

Integer Multicommodity Flow :

Input: - $G = (V, E)$
- $(s_1, t_1), (s_2, t_2), \dots, (s_k, t_k)$

Feasible: Paths $p_1, p_2, \dots, p_k$ s.t. $p_i \in P_i$

P_i : all $s_i \rightarrow t_i$ paths in G

Objective: $\min \max_{e \in E} \text{cong}(e)$

$$\text{cong}(e) = |\{i \in [k] : e \in p_i\}|$$

LP Relaxation:

$\min \quad W$

$$\text{s.t.} \quad \sum_{p \in P_i} x_p = 1 \quad \forall i \in [k]$$

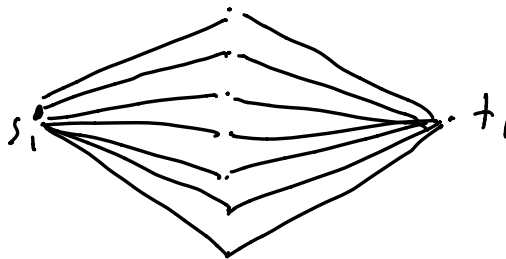
$$\sum_{i=1}^k \sum_{p \in P_i : e \in p} x_p \leq W \quad \forall e \in E$$

$$0 \leq x_p \leq 1 \quad \forall i \in [k], \forall p \in P_i$$

Problem : Too many variables !

Problem : Integrality gap $\geq n-2$

PF:



$$\text{OPT} = 1$$

LP:

$$x_p = \frac{1}{n-2} \quad \forall p$$

$$W = \frac{1}{n-2}$$

Algorithm: [Raghavan - Thompson '87]

- Solve LP to get (x^*, W^*)
- For each $i \in [k]$,
 - $\{x_p^*\}_{p \in P_i}$ form a distribution
 - Choose p_i from this distribution

Thm: With high probability, $ALG \leq O(\log n) \cdot OPT$

not LP!

Note: can improve to $O(\frac{\log n}{\log \log n})$

Def: Random variables:

$$- Z_p = \begin{cases} 1 & \text{if path } p \text{ chosen} \\ 0 & \text{otherwise} \end{cases}$$

$$- Y_e^i = \begin{cases} 1 & \text{if } p_i \text{ (path chosen for } i) \text{ uses } e \\ 0 & \text{otherwise} \end{cases}$$

$$- Y_e = \sum_{i=1}^k Y_e^i = \# \text{ chosen paths using } e = \deg(e)$$

Lemma: $E[Y_e] \leq w^* \quad \forall e \in E$

Pf:

$$E[Y_e] = E\left[\sum_{i=1}^k Y_e^i\right]$$

$$= E\left[\sum_{i=1}^k \sum_{p \in P_i: e \in p} Z_p\right]$$

$$(Y_e^i = \sum_{p \in P_i: e \in p} Z_p)$$

$$= \sum_{i=1}^k \sum_{p \in P_i: e \in p} E[Z_p]$$

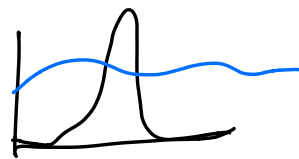
$$= \sum_{i=1}^k \sum_{p \in P_i: c \in p} x_p^+ \quad (E[Z_p] = x_p^+)$$

$$\leq W^* \quad (LP \text{ constraint})$$

Not done! Prove $\max_{c \in E} E[Y_c] \leq W^*$

Not the same as $E[\max_{c \in E} Y_c]$

Need Concentration of Measure: close to expectation with high probability



Thm [Chernoff Bounds]:

Let $X_1, \dots, X_n$ be independent random variables distributed over $[0, 1]$ (not necessarily uniformly).

Let $X = \sum_{i=1}^n X_i$. Note: $E[X] = \sum_{i=1}^n E[X_i]$

$$- \forall 0 < \epsilon < 1: \Pr[X > (1+\epsilon)E[X]] \leq e^{-\frac{\epsilon^2}{3} E[X]}$$

$$\Pr[X < (1-\epsilon)E[X]] \leq e^{-\frac{\epsilon^2}{2} E[X]}$$

$$\forall t > 2e \mathbb{E}[X], \Pr[X > t] \leq 2^{-t}$$

Use Chernoff to analyze algorithm:

