

# Maximum Flow

Reading: CLRS Chapter 26.

CSE 6331 Algorithms

Steve Lai

# Flow Network

- A flow network  $G = (V, E)$  is a directed graph with
  - a source node  $s \in V$ ,
  - a sink node  $t \in V$ ,
  - a capacity function  $c$ .
- Each edge  $(u, v) \in E$  has a nonnegative capacity  $c(u, v) \geq 0$ .
- If  $(u, v) \notin E$ , assume  $c(u, v) = 0$ .
- Also, assume that every node  $v$  is on some path from  $s$  to  $t$ .
  - This implies  $O(V + E) = O(E)$ .
  - A maxflow may only go through such nodes.

# Flow

- Let  $G = (V, E)$  be a flow network with capacity function  $c$ , source node  $s$ , and sink node  $t$ .
- A flow is a real-valued function  $f : V \times V \rightarrow \mathbb{R}$  satisfying
  - Capacity constraint:  $\forall u, v \in V, f(u, v) \leq c(u, v)$ .
  - Skew symmetry:  $\forall u, v \in V, f(u, v) = -f(v, u)$ .
  - Flow conservation:  $\forall u \in V - \{s, t\}$ ,

$$f(u, V) \triangleq \sum_{v \in V} f(u, v) = 0$$

- The **value of a flow**  $f$  is  $|f| \triangleq f(s, V) = \sum_{v \in V} f(s, v)$ .
- The **maxflow problem** is to find a flow of maximum value.

## Some Properties of Flows

- If no edge between  $u$  and  $v$ , then  $f(u, v) = f(v, u) = 0$ .
- Flow conservation implies:  $\forall u \in V - \{s, t\}$ ,  
Total positive flow into  $u$  = Total positive flow out of  $u$ .
- For  $X, Y \subseteq V$ , define  $f(X, Y) = \sum_{x \in X} \sum_{y \in Y} f(x, y)$ .
- $f(X, X) = 0$ .
- $f(X, Y) = -f(Y, X)$ .
- $f(X \cup Y, Z) = f(X, Z) + f(Y, Z)$ , if  $X \cap Y = \emptyset$ .
- $f(Z, X \cup Y) = f(Z, X) + f(Z, Y)$ , if  $X \cap Y = \emptyset$ .

# Residual networks and augmenting paths

- Let  $G = (V, E)$  be a flow network and  $f$  a flow.

- Residual capacity of  $(u, v)$  is

$$c_f(u, v) = c(u, v) - f(u, v).$$

- Residual network induced by  $f$  is  $G_f = (V, E_f)$ , where

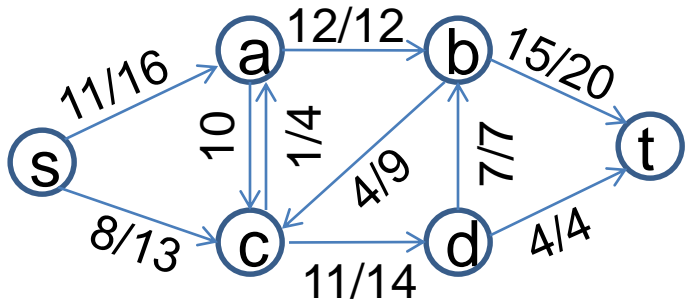
$$E_f = \{(u, v) \in V \times V : c_f(u, v) > 0\}.$$

- Augmenting path: a simple path  $p$  from  $s$  to  $t$  in  $G_f$ .

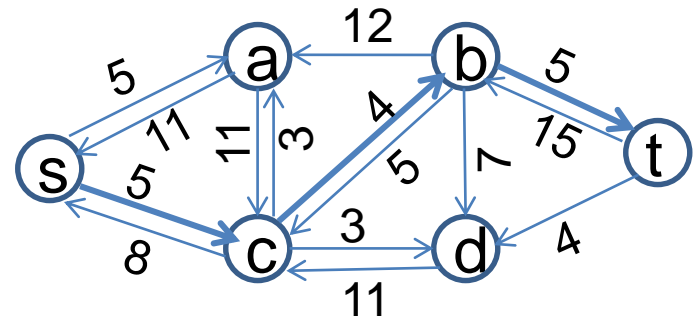
- Residual capacity of  $p$ :

$$c_f(p) = \min \{c_f(u, v) : (u, v) \text{ is in } p\}.$$

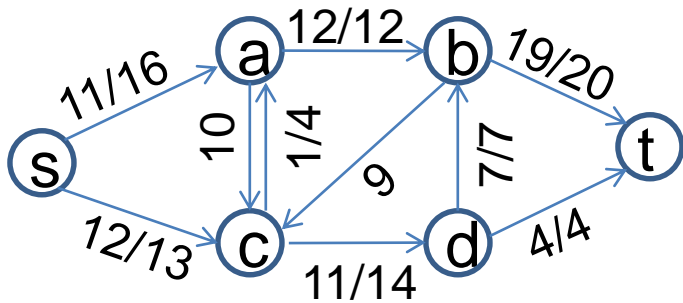
# Example



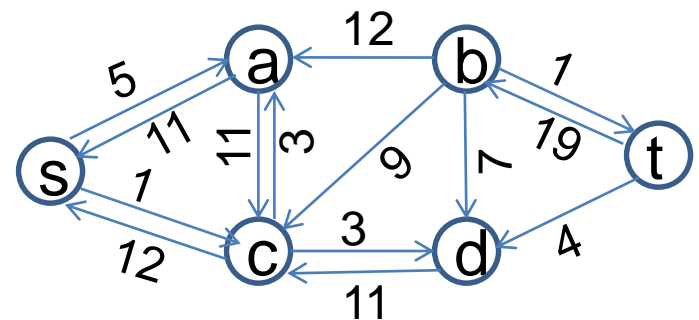
(a) Flow network and flow



(b) Residual network and augmenting path  $p$  with  $c_f(p) = 4$



(c) Augmented flow



(d) No augmenting path

## Flow: an alternative definition (CLRS, 3rd ed.)

- Let  $G = (V, E)$  be a flow network with a capacity function  $c$ , source  $s$ , and sink  $t$ . Assume  $G$  has no parallel edges, i.e., if  $(u, v) \in E$  then  $(v, u) \notin E$ .

- A flow is a real-valued function  $f : V \times V \rightarrow R$ , satisfying

- Capacity constraint:  $\forall u, v \in V, 0 \leq f(u, v) \leq c(u, v)$ .

- Flow conservation:  $\forall u \in V - \{s, t\}$ ,

$$\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v) \quad (\text{i.e. } f(V, u) = f(u, V))$$

- The value of a flow is  $|f| = f(s, V) - f(V, s)$ .

