

CS 577- Intro to Algorithms

Network Flow (Part 2)

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Recap - notions

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Network

- ▶ a digraph (V, E)
- ▶ edge capacities $c : E \rightarrow [0, \infty)$
- ▶ the source $s \in V$, which has indegree 0, and
- ▶ the sink $t \in V$, which has outdegree 0.

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Flow

A mapping $f : E \rightarrow [0, \infty)$ satisfying

- ▶ [capacity constraints] $(\forall e \in E) f(e) \leq c(e)$
- ▶ [conservation constraints] $(\forall v \in V \setminus \{s, t\}) f_{\text{in}}(v) = f_{\text{out}}(v)$
where $f_{\text{in}}(v) \doteq \sum_{u \in V: e \doteq (u, v) \in E} f(e)$ and
 $f_{\text{out}}(v) \doteq \sum_{u \in V: e \doteq (v, u) \in E} f(e)$.

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st -Cut

A partition (S, T) of V such that $s \in S$ and $t \in T$.

Recap - computational problems

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Max flow

Input: network $N = (V, E, c, s, t)$

Output: flow f such that $\nu(f) \doteq f_{\text{out}}(s)$ is maximized

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Min cut

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Output: st -cut (S, T) such that $c(S, T) \doteq \sum_{e \in S \times T} c(e)$ is minimized

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- $(\forall e \in E \cap S \times T) f(e) = c(e)$
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- Strong duality (today):

$$\max_{\text{flow } f} \nu(f) = \min_{st\text{-cut } (S, T)} c(S, T)$$

Residual Network

Consider a flow f in $N = (V, E, c, s, t)$.

Definition

The residual network $N_f = (V, E_f, c_f, s, t)$ has:

- ▶ For each $e \in E$ with $f(e) < c(e)$, an edge e in E_f with $c_f(e) \doteq c(e) - f(e)$.
- ▶ For each $e = (u, v) \in E$ with $f(e) > 0$, an edge $e' \doteq (v, u)$ in E_f with $c_f(e') \doteq f(e)$.

Path Augmentation

Scheme

1. Start with $f \equiv 0$.
2. While there is an st -path in N_f
 - ▶ Pick such a path P .
 - ▶ $f \leftarrow f +$ flow along P of value $\min_{e \in P}(c_f(e))$
3. Return f .

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If the algorithm produces an output, it is correct.

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 - $(\forall e \in E \cap S \times T) f(e) = c(e)$
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- (3) \Rightarrow (1) Follows from weak duality.

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Corollary: Strong duality

$$\max_{\text{flow } f} \nu(f) = \min_{st\text{-cut } (S, T)} c(S, T)$$

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- ▶ $\therefore (S, T)$ is an st -cut with $c(S, T) = \nu(f)$.
By weak duality $\nu(f)$ is maximized and $c(S, T)$ minimized.
- ▶ Construction of (S, T) from f runs in linear time.

Path Augmentation - recap

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If the algorithm produces an output, it is correct.

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- ▶ If each augmenting path is picked using linear-time graph traversal, running time is $O((n + m) \cdot F)$.
- ▶ Running time is *pseudopolynomial*, i.e., bounded by polynomial in size parameters (n and m) and *value* of numbers involved (capacities). [$F \leq \sum_{e \in E} c(e)$.]

Slow convergence

Better Algorithms

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Augmentation along path of maximum residual capacity

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- ▶ Finding one path: $O(n + m)$ time
- ▶ Overall running time: $O((n + m)m \cdot \log F)$

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- ▶ # augmentations = $O(m \cdot \log F)$ for integral instances
- ▶ Finding one path: $O(n + m)$ time
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Augmentation along path with smallest number of edges

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Augmentation along path with smallest number of edges

- ▶ # augmentations = $O(nm)$ for all instances

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- ▶ Finding one path: $O(n + m)$ time
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Augmentation along path with smallest number of edges

- ▶ # augmentations = $O(nm)$ for all instances
- ▶ Finding one path: $O(n + m)$ time (BFS)

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- ▶ # augmentations = $O(m \cdot \log F)$ for integral instances
- ▶ Finding one path: $O(n + m)$ time
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- ▶ # augmentations = $O(nm)$ for all instances
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- ▶ Finding one path: $O(n + m)$ time
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Augmentation along path with smallest number of edges

- ▶ # augmentations = $O(nm)$ for all instances
- ▶ Finding one path: $O(n + m)$ time (BFS)
- ▶ Overall running time: $O(nm(n + m))$
- ▶ Running time is *strongly polynomial*, i.e., bounded by polynomial in size parameters only.

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Augmentation along path of maximum residual capacity

- ▶ # augmentations = $O(m \cdot \log F)$ for integral instances
- ▶ Finding one path: $O(n + m)$ time
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Augmentation along path with smallest number of edges

- ▶ # augmentations = $O(nm)$ for all instances
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Other approaches: $O(nm)$ time for all instances