

9 Oct 2024

Max Flow

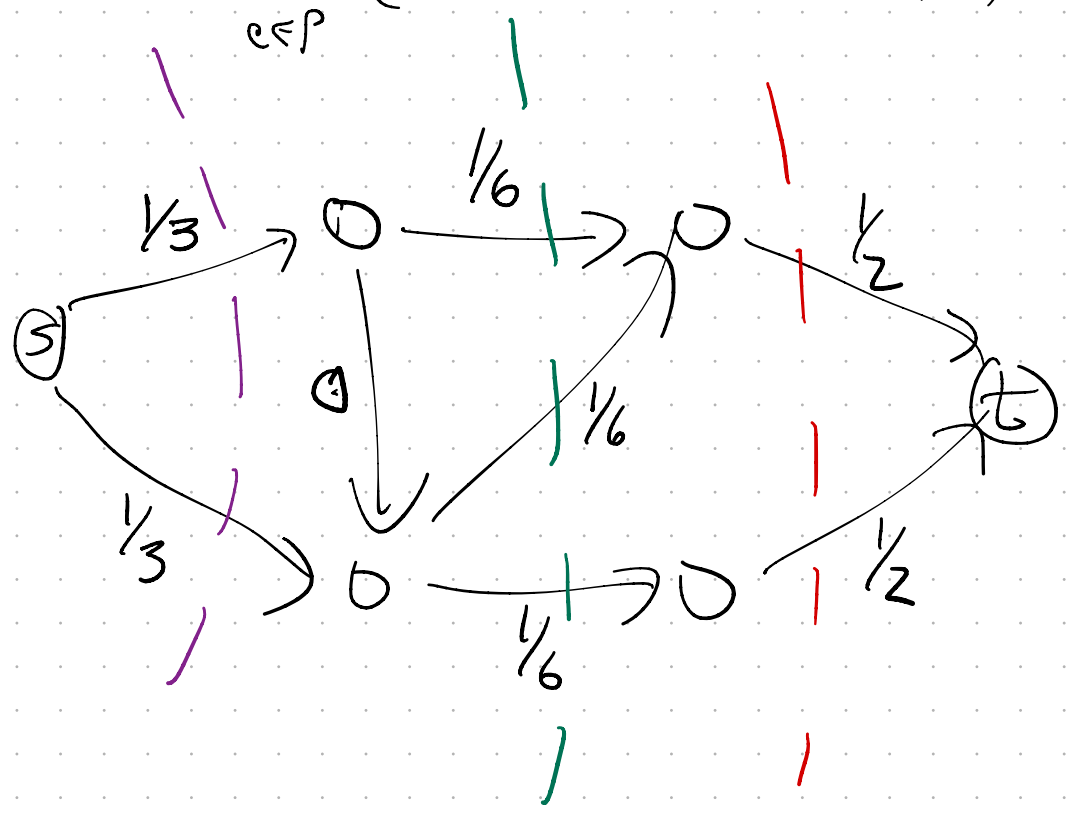
RECALL: Exact s-t cut set = $\{(u,v) \mid u \in A, v \in B\}$
 where $s \in A, t \in B$

s-t cut set = superset of exact s-t cut set

fractional s-t cut set = $x \in \mathbb{R}^E$ s.t.

$$x_e \geq 0 \quad \forall e,$$

$$\sum_{e \in P} x_e \geq 1 \quad \forall P \in \mathcal{P}(s,t)$$



This fractional s-t cut is

$$\frac{1}{3} \left(\text{cut } 1 \right) + \frac{1}{6} \left(\text{cut } 2 \right) + \frac{1}{2} \left(\text{cut } 3 \right)$$

LEMMA (last time): Whenever $\vec{x} \in \mathbb{R}^E$ is a fractional s-t cut set, there is a convex combination of exact s-t cut sets, x^i , such that $x^i_e \leq x_e \quad \forall e$.

This shows...

Cor. For any vector $\vec{c} \in \mathbb{R}_{\geq 0}^E$ ("capacities")

$$\begin{aligned} \min & \sum_e c(e) x_e \\ \text{s.t.} & \sum_{e \in P} x_e \geq 1 \quad \forall P \in \mathcal{P}(s,t) \\ & x_e \geq 0 \end{aligned}$$

the minimum is attained at a $\{0,1\}^E$ vector corresponding to an exact s-t cut.

Proof. Suppose $\vec{x} \in \mathbb{R}^E$ is a vector that attains the minimum value.

Using lemma write $\vec{x} \succeq \vec{x}'$ and

$$x'_e = \sum_{i=1}^M w_i x_{\text{cut}(A_i, B_i)} \leftarrow \begin{array}{l} \{0,1\} \text{ vector} \\ \text{corresponding} \\ \text{to s-t cut} \\ (A_i, B_i). \end{array}$$

where $w_i \geq 0 \quad \forall i, \quad \sum_i w_i = 1$.