Claim: $\Pr[Y_e > 3 \log n \cdot \max(w^*, 1)] \leq \frac{1}{n^3}$

pf:

$$3 \log n \cdot \max(w^*, 1) > 2e \cdot w^* \geq 2e \mathbb{E}[Y_e]$$

$$\uparrow$$

$$n \geq 4 \Rightarrow 3 \log n > 2e$$

Apply Chernoff:

$$\Pr[Y_e > 3 \log n \cdot \max(w^*, 1)]$$

$$\leq 2^{-3 \log n \cdot \max(w^*, 1)}$$

$$\leq 2^{-3 \log n}$$

$$= \frac{1}{n^3}$$



Union bound over edges:

$$P\left[\max_{e \in E} Y_e > 3 \log n \cdot \underbrace{\max(w^*, 1)}_{\leq OPT}\right] \leq n^2 \cdot \frac{1}{n^3} = \frac{1}{n}$$

$\Rightarrow$ With high probability, $\max_{e \in E} \text{cost}(e) \leq O(\log n) \cdot OPT$

Extensions:

- Better version of Chernoff: $O\left(\frac{\log n}{\log \log n}\right)$

- If w^* larger, can do better!

Ex: If $w^* \geq \frac{18e}{\epsilon^2} \ln n$

If $E[Y_e] \geq \frac{q}{\epsilon^2} \ln n$

$$\Rightarrow P[Y_e > (4 + \epsilon) E[Y_e]] \leq e^{-\frac{\epsilon^2}{3} E[Y_e]} \leq e^{-3 \ln n} = \frac{1}{n^3}$$

If $E[Y_e] < \frac{q}{\epsilon^2} \ln n$

$$\Rightarrow w^* \geq 2e E[Y_e]$$

$$\Rightarrow P[\sum Y_e > W^*] \leq 2^{-W^*} \leq \frac{1}{n^3}$$

$\Rightarrow (1+\epsilon)$ -approx w.h.p.

Solving the LP:

LP intuition: flow along paths

Main idea:

solve equivalent compact representation (polynomial size)

x_{uv}^i : flow from $s_i \rightarrow t_i$ using edge $u \rightarrow v$

min W

$$\text{s.t.} \quad \sum_{v \in N(u)} x_{uv}^i - \sum_{v \in N(u)} x_{vu}^i = 0 \quad \forall i \in [k], \forall u \notin \{s_i, t_i\}$$

$$\sum_{v \in N(s_i)} x_{s_i v}^i - \sum_{v \in N(s_i)} x_{vs_i}^i = 1 \quad \forall i \in [k]$$

$$\sum_{i=1}^k (x_{uv}^i + x_{vu}^i) \leq W \quad \forall \{u, v\} \in E$$

$$0 \leq x_{uv}^i \leq 1$$

$$\forall i \in C(k), \forall \{u, v\} \in E$$

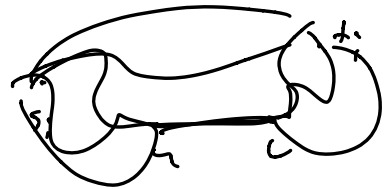
Poly size!

Lemma: Let $(\{x_p\}, w)$ be solution to original LP.

There is a solution to compact LP with same cost w .

Pf Sketch:

$$\text{Let } x_{uv}^i = \sum_{p \in P: \{u, v\} \in p} x_p.$$



$\Rightarrow$ All constraints satisfied $\Rightarrow$ feasible solution

Lemma: Let $(\{x_{uv}^i\}, w)$ be a solution to compact LP. Then there is a solution to original LP of same cost, which can be found in polynomial time

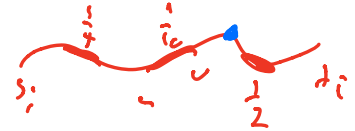
Pf Sketch:

"Flow-Path decomposition"

for $i = 1$ to k {

while $\exists P \in P_i$ s.t. $x_{uv}^i > 0 \ \forall (u,v) \in P$ {

- set $x_p = \min_{(u,v) \in P} x_{uv}^i$,



- set $x_{uv}^i = x_{uv}^i - x_p \ \forall (u,v) \in P$

}

}

(correctness):

induction $\Rightarrow$ flow in = flow out always true,

$(s_i \rightarrow t_i \text{ flow in } x_p's) + (s_i \rightarrow t_i \text{ flow in } x_{uv}^i's) = 1$

Running time:

Each time find path ≥ 1 edge gets $x_{uv}^i = 0$

$\Rightarrow \leq K \cdot n^2$ iterations

Each iteration just connectivity $\Rightarrow$ poly time