- Note: when  $(u, v) \notin E, f(u, v) = 0$ .

## Some of these properties do not hold any more (when using the second definition of flows)

- If no edge between  $u$  and  $v$ , then  $f(u, v) = f(v, u) = 0$ .
- Flow conservation implies:  $\forall u \in V - \{s, t\}$ ,  
Total positive flow into  $u$  = Total positive flow out of  $u$ .
- For  $X, Y \subseteq V$ , define  $f(X, Y) = \sum_{x \in X} \sum_{y \in Y} f(x, y)$ .
- $f(X, X) = 0$ .
- $f(X, Y) = -f(Y, X)$ .
- $f(X \cup Y, Z) = f(X, Z) + f(Y, Z)$ , if  $X \cap Y = \emptyset$ .
- $f(Z, X \cup Y) = f(Z, X) + f(Z, Y)$ , if  $X \cap Y = \emptyset$ .



# Residual networks and augmenting paths (using the second definition of flows)

- Let  $G = (V, E)$  be a flow network and  $f$  a flow.
- **Residual capacity** of  $(u, v)$  is

$$c_f(u, v) = \begin{cases} c(u, v) - f(u, v) & \text{if } (u, v) \in E \\ f(v, u) & \text{if } (v, u) \in E \\ 0 & \text{otherwise} \end{cases}$$

- Residual network induced by  $f$  is  $G_f = (V, E_f)$ , where

$$E_f = \{(u, v) \in V \times V : c_f(u, v) > 0\}.$$

- Augmenting path: a simple path  $p$  from  $s$  to  $t$  in  $G_f$ .

## Ford-Fulkerson Method

Given a flow network  $G = (V, E)$  with source  $s$  and sink  $t$ ,

Initialize flow  $f$  to 0

**while** there exists an augmenting path  $p$  **do**

**do** augment flow  $f$  along  $p$

**return**  $f$

## Ford-Fulkerson( $G, s, t$ )

**for** each edge  $(u, v) \in E(G)$

**do**  $f(u, v) \leftarrow 0$

$f(v, u) \leftarrow 0$

**while** there exists an augmenting path  $p$  in residual network  $G_f$

**do**  $c_f(p) = \min \{c_f(u, v) : (u, v) \text{ is on } p\}$

**for** each edge  $(u, v)$  is in  $p$

**do**  $f(u, v) \leftarrow f(u, v) + c_f(p)$

$f(v, u) \leftarrow -f(u, v)$

# Analysis

- The running time depends on how the augmenting path  $p$  is determined.
- If capacities are integers, the running time is  $O(E |f^*|)$ , where  $|f^*|$  is the value of the maxflow.
  - Each iteration can be done in  $O(E)$  time.
  - There are at most  $|f^*|$  iterations.

**Integrality Theorem.** If all capacities are integers, the flow  $f$  produced by the Ford-Fulkerson method has the property that  $f(u, v)$  is an integer for all  $u, v \in V$ .

**Lemma 1.** Let  $G_f$  be the residual network induced by flow  $f$ .

Let  $f'$  be a flow in  $G_f$ . Then  $f + f'$  is a flow in  $G$  with

$$|f + f'| = |f| + |f'|.$$

**Proof.** • Skew symmetry:

$$\begin{aligned}(f + f')(u, v) &= f(u, v) + f'(u, v) \\ &= -f(v, u) - f'(v, u) \\ &= -(f + f')(v, u)\end{aligned}$$

• Capacity constraint:

$$\begin{aligned}(f + f')(u, v) &= f(u, v) + f'(u, v) \\ &\leq f(u, v) + c_f(u, v) \\ &= f(u, v) + (c(u, v) - f(u, v)) \\ &= c(u, v)\end{aligned}$$

- Flow conservation: for all  $u \in V - \{s, t\}$ ,

$$\begin{aligned}(f + f')(u, V) &= f(u, V) + f'(u, V) \\ &= 0 + 0 = 0\end{aligned}$$

- Finally,

$$\begin{aligned}|f + f'| &= (f + f')(s, V) \\ &= f(s, V) + f'(s, V) \\ &= |f| + |f'|\end{aligned}$$

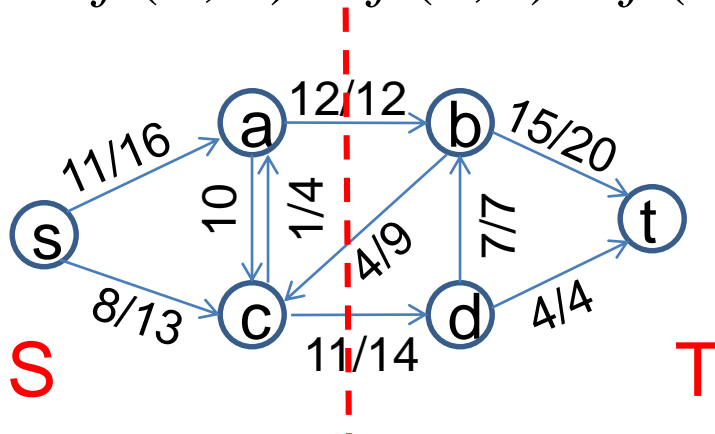
**Lemma 2.** If  $p$  is an augmenting path in  $G_f$ , then augmenting  $f$  along  $p$  yields a flow in  $G$  with value  $|f| + c_f(p) > |f|$ .

**Corollary 3.** The  $f$  produced by Ford-Fulkerson is a flow.

# Cuts

- A **cut**  $(S, T)$  of a flow network  $G = (V, E)$  is a partition of  $V$  into  $S$  and  $T = V - S$  such that  $s \in S$  and  $t \in T$ .
- If  $f$  is a flow,  $f(S, T)$  denotes the **net flow** across the cut  $(S, T)$ .
- The **capacity** of  $(S, T)$  is  $c(S, T) = \sum_{u \in S} \sum_{v \in T} c(u, v)$ .
- Example:  $c(S, T) = c(a, b) + c(c, d) = 12 + 14 = 26$ .

$$f(S, T) = f(a, b) + f(c, d) + f(c, b) = 12 + 11 - 4 = 19.$$





Lemma 4. For any cut  $(S, T)$ ,  $|f| = f(S, T)$ .

Proof. Note that  $f(u, V) = 0 \quad \forall u \neq s, t$ .

$$\begin{aligned} f(S, T) &= f(S, V) - f(S, S) \\ &= f(S, V) \\ &= f(s, V) + f(S - s, V) \\ &= f(s, V) = |f| \end{aligned}$$

Corollary 5.  $|f| \leq c(S, T)$ .

