

Lecture13



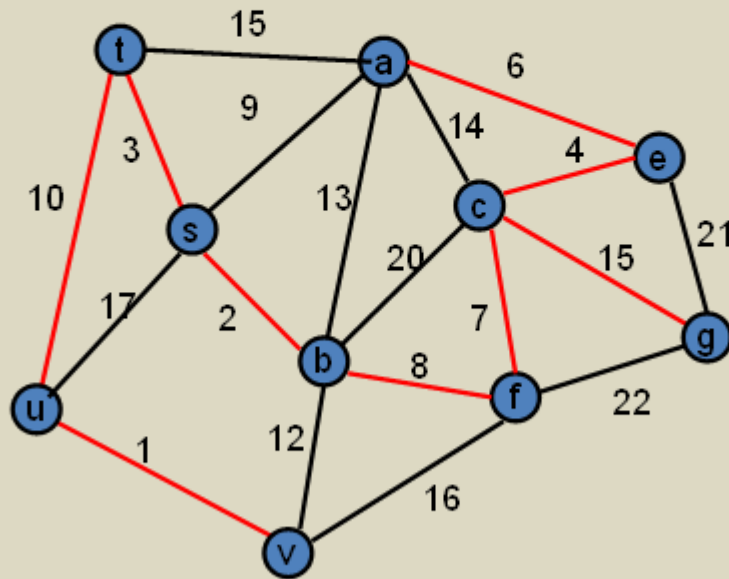
CSE 417 Algorithms and Complexity

Winter 2023
Lecture 13
Minimum Spanning Trees

Announcements

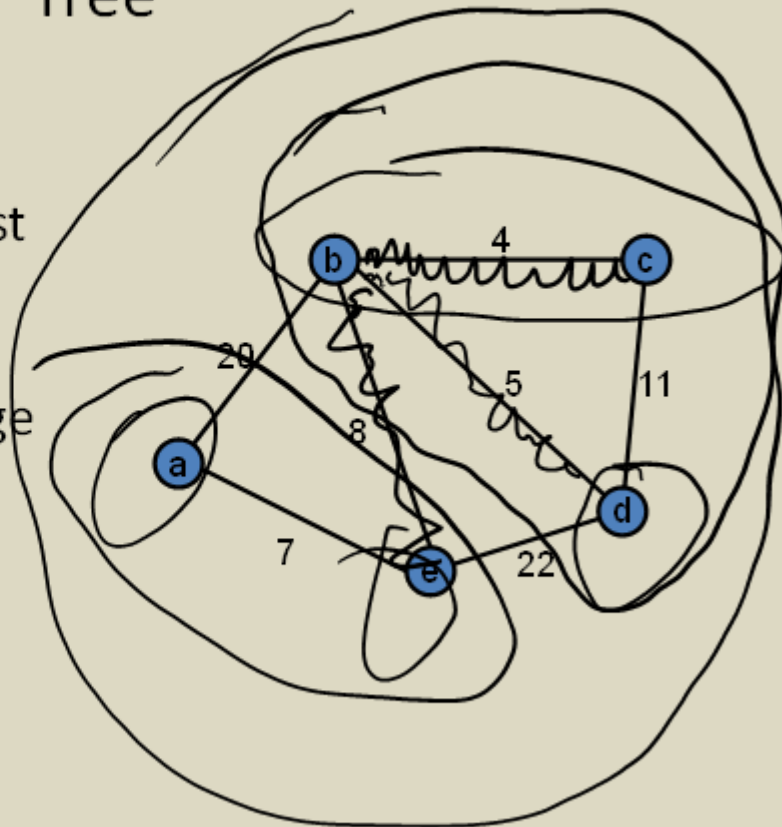
- Midterm, Wednesday, Feb 8
- Topics: Material Presented in Lecture
 - Stable Matching
 - Graphs and simple graph algorithms
 - Breadth First Search
 - Topological Sort
 - Greedy Algorithms
 - Interval Scheduling Problems
 - Shortest Paths Algorithms
 - Minimum Spanning Tree Algorithms

Minimum Spanning Tree



Greedy Algorithms for Minimum Spanning Tree

- Prim's Algorithm:
Extend a tree by including the cheapest outgoing edge
- Kruskal's Algorithm:
Add the cheapest edge that joins disjoint components