Each $\vec{x}_{\text{cut}(A_i, B_i)}$ is feasible for the LP above. By def'n of \vec{x}

$$\forall i \quad \sum_e c(e) x_{\text{cut}(A_i, B_i), e} \geq \sum_e c(e) x_e$$

$$\sum_e \sum_i w_i c(e) x_{\text{cut}(A_i, B_i), e} \geq \sum_e c(e) x_e$$

$$\sum_e c(e) x'_e \geq \sum_e c(e) x_e$$

In the other hand

$$\forall e \quad x_e \geq x'_e \quad \text{and} \quad c(e) \geq 0$$

$$\Rightarrow c(e) x_e \geq c(e) x'_e$$

$$\sum_e c(e) x_e \geq \sum_e c(e) x'_e$$

If any $x_{\text{cut}(A_i, B_i)}$ had greater obj. value than x , that inequality would have been strict and $\text{OBJ}(x') > \text{OBJ}(x)$ would not hold.

\therefore each $x_{\text{cut}(A_i, B_i)}$ attains the LP min.

Strong duality:

$$\min \sum_e c(e) x_e$$

$$= c^T x$$

$$c = \begin{bmatrix} c(e_1) \\ c(e_2) \\ \vdots \\ c(e_n) \end{bmatrix}$$

st

$$\sum_{e \in P} x_e \geq 1$$

$$\forall P \in \mathcal{P}(st) \Rightarrow Ax \geq \vec{1}$$

$$x_e \geq 0 \quad \forall e \in E$$

$$A = \left. \begin{matrix} P_1 \\ \vdots \\ P_k \end{matrix} \right\} \text{Paths}$$

Edges

$x_{e \in P}$

$$\max \sum_{P \in \mathcal{P}(s,t)} y_P$$

$$\text{s.t.} \quad \sum_{P: e \in P} y_P \leq c(e) \quad \forall e \in E$$

$$y_P \geq 0 \quad \forall P \in \mathcal{P}(s,t)$$

This is a path-packing problem called "maximum flow."

$$\begin{aligned} \max(\text{flow LP}) &\stackrel{\text{strong duality}}{=} \min(\text{frac st cut set}) \\ &\stackrel{\text{Corollary of Monday's lemma}}{=} \min(\text{capacity of exact st cut set}) \end{aligned}$$

Flows are usually represented more succinctly as functions mapping edges (u,v) to numbers, $f(u,v)$.

For directed graph $G = (V, E)$, let

$$\vec{E} = E \cup \{ (v,u) \mid (u,v) \in E \}$$

Def. An st flow is a function

$$f: \vec{E} \rightarrow \mathbb{R} \quad \text{satisfying}$$

- skew symmetry $f(v,u) = -f(u,v)$

- flow conservation

$$\forall u \neq s, t \quad \sum_v f(u, v) = 0$$

A flow is feasible for capacity function

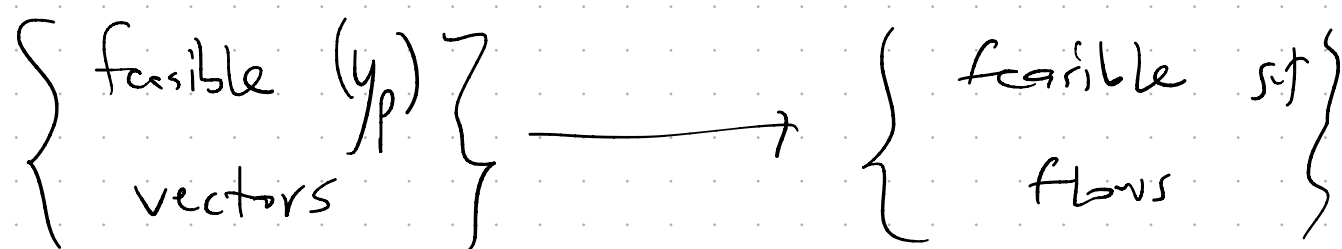
$$c: \vec{E} \rightarrow \mathbb{R} \quad \text{if} \quad f(u, v) \leq c(u, v) \quad \forall (u, v) \in \vec{E}.$$

The value of a flow f is

$$\text{val}(f) = \sum_v f(s, v).$$

The maximum flow problem is: given $G = (V, E)$
and $c: \vec{E} \rightarrow \mathbb{R}_{\geq 0}$ find a
feasible flow of maximum value.

Relation to $\max \left\{ \sum y_p \mid \sum_{p: e \in P} y_p \leq c(e) \quad \forall e, y_p \geq 0 \right\}$.



y \longmapsto f def'd by

* not surjective!

$$f(u, v) = \sum_{p: (u, v) \in P} y_p - \sum_{p: (v, u) \in P} y_p.$$

$$\sum_v f(u, v) = 0 \quad \text{when} \quad u \notin \{s, t\}.$$

$$\sum_v \sum_{p: (u, v) \in P} y_p \leftarrow \sum_v \sum_{p: (v, u) \in P} y_p$$