Proof.  $|f| = f(S, T) \leq c(S, T)$ .

## The Max-flow Min-cut Theorem.

**Theorem.** The following conditions are equivalent:

1.  $f$  is a maxflow in  $G$ .
2. The residual network  $G_f$  contains no augmenting paths.
3.  $|f| = c(S, T)$  for some cut  $(S, T)$  in  $G$ . //minimum cut//

**Proof.** (3)  $\Rightarrow$  (1): Immediately follows from Corollary 5.

(1)  $\Rightarrow$  (2): Immediately follows from Lemma 2. (If  $G_f$  contains an augmenting path  $p$ , augmenting  $f$  along  $p$  will increase the flow.)

2. The residual network  $G_f$  contains no augmenting paths.
3.  $|f| = c(S, T)$  for some cut  $(S, T)$  in  $G$ .

(2)  $\Rightarrow$  (3):

Suppose  $G_f$  contains no augmenting path. Define

$$S = \{v : \text{there is a path from } s \text{ to } v \text{ in } G_f\},$$

$$T = V - S.$$

$(S, T)$  is a cut since  $s \in S$  and  $t \in T$  (no path from  $s$  to  $t$  in  $G_f$ ).

For all  $u \in S$ ,  $v \in T$ , we have  $f(u, v) = c(u, v)$ , since otherwise  $(u, v) \in E_f$  and  $v$  would be in  $S$ . So,  $f(S, T) = c(S, T)$ .

By Lemma 4,  $|f| = f(S, T) = c(S, T)$ .

## Edmonds-Karp Algorithm

- In the while loop of Ford-Fulkerson, find the augmenting path  $p$  with a breadth-first search, that is, the augmenting path is a shortest path from  $s$  to  $t$  in the residual network, where "shortest" is in terms of number of edges.
- Running time:  $O(VE^2)$  (to be shown).

# Analysis of the Edmonds-Karp Algorithm

**Lemma 6.** In the execution of Edmonds-Karp algorithm, for all  $v \neq s, t$ ,  $\delta_f(v)$  is nondecreasing with each flow augmentation where  $\delta_f(v) =$  shortest distance (# edges) from  $s$  to  $v$  in  $G_f$ .

**Proof.** By contradiction. Assume the lemma is not true.

Consider the first augmentation that decreases some  $\delta_f(\cdot)$ .

Let  $f$  and  $f'$  be the flows just **before** and **after** the augmentation.

Let  $v$  be the vertex s.t.  $\delta_f(v) > \delta_{f'}(v)$  and  $\delta_{f'}(v)$  is minimum among those nodes  $x$  with  $\delta_f(x) > \delta_{f'}(x)$ . Let  $p$  be a shortest

path from  $s$  to  $v$  in  $G_{f'}$ , and let  $(u, v)$  be the last edge of  $p$ .

So,  $(u, v) \in E_{f'}$ ,  $\delta_{f'}(u) + 1 = \delta_{f'}(v)$ , and  $\delta_f(u) \leq \delta_{f'}(u)$ .

- Case 1:  $(u, v) \in E_f$ . Then,  $\delta_f(u) + 1 \geq \delta_f(v)$ , and then

$$\delta_{f'}(v) = \delta_{f'}(u) + 1 \geq \delta_f(u) + 1 \geq \delta_f(v) > \delta_{f'}(v),$$

a contradiction.

- Case 2:  $(u, v) \notin E_f$ . Now,  $(u, v) \notin E$ , but  $(u, v) \in E_{f'}$ .

This means, the augmenting path contains edge  $(v, u)$ .

As Edmonds-Karp always augments flow along shortest paths,  $(v, u)$  is the last edge of a shortest path from  $s$  to  $u$  in  $G_f$ .

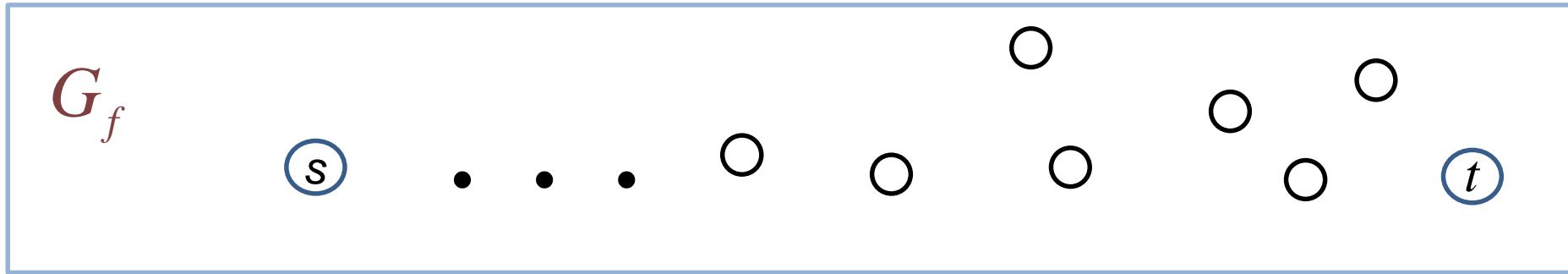
Therefore,  $\delta_f(u) = \delta_f(v) + 1 \Rightarrow \delta_f(u) + 1 \geq \delta_f(v)$ .

As in case 1, this will lead to a contradiction.

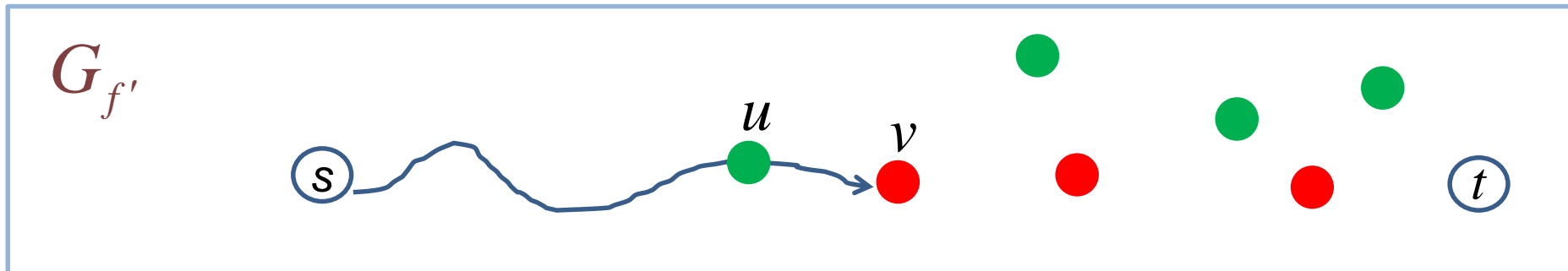
$\delta(\cdot)$  decreases for red nodes; does not decrease for green nodes.