2/3/2002

CSE 417

4

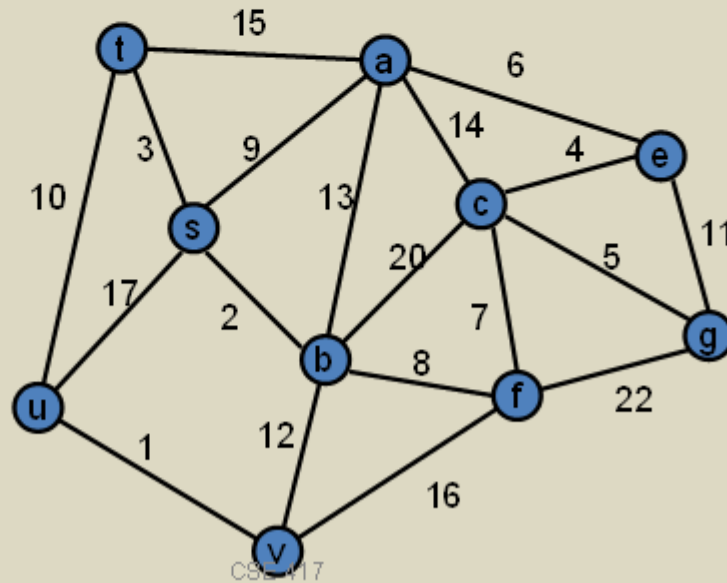
Greedy Algorithm 1

Prim's Algorithm

- Extend a tree by including the cheapest out going edge

Construct the MST
with Prim's
algorithm starting
from vertex a

Label the edges in
order of insertion



2/3/2002

CSE 417

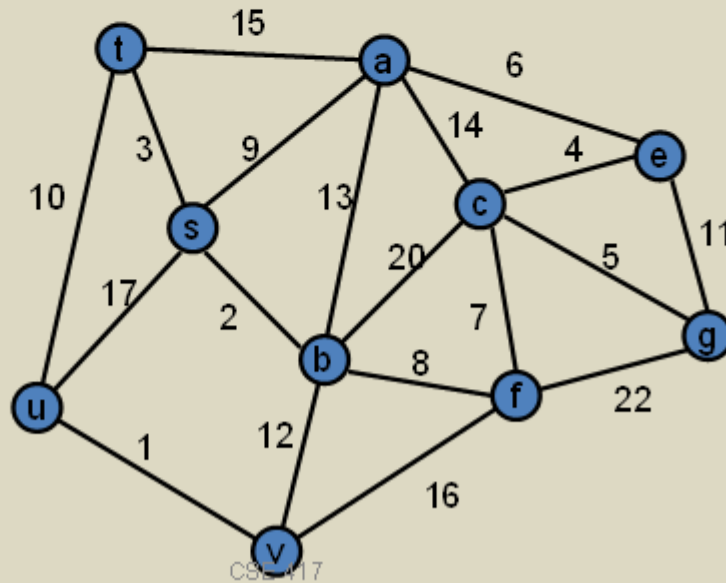
5

Greedy Algorithm 2 Kruskal's Algorithm

- Add the cheapest edge that joins disjoint components

Construct the MST
with Kruskal's
algorithm

Label the edges in
order of insertion

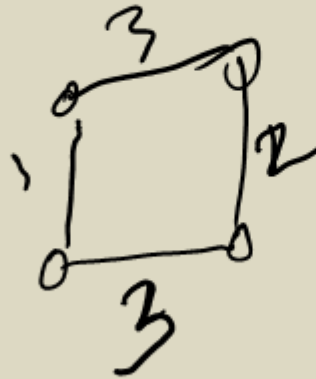


CSE 417

6

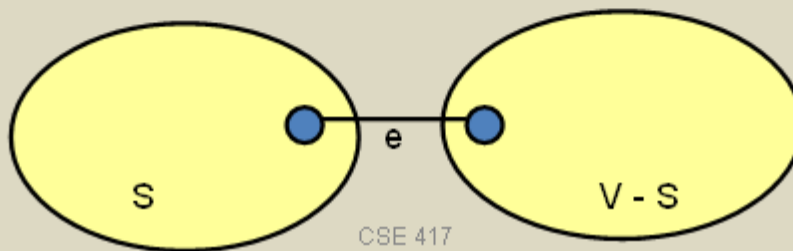
Why do the greedy algorithms work?

