

# Foundations of Operations Research

Master of Science in Computer Engineering

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Tuesday 13.15 - 15.15

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<http://homes.di.unimi.it/~cordone/courses/2014-for/2014-for.html>

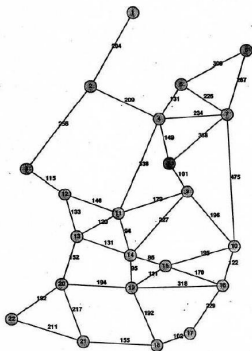


# A network design model

*A telecommunication company wants to build a new fiberoptic network between some major European cities.*

*All cities should be connected to each other, directly or indirectly.  
A set of potential connections and the cost of building each link (proportional to the distance) are known.*

*Design the fiberoptic network of minimum total cost.*



A natural combinatorial model is given by an **edge-weighted undirected graph**  $(V, E, c)$

- $V$  includes the **cities**
- $E$  includes the **potential links**
- $c : E \rightarrow \mathbb{R}^+$  provides the **cost of a link**

We are looking for a subgraph  $T = (U, X)$

*What kind of subgraph?*

# A network design model

Let us denote by  $T = (V, X)$  any feasible subgraph:

- it must **include all vertices**: it must be **spanning**
- it must **include a path between any pair of vertices**: it must be **connected**
- its **total cost** must be **minimum**:  $X = \arg \min \sum_{e \in X} c_e$

*Can it include cycles?*

Given a cyclic connected subgraph, **remove one edge  $e$  from a cycle**:

- **the result is connected**
- if all cycles include an edge with  $c_e \geq 0$ , **the result is not more expensive**

The optimal solution includes no cycle:  $T = (V, X)$  is an acyclic subgraph

Definitions

- a **forest** is an **acyclic graph**
- a **tree** is an **acyclic connected graph**
- a **spanning tree** is an **acyclic connected spanning subgraph**

If all cycles of  $G$  include an edge  $e$  with  $c_e \geq 0$ ,  $T$  is a **minimum spanning tree**

# Minimum spanning tree problem

Given

- an undirected connected graph  $G = (V, E)$  with  $n = |V|$  vertices and  $m = |E|$  edges
- a cost function  $c : E \rightarrow \mathbb{R}$

find a subgraph  $T^* = (U^*, X^*)$

- 1 spanning:  $U^*$  contains all vertices ( $U^* = V$ )
- 2 connected:  $X^*$  includes a path between each pair of vertices  $u$  and  $v$
- 3 acyclic:  $X^*$  does not contain any cycle
- 4 of minimum total cost:

$$c_{X^*} \leq c_X \text{ for all } T = (U, X) \text{ enjoying properties 1, 2 e 3}$$

$$\text{where } c_X = \sum_{e \in X} c_e$$

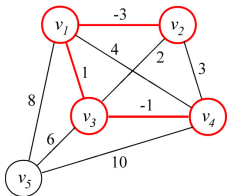
If  $G$  includes a cycle with all edges of negative cost, the minimum spanning tree problem is not a good model for the previous problem

*But you can apply a simple adaptation: which one?*

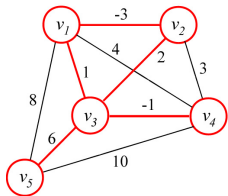
# Examples

$T$  is not an optimal solution because it is...

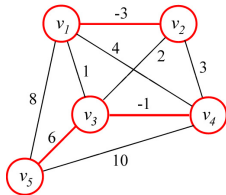
... nonspanning



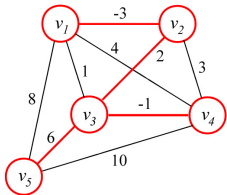
... cyclic



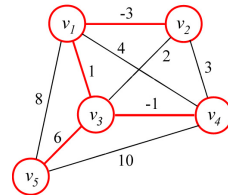
... nonconnected



... nonminimal ( $c_X = 4$ )



$T^*$  is optimal ( $c_{X^*} = 3$ )



# A second model: secure message transmission

*Broadcast to all stations of a communication network a secret message, minimizing the probability of interception at the links.*

We model the network as a graph  $G = (V, E)$ :

- vertices for the stations
- edges for the links
- a probability of interception  $p_e \in [0; 1)$  for each edge

What is the probability of interception using a subset  $X$  of the edges?

$$f(X) = 1 - \prod_{e \in X} (1 - p_e)$$

i. e. the complement of the probability not to be intercepted at any link

$$\min_X f(X) \Leftrightarrow \max_X \log \prod_{e \in X} (1 - p_e) = \sum_{e \in X} \log(1 - p_e) \Leftrightarrow \min_X \sum_{e \in X} \log \frac{1}{(1 - p_e)}$$

We are looking for a connected spanning subgraph  $(V, X)$   
(nonnegative costs: the optimal subgraph is also acyclic)