$v$ : the red node closest to  $s$  in  $G_{f'}$ .

$u$ : predecessor of  $v$  on shortest path  $s$  to  $v$  in  $G_{f'}$ ; a green node.

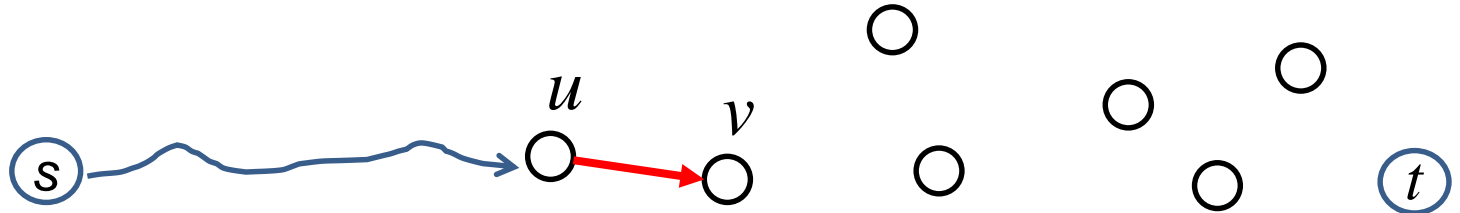


augmentation



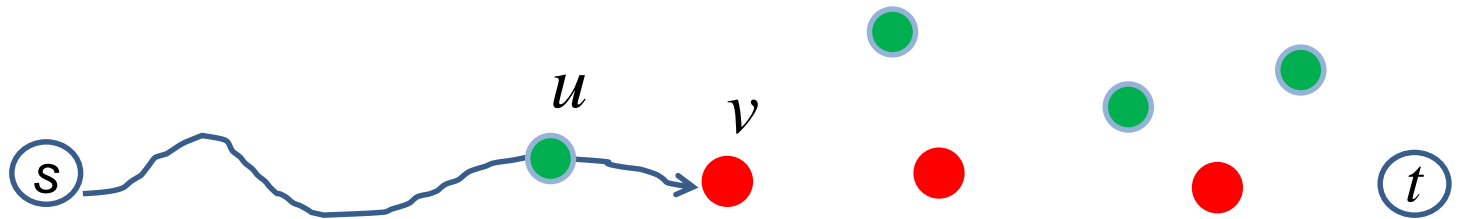
If edge  $(u, v)$  exists in  $G_f$

$G_f$



$$\delta_f(u) + 1 \geq \delta_f(v)$$

$G_{f'}$

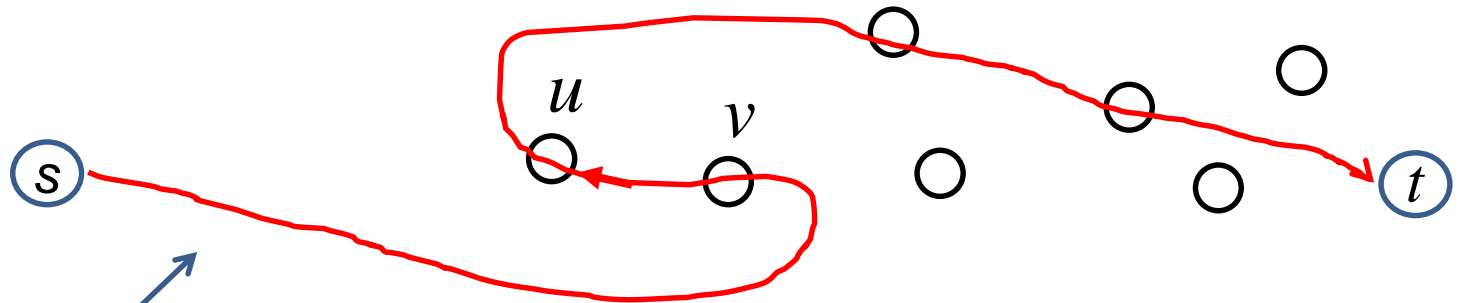


$$\delta_{f'}(u) + 1 = \delta_{f'}(v)$$



If edge  $(u, v)$  does not exist in  $G_f$

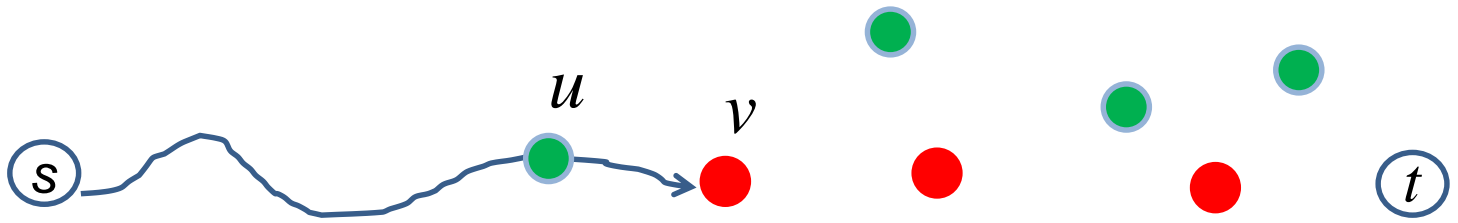
$G_f$



augmenting path

$$\delta_f(u) - 1 = \delta_f(v)$$

$G_{f'}$



$$\delta_{f'}(u) + 1 = \delta_{f'}(v)$$

**Theorem 7.** If Edmonds-Karp Algorithm runs on  $G = (V, E)$ , then the total number of flow augmentations is  $O(VE)$  and hence the total running time is  $O(VE^2)$ .

**Proof.** An edge  $(u, v)$  in  $G_f$  is **critical** on an augmenting path  $p$  if  $c_f(p) = c_f(u, v)$ . Every augmenting path has a critical edge. An edge  $(u, v)$  may become critical only if  $(u, v) \in E$  or  $(v, u) \in E$ . So there are at most  $2|E|$  edges that may become critical during the algorithm's execution. We will show that each of these edges may become critical at most  $|V|/2$  times, which will imply that during the execution of the Edmonds-Karp algorithm there are at most  $O(VE)$  augmentations.

