



Algorithm Analysis and Data Structures

CSCI 7432 - Fall 2022

Advanced Graph Algorithms: Flow Networks

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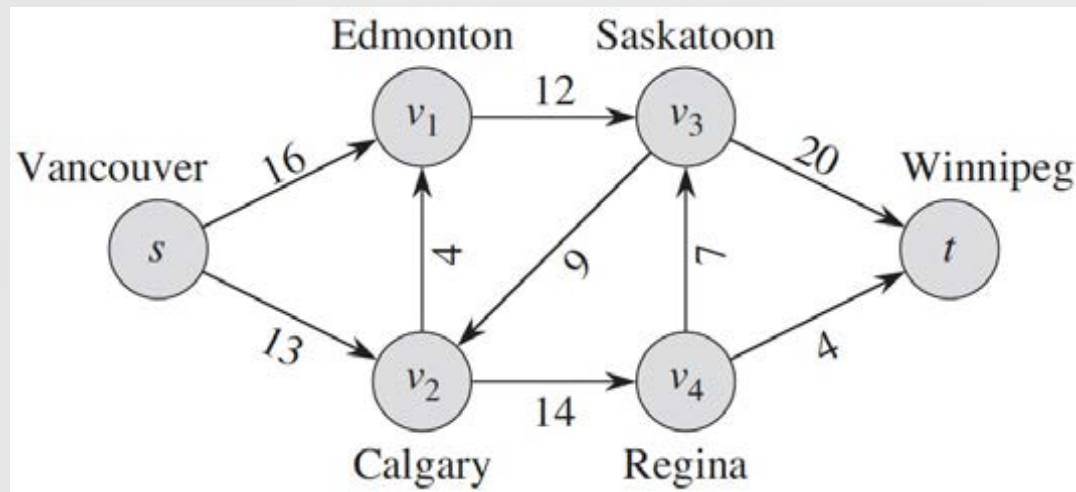


Flow Networks

The Maximum-Flow Problem

Flow Networks

- A **flow network** is a directed graph $G = (V, E)$ in which
 - each edge $(u, v) \in E$ has a nonnegative **capacity** $c(u, v) \geq 0$;
 - if $(u, v) \notin E$, then for convenience $c(u, v) = 0$;
 - there are two distinguished vertices: a **source** $s \in V$ and a **sink** $t \in V$.
- **Intuition:** Material originates at **source** s and is sent to **sink** t .
- **Example:** A flow network for the Lucky Puck Company's trucking problem.



Flow

- A **flow** in a **flow network** $G = (V, E)$ is a function $f: V \times V \rightarrow \mathbb{R}$ satisfying:

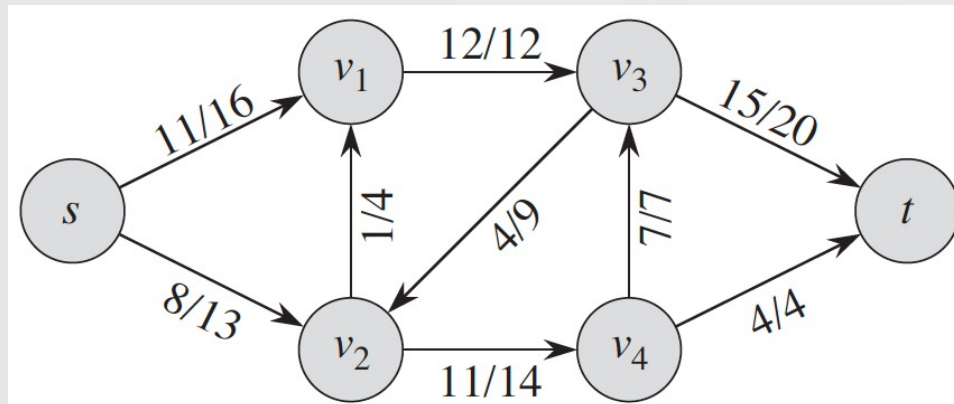
- **Capacity constraint**: For all $u, v \in V$, $0 \leq f(u, v) \leq c(u, v)$;

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

$$\underbrace{\sum_{v \in V} f(v, u)}_{\text{flow into } u} = \underbrace{\sum_{v \in V} f(u, v)}_{\text{flow out of } u}.$$

- **Value** of a flow f is: $|f| = \sum_{v \in V} f(s, v) - \sum_{v \in V} f(v, s)$.

