



Design and analysis of algorithms

Lecture 18 & 19

Edyta Szymańska

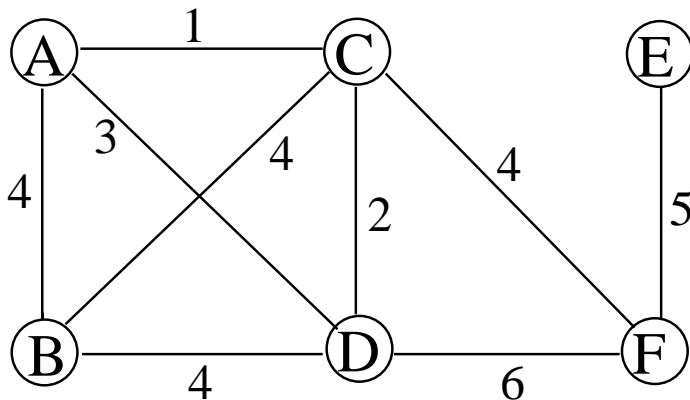
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Minimum spanning tree (MST)

Problem: given a set of computers and a maintenance cost associated with every potential link, build a communication network which is connected and has the cheapest cost.

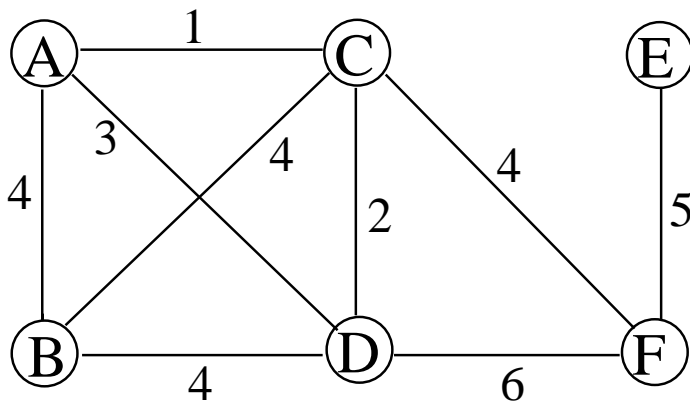
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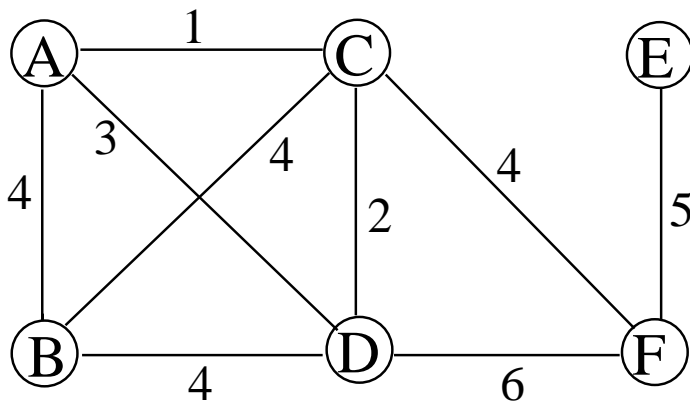
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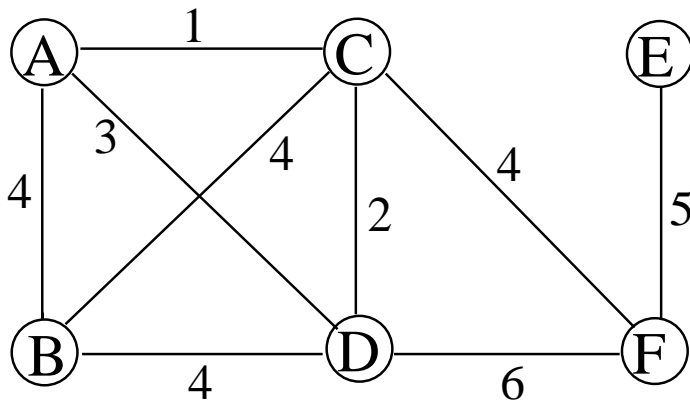
Thus, the solution must be connected and acyclic - TREE.

Minimum spanning tree (MST)

INPUT: an undirected, connected graph $G = (V, E)$ with edge weights $w(e)$.

