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Multi-commodity Flow

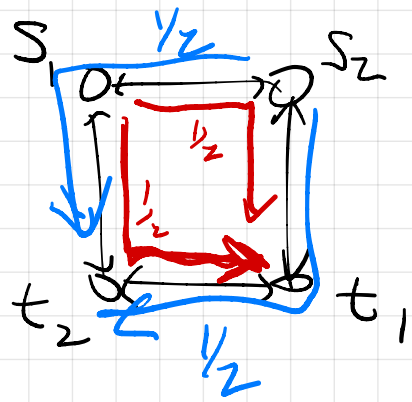
For today, all edges have capacity 1,

Problem: Given $G = (V, E)$, along with pairs $(s_1, t_1), \dots, (s_k, t_k)$,

Pack as many (fractional) paths into G as possible, given each path P must belong to $\mathcal{P} = \{ \text{paths with endpoints } s_i, t_i \}$ for some i

and each edge used by ≤ 1 path in total.

Ex.



(PRIMAL)

$$\max \sum_{P \in \mathcal{P}} x_P$$

$$\text{s.t.} \sum_{P: e \in P} x_P \leq 1 \quad \forall e$$

$$x_P \geq 0$$

(DUAL)

$$\min \sum_{e \in E} y_e$$

$$\text{s.t.} \sum_{P \in \mathcal{P}} y_e \geq 1 \quad \forall P \in \mathcal{P}$$

$$y_e \geq 0$$

Game theoretic interpretation:

- Kicker chooses a src-sink pair s_i, t_i and a path P from s_i to t_i
 - Goalie chooses $e \in E$, wins if $e \in P$.
- Simultaneously*

Algorithm

1. Initialize $w_e^1 = 1 \quad \forall \text{ edge } e$ // w_e^t is graph's un-normalized weights at start of round t .
2. for $t = 1, \dots, T$:

choose P_t to minimize $\sum_{e \in P_t} w_e^t$

for all $e \in E$:

$$w_e^{t+1} = w_e^t \cdot \begin{cases} (1+\epsilon) & \text{if } e \in P_t \\ 1 & \text{if } e \notin P_t \end{cases}$$

endfor

endfor

output $f_{\text{ALG}} = \frac{1}{T} \sum_{t=1}^T f^{P_t}$ ← elementary flow sending one unit on P_t .

... scaled as high as possible without overloading edges.

Analysis. Let $W^t = \sum_{e \in E} w_e^t$ for all t .

$$\frac{1}{T} \sum_{t=1}^T \sum_{e \in E} \underbrace{\left(\frac{w_e^t}{W_t} \right)}_{\text{Graph exp. payoff @ } t} \mathbb{1}[e \in P_t] \geq (1-\epsilon) \max_{e \in E} \left\{ \underbrace{\frac{1}{T} \sum_{t=1}^T \mathbb{1}[e \in P_t]}_{f_{\text{ALG}}(e)} \right\} - O\left(\frac{\ln m}{\epsilon T}\right)$$

(by performance guarantee of MW algorithm)

$$\max_{e \in E} \left\{ f_{\text{ALG}}(e) \right\} \leq \frac{1}{(1-\epsilon)T} \sum_{t=1}^T \left[\sum_{e \in E} \left(\frac{w_e^t}{W_t} \right) \mathbb{1}[e \in P_t] \right], \quad O\left(\frac{\ln m}{(1-\epsilon)\epsilon T}\right)$$

← P_t was chosen to minimize this sum.

If $f_{\text{OPT}} = \frac{\text{optimum MCF}}{\text{val}(\text{optimum MCF})}$ then

$$v^* :=$$

$$\forall e \quad f_{\text{opt}}(e) \leq \frac{1}{v^*}$$

then

$$\forall t \quad \sum_{e \in E} \left(\frac{w_e^t}{W_t} \right) \mathbb{1}[e \in P_t]$$

$$\leq \sum_{e \in E} \left(\frac{w_e^t}{W_t} \right) f_{\text{OPT}}(e)$$

... by greedy property of P_t .

$$\max_{e \in E} \left\{ f_{\text{ALG}}(e) \right\} \leq \frac{1}{(1-\epsilon)T} \sum_{t=1}^T \sum_{e \in E} \left(\frac{w_e^t}{W_t} \right) f_{\text{OPT}}(e) + O\left(\frac{\ln m}{(1-\epsilon)\epsilon T} \right)$$

$$\leq \frac{1}{(1-\epsilon)T} \sum_{t=1}^T \sum_{e \in E} \left(\frac{w_e^t}{W_t} \right) \frac{1}{v^*} + O\left(\frac{\ln m}{(1-\epsilon)\epsilon T} \right)$$

$$= \frac{1}{(1-\epsilon)T v^*} \sum_{t=1}^T \left[\sum_{e \in E} \left(\frac{w_e^t}{W_t} \right) \right] + O\left(\frac{\ln m}{(1-\epsilon)\epsilon T} \right)$$

$$= \frac{1}{(1-\epsilon)T v^*} + O\left(\frac{\ln m}{(1-\epsilon)\epsilon T} \right) \leq \frac{1}{(1-2\epsilon) v^*}$$

for large enough T .