

MA/AMA 514 :

Networks & Combinatorial Optimization

Today: §1.4, §2.1, §2.2

From last time :

Given a connected graph $G = (V, E)$
and length function $l: E \rightarrow \mathbb{R}$, want to
find minimum length spanning tree :

$$l(T) = \sum_{e \in T} l(e).$$

Thm 1.11 Suppose

- F is a greedy forest of G
- U is the vertex set of a conn. component of F
- $e \in \delta(U)$ has min. length in $\delta(U)$

then $F \cup \{e\}$ is a greedy forest of G .

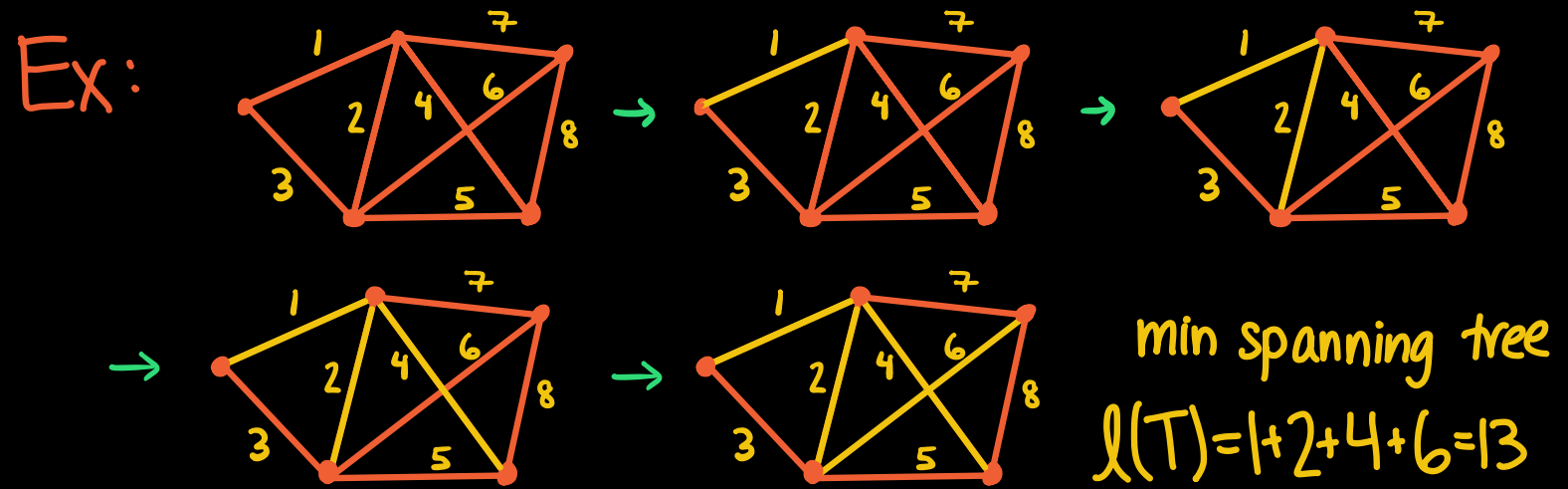
Kruskal's Algorithm (1956)

- 1) Sort edges s.t. $l(e_1) \leq l(e_2) \leq \dots \leq l(e_m)$ ($|E| = m$)
- 2) Set $F_0 = (V, \emptyset)$

3) For $i=1, \dots, m$, if $F_{i-1} \cup \{e_i\}$ acyclic, $F_i = F_{i-1} \cup \{e_i\}$
 otherwise $F_i = F_{i-1}$.

Output $T = F_m$.

Run time:
 $O(|E| \log(|E|))$
 sorting edges!



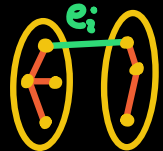
(Proof of correctness) Start F_0 is a greedy forest
 If $F_{i-1} \cup \{e_i\}$ is acyclic, e_i connects two connected components U, U' of $F_{i-1} \Rightarrow e_i \in \delta(U)$.

If $l(f) < l(e_i)$, end pts of f are connected in F_{i-1} .

$\Rightarrow e_i$ has min length among $\delta(U)$.

$\Rightarrow$ By Thm 1.1, F_i is a greedy forest.

Q: How do we implement step (3)?

One way is to keep track of the connected components of F_i . $F_{i-1} \cup \{e_i\}$ acyclic $\Leftrightarrow$ 
 endpts of e_i lie in different conn. comp.