# A third model: compact binary sequence representation

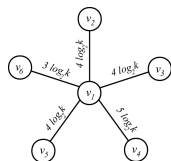
You have a large number  $n$  of binary sequences of huge length  $k$ ,  
and you want to represent them in a compact way

$$\begin{array}{lll} s_1 : [011100011101] & s_2 : [101101011001] & s_3 : [110100111001] \\ s_4 : [101001111101] & s_5 : [100100111101] & s_6 : [010101011100] \end{array}$$

An idea is to select a reference sequence and provide the differences (“bit flips”) between the other ones and it

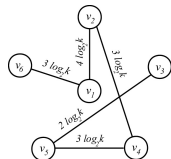
$$\begin{array}{lll} s_1 : [011100011101] & s_2 - s_1 : [1\ 2\ 6\ 10] & s_3 - s_1 : [1\ 3\ 7\ 10] \\ s_4 - s_1 : [1\ 2\ 4\ 6\ 7] & s_5 - s_1 : [1\ 2\ 3\ 7] & s_6 - s_1 : [3\ 6\ 12] \end{array}$$

This pays if many sequences are similar to the reference



A better idea is to allow a connected set of differences

$$\begin{array}{lll} s_6 : [010101011100] & s_1 - s_6 : [3\ 6\ 12] & s_2 - s_1 : [1\ 2\ 6\ 10] \\ s_4 - s_2 : [4\ 7\ 10] & s_5 - s_4 : [3\ 4\ 6] & s_3 - s_5 : [2\ 10] \end{array}$$



# A third model: compact binary sequence representation

Consider a complete undirected weighted graph:

- the vertices represent sequences
- the edges represent pairs of sequences
- the cost function is the number of bit flips between two sequences

$c_{uv}$	1	2	3	4	5	6
1	0	4	4	5	4	3
2	4	0	4	3	4	5
3	4	4	0	5	2	5
4	5	3	5	0	3	6
5	4	4	2	3	0	5
6	3	5	5	6	5	0

We look for a subgraph, which must be

- spanning, to represent all sequences
- connected, to allow reconstructing any sequence from the reference
- of minimum cost, to save memory space

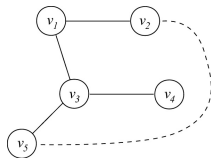
Since the costs are nonnegative, the subgraph is acyclic



# Useful properties on trees

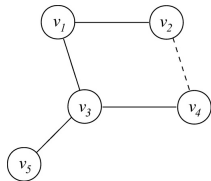
A tree contains exactly one path  $P_{uv}$  between any pair of vertices  $u$  and  $v$

- a tree is connected  $\Rightarrow$  there is at least one path
- two paths form a cycle, but a tree is acyclic  $\Rightarrow$  there is at most one path



Adding an edge  $[u, v]$  to a spanning tree yields exactly one cycle

- the tree spans  $u$  and  $v$  and contains a path  $P_{uv} \Rightarrow [u, v] \cup P_{uv}$  is a cycle  $\Rightarrow$  there is at least one cycle
- if adding  $[u, v]$  yields at least two cycles, the original tree had two different paths between  $u$  and  $v$  (contrary to the previous thesis) adding  $\Rightarrow [u, v]$  yields at most one cycle



# A general scheme

The vertex set of an optimal spanning tree is obviously  $V$

We want to build the edge set with a scheme of this kind:

- 1 Find a set of edges  $X$  certainly included in the edge set of an optimal solution
- 2 If  $(V, X)$  is an optimal solution, terminate
- 3 Otherwise, find an edge  $e^*$  such that  $X \cup \{e^*\}$  is still included in the edge set of an optimal solution and go back to point 2

The scheme provides an optimal solution in a finite number of steps, provided that we can always find  $e^*$

The optimal spanning tree problem is one of the few problems which admits such a scheme

*How is it possible, and why?*

# A fundamental theorem

Given the following assumptions:

- 1  $S \subset V$  is a nonempty proper subset of vertices and  $\Delta_S = \{[u, v] \in E : |\{u, v\} \cap S| = 1\}$  is its induced cut
- 2  $e^* = \arg \min_{e \in \Delta(S)} c_e$  is one of the edges of minimum cost in  $\Delta(S)$

there exists an optimal spanning tree whose edge set includes  $e^*$

- 3  $T^* = (V, X^*)$  is an optimal spanning tree and  $X \subseteq X^*$  is a subset of its edges
- 4  $\Delta_S \cap X = \emptyset$

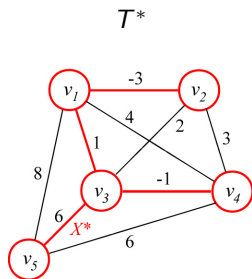
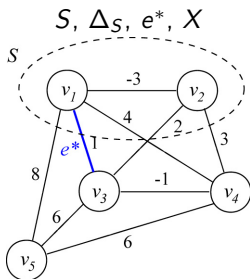
there exists an optimal spanning tree whose edge set includes  $X \cup \{e^*\}$

