Maximum Flow Problems III.

Review: G = (V, E); (s, t)-cut $\delta(A)$; edge capacity $u : E \to \mathbb{R}_{\geq 0}$; (s, t)-flow $x : E \to \mathbb{R}_{\geq 0}$ Goal: <u>Maximum Flow Problem</u>

Maximize
$$f_x(t) = \sum_{\substack{w \in V \\ wt \in E}} x_{wt} - \sum_{\substack{w \in V \\ tw \in E}} x_{tw}$$

subject to $f_x(v) = \sum_{\substack{w \in V \\ wv \in E}} x_{wv} - \sum_{\substack{w \in V \\ vw \in E}} x_{vw} = 0 \qquad \forall v \in V \setminus \{s, t\}$
$$0 \le x_e \le u_e \qquad \forall e \in E$$

1 Cuts

 $\text{Recall } \delta(A) = \{ vw \mid v \in A, w \in \overline{A} \}, \text{ and } (s,t) \text{-cut } \delta(A) \text{ if } s \in A \text{ and } t \in \overline{A} = V \setminus A \}$

Theorem 1. Every (s,t)-cut $\delta(A)$ and every (s,t)-flow x satisfy:

$$\underbrace{\sum_{e \in \delta(A)} x_e - \sum_{e \in \delta(\overline{A})} x_e}_{\text{flow across the cut } \delta(A)} = f_x(t)$$
denoted by $x(\delta(A))$

Proof. x is a flow $\Rightarrow f_x(v) = 0$ for all $v \in V \setminus \{s, t\}$. Summing up over all $v \in \overline{A} \setminus \{t\}$:

$$\sum_{v \in \overline{A}} f_x(v) = f_x(t) + \underbrace{\sum_{v \in \overline{A} \setminus \{t\}}^{=0} f_x(v)}_{v \in \overline{A} \setminus \{t\}} f_x(v) = f_x(t) \qquad \text{Recall: } f_x(v) = \underbrace{\sum_{w \in V}^{\text{incoming}} x_{wv}}_{wv \in E} - \underbrace{\sum_{w \in V}^{\text{incoming}} x_{vw}}_{vw \in E} + x_{vw} \text{ contribution to } f_x(w) - x_{vw} \text{ contribution to } f_x(v)$$

Contribution of vw to the left-hand-side (LHS)

$$\begin{array}{ll} v \in A & w \in A & \text{none because LHS sums-up } f_x(v) \text{ for } v \in A \\ v \in A & w \in \overline{A} & +x_{vw} \text{ from } f_x(w) & e = vw \in \delta(A) \\ v \in \overline{A} & w \in A & -x_{vw} \text{ from } f_x(v) & e = vw \in \delta(\overline{A}) \\ v \in \overline{A} & w \in \overline{A} & +x_{vw} + -x_{vw} \\ & & from f_x(v) & from f_x(w) & \Box \end{array}$$

Corollary 1. Every (s,t)-cut $\delta(A)$ and every feasible (s,t)-flow x satisfy:

$$f_x(t) \le \sum_{e \in \delta(A)} u_e$$

Proof. x is a feasible flow $\Rightarrow 0 \le x_e \le u_e$ for $\forall e \in E$. Thus, by Theorem 1 \Rightarrow

$$f_x(t) = \sum_{e \in \delta(A)} \underbrace{x_e}_{\leq u_e} - \sum_{e \in \delta(\overline{A})} \underbrace{x_e}_{\geq 0} \leq \underbrace{\sum_{e \in \delta(A)} u_e}_{capacity of the cut \delta(A)}$$
denoted by $u(\delta(A))$

 $(v_0,...,v_6)$ is an undirected path $v_0v_1,\,v_2v_3,\,v_3v_4$ "forward" edges $v_1v_2,\,v_4v_5,\,v_5v_6$ "backward" edges

 V_0 V_2 V_3 V_4 V_5 V_6 (V_2, V_3, V_4) is a directed path

i.e., the value of a feasible flow is at most the capacity of a cut.

Theorem 2. (Max-Flow Min-Cut Theorem) [Ford-Fulkerson 1956], [Kotzig 1956] The maximum value of a feasible (s,t)-flow is equal to the minimum capacity of an (s,t)cut. If all capacities are integral, then there exists an integral maximum feasible flow.