- 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



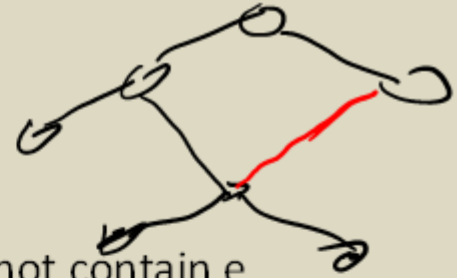
2/3/2002

CSE 417

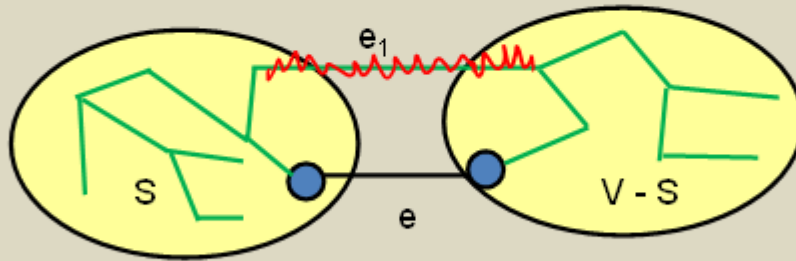
8

**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



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

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

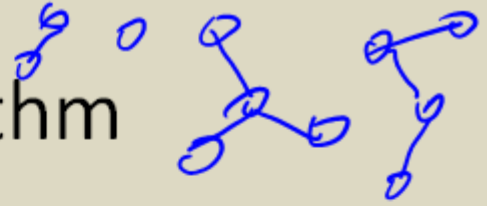
 add v to S

Prove Prim's algorithm computes an MST

- Show an edge $e = (u, v)$ is in the MST when it is added to T



Kruskal's Algorithm



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

while $|C| > 1$

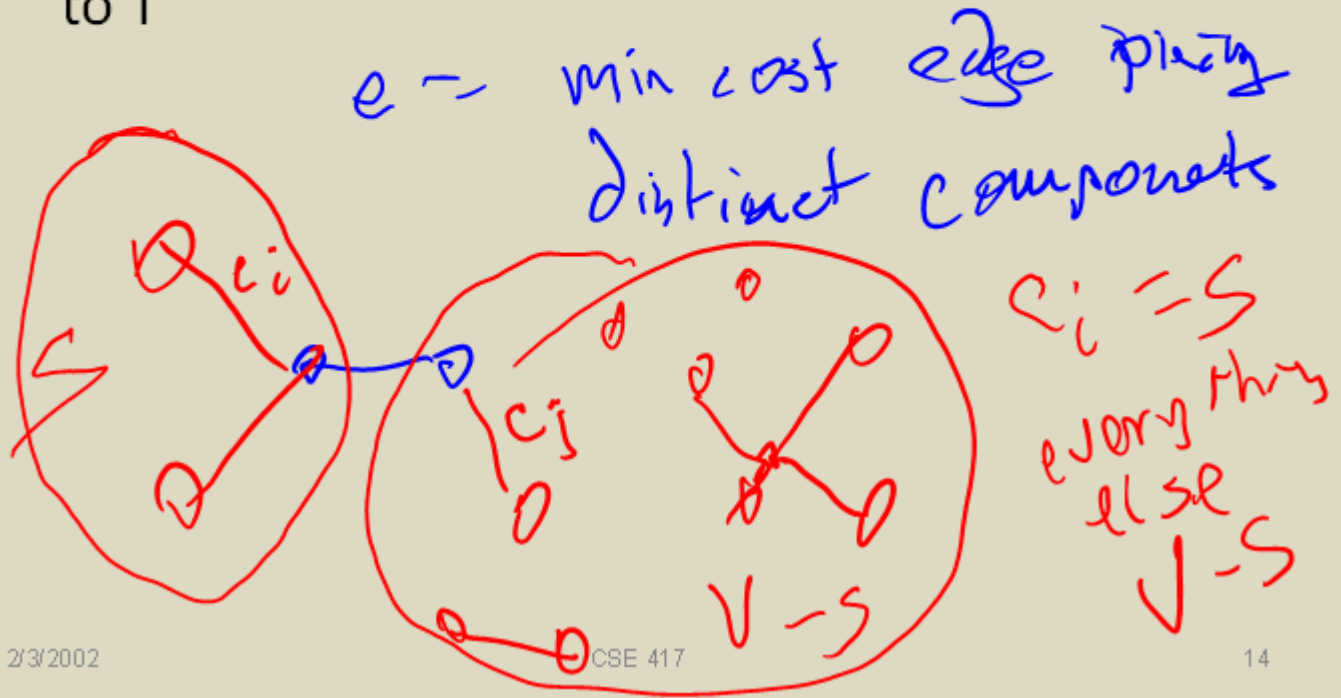
Let $e = (u, v)$ with u in C_i and v in C_j be the minimum cost edge joining distinct sets in C