*Such a tree can be different from  $T^*$ !*

One can always enrich a subset of the edges of an optimal spanning tree with a minimum cost edge of a cut not intersecting the subset

The only condition is that the graph be connected

# Examples (1)



$$S = \{v_1, v_2\}$$

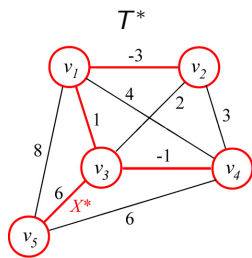
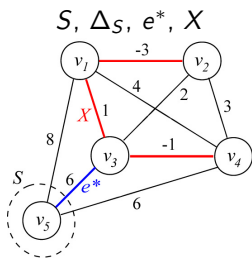
$$\Delta_S = \{(v_1, v_3), (v_1, v_4), (v_1, v_5), (v_2, v_3), (v_2, v_4)\}$$

$$e^* = (v_1, v_3)$$

$$X = \emptyset$$

$$\Rightarrow X \cup \{e^*\} = \{(v_1, v_3)\} \subseteq X^*$$

## Examples (2)



$$S = \{v_5\}$$

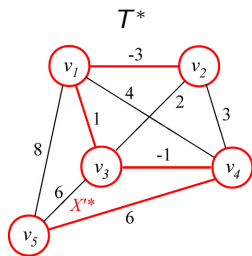
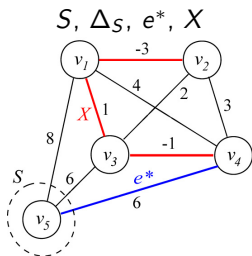
$$\Delta_S = \{(v_1, v_5), (v_3, v_5), (v_4, v_5)\}$$

$$e^* = (v_3, v_5)$$

$$X = \{(v_1, v_2), (v_1, v_3), (v_3, v_4)\}$$

$$\Rightarrow X \cup \{e^*\} = \{(v_1, v_2), (v_1, v_3), (v_3, v_4), (v_3, v_5)\} \subseteq X^*$$

# Examples (3)



$$S = \{v_5\}$$

$$\Delta_S = \{(v_1, v_5), (v_3, v_5), (v_4, v_5)\}$$

$$e^* = (v_4, v_5)$$

$$X = \{(v_1, v_2), (v_1, v_3), (v_3, v_4)\}$$

$$\Rightarrow X \cup \{e^*\} = \{(v_1, v_2), (v_1, v_3), (v_3, v_4), (v_4, v_5)\}$$

$X \cup \{e^*\} \not\subseteq X^*$ , but it is included in another optimal spanning tree  $X'^*$

But what if you consider  $S = \{v_5\}$  and  $X = \{(v_3, v_4), (v_3, v_5)\}$ ?

# Proof (1)

There are two possible cases

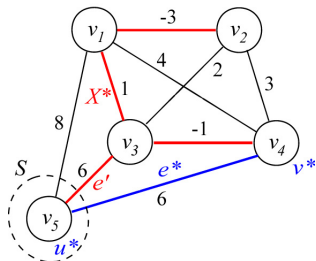
①  $e^* \in X^*$ : since  $X \subseteq X^*$ , then  $X \cup \{e^*\} \subseteq X^*$  and the thesis follows

②  $e^* = [u^*, v^*] \notin X^*$ :

the optimal solution  $T = (V, X^*)$  is spanning and connected

$X^*$  includes a path  $P_{u^*v^*}$  between  $u^*$  and  $v^*$

$P_{u^*v^*}$  intersects  $\Delta_S$  in at least one edge  $e'$



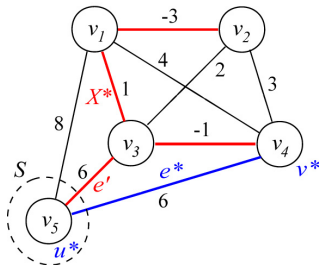
adding  $e^*$  to  $X^*$  produces a cycle

removing  $e'$  from this cycle yields another spanning tree

(the extreme vertices of  $e'$  are now connected through  $e^*$ )

## Proof (2)

$(V, X^* \cup \{e^*\} \setminus \{e'\})$  is another spanning tree  
and its cost is  $c_{X^*} + c_{e^*} - c_{e'}$  (where  $c_{X^*} = \sum_{e \in X^*} c_e$ )



Notice that

- $T = (V, X^*)$  is optimal  $\Rightarrow c_{X^*} + c_{e^*} - c_{e'} \geq c_{X^*} \Rightarrow c_{e^*} \geq c_{e'}$
- $e^* = \arg \min_{e \in \Delta_S} c_e$  and  $e' \in \Delta_S \Rightarrow c_{e^*} \leq c_{e'}$

which implies that  $c_{e^*} = c_{e'}$  (the two edges have the same cost)

The two spanning trees have equal cost: **the new spanning tree is optimal**



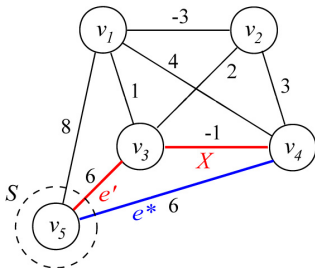
# Proof (3)