Application 1.7: Maximum Reliability Problem

Want to design a network that maximizes "reliability"

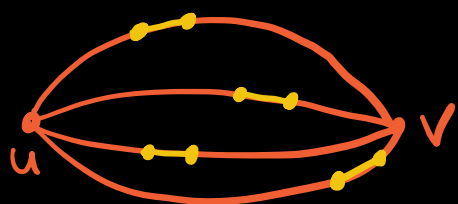
Input: Graph $G=(V,E)$, $s:E \rightarrow \mathbb{R}_+$

$s(e)$ = "strength" of edge

The reliability of a path P is $r(P) = \min_{e \in E(P)} s(e)$
(strength of weakest link in P) $\rightarrow$

For vertices $u, v \in V$, define

$r_G(u,v) = \max \{ r(P) : P \text{ is a path in } G \text{ from } u \text{ to } v \}$

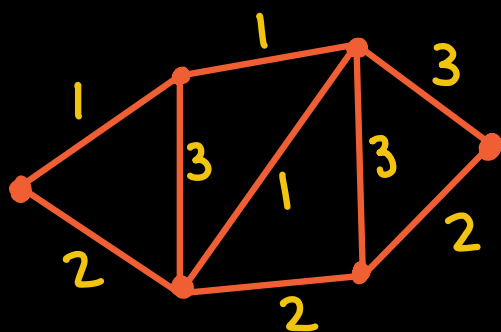


(Strongest weakest link!)

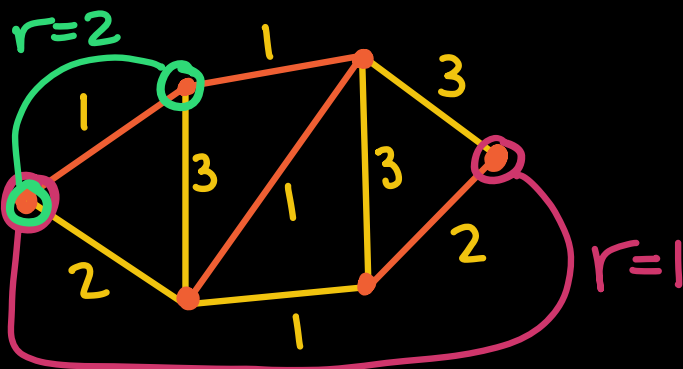
Let T be a maximum strength spanning tree
(minimum length with $l(e) = -s(e)$).

Claim: $r_G(u,v) = r_T(u,v)$ for all $u,v \in V$

Proof in
Hwk 1!

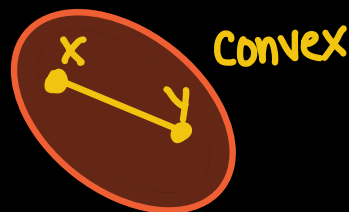


Kruskal
 $\rightarrow$



§2.1/§2.2 Convex Sets

A set $C \subseteq \mathbb{R}^n$ is convex if for every $x, y \in C$, $\lambda \in [0, 1]$, $\lambda x + (1 - \lambda)y \in C$. That is, C contains the line segment between any two points in C .



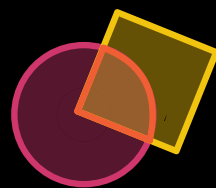
Remark: The intersection of convex sets is convex.

$C_i \subseteq \mathbb{R}^n$ convex for all $i \in I \Rightarrow \bigcap_{i \in I} C_i$ convex

$$x, y \in \bigcap_i C_i \Rightarrow x, y \in C_i \quad \forall i \in I$$

$$\Rightarrow \lambda x + (1 - \lambda)y \in C_i \quad \forall i \in I$$

$$\Rightarrow \lambda x + (1 - \lambda)y \in \bigcap_{i \in I} C_i$$



The convex hull of a set $X \subseteq \mathbb{R}^n$, denoted $\text{conv}(X)$ is the smallest convex set containing X , i.e.

$$\text{conv}(X) = \bigcap_{\substack{C \text{ convex} \\ X \subseteq C}} C$$



Alternatively,

$$\text{conv}(X) = \left\{ \sum_{i=1}^m \lambda_i x_i : x_1, \dots, x_m \in X, \lambda_i \geq 0, \sum_{i=1}^m \lambda_i = 1 \right\}$$

↖ a "convex combination" of $x_1, \dots, x_m$

The convex hull of a finite set is a polytope

Ex 1: $X = \{(0,0), (1,0), (0,1)\}$



$\Rightarrow \text{conv}(X) = \{(x,y) \in \mathbb{R}^2 : x \geq 0, y \geq 0, 1 \geq x+y\}$

$\hookrightarrow = \underbrace{(1-x-y)}_{\text{coeff}}(0,0) + \underbrace{x}_{\text{coeff}}(1,0) + \underbrace{y}_{\text{coeff}}(0,1)$

coeff are ≥ 0 and sum to 1

Ex 2: $X = \{(0,0,0)\} \cup \{(x,y,1) : x^2+y^2=1\}$



$\text{conv}(X) = \{(x,y,z) \in \mathbb{R}^3 : x^2+y^2 \leq z^2, 0 \leq z \leq 1\}$

Claim: For $X \subseteq \mathbb{R}^n$, $c \in \mathbb{R}^n$,

$\max\{c^T x : x \in X\} = \max\{c^T y : y \in \text{conv}(X)\}$

(Proof) " $\leq$ " follows from $X \subseteq \text{conv}(X)$.