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



MST Implementation and runtime

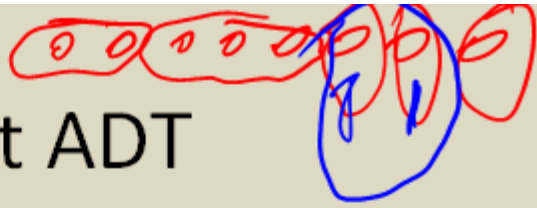
- Prim's Algorithm
 - Implementation, runtime: just like Dijkstra's algorithm
 - Use a heap, runtime $O(m \log n)$
- Kruskal's Algorithm
 - Sorting edges by cost: $O(m \log n)$
 - Managing connected components uses the Union-Find data structure
 - Amazing, pointer based data structure
 - Very interesting mathematical result

2m - Find
n-1 - Union

$O(\log n)$

$O(\alpha(n))$

Disjoint Set ADT

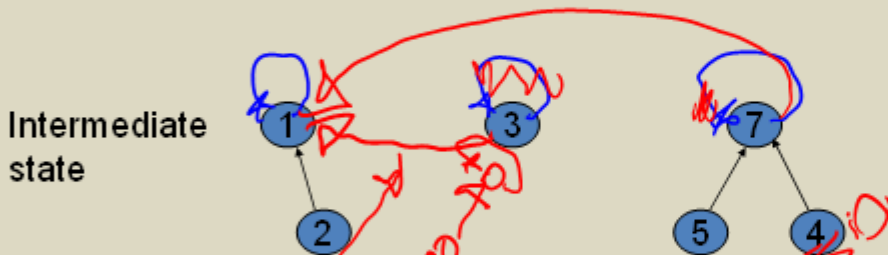


- Data: set of pairwise **disjoint sets**.
- Required operations
 - **Union** – merge two sets to create their union
 - **Find** – determine which set an item appears in
- Check $\text{Find}(v) \neq \text{Find}(w)$ to determine if (v,w) joins separate components
- Do $\text{Union}(v,w)$ to merge sets

Up-Tree for DS Union/Find

Observation: we will only traverse these trees upward from any given node to find the root.

Idea: reverse the pointers (make them point up from child to parent). The result is an **up-tree**.



Roots are the names of each set.

