Termination of the Ford-Fulkerson Algorithm

Theorem 7.6.

- a If all capacities are rational, then the algorithm terminates with a maximum s-t-flow.
- **b** If all capacities are integral, it finds an integral maximum s-t-flow.

Termination of the Ford-Fulkerson Algorithm

Theorem 7.6.

- a If all capacities are rational, then the algorithm terminates with a maximum s-t-flow.
- **b** If all capacities are integral, it finds an integral maximum s-t-flow.

When an arbitrary *x*-augmenting path is chosen in every iteration, the Ford-Fulkerson Algorithm can behave badly:



Theorem 7.7.

If all capacities are integral and the maximum flow value is $K < \infty$, then the Ford-Fulkerson Algorithm terminates after at most K iterations. Its running time is $O(m \cdot K)$ in this case.

Theorem 7.7.

If all capacities are integral and the maximum flow value is $K < \infty$, then the Ford-Fulkerson Algorithm terminates after at most K iterations. Its running time is $O(m \cdot K)$ in this case.

Proof: In each iteration the flow value is increased by at least 1.

Theorem 7.7.

If all capacities are integral and the maximum flow value is $K < \infty$, then the Ford-Fulkerson Algorithm terminates after at most K iterations. Its running time is $O(m \cdot K)$ in this case.

Proof: In each iteration the flow value is increased by at least 1.

A variant of the Ford-Fulkerson Algo. is the Edmonds-Karp Algorithm:

▶ In each iteration, choose shortest *s*-*t*-dipath in D_x (edge lengths=1)

Theorem 7.7.

If all capacities are integral and the maximum flow value is $K < \infty$, then the Ford-Fulkerson Algorithm terminates after at most K iterations. Its running time is $O(m \cdot K)$ in this case.

Proof: In each iteration the flow value is increased by at least 1.

A variant of the Ford-Fulkerson Algo. is the Edmonds-Karp Algorithm:

▶ In each iteration, choose shortest *s*-*t*-dipath in D_x (edge lengths=1)

Theorem 7.8.

The Edmonds-Karp Algorithm terminates after at most $n \cdot m$ iterations; its running time is $O(n \cdot m^2)$.

Remark: The Edmonds-Karp Algorithm can be implemented with running time $O(n^2 \cdot m)$.

Arc-Based LP Formulation

Straightforward LP formulation of the maximum *s*-*t*-flow problem:

$$\begin{array}{rcl} \max & \sum_{a \in \delta^{+}(s)} x_{a} - \sum_{a \in \delta^{-}(s)} x_{a} & \text{nef flow out of } S \\ \text{s.t.} & \sum_{a \in \delta^{-}(v)} x_{a} - \sum_{a \in \delta^{+}(v)} x_{a} = 0 & \text{flow conservation} \\ & x_{a} \leq u(a) & \text{for all } v \in V \setminus \{s, t\} & Y_{v} \\ & x_{a} \geq 0 & \text{for all } a \in A & \mathbb{Z}_{a} \\ & x_{a} \geq 0 & \text{for all } a \in A & \mathbb{Z}_{a} \\ \text{Cohmen of primal} & \text{Define } Y_{S} = [\\ & u(a) \cdot \mathbb{Z}_{a} & Y_{t} = 0 \\ & y_{v} + 2 \sin & \ge 0 & \forall (v_{i}w) \in A \\ & y_{v} + 2 \sin & \ge 1 & \forall (s_{i}w) \in A \\ & -Y_{v} + 2v_{s} \geq -1 & \forall (v_{i}s) \in A \\ & -Y_{v} + 2v_{s} \geq 0 & \forall (v_{i}t) \in A \\ & y_{w} + 2 \sin & \ge 0 & \forall (v_{i}t) \in A \\ & y_{w} + 2 \oplus & 0 & \forall (v_{i}t) \in A \\ & y_{w} + 2 \oplus & 0 & \forall (v_{i}t) \in A \\ & y_{w} + 2 \oplus & 0 & \forall (v_{i}t) \in A \\ & y_{w} + y_{w} \in A \\ & y_{w} = 0 & \forall (v_{w} \in A \\ & y_{w} = 0 \\ &$$

