neq s$

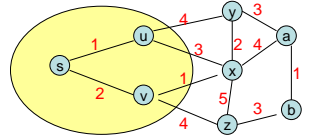
While $S \neq V$

Choose v in $V-S$ with minimum $d[v]$

Add v to S

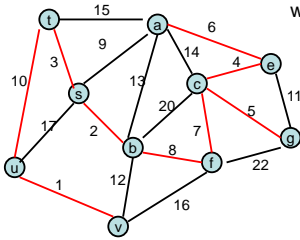
For each w in the neighborhood of v

$d[w] = \min(d[w], c(v, w))$



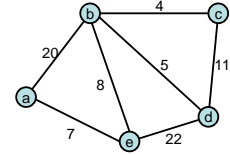
Minimum Spanning Tree

Undirected Graph $G=(V,E)$ with edge weights



Greedy Algorithms for Minimum Spanning Tree

- [Prim] Extend a tree by including the cheapest outgoing edge
- [Kruskal] Add the cheapest edge that joins disjoint components
- [ReverseDelete] Delete the most expensive edge that does not disconnect the graph

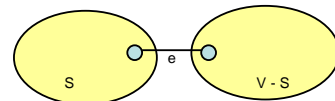


Why do the greedy algorithms work?

- For simplicity, assume all edge costs are distinct

Edge inclusion lemma

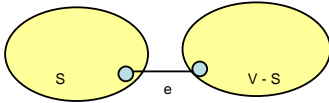
- Let S be a subset of V , and suppose $e = (u, v)$ is the minimum cost edge of E , with u in S and v in $V-S$
- e is in every minimum spanning tree of G
– Or equivalently, if e is not in T , then T is not a minimum spanning tree



e is the minimum cost edge between S and V-S

Proof

- Suppose T is a spanning tree that does not contain e
- Add e to T, this creates a cycle
- The cycle must have some edge $e_1 = (u_1, v_1)$ with u_1 in S and v_1 in V-S



- $T_1 = T - \{e_1\} + \{e\}$ is a spanning tree with lower cost
- Hence, T is not a minimum spanning tree

Optimality Proofs

- Prim's Algorithm computes a MST
- Kruskal's Algorithm computes a MST
- Show that when an edge is added to the MST by Prim or Kruskal, the edge is the minimum cost edge between S and V-S for some set S.

Prim's Algorithm

$S = \{ \}; T = \{ \};$

while $S \neq V$

 choose the minimum cost edge
 $e = (u, v)$, with u in S, and v in V-S
 add e to T
 add v to S

Prove Prim's algorithm computes an MST

- Show an edge e is in the MST when it is added to T

Kruskal's Algorithm

Let $C = \{ \{v_1\}, \{v_2\}, \dots, \{v_n\} \}; T = \{ \}$

while $|C| > 1$

 Let $e = (u, v)$ with u in C_i and v in C_j be the
 minimum cost edge joining distinct sets in C
 Replace C_i and C_j by $C_i \cup C_j$
 Add e to T

Prove Kruskal's algorithm computes an MST

- Show an edge e is in the MST when it is added to T

Reverse-Delete Algorithm

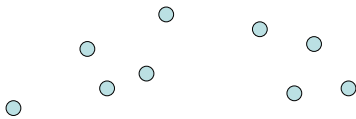
- Lemma: The most expensive edge on a cycle is never in a minimum spanning tree

Dealing with the assumption of no equal weight edges

- Force the edge weights to be distinct
 - Add small quantities to the weights
 - Give a tie breaking rule for equal weight edges

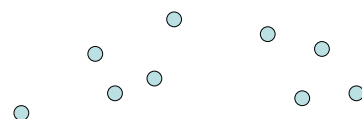
Application: Clustering

- Given a collection of points in an r-dimensional space, and an integer K, divide the points into K sets that are closest together

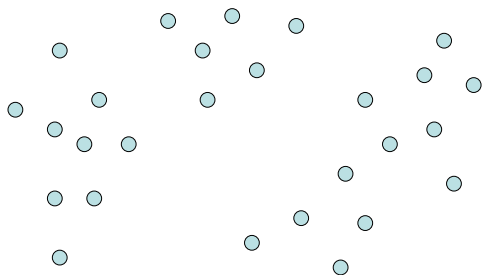


Distance clustering

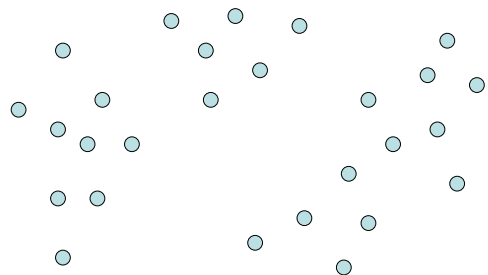
- Divide the data set into K subsets to maximize the distance between any pair of sets
 - $\text{dist}(S_1, S_2) = \min \{ \text{dist}(x, y) \mid x \in S_1, y \in S_2 \}$



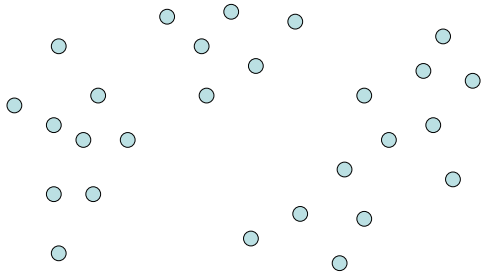
Divide into 2 clusters



Divide into 3 clusters



Divide into 4 clusters



Distance Clustering Algorithm

Let $C = \{\{v_1\}, \{v_2\}, \dots, \{v_n\}\}; T = \{\}$

while $|C| > K$

Let $e = (u, v)$ with u in C_i and v in C_j be the minimum cost edge joining distinct sets in C

Replace C_i and C_j by $C_i \cup C_j$

K-clustering