Given a partial optimal solution  $(V, X)$ , if we find a vertex set  $S \subset V$  whose induced cut  $\Delta_S$  does not intersect  $X$ , we can augment  $X$  obtaining a partial optimal solution  $(V, X \cup \{e^*\})$

*Sooner or later, we will obtain a complete optimal solution*

If  $\Delta_S \cap X \neq \emptyset$ , one cannot correctly enlarge set  $X$ :

either  $e^* \in X$  (and  $X$  does not grow) or  $e^*$  closes a cycle with  $X$  (and the new tree includes  $e^*$  and  $X \setminus \{e'\}$ , but not  $X$ )



# A general scheme (revisited)

- 1 Set  $X := \emptyset$  (to be included in an optimal solution)
- 2 Find a cut  $\Delta_S$  not intersecting  $X$ ; if there is none, terminate
- 3 Otherwise, set  $X := X \cup \arg \min_{e \in \Delta_S} c_e$  and go to step 2

The scheme works because

- $X$  is always included in an optimal solution (theorem)
- $X$  is augmented step by step (since  $\Delta_S$  does not intersect  $X$ )
- when every cut intersects  $X$ ,  $(V, X)$  is a spanning tree  
 $\Rightarrow$  in the end,  $(V, X)$  is an optimal spanning tree

Different algorithms apply this scheme

**Prim's** algorithm (1957)

- $S$  collects the extreme vertices of the edges of  $X$   
(necessarily,  $\Delta_S$  does not intersect  $X$ )  
at first  $S$  is a single vertex, chosen *ad libitum*
- $e^* := \arg \min_{e \in \Delta_S} c_e$

**Kruskal's** algorithm (1956)

- first find the minimum cost edge  $e^* := \arg \min_{e \in E \setminus X} c_e$
- if there is a cut including  $e$  and not intersecting  $X$ , add  $e^*$  to  $X$   
(i. e. if the extreme vertices of  $e^*$  are disconnected in  $X$ ),  
otherwise, remove  $e^*$  from  $E$

# Prim's algorithm

**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

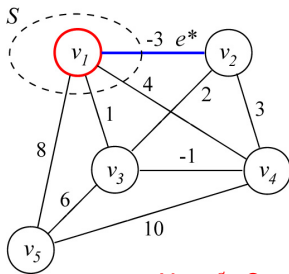
$e^* = [u^*, v^*] := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$       { One of the two extremes is already in  $S$  }

**Return**  $(S, X);$

# Application of Prim's algorithm (1)



**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

$e^* := (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**Return**  $(S, X);$

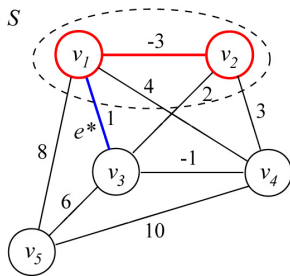
$X := \emptyset; S := \{v_1\};$

$e^* := (v_1, v_2);$

$X := \{(v_1, v_2)\};$

$S := \{v_1, v_2\};$

# Application of Prim's algorithm (2)



**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

$e^* := (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**Return**  $(S, X);$

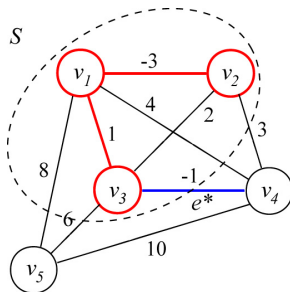
$X := \{(v_1, v_2)\}; S := \{v_1, v_2\};$

$e^* := (v_1, v_3);$

$X := \{(v_1, v_2), (v_1, v_3)\};$

$S := \{v_1, v_2, v_3\};$

# Application of Prim's algorithm (3)



**Prim**( $V, E, c$ )

$X := \emptyset$ ;  $S := \{\bar{v}\}$ ;

**While**  $S \subset V$

$e^* := (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e$ ;

$X := X \cup \{e^*\}$ ;

$S := S \cup \{u^*, v^*\}$ ;

**Return**  $(S, X)$ ;

$X := \{(v_1, v_2), (v_1, v_3)\}$ ;

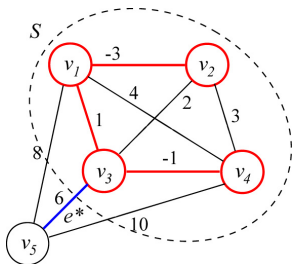
$S := \{v_1, v_2, v_3\}$ ;

$e^* := (v_3, v_4)$ ;

$X := \{(v_1, v_2), (v_1, v_3), (v_3, v_4)\}$ ;

$S := \{v_1, v_2, v_3, v_4\}$ ;

# Application of Prim's algorithm (4)



**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

$e^* := (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**Return**  $(S, X);$

