

Discrete Optimization

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Polynomial running time algorithms for the max flow problem

- The generic augmenting path algorithm runs in $\mathcal{O}(mnU)$
- Ways to improve the algorithm and make it run in polynomial time
 - ▶ Augmenting in **large** increments of flows
→ **Capacity scaling** algorithm
 - ▶ Augment on **shortest paths** in the residual network
 - ▶ Relax the **mass balance constraints** and only augment **locally**
→ The **push-relabel** algorithm which is the most efficient one!

In order to implement the last two algorithms, we need to rely on **distance labels**.

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Distance labels

Definition

A distance function gives a **numeric label** to each node.

The distance function is **valid** if

$$d(t) = 0$$

$$d(i) \leq d(j) + 1 \quad \text{for every arc } (i,j) \text{ in } G(x)$$

Property

If d is valid, $d(i)$ is a **lower bound** on the length of the shortest path from i to t in the **residual graph**.

Corollary

If $d(s) \geq n$, there exists no path from s to t in the residual graph.

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Distance labels

The distance labels are **exact** if each label indicates the **exact length** of the shortest path to t in the residual graph.

An exact labeling can be determined in $\mathcal{O}(m)$ by **backward breadth-first search**.

An arc $(i, j) \in G(x)$ is **admissible** if

$$d(i) = d(j) + 1,$$

otherwise it is **inadmissible**.

An **admissible path** is a path consisting only of admissible arcs.

An admissible path is a **shortest augmenting path** !

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The shortest augmenting path algorithm

- Augmenting on shortest paths in the residual network guarantees a **polynomial running time**!
- We could rerun backward breadth-first search at **each iteration** → very inefficient but runs in $\mathcal{O}(nm^2)$
- The minimum distance from each node to the sink is **monotonically increasing**.

The shortest augmenting path algorithm

- Perform an **initial labeling** by backward breadth-first-search
- Repeat : Perform an **advance operation** (finding an admissible arc from the current last node in the admissible path)
- If we find an augmenting path, then we augment along it !
- If at some node i , we do not find any admissible arc, we **relabel** the node

$$\min\{d(j) + 1 \mid (i, j) \in \delta(\{i\}) \text{ and } r_{ij} > 0\}$$

and remove the node i from the current admissible path.

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Correctness of the algorithm

- Every operation maintains a **valid distance labeling**.
- Each relabel strictly increases the distance label of a node.
- The shortest augmenting path algorithm correctly computes a maximum flow.

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Complexity of the algorithm

- Between two **relabels**, if an arc becomes inadmissible, it stays inadmissible.
- If the algorithm relabels any node at most k times, the complexity of finding admissible arcs and relabeling the nodes is $\mathcal{O}(km)$ and the algorithm saturates arcs at most $\frac{km}{2}$ times.
- Each distance label increases at most n times. The total number of relabeling is n^2 and the total number of augment operations is $\frac{nm}{2}$.

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Preflow-push algorithm

Drawback of the augmenting path algorithm : the expensive operation of **sending flow** along a path.

Many of these augmentations may share **the same subpath** !

Idea of preflow-push : augment along **arcs** and we therefore have to **relax the flow balance constraints**.

Preflows

Definition

A **preflow** is a function $x : A \rightarrow \mathbb{R}_+$ such that

$$\sum_{j|(j,i) \in A} x_{ji} - \sum_{j|(i,j) \in A} x_{ij} \geq 0. \quad \text{for all } i \in V \setminus \{s, t\}.$$

Definition

The **excess** at a node i of a given preflow x is given by

$$e(i) := \sum_{j|(j,i) \in A} x_{ji} - \sum_{j|(i,j) \in A} x_{ij}.$$

A node with positive excess is said to be **active** because we need to recover the mass-balance constraint at some point.

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Preflow-push algorithm

- Compute exact **distance labels** $d(i)$
- Push flows along all arcs emanating from s : $x_{sj} := u_{sj}$
- $d(s) := n$
- If a node is active, **push-relabel**
 - ▶ If (i, j) is admissible, push $\delta := \min\{e(i), r_{ij}\}$
 - ▶ Else $d(i) := \min\{d(j) + 1 \mid (i, j) \in \delta(\{i\}) \text{ and } r_{ij} > 0\}$

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Complexity of the algorithm

- Distance labels are always **valid** during the algorithm
- At any stage, a node i with excess $e(i) > 0$ has a path from i to s in the residual network
- For each i , $d(i) < 2n$
- The total number of **relabels** is $2n^2$
- The algorithm performs at most nm **saturating** pushes.
- The algorithm performs at most n^2m **nonsaturating** pushes
- The algorithm runs in $\mathcal{O}(n^2m)$

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