Find —
Traversing
up to
root

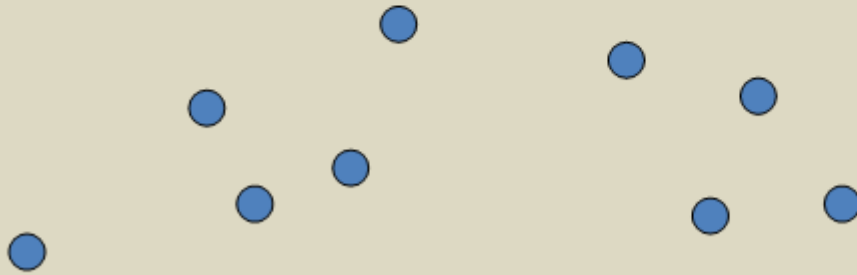
Union
One root
to
the other

Path compression



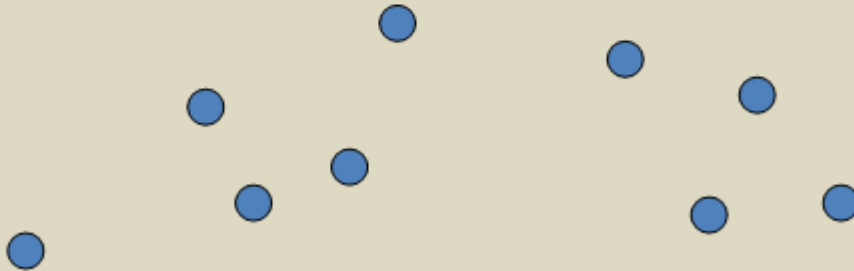
Application: Clustering

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

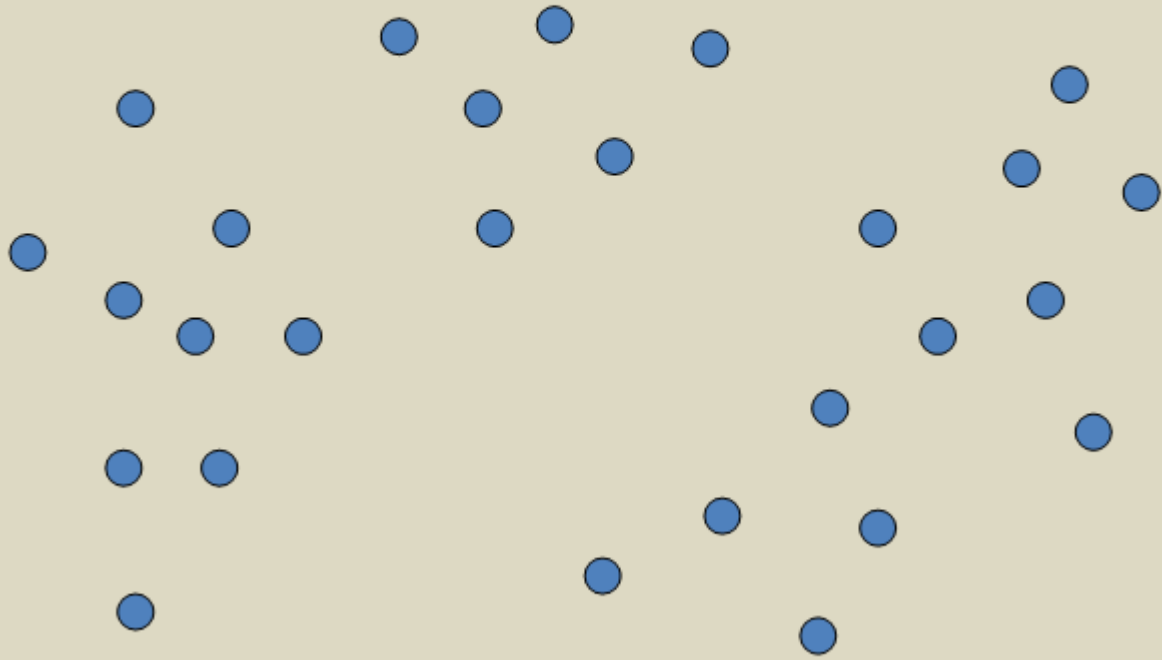


Distance clustering

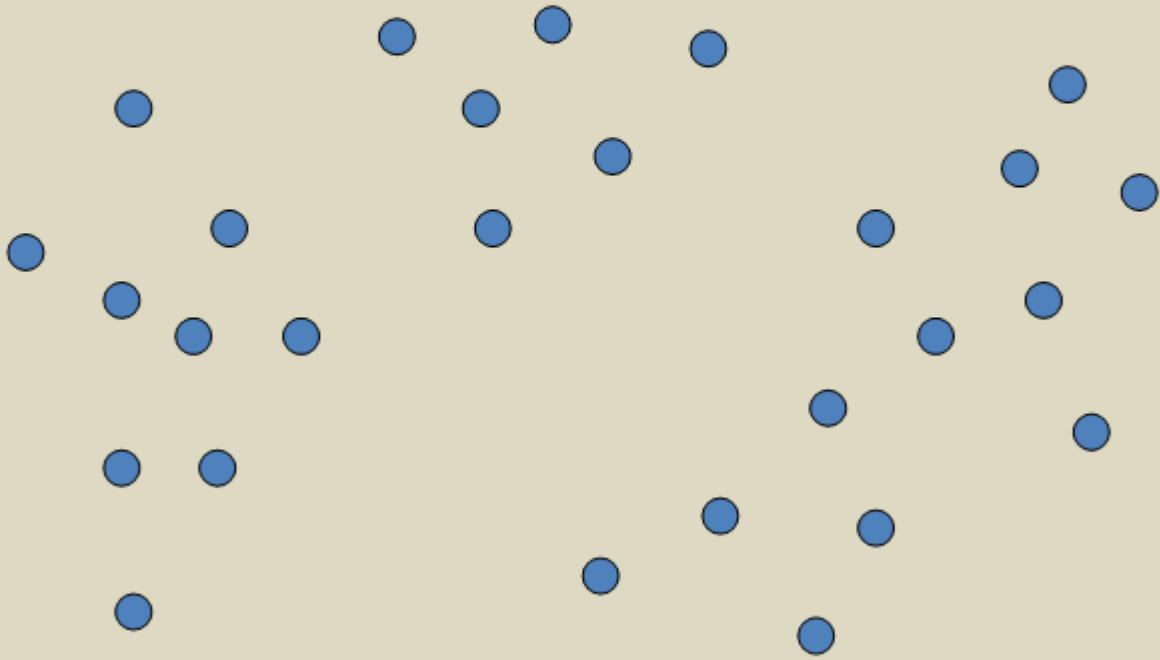
- Divide the data set into K subsets to maximize the distance between any pair of sets
 - $\text{dist}(S_1, S_2) = \min \{ \text{dist}(x, y) \mid x \text{ in } S_1, y \text{ in } S_2 \}$