$X := \{(v_1, v_2), (v_1, v_3), (v_3, v_4)\};$

$S := \{v_1, v_2, v_3, v_4\};$

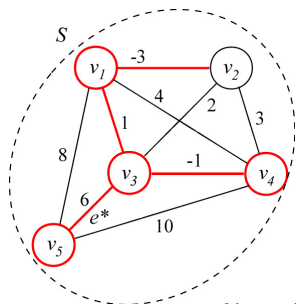
$e^* := (v_3, v_5);$

$X := \{(v_1, v_2), (v_1, v_3), (v_3, v_4), (v_3, v_5)\};$

$S := \{v_1, v_2, v_3, v_4, v_5\};$



# Application of Prim's algorithm (5)



**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

$e^* = (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**Return**  $(S, X);$

$X := \{(v_1, v_2), (v_1, v_3), (v_3, v_4), (v_3, v_5)\};$

$S := \{v_1, v_2, v_3, v_4, v_5\} = V;$

# Complexity of Prim's algorithm

Prim's algorithm consists of an **initial step** of complexity  $T_{\text{in}}$  and a certain number  $i_{\text{max}}$  of **iterations** of complexity  $T_{\text{iter}}^{(i)}$

$$T = T_{\text{in}} + \sum_{i=1}^{i_{\text{max}}} T_{\text{iter}}^{(i)}$$

**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

$e^* := (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**Return**  $(S, X);$

$T_{\text{in}} \in O(1)$

$i_{\text{max}} = n - 1$  (one vertex at a time)

$T_{\text{iter}}^{(i)} = \alpha$  (to be determined)

Overall  $T \in O(\alpha n)$

# Minimum cost edge identification (1)

Possible implementations

- 1 Scan all the edges and verify which ones belong to  $\Delta_S$ :  $O(m)$
- 2 Maintain subset  $\Delta_S$ 
  - build it:  $O(n)$
  - find the minimum cost element:  $O(m)$
  - update it:  $O(n)$
- 3 Maintain for each  $v \in V \setminus S$  the cheapest edge in  $\Delta_S \cap \Delta_{\{v\}}$

$$\tilde{e}_v = \arg \min_{[u,v] \in \Delta_S \cap \Delta_{\{v\}}} c_e$$

- build  $\tilde{e}_v$ :  $O(n)$
- find the minimum  $\tilde{e}_v$ :  $O(n)$
- update  $\tilde{e}_v$ :  $O(n)$

→ the complexity of the first implementation is  $T \in O(mn)$

**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**While**  $S \subset V$

$e^* = (u^*, v^*) := \arg \min_{e \in \Delta_S} c_e;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**Return**  $(S, X);$

# Minimum cost edge identification (2)

## Possible implementations

- 1 Scan all the edges and verify which ones belong to  $\Delta_S$ :  $O(m)$
- 2 Maintain subset  $\Delta_S$ 
  - build it:  $O(n)$
  - find the minimum cost element:  $O(m)$
  - update it:  $O(n)$
- 3 Maintain for each  $v \in V \setminus S$  the cheapest edge in  $\Delta_S \cap \Delta_{\{v\}}$

$$\tilde{e}_v = \arg \min_{[u,v] \in \Delta_S \cap \Delta_{\{v\}}} c_e$$

- build  $\tilde{e}_v$ :  $O(n)$
- find the minimum  $\tilde{e}_v$ :  $O(n)$
- update  $\tilde{e}_v$ :  $O(n)$

→ the complexity of the second implementation is  $T \in O(mn)$

**Prim**( $V, E, c$ )

$X := \emptyset$ ;  $S := \{\bar{v}\}$ ;  $D := \Delta_{\bar{v}}$ ;

**While**  $S \subset V$

$e^* = (u^*, v^*) := \arg \min_{e \in D} c_e$ ;

$X := X \cup \{e^*\}$ ;

$S := S \cup \{u^*, v^*\}$ ;

**For each**  $w \in S$

$D := D \setminus \{(w, v^*)\}$ ;

**For each**  $w \in V \setminus S$

$D := D \cup \{(w, v^*)\}$ ;

**Return**  $(S, X)$ ;

# Minimum cost edge identification (3)

## Possible implementations

- 1 Scan all the edges and verify which ones belong to  $\Delta_S$ :  $O(m)$
- 2 Maintain subset  $\Delta_S$ 
  - build it:  $O(n)$
  - find the minimum cost element:  $O(m)$
  - update it:  $O(n)$
- 3 Maintain for each  $v \in V \setminus S$  the cheapest edge in  $\Delta_S \cap \Delta_{\{v\}}$

$$\tilde{e}_v = \arg \min_{[u,v] \in \Delta_S \cap \Delta_{\{v\}}} c_e$$

- build  $\tilde{e}_v$ :  $O(n)$
- find the minimum  $\tilde{e}_v$ :  $O(n)$
- update  $\tilde{e}_v$ :  $O(n)$