OUTPUT: a tree $T = (V, E')$ spanning all the vertices of G and minimizing

$$\text{weight}(T) = \sum_{e \in E'} w(e)$$

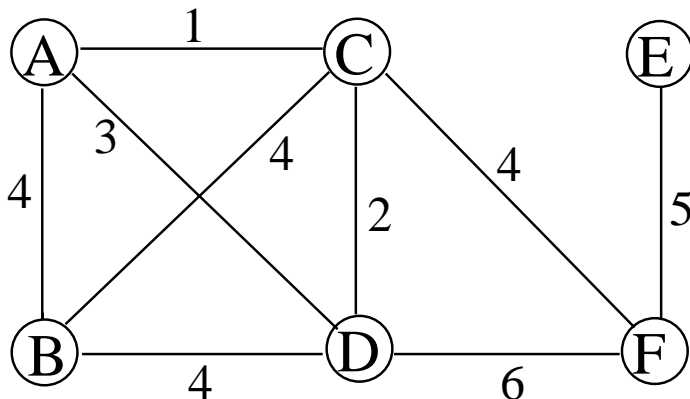


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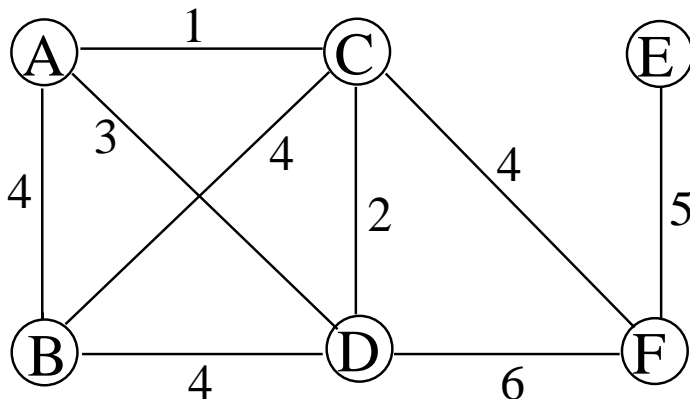
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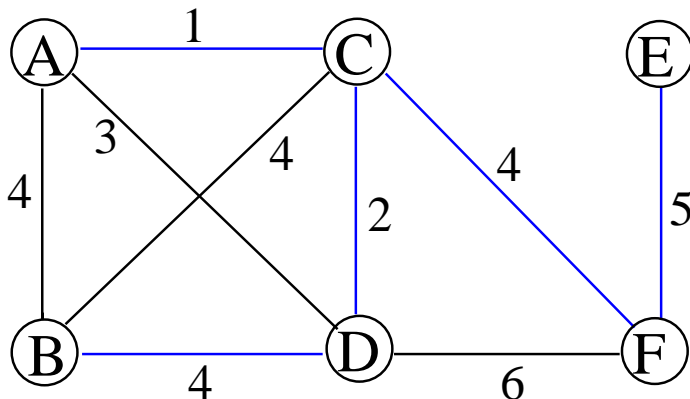
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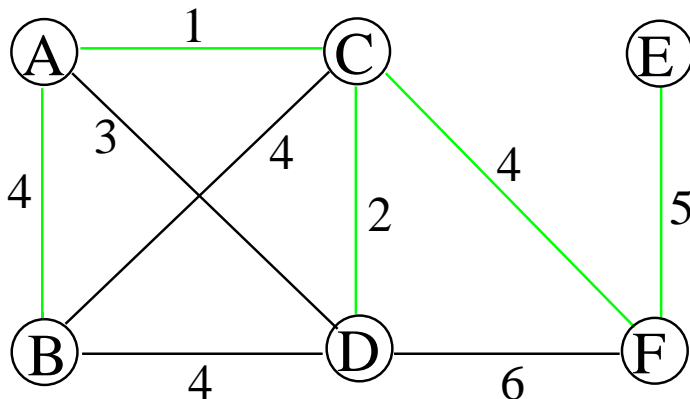
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build a spanning tree by successively adding edges in order of increasing weights and in such a way that they do not form a cycle with those edges already chosen.

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This a **greedy** method: makes a decision at each step myopically to optimize the underlying criterion.

Why is Kruskal's algorithm correct ?

We will prove a more general property first:

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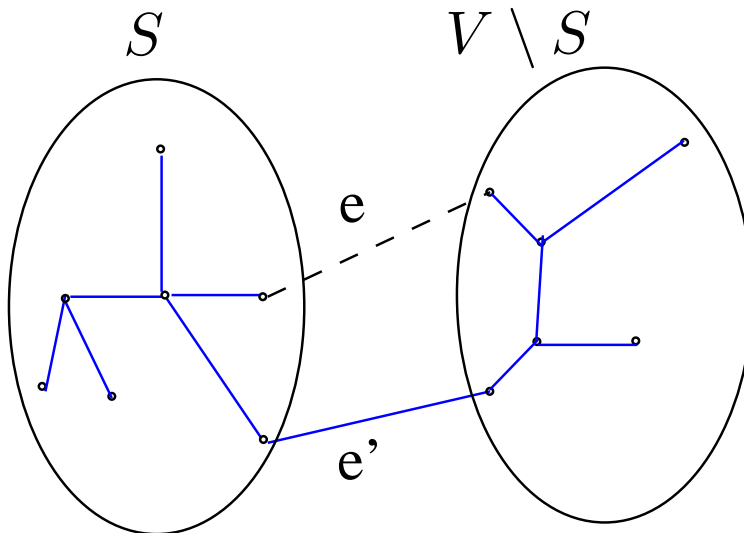
Cut property: Let $X \subseteq E$ be part of some minimum spanning tree T of $G = (V, E)$. Pick any set of nodes $S \subset V$ such that there is no edge in X which connects a node in S to one in $V \setminus S$. If $e \in E$ is the minimum-weight edge between S and $V - S$, then $X \cup \{e\}$ is also part of some MST.