- **Example**: Q: $|f| = ?$



A: $|f| = 19$

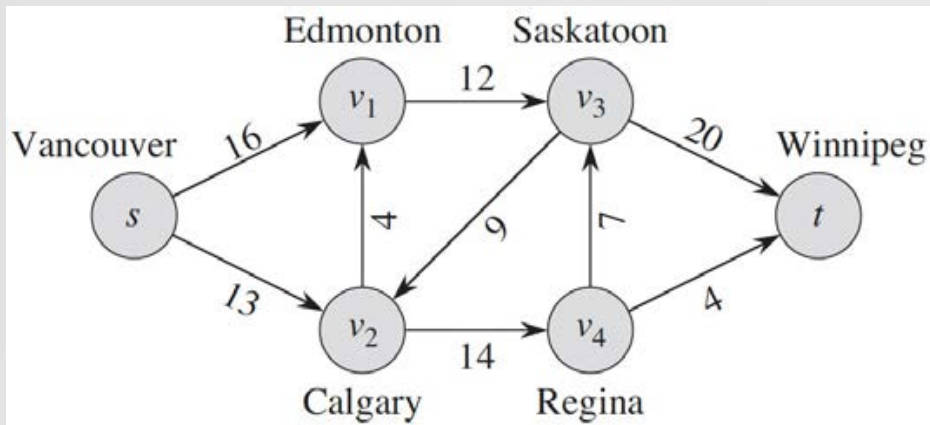
The Maximum-Flow Problem

Input: A flow network $G = (V, E)$ with source $s \in V$, sink $t \in V$, and capacity $c(u, v) \geq 0$ for each edge $(u, v) \in E$.

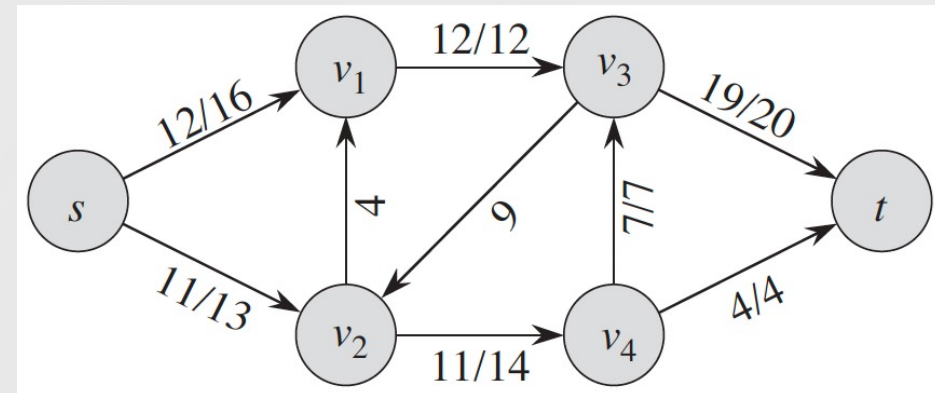
Output: A flow of **maximum value**.

- **Goal:** Find flow $f(u, v)$ for each edge $(u, v) \in E$ s.t.
 - For all $(u, v) \in E$, $0 \leq f(u, v) \leq c(u, v)$ – **Capacity constraint**
 - For all $u \in V - \{s, t\}$, $\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v)$ – **Flow conservation**
 - $|f| = \sum_{v \in V} f(s, v) - \sum_{v \in V} f(v, s)$ is **maximized**

- **Example:**



Max
flow:

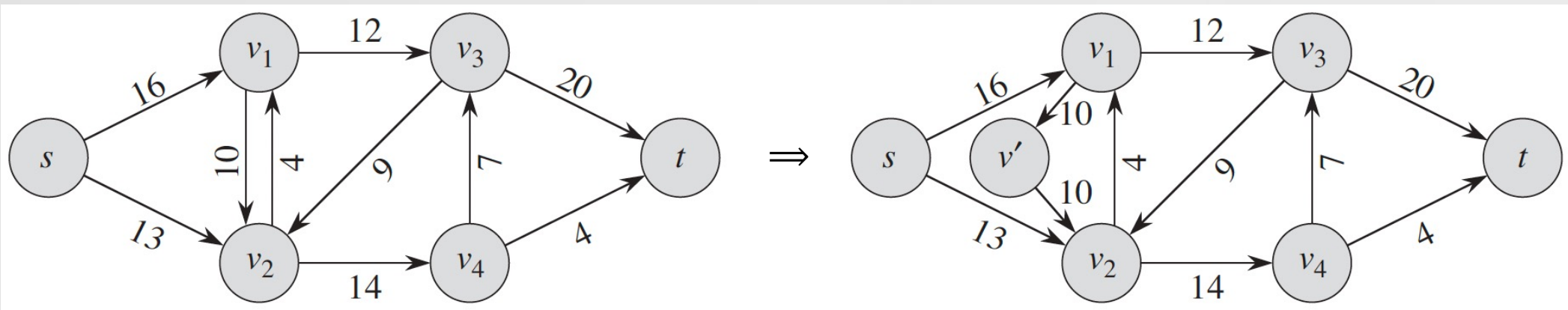


Flow Network Assumptions

Assumption 1: Self-loops are disallowed.

Assumption 2: If $(u, v) \in E$, then the reverse edge $(v, u) \notin E$.