For " $\geq$ ", let $c_0 = \max\{c^T x : x \in X\}$ and $y \in \text{conv}(X)$

$\Rightarrow y = \sum_{i=1}^m \lambda_i x_i$ for some $x_1, \dots, x_m \in X$, $\lambda_i \geq 0$ s.t. $\sum \lambda_i = 1$

$\Rightarrow c^T y = \sum_{i=1}^m \lambda_i c^T x_i \leq \sum_{i=1}^m \lambda_i c_0 = c_0$

↑
linearity

↑
 $\lambda_i \geq 0 \forall i$

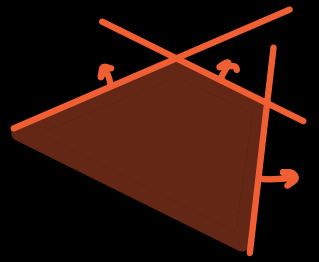
↑
 $\sum \lambda_i = 1$

Ex/Def: a closed halfspace is a set of the form $\{x \in \mathbb{R}^n : a^T x \leq b\}$ where $a \in \mathbb{R}^n$, $b \in \mathbb{R}$



Convexity: $a^T x \leq b, a^T y \leq b \Rightarrow \lambda a^T x \leq \lambda b$
 $\quad \quad \quad (1-\lambda) a^T y \leq (1-\lambda) b$
 $\Rightarrow a^T (\lambda x + (1-\lambda) y) \leq b$

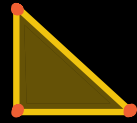
A polyhedron is the intersection of finitely-many halfspaces, i.e.



$$P = \{x \in \mathbb{R}^n : a_1^T x \leq b_1, \dots, a_m^T x \leq b_m\}$$

where $a_1, \dots, a_m \in \mathbb{R}^n$, $b_1, \dots, b_m \in \mathbb{R}$

Ex: $\{(x, y) \in \mathbb{R}^2 : x \geq 0, y \geq 0, 1 \geq x + y\}$



Thm 2.4 Every polytope is a bounded polyhedron.

A set $X \subseteq \mathbb{R}^n$ is bounded if X is contained in some ball, i.e. $\exists R \in \mathbb{R}_{>0}$ s.t. $\|x\|_2 \leq R$ for all $x \in X$.

You check: If X is bounded, so is $\text{conv}(X)$.

Thm: Every bounded polyhedron is a polytope. ^(Cor. 2.3a)

We will translate comb. opt. problems to the form $\max\{c^T x : x \in X\}$ with X finite and then replace X with $\text{conv}(X)$.

Teaser for the future:

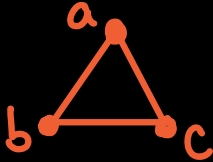
Spanning trees as a polytope

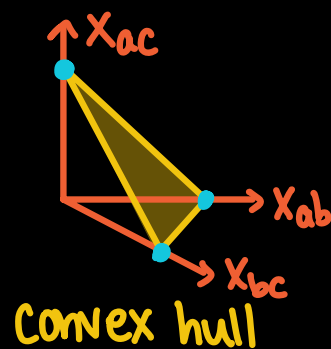
$G = (V, E)$ connected graph, $\ell: E \rightarrow \mathbb{R}$

For $T \subseteq E$, define $\mathbb{1}_T \subseteq \mathbb{R}^E$ by

$$(\mathbb{1}_T)_e = \begin{cases} 1 & \text{if } e \in T \\ 0 & \text{if } e \notin T \end{cases}$$

Then $\text{conv} \{ \mathbb{1}_T : T \text{ is a span. tree of } G \} = P_G$
is a polytope.

Ex:  $T = \{ab, ac\} \rightarrow \mathbb{1}_T = (\overset{ab}{1}, \overset{ac}{1}, \overset{bc}{0})$
 $T = \{ab, bc\} \rightarrow \mathbb{1}_T = (1, 0, 1)$
 $T = \{ac, bc\} \rightarrow \mathbb{1}_T = (0, 1, 1)$



Claim: length of min sp. tree

$$= \min \left\{ \sum_{e \in E} l(e) x_e : x \in P_G \right\}$$