→ the complexity of the third implementation is  $T \in O(n^2)$

**Prim**( $V, E, c$ )

$X := \emptyset; S := \{\bar{v}\};$

**For each**  $w \in V \setminus \{\bar{v}\}$

$\tilde{e}_w := [\bar{v}, w];$

**While**  $S \subset V$

$e^* = (u^*, v^*) := \arg \min_{w \in V \setminus S} \tilde{e}_w;$

$X := X \cup \{e^*\};$

$S := S \cup \{u^*, v^*\};$

**For each**  $w \in V \setminus S$

**If**  $c_{wv^*} < c_{\tilde{e}_w}$

**then**  $\tilde{e}_w := (w, v^*);$

**Return**  $(S, X);$

# Kruskal's algorithm

Start with  $X = \emptyset$

Find the minimum cost edge  $e^*$  not in  $X$  and not discarded

- if there is a cut  $\Delta_S$  including  $e^*$  and not intersecting  $X$  add  $e^*$  to  $X$

Notice that it is not required to determine  $S$ , because

$$\exists S \subset V : e^* \in \Delta_S \text{ and } \Delta_S \cap X = \emptyset \Leftrightarrow X \cup \{e^*\} \text{ is acyclic}$$

- if  $\nexists S$ , it will not exist for any larger  $X \Rightarrow$  discard  $e^*$  permanently

**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;  $S := V$ ;

$E' := E$ ;                    { Not yet discarded edges }

**While**  $E' \neq \emptyset$

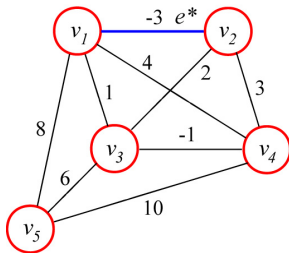
$e^* := \arg \min_{e \in E'} c_e$ ;

$E' := E' \setminus \{e^*\}$ ;

**If** Acyclic( $X \cup \{e^*\}$ ) **then**  $X := X \cup \{e^*\}$ ;

**Return** ( $S, X$ );

# Application of Kruskal's algorithm (1)



**Kruskal**( $V, E, c$ )

$X := \emptyset; S := V;$

$E' := E;$

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$E' := E' \setminus \{e^*\};$

**If**  $\text{Acyclic}(X \cup \{e^*\})$

**then**  $X := X \cup \{e^*\};$

**Return**  $(S, X);$

$X := \emptyset;$

$|E'| = 9;$

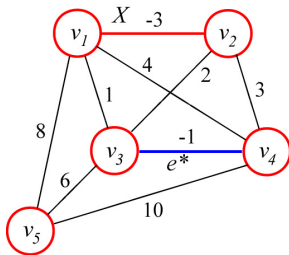
$e^* := (v_1, v_2);$

$|E'| = 8;$

$X \cup \{e^*\} := \{(v_1, v_2)\}$  acyclic

$X := \{(v_1, v_2)\};$

# Application of Kruskal's algorithm (2)



**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;  $S := V$ ;

$E' := E$ ;

$|E'| = 8$ ;

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$e^* := (v_3, v_4)$ ;

$E' := E' \setminus \{e^*\}$ ;

$|E'| = 7$ ;

**If**  $\text{Acyclic}(X \cup \{e^*\})$

$X \cup \{e^*\} := \{(v_1, v_2), (v_3, v_4)\}$  acyclic

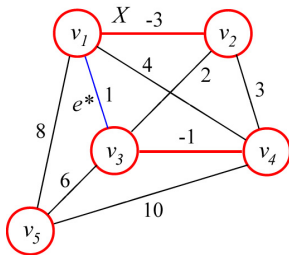
**then**  $X := X \cup \{e^*\}$ ;

$X := \{(v_1, v_2), (v_3, v_4)\}$ ;

**Return**  $(S, X)$ ;



# Application of Kruskal's algorithm (3)



**Kruskal**( $V, E, c$ )

$X := \emptyset; S := V;$

$E' := E;$

$|E'| = 7;$

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$e^* := (v_1, v_3);$

$E' := E' \setminus \{e^*\};$

$|E'| = 6;$

**If**  $\text{Acyclic}(X \cup \{e^*\})$

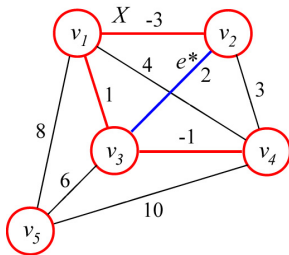
$X \cup \{e^*\}$  acyclic

**then**  $X := X \cup \{e^*\};$

$X := \{(v_1, v_2), (v_3, v_4), (v_1, v_3)\};$

**Return**  $(S, X);$

# Application of Kruskal's algorithm (4)



**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;  $S := V$ ;

$E' := E$ ;

$|E'| = 6$ ;

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$e^* := (v_2, v_3)$ ;

$E' := E' \setminus \{e^*\}$ ;

$|E'| = 5$ ;

