

Flows in networks, III

Lecture 27

A function $f : E \rightarrow \mathbf{R}$ is called **a flow in G** if for every vertex $v \in V - s - t$,

$$\text{div}_f(v) = \sum_{e \in E^+(v)} f(e) - \sum_{e \in E^-(v)} f(e) = 0. \quad (1)$$

If $0 \leq f(e) \leq c(e)$ for every $e \in E$, then the flow is called **feasible** (for G).

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The **capacity** of (S, \bar{S}) is

$$\mathbf{c}(S, \bar{S}) = \sum_{xy \in E: x \in S, y \in \bar{S}} \mathbf{c}(xy). \quad (2)$$

Important inequality

Claim 4.17. For every feasible flow f in a network $G = (V, E, s, t, \{\mathbf{c}(e)\}_{e \in E})$ and every s, t -cut (S, \bar{S}) of V ,

$$M(f) \leq \mathbf{c}(S, \bar{S}). \quad (3)$$

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Proof. Consider $F(f, S) = \sum_{v \in S} \text{div}_f(v)$.

By the definition of a flow, since $s \in S$ and $t \notin S$,

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On the other hand, by (1),

$$F(f, S) = \sum_{v \in S} \left(\sum_{e \in E^+(v)} f(e) - \sum_{e \in E^-(v)} f(e) \right).$$

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Let $e = xy$. Then e contributes $f(e)$ to $\text{div}_f(x)$ and $-f(e)$ to $\text{div}_f(y)$.

So, if $\{x, y\} \subset S$, then the net contribution of e is $f(e) - f(e) = 0$.

Also, if $\{x, y\} \subset \bar{S}$, then the net contribution of e is 0.

If $x \in S$ and $y \in \bar{S}$, then e contributes $f(e)$ into the RHS of (4).

Finally, if $y \in S$ and $x \in \bar{S}$, then e contributes $-f(e)$.

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Hence $F(f, S) = \sum_{xy \in E: x \in S, y \in \bar{S}} f(xy) - \sum_{yx \in E: x \in S, y \in \bar{S}} f(yx)$.

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Hence $F(f, S) = \sum_{xy \in E: x \in S, y \in \bar{S}} f(xy) - \sum_{yx \in E: x \in S, y \in \bar{S}} f(yx)$.

Since $0 \leq f(e) \leq c(e)$ for each $e \in E$, the last sum is at most

$$\sum_{xy \in E: x \in S, y \in \bar{S}} c(xy) = c(S, \bar{S}).$$

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$$\sum_{xy \in E: x \in S, y \in \bar{S}} c(xy) = c(S, \bar{S}). \quad \text{So, } M(f) \leq c(S, \bar{S}).$$

Max-flow min-cut Theorem

Theorem 4.18. For every network $G = (V, E, s, t, \{\mathbf{c}(e)\}_{e \in E})$, maximum of $M(f)$ over all feasible flows equals to the minimum of $\mathbf{c}(S, \bar{S})$ over all s, t -cuts (S, \bar{S}) .

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An idea for **Ford-Fulkerson Algorithm**:

1. Finding the **maximum $M(f)$ for a feasible flow** in a network G is finding the maximum of a **linear function $div_f(s)$** on the **closed bounded set** formed by the inequalities $0 \leq f(e) \leq c(e)$ for every $e \in E$ and equations $div_f(v) = 0$ for each $v \in V - s - t$. Since each **continuous function** on a **compact set** achieves its maximum, the maximum flow **does exist**.

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2. Let f be a **feasible flow** in G with the **maximum $M(f)$** . By Theorem 4.17, $f = \sum_{i=1}^k f_i$, where each f_i is either

- (a) a **flow along a cycle**, or
- (b) a flow along a **t, s -path**, or
- (c) a **flow along an s, t -path**.

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 - (a) a **flow along a cycle**, or
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If some f_{i_0} satisfies (a), then $f - f_{i_0}$ has the same value $M(f - f_{i_0})$ and is **feasible**.

If some f_{i_0} satisfies (b), then $f - f_{i_0}$ has even larger $M(f - f_{i_0})$ and is **feasible**.

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Thus, it is enough to seek f in the form $f = \sum_{i=1}^k f_i$, where each f_i is a **flow along an s, t -path**.

Ford-Fulkerson Algorithm with an example

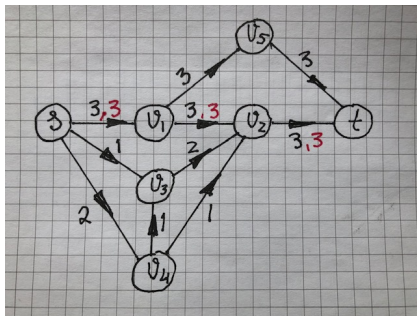


Figure: Network and a starting flow f_1 ; $M(f_1) = 3$.

Given a **feasible flow** f in a network G , the **residue network** $G(f)$ has $V(G(f)) = V$ and $xy \in E(G(f))$ iff

either $xy \in E$ and $f(xy) < c(xy)$ (then $c_f(xy) = c(xy) - f(xy)$),

or $yx \in E$ and $f(yx) > 0$ (in this case, $c_f(xy) = f(yx)$).

Ford-Fulkerson Algorithm with an example, 2

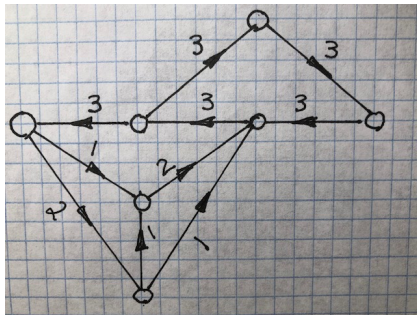


Figure: First residue network G_{f_1} .