Arc-Based LP Formulation

Straightforward LP formulation of the maximum *s*-*t*-flow problem:

$$\begin{array}{ll} \max & \sum_{a \in \delta^+(s)} x_a - \sum_{a \in \delta^-(s)} x_a \\ \text{s.t.} & \sum_{a \in \delta^-(v)} x_a - \sum_{a \in \delta^+(v)} x_a = 0 & \text{for all } v \in V \setminus \{s, t\} \\ & x_a \leq u(a) & \text{for all } a \in A \\ & x_a \geq 0 & \text{for all } a \in A \end{array}$$

Dual LP:

$$\begin{array}{ll} \min & \sum_{a \in A} u(a) \cdot z_a \\ \text{s.t.} & y_w - y_v + z_{(v,w)} \ge 0 & \text{for all } (v,w) \in A \\ & y_s = 1, \quad y_t = 0 \\ & z_a \ge 0 & \text{for all } a \in A \end{array}$$

$$\begin{array}{ll} \min & \sum_{a \in A} u(a) \cdot z_a \\ \text{s.t.} & y_w - y_v + z_{(v,w)} \ge 0 & \text{for all } (v,w) \in A \\ & y_s = 1, \quad y_t = 0 \\ & z_a \ge 0 & \text{for all } a \in A \end{array}$$

Observation: An s-t-cut $\delta^+(U)$ (with $U \subseteq V \setminus \{t\}$, $s \in U$) yields feasible dual solution (y, z) of value $u(\delta^+(U))$:

$$\begin{array}{ll} \min & \sum_{a \in A} u(a) \cdot z_a \\ \text{s.t.} & y_w - y_v + z_{(v,w)} \ge 0 & \text{for all } (v,w) \in A \\ & y_s = 1, \quad y_t = 0 \\ & z_a \ge 0 & \text{for all } a \in A \end{array}$$

Observation: An s-t-cut $\delta^+(U)$ (with $U \subseteq V \setminus \{t\}$, $s \in U$) yields feasible dual solution (y, z) of value $u(\delta^+(U))$:

let y be the characteristic vector x^U of U
 (i. e., y_v = 1 for v ∈ U, y_v = 0 for v ∈ V \ U)

min
$$\sum_{a \in A} u(a) \cdot z_a$$

s.t.
$$y_w - y_v + z_{(v,w)} \ge 0$$

$$y_s = 1, \quad y_t = 0$$

$$z_a \ge 0$$
 for all $a \in A$

Observation: An s-t-cut $\delta^+(U)$ (with $U \subseteq V \setminus \{t\}$, $s \in U$) yields feasible dual solution (y, z) of value $u(\delta^+(U))$:

- Int y be the characteristic vector x^U of U (i. e., y_v = 1 for v ∈ U, y_v = 0 for v ∈ V \ U)
- let z be the characteristic vector χ^{δ+(U)} of δ⁺(U)
 (i. e., z_a = 1 for a ∈ δ⁺(U), z_a = 0 for a ∈ A \ δ⁺(U))

2,

$$\begin{array}{ll} \min & \sum_{a \in A} u(a) \cdot z_a \\ \text{s.t.} & y_w - y_v + z_{(v,w)} \ge 0 & \text{for all } (v,w) \in A \\ & y_s = 1, \quad y_t = 0 \\ & z_a \ge 0 & \text{for all } a \in A \end{array}$$

Observation: An s-t-cut $\delta^+(U)$ (with $U \subseteq V \setminus \{t\}$, $s \in U$) yields feasible dual solution (y, z) of value $u(\delta^+(U))$:

- Int y be the characteristic vector x^U of U (i. e., y_v = 1 for v ∈ U, y_v = 0 for v ∈ V \ U)
- let z be the characteristic vector χ^{δ+(U)} of δ⁺(U)
 (i. e., z_a = 1 for a ∈ δ⁺(U), z_a = 0 for a ∈ A \ δ⁺(U))

Theorem 7.9.

There exists an *s*-*t*-cut $\delta^+(U)$ (with $U \subseteq V \setminus \{t\}$, $s \in U$) such that the corresponding dual solution (y, z) is an optimal dual solution.

Proof Thm Y.g: