

Algorithms for Network Flows

Lecture 4:

Generalized flows II: a strongly polynomial algorithm

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Slides will be available at: <http://nolver.net/home/valparaiso>

Today's lecture

- ▶ The first strongly polynomial algorithm was given only quite recently by Véggh '14. Unfortunately it's **very** complicated!
- ▶ Here we discuss a much simpler (and faster) algorithm by O.-Véggh '16.

Our algorithm	$O((m + n \log n)mn \log(n^2/m))$
Radzik '04	$O((m + n \log n)mn \log B)$
Véggh '12	$O(m^2 n^3)$

Primal and dual again

$$\begin{aligned} \max \quad & \nabla f_t \\ \text{s.t.} \quad & \nabla f_i \geq b_i \quad \forall i \neq t \\ & f \geq 0 \end{aligned}$$

$$\begin{aligned} \max \quad & \sum_{j \in V \setminus \{t\}} b_j^\mu \\ \text{s.t.} \quad & \gamma_e^\mu \leq 1 \quad \forall e \in E \\ & \mu_t = 1 \\ & \mu_i \in \mathbb{R}_{++} \quad \forall i \in V \end{aligned}$$

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$$\begin{aligned} & \max \quad \mu_t \sum_{j \in V \setminus \{t\}} b_j^\mu \\ \text{s.t.} \quad & \gamma_e^\mu \leq 1 \quad \forall e \in E \\ & \mu_i \in \mathbb{R}_{++} \quad \forall i \in V \end{aligned}$$

Reminder

Given $f \in \mathbb{R}_+^E$, $\mu \in \mathbb{R}_{++}^V$, (f, μ) is called a **fitting pair** if:

- ▶ μ is dual feasible
- ▶ $f_e > 0$ implies $\gamma_e^\mu = 1$.

If (f, μ) is a fitting pair and $\nabla f_i = b_i$ for all $i \neq t$, then f and μ are both optimal.

- ▶ Our algorithm will always maintain a fitting pair.

Goal: find a contractible edge

An edge $e \in E$ is **contractible** if $\gamma_e^{\mu^*} = 1$ for any dual optimum μ^* .

- ▶ Precisely as we saw for min cost flow, if e is contractible, can extend a dual optimum to $G/\{e\}$ to a dual optimum for G .
- ▶ So our goal is to produce a contractible arc in strongly polynomial time.

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$$\text{Ex}(f, \mu) := \sum_{i \neq t} \max\{\nabla f_i^\mu - b_i^\mu, 0\}.$$

Lemma

Suppose f is feasible, and (f, μ) is a fitting pair.
Then if $f^\mu(\hat{e}) > \text{Ex}(f, \mu)$, \hat{e} is contractible.

Plentiful nodes

- ▶ Our algorithm will maintain the invariant that $\nabla f_i^\mu < b_i^\mu + 2$, so

$$\text{Ex}(f, \mu) < 2n.$$

Given a feasible dual μ , we say i is **plentiful** if

$$|b_i^\mu| \geq 2n^2.$$

Lemma

If (f, μ) is a fitting pair with f feasible, and i is a plentiful node, then some edge adjacent to i is contractible.

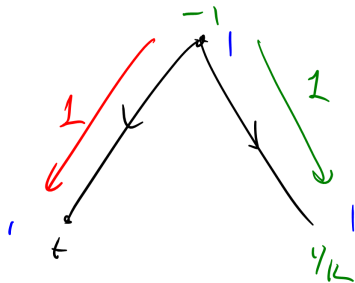


Pf. If $S_i^u \leq -2n^2$, $\partial f_i^u \leq -2n^2 + 2$

$\exists e \in \delta^+(i)$ with $f(e) \geq \frac{2n^2 - 2}{n - 1} > 2n$.
 $\Rightarrow E_x(f, \mu)$

A key new idea

- ▶ All previous algorithms maintain (roughly) a fitting pair (f, μ) with f feasible.
- ▶ We keep a fitting pair (f, μ) —but do **not** require f to be feasible!



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Definition

μ is **safe** if there **exists** a feasible g s.t. (g, μ) is a fitting pair.

- ▶ Instead, we maintain a fitting pair (f, μ) where μ is safe, $\nabla f_i^\mu < b_i^\mu + 2$ for all $i \neq t$.

Lemma

Given such an (f, μ) , if node i is plentiful then we can find a contractible arc adjacent to i .

Pf.: μ safe $\Rightarrow \exists g$ st. $\text{supp}(g) \subseteq \text{right}(\mu)$

\uparrow
 $\{e: \nu_e = 1\}$

$$\nabla g_i \geq \delta_i^M$$

Also $\text{supp}(f) \subseteq \text{right}(\mu)$

$$\nabla f_i^M \leq \delta_i^M + 2$$

$$\text{Th s.t. } \delta_i^M \leq \nabla h_i^M \leq \nabla \delta_i^M + 2.$$

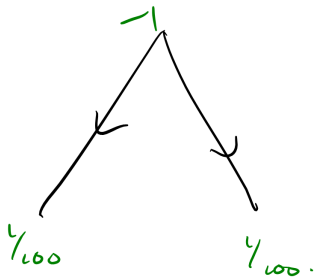
$\forall i \neq t.$

Q.E.D.

Integrality vs feasibility

- ▶ A **big** benefit we gain from working with f infeasible is that we will maintain that f^μ is **integral**.

Feasibility and integrality are not compatible:



Invariants

Our algorithm maintains:

- ▶ (f, μ) a fitting pair, with f^μ integral

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- ▶ $\nabla f_i^\mu < b_i^\mu + 2$ for all i (we don't require feasibility)
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- ▶ μ is safe

Goal

Adjusting f, μ satisfying above, produce j s.t. $|b_j^\mu| \geq 2n^2$.

Our algorithm

- ▶ **Initialization:** Find an initial fitting pair (f, μ) , with f feasible.
 - ▶ Can do this with cycle cancelling, using strongly polynomial result of Radzik.
- ▶ Round f so that f^μ is integral, $-1 < \nabla f_i^\mu - b_i^\mu < 2$ for all $i \neq t$.

$$\bar{\mu} \leftarrow \lambda \mu \quad \text{so that} \quad \delta_i^{\bar{\mu}} \leq \nabla f_i^{\bar{\mu}} \leq \delta_i^{\bar{\mu}} + 1.$$

Now let $\bar{f}^{\bar{\mu}}$ be an integral (regular) flow satisfying $\text{supp}(\bar{f}^{\bar{\mu}}) \subseteq \text{tight}(\bar{\mu})$

$$\lfloor \delta_i^{\bar{\mu}} \rfloor \leq \nabla f_i^{\bar{f}^{\bar{\mu}}} \leq \lceil \delta_i^{\bar{\mu}} + 1 \rceil$$

$$f \leftarrow \bar{f}^{\bar{\mu}}, \quad \mu \leftarrow \bar{\mu}$$

Our algorithm

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While $|b_j^\mu| < 2n^2$ for all j :

1. Augment f^μ
(μ won't change)
2. Rescale μ
(f^μ won't change)

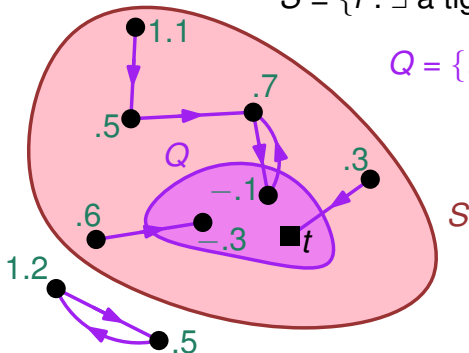
Augmentation step

while $\exists i \in S \cap V^-$ with $\nabla f_i^\mu \geq b_i^\mu + 1$ **do**

Send 1 unit of relabelled flow from i to Q

$S = \{i : \exists \text{ a tight } i\text{-}Q\text{-path in } E^f\}$

$Q = \{t\} \cup \{i : \nabla f_i^\mu < b_i^\mu\}$



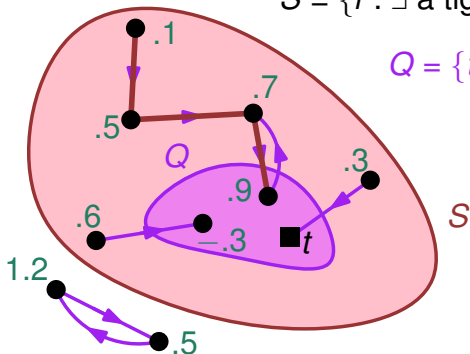
- ▶ Only augment on tight arcs, so (f, μ) stays a fitting pair.
- ▶ After augmenting, $\nabla f_i^\mu < b_i^\mu + 1$ for all $i \in S$.

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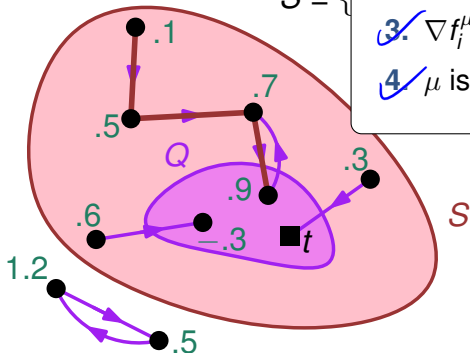
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Augmentation step

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Send 1 unit of relabelled flow

$S = \{$



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2. $\nabla f_i^\mu < b_i^\mu + 2$ for all i

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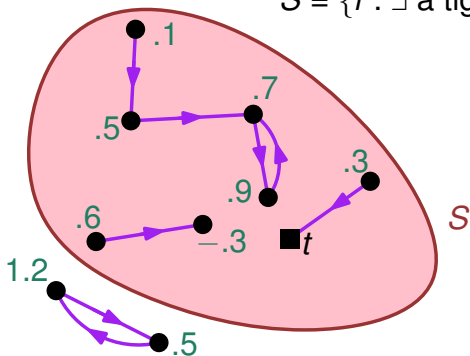
4. μ is safe

► Only augment on tight arcs, so (f, μ) stays a fitting pair.

► After augmenting, $\nabla f_i^\mu < b_i^\mu + 1$ for all $i \in S$.

Rescaling step

$$S = \{i : \exists \text{ a tight } i\text{-}Q\text{-path in } E^f\}$$



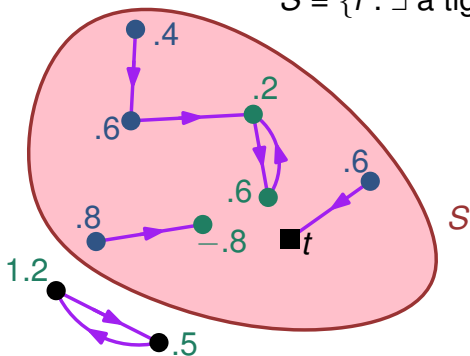
f unchanged!

$$\mu'_i = \begin{cases} \mu_i / \alpha & \text{for } i \in S \\ \mu_i & \text{for } i \notin S \end{cases}$$

$$f'_e = \begin{cases} f_e / \alpha & \text{for } e \in E(S) \\ f_e & \text{for } e \notin E(S) \end{cases}$$

Rescaling step

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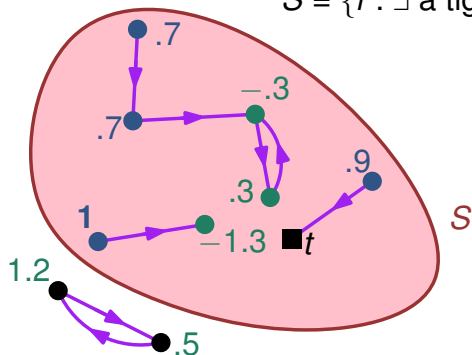


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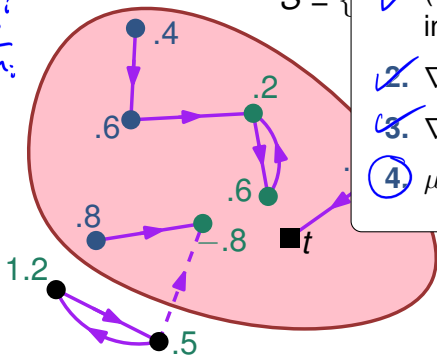
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α chosen maximally s.t. $\gamma_e^\mu \leq 1$ for all $e \in \delta^-(S)$, $\nabla f_i^\mu \leq b_i^\mu + 1$ for all $i \in S$.

Rescaling step

$$b_i^\mu = \frac{b_i}{\mu_i}$$



$S = \{$

1. (f, μ) a fitting pair, with f^μ integral
2. $\nabla f_i^\mu < b_i^\mu + 2$ for all i
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Key technical lemma

Safety of μ is preserved in the rescaling step.

$$\mu'_i = \begin{cases} \mu_i/\alpha & i \in S \\ \mu_i & i \notin S \end{cases} \quad \begin{array}{l} R = \text{tight arcs w.r.t. } \mu \\ R' = \text{tight arcs w.r.t. } \mu' \end{array}$$

Suppose for a contradiction that μ is safe, but μ' is not.

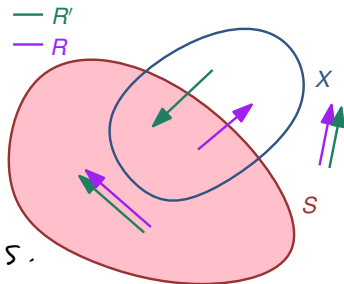
$\therefore \nexists$ flow on R' satisfying demands $b_i^{\mu'}$.

$\therefore \exists X, t \notin X$, with $\sum_{i \in X} b_i^{\mu'} > 0$

$$\delta_{R'}^-(X) = \emptyset.$$

Claim: $b^{\mu'}(X \setminus S) > 0$

Pf: $\delta^-(S) \cap R = \emptyset$ by defⁿ of S .



$R = R'$ inside of S

$$\therefore \delta^-(x \cap S) \cap R = \emptyset.$$

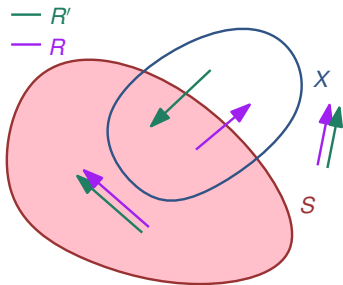
$$\therefore \delta^+(x \cap S) \leq 0, \text{ since } \mu \text{ safe.}$$

$$\Rightarrow \delta^+(x \setminus S) = \delta^{m'}(x \setminus S) \geq 0.$$

Claim: $f(\delta^-(x \setminus S)) = 0$

Pf: $R = R'$ outside of S , so nothing from $v \setminus (x \cup S)$ into $x \setminus S$.

No flow from S into $x \setminus S$.



Bounding the number of augmentations

- ▶ Keep track of potentials

$$\Phi := \sum_{i \in V} \nabla f_i^\mu - b_i^\mu, \quad \Psi := - \sum_{i \in V^-} b_i^\mu$$

$$\{i: b_i < 0\}.$$

$$-n \leq \Phi \leq 2n.$$

Rescale: $\Delta \Phi = \Delta \Psi$.

Augmentations: Ψ unchanged,

Helpful augmentation: from V^- to $V \setminus V^-$.

A missing detail

An algorithm is **strongly polynomial** if:

1. Number of arithmetic operations is polynomial in the number of integers in the input (e.g., size of the graph)
2. **Encoding lengths of numbers computed during execution are polynomial in input encoding length**

- ▶ We need to show that the μ_i values stay under control.
- ▶ **Dramatically** easier for our algorithm than for the algorithm of Végh '14, but still not light entertainment. . .
- ▶ We can exploit the flexibility in some of the rescaling steps to ensure that some μ_i 's stay “nice”, and for any j , $\mu_j = \gamma(P)\mu_i$ for some “nice” μ_i and path P .

Extra ingredients for a faster running time

With some extra work, the running time of a souped up version of this algorithm is $O((m + n \log n)mn \log(n^2/m))$.

- ▶ Strongly polynomial cycle cancelling algorithm of Radzik is too slow. Replace with an execution of our algorithm on an auxiliary instance.
- ▶ Algorithm needs to be implemented efficiently so that not too much time is spent updating labels.
- ▶ Don't start from scratch after a contraction.
- ▶ A refined potential analysis is needed.

Open question

Strongly polynomial algorithm for minimum cost generalized flow?

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Thank you!

References



N. Olver and L. Végh.

A simpler and faster strongly polynomial algorithm for generalized flow maximization.

arXiv:1611.01778, 2016.



M. Shigeno.

A survey of combinatorial maximum flow algorithms on a network with gains.

Journal of the Operations Research Society of Japan,
47(4):244–264, 2004.



L. Végh.

A strongly polynomial algorithm for generalized flow maximization.

Mathematics of Operations Research.