**If**  $\text{Acyclic}(X \cup \{e^*\})$

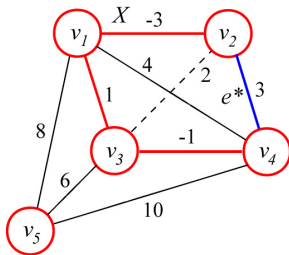
$X \cup \{e^*\}$  cyclic: **discard**  $e^*$

**then**  $X := X \cup \{e^*\}$ ;

$X := \{(v_1, v_2), (v_3, v_4), (v_1, v_3)\}$ ;

**Return**  $(S, X)$ ;

# Application of Kruskal's algorithm (5)



**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;  $S := V$ ;

$E' := E$ ;

$|E'| = 5$ ;

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$e^* := (v_2, v_4)$ ;

$E' := E' \setminus \{e^*\}$ ;

$|E'| = 4$ ;

**If**  $\text{Acyclic}(X \cup \{e^*\})$

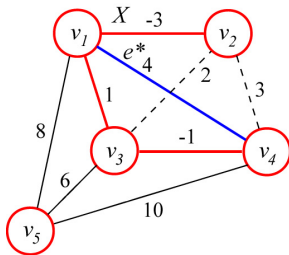
$X \cup \{e^*\}$  cyclic: **discard**  $e^*$

**then**  $X := X \cup \{e^*\}$ ;

$X := \{(v_1, v_2), (v_3, v_4), (v_1, v_3)\}$ ;

**Return**  $(S, X)$ ;

# Application of Kruskal's algorithm (6)



**Kruskal**( $V, E, c$ )

$X := \emptyset; S := V;$

$E' := E;$

$|E'| = 4;$

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$e^* := (v_1, v_4);$

$E' := E' \setminus \{e^*\};$

$E' = \{(v_1, v_5), (v_3, v_5), (v_4, v_5)\};$

**If**  $\text{Acyclic}(X \cup \{e^*\})$

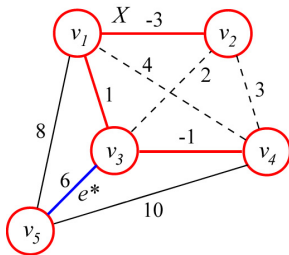
$X \cup \{e^*\}$  cyclic: **discard**  $e^*$

**then**  $X := X \cup \{e^*\};$

$X := \{(v_1, v_2), (v_3, v_4), (v_1, v_3)\};$

**Return**  $(S, X);$

# Application of Kruskal's algorithm (7)



**Kruskal**( $V, E, c$ )

$X := \emptyset; S := V;$

$E' := E;$

**While**  $E' \neq \emptyset$

$e^* := \arg \min_{e \in E'} c_e$

$E' := E' \setminus \{e^*\};$

**If**  $\text{Acyclic}(X \cup \{e^*\})$

**then**  $X := X \cup \{e^*\};$

**Return**  $(S, X);$

$E' = \{(v_1, v_5), (v_3, v_5), (v_4, v_5)\};$

$e^* := (v_3, v_5);$

$E' = \{(v_1, v_5), (v_4, v_5)\};$

$X \cup \{e^*\}$  acyclic

$X := \{(v_1, v_2), (v_3, v_4), (v_1, v_3), (v_3, v_5)\};$

# Anticipated termination

Given a tree, a **leaf** is a **vertex with a single incident arc**

- any acyclic graph with  $n > 1$  vertices includes at least one leaf

*Proof by contradiction: otherwise, the visit of the tree would never terminate...*

Consequently

- an acyclic graph with  $n$  vertices has  $m \leq n - 1$  edges

*Proof by induction*

- an acyclic graph with  $n = 1$  vertex has  $m = 0$  leaves
- a generic acyclic graph with  $n > 1$  vertices has a leaf; removing it produces an acyclic graph with  $n'$  vertices and  $m'$  edges (where  $n' = n - 1$  and  $m' = m - 1$ ); if for that graph  $m' \leq n' - 1 \Rightarrow m \leq n - 1$

Therefore **Kruskal's algorithm can terminate as soon as  $|X| = n - 1$**

# Complexity of Kruskal's algorithm

Kruskal's algorithm consists of an **initial step of complexity**  $T_{\text{in}}$  and a certain number  $i_{\text{max}}$  of **iterations of complexity**  $T_{\text{iter}}^{(i)}$

$$T = T_{\text{in}} + \sum_{i=1}^{i_{\text{max}}} T_{\text{iter}}^{(i)}$$

**Kruskal**( $V, E, c$ )

$X := \emptyset;$

$E' := E;$

**While**  $|X| < |V| - 1$

$e^* := \arg \min_{e \in E'} c_e$

$E' := E' \setminus \{e^*\};$

**If** **Acyclic**( $X \cup \{e^*\}$ )

**then**  $X := X \cup \{e^*\};$

**Return** ( $V, X$ );

$T_{\text{in}} \in O(1)$

$i_{\text{max}} \leq m$  (one edge at a time)