**Claim:** an edge  $(u, v)$  can become critical at most  $|V|/2$  times.

- When  $(u, v)$  becomes critical in a residual network  $G_f$ ,

$$\delta_f(v) = \delta_f(u) + 1. \quad (1)$$

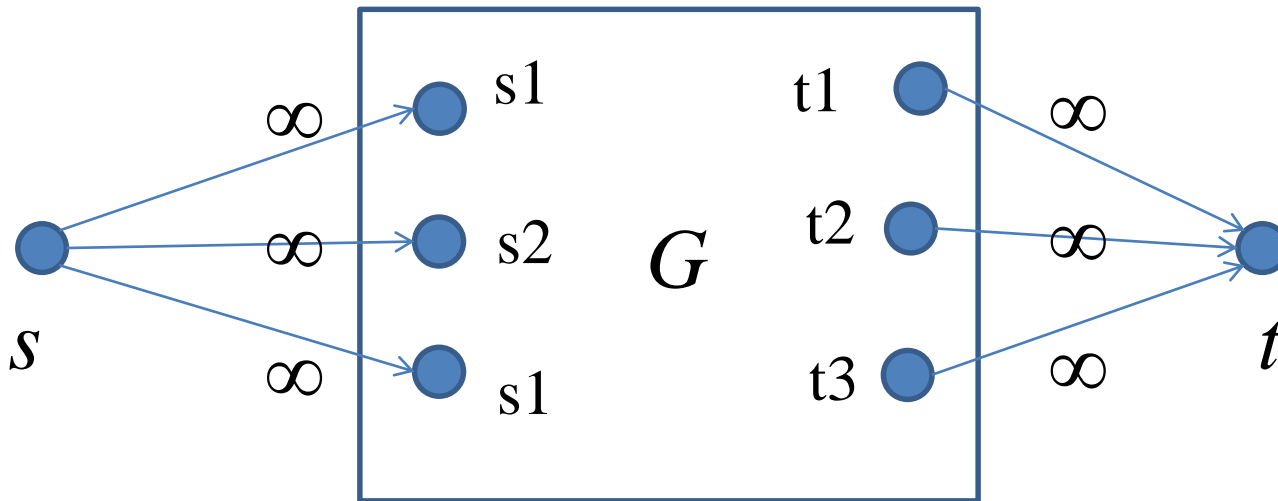
- After flow augmentation along the augmenting path,  $(u, v)$  disappears from the residual network.
- It may reappear later on another augmenting path only **after  $f(u, v)$  decreases**, which occurs only if edge  $(v, u)$  appears on an augmenting path in some  $G_{f'}$ , in which case,

$$\begin{aligned} \delta_{f'}(u) = \delta_{f'}(v) + 1 &\geq \delta_f(v) + 1 && \text{by Lemma 6} \\ &\geq \delta_f(u) + 2 && \text{by (1).} \end{aligned}$$

- Thus, if  $(u, v)$  becomes critical more than once, then for each additional time  $(u, v)$  becomes critical,  $\delta(u)$  increases by at least 2.
  - When  $(u, v)$  becomes critical for the last time,  $\delta(u) \leq |V| - 2$ .
  - Thus,  $(u, v)$  can become critical no more than  $|V|/2$  times.
- This proves the claim and the theorem.

# Networks with multiple sources and sinks

- $G = (V, E)$ : flow network with  
 $m$  sources:  $\{s_1, s_2, \dots, s_m\}$   
 $n$  sinks:  $\{t_1, t_2, \dots, t_m\}$



# Maximum Bipartite Matching

- $G = (V, E)$ : undirected graph
- Bipartite graph: if  $V$  can be partitioned into  $L$  and  $R$  such that all edges in  $E$  go between  $L$  and  $R$ .
- Theorem:  $G$  is bipartite iff it has no cycles of odd length.
- Matching: a set of edges  $M \subseteq E$  such that every vertex in  $V$  is an endpoint of at most one edge in  $M$ .
- Maximum matching: a matching with the max cardinality.
- A maximum matching of a bipartite graph can be found using the Ford-Fulkerson method.

# Edge-Disjoint Paths

- $G = (V, E)$ : a graph
- Edge-disjoint paths: two paths are edge-disjoint if they do not share any edges.
- Problem: Given a **directed** graph  $G = (V, E)$  and two nodes  $s, t$ , find a maximum number of edge-disjoint paths from  $s$  to  $t$ .
- Problem: Given an **undirected** graph  $G = (V, E)$  and two nodes  $s, t$ , find a maximum number of edge-disjoint paths from  $s$  to  $t$ .

## Node-Disjoint Paths

- $G = (V, E)$ : a graph
- Node-disjoint paths: two paths from  $s$  to  $t$  are node-disjoint if they do not share any intermediate nodes.
- Problem: Given a **directed** graph  $G = (V, E)$  and two nodes  $s, t$ , find a maximum number of node-disjoint paths from  $s$  to  $t$ .
- Problem: Given an **undirected** graph  $G = (V, E)$  and two nodes  $s, t$ , find a maximum number of node-disjoint paths from  $s$  to  $t$ .



# Image Segmentation

- A fundamental problem in computer vision.
- Given a digital image (a set of pixels), we want to partition it into multiple segments.
- In a simple case, we just want to divide the image into two segments: the foreground and the background.
- Represent the image by an undirected graph  $G = (V, E)$ , where  $V$  is the set of pixels and there is an edge between two pixels iff there are neighbors.



- Each pixel  $i$  has a likelihood (goodness)  $a_i > 0$  to belong to the foreground and a likelihood  $b_i > 0$  to belong to the background.
- Each edge  $(i, j) \in E$  is associated with a separation penalty  $p_{ij} = p_{ji} > 0$ , which is incurred if pixels  $i$  and  $j$  are placed in different segments.

- **Problem:** Given a pixel graph  $G = (V, E)$ , likelihood functions  $a, b : E \rightarrow \mathbb{R}^+$  and penalty function  $p : E \rightarrow \mathbb{R}^+$ , we want to partition  $V$  into two sets  $A$  and  $B$  and **maximize**

$$Q(A, B) =$$

$$\sum_{i \in A} a_i + \sum_{i \in B} b_i - \sum \{ p_{ij} : (i, j) \in E, i, j \text{ in different segments} \}$$

- Or, equivalently, **minimize**

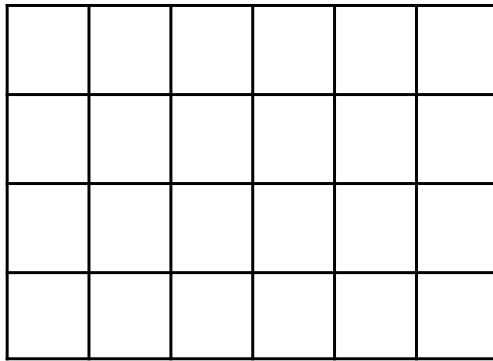
$$Q'(A, B) = \left( \sum_{i \in V} a_i + \sum_{i \in V} b_i \right) - Q(A, B)$$

$$= \sum_{i \in B} a_i + \sum_{i \in A} b_i + \sum \{ p_{ij} : (i, j) \in E, i, j \text{ in different segments} \}$$