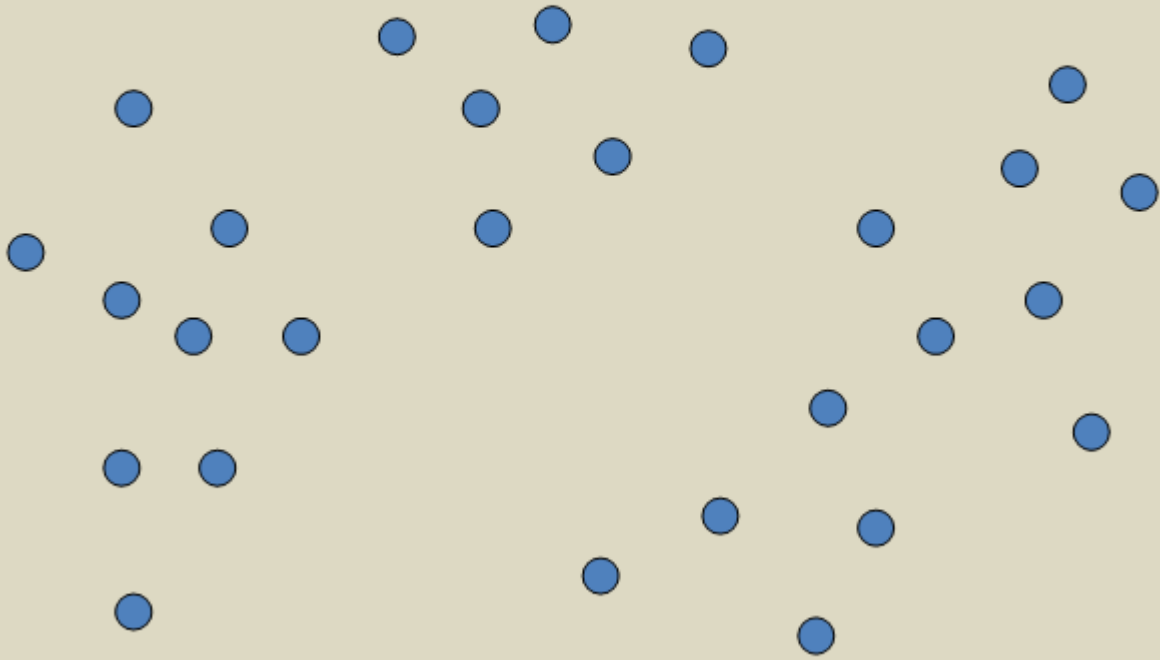
Divide into 2 clusters



Divide into 3 clusters



Divide into 4 clusters



Distance Clustering Algorithm

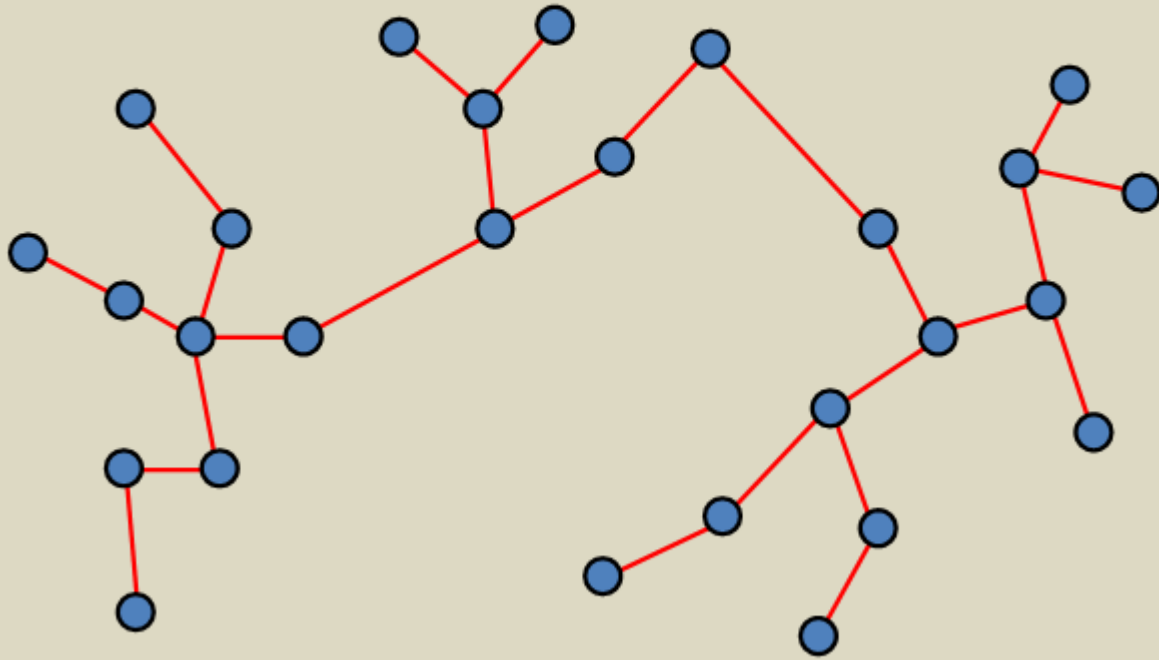
Let $C = \{\{v_1\}, \{v_2\}, \dots, \{v_n\}\}$; $T = \{\}$

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