2 Flow augmentation

Let x be a feasible (s, t)-flow of value k (for instance, x = 0 is a feasible flow of value 0) For an st-path $P = (v_0, \ldots, v_m)$ define $\underline{x \text{-width of } P} = \min_{i \in \{1...m\}} u_{v_{i-1}v_i} - x_{v_{i-1}v_i}$ If P is a path of x-width $\varepsilon > 0$, then $\forall i$ increase $x_{v_{i-1}v_i}$ by $\varepsilon \Rightarrow$ a feasible flow of value $k + \varepsilon$

(... just like in the proof of the flow-paths theorem...)

we may get stuck before reaching the maximum flow \Rightarrow need to allow more general paths

Idea: use also backward edges

directed path = path (as defined before) undirected path = a sequence (v_0, \ldots, v_m) where v_i distinct and for all $i \in \{1 \ldots m\}$ either $v_{i-1}v_i \in E$ ("forward" edge) or $v_iv_{i-1} \in E$ ("backward" edge)

 $\underline{x \text{-width}} \text{ of an undirected path } (v_0, \dots, v_m) = \min_{i \in \{1,\dots,m\}} \begin{cases} u_{v_{i-1}v_i} - x_{v_{i-1}v_i} & \text{if } v_{i-1}v_i \in E \\ x_{v_iv_{i-1}} & \text{if } v_iv_{i-1} \in E \end{cases}$ $\underline{x \text{-increasing path}} = \text{undirected path of positive } x \text{-width}$ $\overline{x \text{-augmenting path}} = x \text{-increasing path from } s \text{ to } t$

If $P = (v_0, \ldots, v_m)$ is an x-augmenting path of width $\varepsilon > 0$, then $\forall i \in \{1 \ldots m\}$ if $v_{i-1}v_i \in E$, increase $x_{v_{i-1}v_i}$ by ε , $v_iv_{i-1} \in E$, decrease $x_{v_iv_{i-1}}$ by ε . $\} \Rightarrow$ a feasible flow of value $k + \varepsilon$

No x-augmenting path \Rightarrow maximum flow (we now prove)

<u>Proof of Max-Flow Min-Cut Theorem.</u> Let x be a feasible flow of maximum falue. Let $U = \{z \mid \exists \text{ an } x \text{-increasing path from } s \text{ to } z\}$. Note that $s \in U$. If $t \in U$, then \exists an x-augmenting path $\Rightarrow x$ is not maximum flow, a contradiction.

So
$$s \in U$$
 and $t \in \overline{U} \Rightarrow \delta(U)$ is an (s, t) -cut. Moreover,
- every $e = vw \in \delta(U)$ satisfies $u_e - x_e = 0$, otherwise $w \in U$.
- every $e = vw \in \delta(\overline{U})$ satisfies $x_e = 0$, otherwise $v \in U$.

$$\int_{value of x} f_x(t) = \sum_{e \in \delta(U)} x_e - \sum_{e \in \delta(\overline{U})} x_e = \sum_{e \in \delta(U)} u_e = \underbrace{u(\delta(U))}_{capacity of \delta(U)} (\delta(U)) (vx_w > 0)$$

The value of x is equal to the capacity of $\delta(U)$. By Corollary 1, the value of a feasible (s, t)-flow is at most the capacity of an (s, t)-cut $\Rightarrow \delta(U)$ is a minimum cut.

Integral capacities \Rightarrow integral widths of augmenting paths \Rightarrow integral flow. \Box

3 Closing remarks

Notice the similarity of the above proof with that of the theorem about cuts and the existence of an st-path. This is no coincidence, as we shall see, and this correspondence will allow us to reduce the problem of finding augmenting paths to simple (s, t)-connectivity question on an auxiliary graph.

Advance note: similar situation occurs with the minimum-cost flow problem which reduces to maximum flow and iterations of shortest path question in an auxiliary graph with general weights (Bellman-Ford); this phenomenon is more generally captured by the so-called Primal-Dual method and is related to Linear Programming (LP) formulations of these problems...