Why is Kruskal's algorithm correct ?

Proof: Assume $e \notin T$. Then $T \cup \{e\}$ contains a cycle C .
Now C contains another edge e' across the cut $(S, V \setminus S)$.

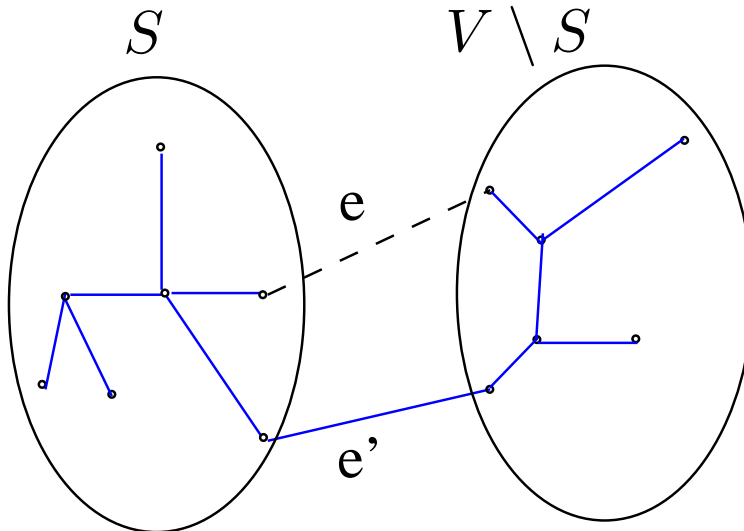
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Consider $T' = T \cup \{e\} - \{e''\}$. It is a tree with $w(T') = w(T) + w(e) - w(e') \leq w(T)$, because $w(e) \leq w(e')$. Since T is MST, so is T' .

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In Kruskal's alg.: when we pick an edge connecting two components C and C' , it is the lightest edge in the cut $(C, V - S)$ and thus the final tree is optimal, MST.

Implementation of Kruskal's algorithm

Avoiding cycles: every time we want to add edge (u, v) to MST we need to check whether its endpoints u and v are in different connected components.

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Set operations:

- ⑥ **MAKESET** (x) : creates set $\{x\}$
- ⑥ **FIND** (x) : to which set does x belong?
- ⑥ **UNION** (x, y) : merge sets containing x and y .

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Kruskal's Algorithm

KRUSKAL(G, w)

for all $v \in V$ MAKESET(v)

$X := \emptyset$

Sort the edges of E by weight

for each edge (u, v) in nondecreasing order of weight

do if FIND(u) \neq FIND(v) then

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$n \times$ MakeSet

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Sort the edges of E by weight

$O(m \log m)$

for each edge (u, v) in nondecreasing order of weight

do if FIND(u) \neq FIND(v) then

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b/c all disjoint-set operations take $O(m \log^ n)$ time, where $\log^* n$ is much smaller than $\log n$.*

Prim's greedy choice

builds an MST by growing a single connected component complying with the cut property, r -initial node

Prim's greedy choice

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PRIM( $G, w, r$ )  
for all  $v \in V$  do  $cost(v) := \infty; prev(v) := nil$   
 $cost(r) := 0$   
 $Q := BuildHeap(V)$  {keys = cost values}  
while  $Q \neq \emptyset$  do  
    ⑥  $u := EXTRACT\_MIN(Q)$   
    ⑥ for each  $v \in Adj[u]$  do  
        if  $v \in Q$  and  $cost(v) > w(u, v)$  then  
             $cost(v) := w(u, v)$   
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Correctness follows from the **Cut Property**.

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⑥ An array:

`Component[s]` := the name of the set containing s

Cost:

`MakeSet(x)`: $O(1)$

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Represent every set by a directed tree, nodes in an arbitrary order, with root as a *representative*

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Example:

Disjoint-sets data structure

- ⑥ A directed tree:
for all x : $\pi(x)$ -parent pointer, $rank(x)$ -the height of the subtree rooted at x

MAKESET(x):

$\pi(x) := x$

$rank(x) := 0$

FIND(x):

while $x \neq \pi(x)$ **do** $x = \pi(x)$

return x

UNION(x, y): ??

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return x $O(\text{height}(\text{Tree}))$

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