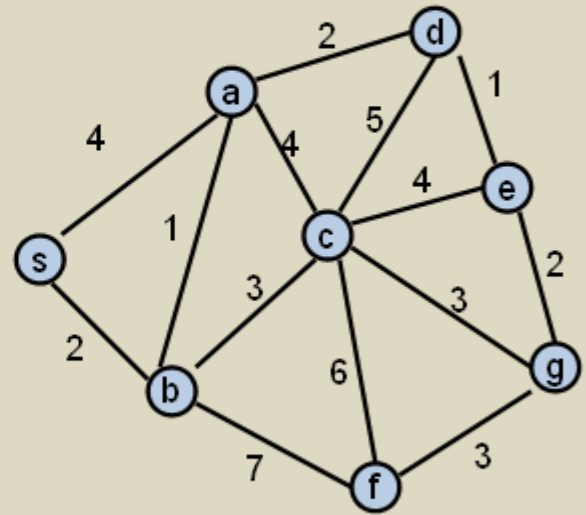
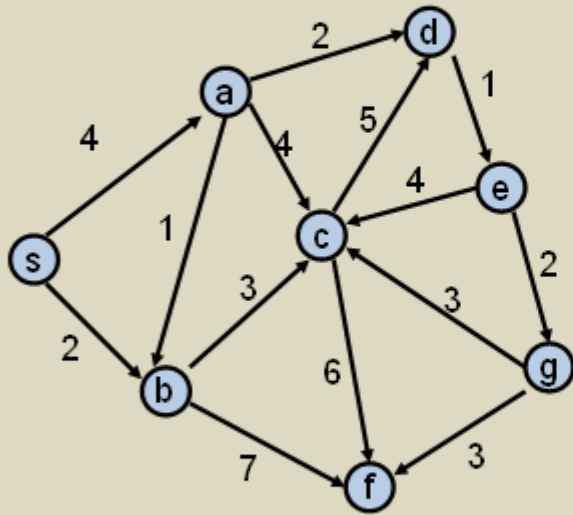


2/3/2002

CSE 417

24

Shortest paths in directed graphs vs undirected graphs



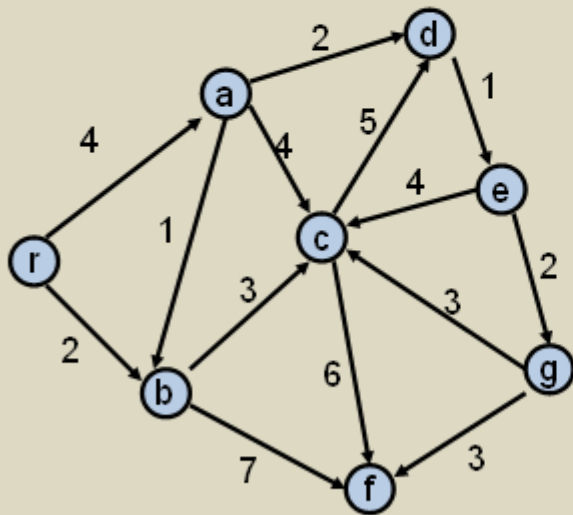
2/3/2002

CSE 417

25

What about the minimum spanning tree of a directed graph?