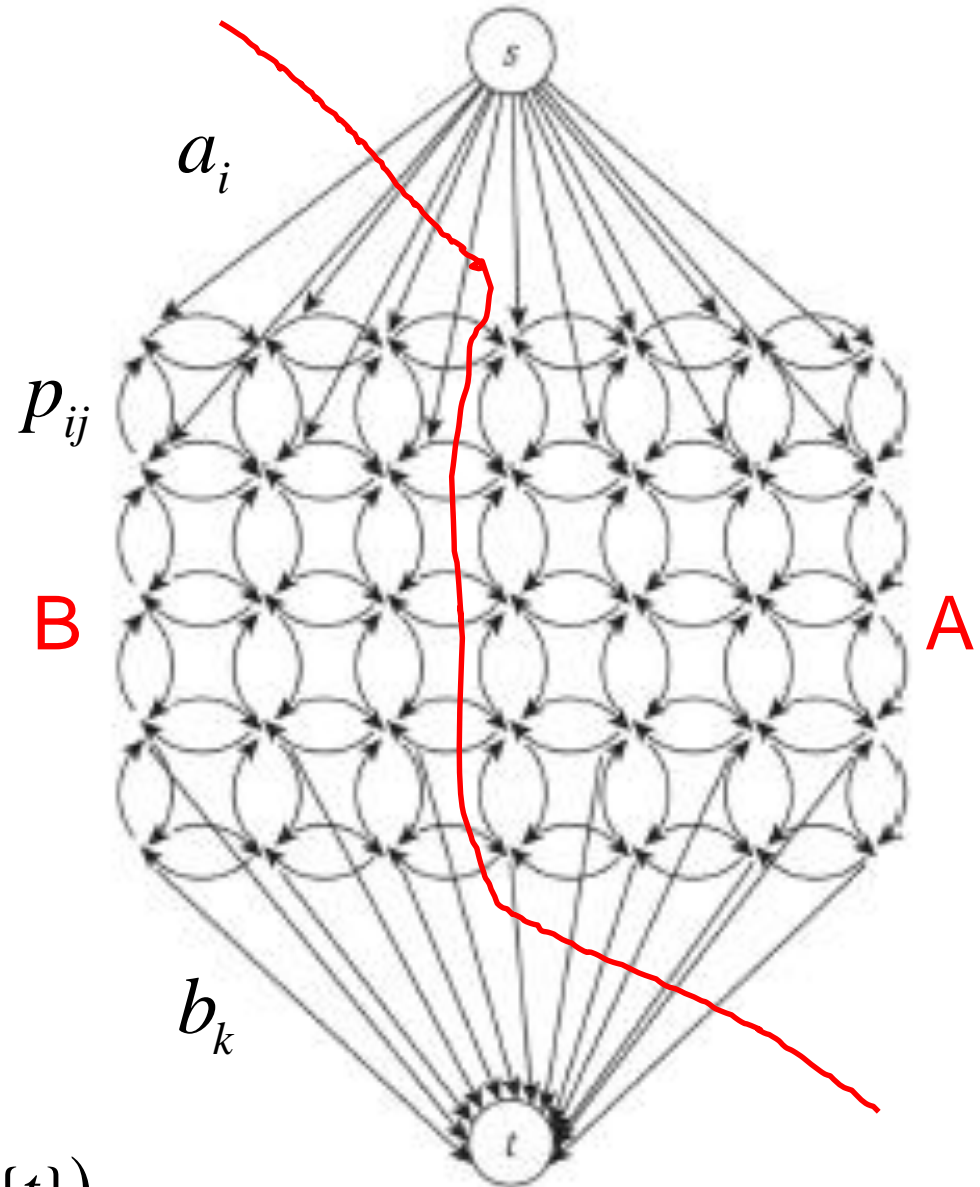
- We can solve the image segmentation problem by converting it to a flow network. Let  $G = (V, E)$  be the pixel graph.
- Introduce two new vertices: a source  $s$  and a sink  $t$ .
- Connect  $s$  to each pixel  $i \in V$  with capacity  $a_i$ .
- Connect  $t$  from each pixel  $i \in V$  with capacity  $b_i$ .
- Replace each edge  $(i, j) \in E$  with two directed edges  $(i, j)$  and  $(j, i)$  with capacities  $p_{ij}$  and  $p_{ji}$ .
- Relationship between the pixel graph  $G = (V, E)$  and the constructed flow network  $G' = (V', E')$ :

Segmentations of  $G \xleftrightarrow{\text{1-1 correspondence}} \text{Cuts of } G'$

$$Q'(A, B) = c(A \cup \{s\}, B \cup \{t\})$$



Pixel graph  $G = (V, E)$



$$Q'(A, B) = c(A \cup \{s\}, B \cup \{t\})$$

# Generic Push-Relabel Algorithms for Maximum Flows

Running time:  $O(V^2 E)$

# Preflows

- Flow net  $G = (V, E)$ , capacity function  $c$ , source  $s$ , sink  $t$ .
- A **preflow** is a function  $f : V \times V \rightarrow \mathbb{R}$ , satisfying
  - Capacity constraint:  $\forall u, v \in V, f(u, v) \leq c(u, v)$ .
  - Skew symmetry:  $\forall u, v \in V, f(u, v) = -f(v, u)$ .
  - **Relaxed** flow conservation:  $\forall u \in V - \{s\}$ ,

$$f(V, u) \geq 0$$

- The quantity  $e(u) = f(V, u)$  is called the **excess flow** into  $u$ .
- Vertex  $u \neq t$  is **overflowing** if  $f(V, u) > 0$ .



- **Height function:** a function  $h : V \rightarrow \mathbb{N}_0$ , satisfying
  - $h(s) = |V|$
  - $h(t) = 0$
  - $h(u) \leq h(v) + 1$  for every residual edge  $(u, v) \in E_f$ .
- **Note:** a height function is defined relative to a preflow.
- **Lemma:** If  $h(u) > h(v) + 1$  then  $(u, v) \notin E_f$ .

## Operation Push( $u, v$ )

- Applicable when:

$u$  is overflowing,  $c_f(u, v) > 0$ , and  $h(u) = h(v) + 1$ .

- **Action:** push  $\Delta_f(u, v) = \min \{e(u), c_f(u, v)\}$  units of flow from  $u$  to  $v$ .

- $f(u, v) \leftarrow f(u, v) + \Delta_f(u, v)$ .