If G_f has an **s, t-path** P and $\rho = \min\{c_f(xy) : xy \in E(P)\}$, then instead of f , consider $f' = f + \phi(P, \rho)$ and repeat. Note $M(f') = M(f) + \rho > M(f)$.

Ford-Fulkerson Algorithm with an example, 2

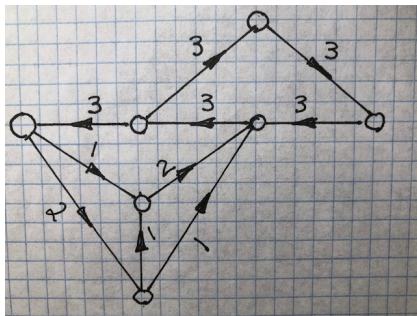


Figure: First residue network G_f .

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Here we have e.g. $P_1 = sv_4v_2v_1v_5t$ and $\rho_1 = \min\{2, 1, 3, 3, 3\} = 1$. So $f_2 = f_1 + \phi(P_1, 1)$.

Second iteration

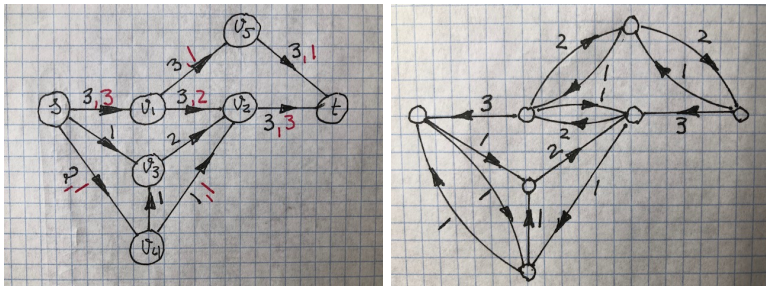


Figure: The second flow f_2 and the second residue network G_{f_2} .

We have $M(f_2) = 4$. The residue network has an s, t -path $P_2 = sv_3v_2v_1v_5t$ and $\rho_2 = \min\{1, 2, 2, 2, 2\} = 1$.

Hence $f_3 = f_2 + \phi(P_2, 1)$.

Third iteration

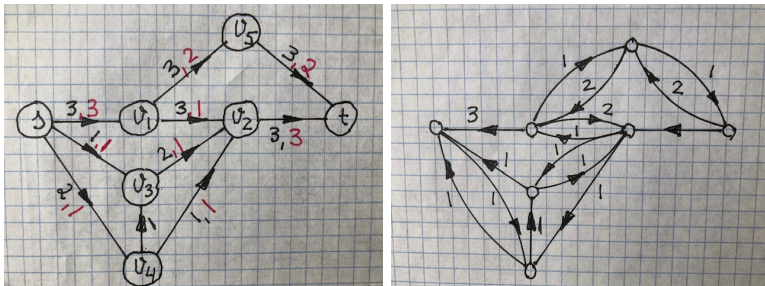


Figure: The third flow f_3 and the third residue network G_{f_3} .

We have $M(f_3) = 5$. The residue network has an s, t -path $P_3 = sv_4v_3v_2v_1v_5t$ and $\rho_3 = \min\{1, 1, 1, 1, 1\} = 1$.

Hence $f_4 = f_3 + \phi(P_3, 1)$.

Last iteration

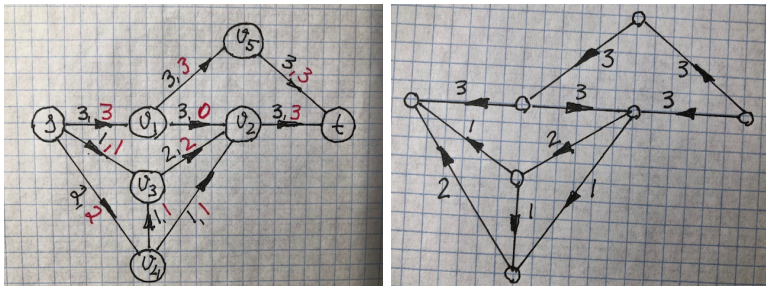


Figure: The fourth flow f_4 and the fourth residue network G_{f_4} .

We have $M(f_4) = 6$. The residue network G_{f_4} has **no s, t -paths**. Hence, there is a set S , namely $S = \{s\}$, such that $s \in S, t \notin S$ and G_{f_4} has **no edges from S to \bar{S}** .

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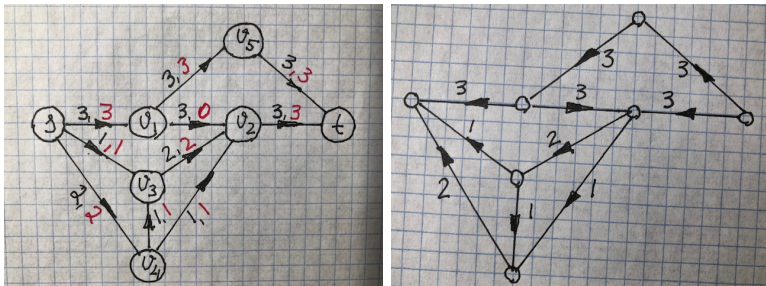


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We check that $\mathbf{c}(S, \bar{S}) = 6 = M(f_4)$. This certifies that **both, f_4 and (S, \bar{S}) are optimal**.