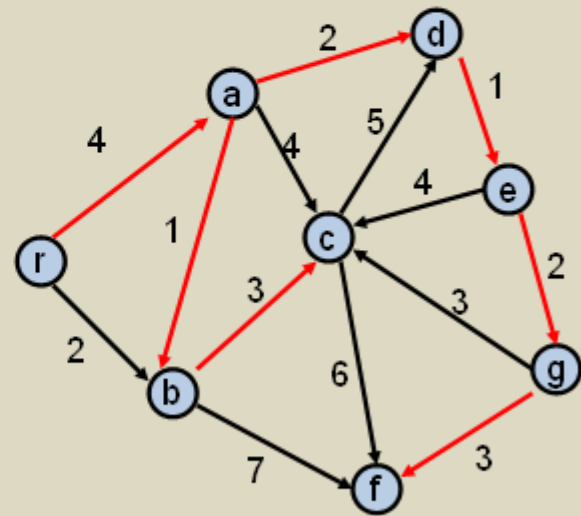
- Must specify the root r
- Branching: Out tree with root r



Assume all vertices reachable from r

2/3/2007

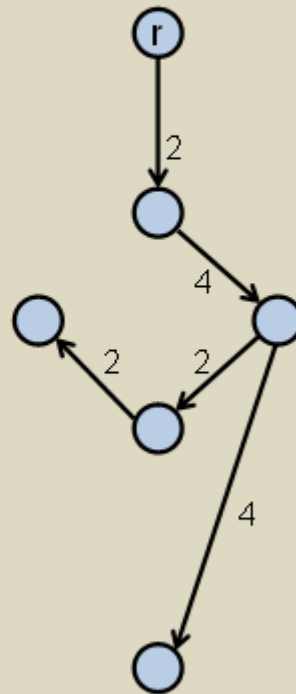
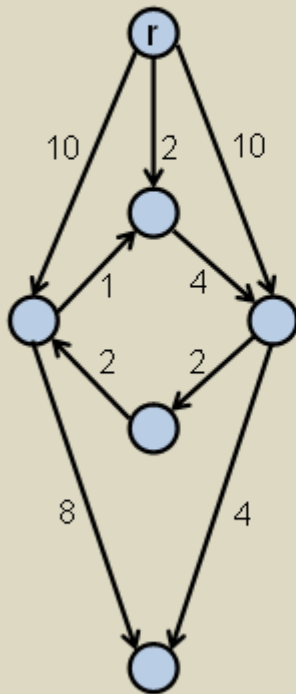
CSE 417



Also called an arborescence

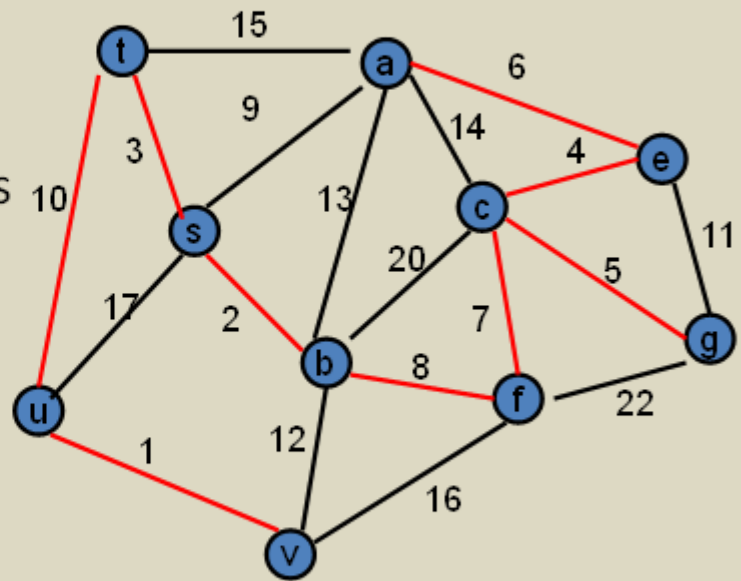
2/3

Finding a minimum branching



Another MST Algorithm

- Choose minimum cost edge into each vertex
- Merge into components
- Repeat until done



Idea for branching algorithm

- Select minimum cost edge going into each vertex
- If graph is a branching then done
- Otherwise collapse cycles and repeat