- $f(v, u) \leftarrow -f(u, v)$ .

- $e(u) \leftarrow e(u) - \Delta_f(u, v)$ .

- $e(v) \leftarrow e(v) + \Delta_f(u, v)$ .

- The operation  $\text{Push}(u, v)$  is called a **push** from  $u$  to  $v$ .
- **Saturating push:** edge  $(u, v)$  becomes saturated (i.e.,  $c_f(u, v) = 0$ ) after the push.
- **Nonsaturating push:**  $c_f(u, v) > 0$  after the push.
- **Lemma:** After a nonsaturating push from  $u$  to  $v$ , vertex  $u$  is no longer overflowing.  
**Proof:** After the push, either  $e(u) = 0$  or  $c_f(u, v) = 0$ .

## Operation Relabel( $u$ )

- **Applicable when:**
  - $u \notin \{s, t\}$  is overflowing and
  - $h(u) \leq h(v)$  for all edges  $(u, v) \in E_f$ .
- **Action:** increase the height of  $u$ .
  - $h(u) \leftarrow 1 + \min \{h(v) : (u, v) \in E_f\}$ .
- **Note:** since  $u$  is overflowing, there is at least one edge  $(u, v) \in E_f$ , so the above min is not over an empty set.

# Initialize-Preflow( $G, s, t$ )

- Initial preflow: For all  $u, v \in V$ ,

$$f(u, v) = \begin{cases} c(u, v) & \text{if } u = s \\ -c(u, v) & \text{if } v = s \\ 0 & \text{otherwise} \end{cases}$$

- Corresponding excess flow function:

$$e(v) = \begin{cases} c(s, v) & \text{if } (s, v) \in E \\ -\sum \{c(s, x) : (s, x) \in E\} & \text{if } v = s \\ 0 & \text{otherwise} \end{cases}$$

- Initial height function:

$$h(u) = \begin{cases} |V| & \text{if } u = s \\ 0 & \text{otherwise} \end{cases}$$

# Generic-Push-Relabel Algorithm

1. Initialize-Preflow( $G, s, t$ ) // initialize a preflow  $f$  //
2. **while** there is an applicable **push** or **relabel** operation  
    **do select** an applicable operation and perform it

# Correctness of Generic-Push-Relabel

**Lemma 1.** If  $u \neq t$  is an overflowing vertex, then either a push or relabel operation can be applied to it.

**Lemma 2.** Whenever a relabel operation is applied to a vertex  $u$ , its height  $h(u)$  increases by at least 1.

**Lemma 3.** During the execution of the algorithm,  $h$  is always a height function.

**Lemma 4.** During the execution of the algorithm,  $f$  is always a preflow.

**Lemma 5.** Let  $f$  be a preflow. If there is a height function  $h$  relative to  $f$ , then there is no path from  $s$  to  $t$  in the residual network  $G_f$ .

**Proof.** Otherwise, if there is a simple path  $p$  in  $G_f$  from  $s$  to  $t$ , then

$$h(s) - h(t) \leq \text{length}(p) \leq |V| - 1$$

contradicting the fact that  $h(s) - h(t) = |V|$ .



**Theorem.** If/when the algorithm terminates, the preflow it computes is a maximum flow.

**Proof.** When the algorithm terminates:

- $f$  is a preflow (by Lemma 4).
- No vertex is overflowing (by Lemma 1).
- So,  $f$  is a flow.
- $h$  is a height function (by Lemma 3).
- There is no augmenting path in  $G_f$  (by Lemma 5).
- So,  $f$  is a maxflow (by Max-flow Min-cut Theorem).

# Analysis of Generic-Push-Relabel

**Lemma 1.** Let  $f$  be a preflow. Then, for any overflowing vertex  $u$ , there is a path from  $u$  to  $s$  in  $G_f$ .

**Lemma 2.** At any time during the execution of the algorithm,  $h(u) \leq 2|V| - 1$  for any node  $u \in V$ .

**Proof.** When a vertex  $u$  is relabeled, it is overflowing and has a simple path to  $s$  (which is still true after the relabel). Since the path has at most  $|V| - 1$  edges,  $h(u) \leq 2|V| - 1$ .

**Corollary (bound on relabel operations).** The total number of relabel operations is at most  $(2|V| - 1)(|V| - 2) < 2|V|^2$ .

**Lemma 3 (bound on saturating pushes).** The total number of saturating pushes is at most  $2|V||E|$ .

**Proof.**  $\text{Push}(u, v)$  may occur only if  $(u, v) \in E$  or  $(v, u) \in E$ . Between two consecutive **saturating** pushes from  $u$  to  $v$ ,  $h(u)$  increases by at least 2. Reasons:

- Between two saturating pushes from  $u$  to  $v$ , there must be a push from  $v$  to  $u$ .
- At the 1st  $\text{Push}(u, v)$ : say  $h(u) = a$ .
- At  $\text{Push}(v, u)$ :  $h(v) = h(u) + 1 \geq a + 1$ .
- At the 2nd  $\text{Push}(u, v)$ :  $h(u) = h(v) + 1 \geq a + 2$ .

So, for each  $(u, v) \in E$  or  $(v, u) \in E$ , saturating  $\text{Push}(u, v)$  may occur no more than  $|V|$  times.

**Lemma 4 (bound on nonsaturating pushes).** The number of nonsaturating pushes is less than  $4|V|^2 (|V| + |E|)$ .

**Proof.** Define  $\Phi = \sum_{e(u)>0} h(u)$ . Initially,  $\Phi = 0$ .

- Relabeling a vertex  $u$  increases  $\Phi$  by less than  $2|V|$ .
- A saturating push from  $u$  to  $v$  increases  $\Phi$  by less than  $2|V|$ .
- Total amount of increase to  $\Phi$  is less than  $2|V| \cdot (2|V|^2 + 2|V||E|) = 4|V|^2 (|V| + |E|)$ .
- A nonsaturating push from  $u$  to  $v$  decreases  $\Phi$  by at least 1.
- Thus, the total number of nonsaturating pushes is less than  $4|V|^2 (|V| + |E|)$ .

**Lemma 5.** Each relabel can be done in  $O(V)$  time and each push can be done in  $O(1)$  time.

**Theorem.** The running time of the generic push-relabel algorithm is  $O(V^2 E)$ .

**Proof.**

- Total time for relabels:  $O(V^3)$ .
- Total time for saturating pushes:  $O(VE)$ .
- Total time for nonsaturating pushes:  $O(V^2 E)$ .