$T_{\text{iter}}^{(i)} \in O(\alpha + \beta)$

( $\alpha$  and  $\beta$  to be determined)

Overall  $T \in O((\alpha + \beta) m)$

# Minimum cost edge identification (1)

## Possible implementations

- 1 Scan all the nondiscarded edges:  
 $O(m)$
- 2 Sort  $E'$  by nondecreasing costs:
  - build it:  $O(m \log m)$
  - extract the minimum:  $O(1)$
- 3 Maintain  $E'$  as a *min-heap*
  - build it:  $O(m)$
  - extract the minimum:  $O(1)$
  - update it:  $O(\log m)$

**Kruskal**( $V, E, c$ )

$X := \emptyset;$

$E' := E;$

**While**  $|X| < |V| - 1$

$e^* := \arg \min_{e \in E'} c_e$

$E' := E' \setminus \{e^*\};$

**If**  $\text{Acyclic}(X \cup \{e^*\})$

**then**  $X := X \cup \{e^*\};$

**Return**  $(V, X);$

→ the complexity is  $T \in O(m^2 + m\beta)$



# Minimum cost edge identification (2)

## Possible implementations

- 1 Scan all the nondiscarded edges:  
 $O(m)$
- 2 Sort  $E'$  by nondecreasing costs:
  - build it:  $O(m \log m)$
  - extract the minimum:  $O(1)$
- 3 Maintain  $E'$  as a *min-heap*
  - build it:  $O(m)$
  - extract the minimum:  $O(1)$
  - update it:  $O(\log m)$

**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;

$E' := E$ ;

Sort( $E'$ );

**While**  $|X| < |V| - 1$

$e^* := \text{First}(E')$ ;

$E' := E' \setminus \{e^*\}$ ;

**If** Acyclic( $X \cup \{e^*\}$ )

**then**  $X := X \cup \{e^*\}$ ;

**Return** ( $V, X$ );

→ the complexity is  $T \in O(m \log m + m\beta)$

# Minimum cost edge identification (3)

Possible implementations

- 1 Scan all the nondiscarded edges:  
 $O(m)$
- 2 Sort  $E'$  by nondecreasing costs:
  - build it:  $O(m \log m)$
  - extract the minimum:  $O(1)$
- 3 Maintain  $E'$  as a *min-heap*:
  - build it:  $O(m)$
  - extract the minimum:  $O(1)$
  - update it:  $O(\log m)$

**Kruskal**( $V, E, c$ )

$X := \emptyset;$

$E' := E;$

**BuildMinHeap**( $E'$ );

**While**  $|X| < |V| - 1$

$e^* := \text{ExtractMinimum}(E');$

$E' := E' \setminus \{e^*\};$

**Heapify**( $E'$ );

**If** **Acyclic**( $X \cup \{e^*\}$ )

**then**  $X := X \cup \{e^*\};$

**Return** ( $V, X$ );

→ the complexity is  $T \in O(m \log m + m\beta)$

# Acyclicity test (1)

Possible implementations

- 1 Visit the graph from  $u^*$  and verify whether  $v^*$  can be reached:  $O(n)$
- 2 Maintain  $X$  as a *merge-find-set*
  - build it:  $O(n)$
  - find and compare the components of  $u^*$  and  $v^*$ :  $\approx O(1)$
  - merge the components:  $O(1)$

**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;

$E' := E$ ;

BuildMinHeap( $E'$ );

**While**  $|X| < |V| - 1$

$e^* := \text{ExtractMinimum}(E')$ ;

$E' := E' \setminus \{e^*\}$ ;

Heapify( $E'$ );

**If not** Reachable( $u^*, v^*, X$ )

**then**  $X := X \cup \{e^*\}$ ;

**Return** ( $V, X$ );

→ the complexity is  $T \in O(m \log m + mn)$

# Acyclicity test (2)

## Possible implementations

- 1 Visit the graph from  $u^*$  and verify whether  $v^*$  can be reached:  $O(n)$
- 2 Maintain  $X$  as a *merge-find-set*
  - build it:  $O(n)$
  - find and compare the components of  $u^*$  and  $v^*$ :  
 $\approx O(1)$
  - merge the components:  $O(1)$

**Kruskal**( $V, E, c$ )

$X := \emptyset$ ;

$\mathcal{C} := \text{BuildMFSet}(X)$ ;

$E' := E$ ;

**BuildMinHeap**( $E'$ );

**While**  $|X| < |V| - 1$

$e^* := \text{ExtractMinimum}(E')$ ;

$E' := E' \setminus \{e^*\}$ ;

**Heapify**( $E'$ );

**If** **DiffComponents**( $u^*, v^*, \mathcal{C}$ )

**then**  $X := X \cup \{e^*\}$ ;

**Merge**( $u^*, v^*, \mathcal{C}$ );

**Return** ( $V, X$ );

→ the complexity is  $T \in O(m \log m)$