# The Relabel-to-Front Algorithm

Running time:  $O(V^3)$

## Admissible edges and networks

- An edge  $(u, v)$  is **admissible** if
$$c_f(u, v) > 0 \text{ and } h(u) = h(v) + 1.$$
- **Admissible network:**  $G_{f,h} = (V, E_{f,h})$ , where  $E_{f,h}$  is the set of admissible edges. It is a subgraph of  $G_f$ .

**Lemma 1.** The admissible network  $G_{f,h}$  is acyclic.

**Proof.** The height function  $h(\cdot)$  is decreasing along any path in  $G_{f,h}$ .

When is  $\text{Push}(u, v)$  applicable? How does it affect  $G_{f,h}$  ?

**Lemma 2.** If a vertex  $u$  is overflowing and edge  $(u, v)$  is admissible, then  $\text{Push}(u, v)$  is applicable. The operation does not create any new admissible edges, but it may cause  $(u, v)$  to become inadmissible.

**Proof.** The  $\text{Push}(u, v)$  operation reduces  $c_f(u, v)$  and increases  $c_f(v, u)$ . If  $c_f(u, v)$  becomes 0,  $(u, v)$  becomes inadmissible. Since  $h(u) = h(v) + 1$ ,  $(v, u)$  cannot become admissible.



When is  $\text{Relabel}(u)$  applicable? How does it affect  $G_{f,h}$  ?

**Lemma 3.** If a vertex  $u \notin \{s,t\}$  is overflowing and there are no admissible edges leaving  $u$ , then  $\text{Relabel}(u)$  is applicable. **After the relabel operation, there is at least one admissible edge leaving  $u$ , but there are no admissible edges entering  $u$ .**

**Proof.** Only the last claim needs a proof.

If, after the relabel,  $(v,u)$  is an admissible edge entering  $u$ , then  $h(v) = h(u) + 1$ . Before the relabel of  $u$ ,  $h(v) > h(u) + 1$  and thus  $(v,u) \notin E_f$ .  $\Rightarrow \Leftarrow$

# Neighbor lists

- Same as the adjacency lists of the flow network  $G = (V, E)$ , except that the list of  $u$  contains  $v$  iff  $(u, v) \in E$  or  $(v, u) \in E$ .
- $N(u)$ : the neighbor list of  $u$ . It contains those vertices  $v$  for which there may be a residual edge  $(u, v)$ .
- $head(N(u))$ : pointing to the first element in  $N(u)$ .
- $current(u)$ : pointing to the vertex currently under consideration in  $N(u)$ . Initially,  $current(u) \leftarrow head(N(u))$ .
- $next-neighbor(\bullet)$ :

# Discharging an overflowing vertex

- Discharge( $u$ ): push **all** excess flow of  $u$  thru admissible edges leaving  $u$ , relabeling  $u$  as necessary.
- Procedure Discharge( $u$ ) //after Discharge( $u$ ),  $e(u) = 0$ //  
**while**  $e(u) > 0$  **do**  
     $v \leftarrow \text{current}(u)$   
    **if**  $v = \text{NIL}$  **then**  
        Relabel( $u$ )  
         $\text{current}(u) \leftarrow \text{head}(N(u))$   
    **elseif**  $(u, v)$  is admissible **then**  
        Push( $u, v$ )  
    **else**  $\text{current}(u) \leftarrow \text{next-neighbor}(v)$

## Algorithm Relabel-to-Front( $G, s, t$ )

- 1 Initialize-Preflow( $G, s, t$ )
- 2  $L \leftarrow V[G] - \{s, t\}$  in any order
- 3 Initialize  $current(u)$  for each  $u \in V[G] - \{s, t\}$
- 4  $u \leftarrow head(L)$
- 5 **while**  $u \neq \text{NIL}$  **do**
- 6     Discharge( $u$ )
- 7     **if**  $u$  has been relabeled during Discharge( $u$ )
- 8         **then** move  $u$  to the front of  $L$
- 9      $u \leftarrow next\text{-neighbor}(u)$

**Lemma 4.** Relabel-to-Front performs push and relabel operations only when they are applicable.

**Lemma 5.** At each test in line 5 of Relabel-to-Front,  $L$  is a topological sort of  $G_{f,h} - \{s, t\}$  and no vertex before  $u$  in the list has excess flow.

**Corollary.** When Relabel-to-Front terminates, there are no applicable push or relabel operations (since by Lemma 5 there is no overflowing node).

**Theorem.** Relabel-to-Front is an implementation of the generic push and relabel algorithm.

Proof of Lemma 5 (part 1). By induction, we show  $L$  is a topological sort of  $G_{f,h} - \{s, t\}$ .

- For iteration 1, it is true, since initially  $E_{f,h} = \phi$ .
- Assume that  $L$  is in topological order at the beginning of an iteration.
- During the iteration, we perform pushes and relables.
  - Pushes do not create any admissible edges (Lemma 2).
  - By Lemma 3,  $\text{Relabel}(u)$  may create admissible edges leaving  $u$ , but after the relabel there will be no admissible edge entering  $u$ . By moving  $u$  to the front of  $L$ ,  $L$  remains in topological order.

Proof of Lemma 5 (part 2). By induction, we show that vertices before  $u$  have no excess flow.

- Initially, it is true since  $u$  is at the front of  $L$ .
- Assume the property holds at the beginning of an iteration.
- Let  $u'$  be the vertex that will be the  $u$  in the next iteration.
- We will show that no vertex before  $u'$  has excess flow.
- If  $u$  is moved to front, it has no excess flow (since it has been discharged), and it is the only vertex before  $u'$ .
- If  $u$  is not moved to front, vertices before  $u$  received no additional flow and thus still have no excess, and  $u$  itself now has no excess.

**Theorem.** The running time of Relabel-to-Front is  $O(V^3)$ .

Proof.

- The running time =

$$O\left(\begin{array}{l} \text{the total number of iterations (discharges)} \\ + \text{the time spent on executing the discharges} \end{array}\right)$$

- We first determine the number of discharges:
  - There are at most  $O(V^2)$  relabels.
  - Preceding each relabel there may be  $O(V)$  calls to Discharge. Similarly,  $O(V)$  discharges after the last relabel.
  - Thus, the total number of calls to Discharge is  $O(V^3)$ .



- Now we determine the total time spent within Discharge.
  - Total time for moving the pointer *current*:  $O(V^3)$ .
    - Preceding each  $\text{Relabel}(u)$ , it takes  $O(V)$  time to move *current*(*u*).
    - There are at most  $O(V^2)$  relabels.
  - Total time for relabels:  $O(V^3)$ .
  - Total time for saturating pushes:  $O(VE) \subseteq O(V^3)$ .
  - Total time for nonsaturating pushes:  $O(V^3)$ .
    - Each discharge has at most 1 nonsaturating push.