- When both $(u, v) \in E$ and $(v, u) \in E$, which are called **antiparallel** edges, we transform G into an equivalent one with no **antiparallel** edges by choosing one of them, say (u, v) ,
 - Adding a new vertex v' , and
 - Replacing (u, v) by (u, v') and (v', v) , with $c(u, v') = c(v', v) = c(u, v)$





Flow Networks

The Minimum Cut Problem

The Minimum s - t Cut Problem

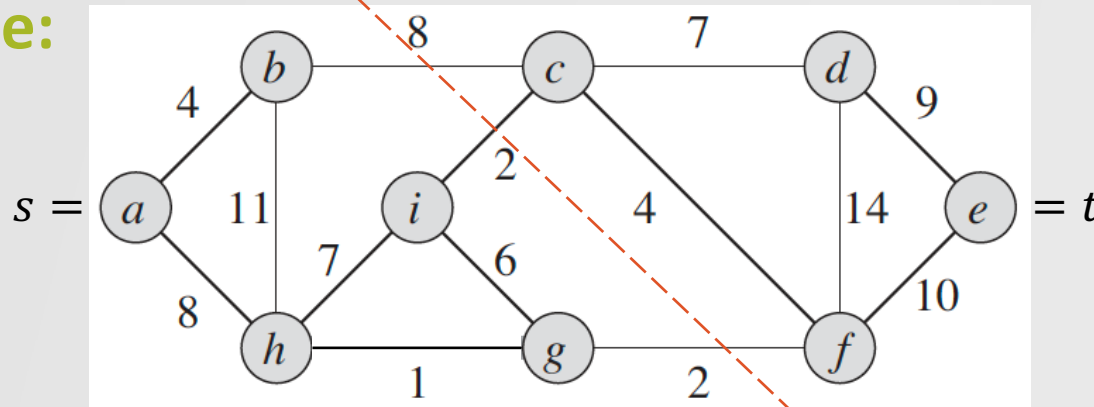
Input: An **undirected** graph $G = (V, E)$, two vertices $s, t \in V$, and weight $w(u, v) \geq 0$ for each edge $(u, v) \in E$.

- A **s - t cut** (S, T) of G is a **partition** of V into two sets S and $T = V - S$ such that $s \in S$ and $t \in T$.

Output: Find a **s - t cut** (S, T) of G with **minimum** weight:

$$w(S, T) = \sum_{u \in S} \sum_{v \in T} w(u, v).$$

- **Example:**



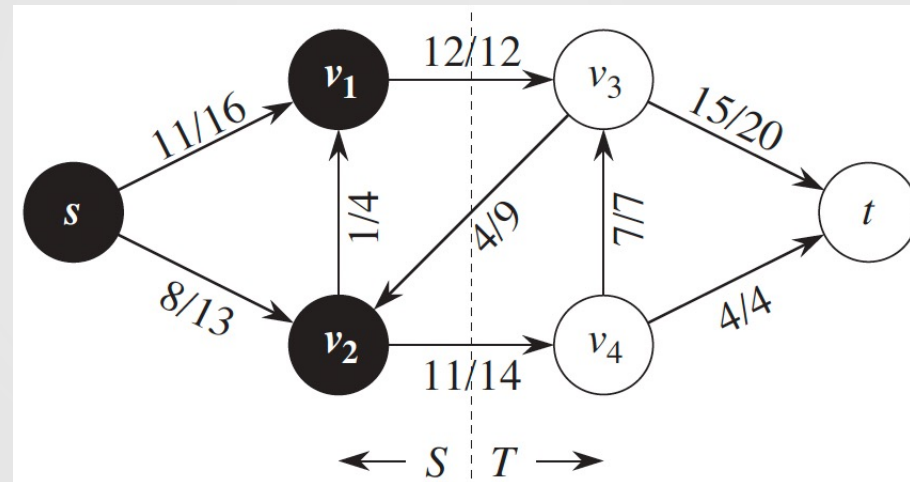
- A minimum s - t cut:
 $S = \{a, b, i, h, g\}$
 $T = \{c, d, e, f\}$
- $w(S, T) = 8 + 2 + 2 = 12$

Cuts of Flow Networks

- A **cut** (S, T) of a **flow network** $G = (V, E)$ is also a partition of V into two sets S and $T = V - S$ such that $s \in S$ and $t \in T$.
- For a flow f , the **net flow** across cut (S, T) is

$$f(S, T) = \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{u \in S} \sum_{v \in T} f(v, u).$$

- **Example:** $f(S, T) = ?$



A: $f(S, T) = 19$

Net Flow Across A Cut ^(1/2)

For a flow f , the **net flow** across cut (S, T) is

$$f(S, T) = \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{u \in S} \sum_{v \in T} f(v, u).$$

Claim:

- $f(X, X) = 0$
- If $X \cap Y = \emptyset$, then

$$f(X \cup Y, Z) = f(X, Z) + f(Y, Z)$$

$$f(Z, X \cup Y) = f(Z, X) + f(Z, Y)$$

Net Flow Across A Cut ^(2/2)

For a flow f , the **net flow** across cut (S, T) is

$$f(S, T) = \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{u \in S} \sum_{v \in T} f(v, u).$$

▷ **Lemma:** For any flow f and any cut (S, T) , we have $f(S, T) = |f|$.^{*}

Proof. *Recall:* $|f| = \sum_{v \in V} f(s, v) - \sum_{v \in V} f(v, s) = f(\{s\}, V)$

$$\begin{aligned} f(S, T) &= f(S, V) - f(S, S) \\ &= f(S, V) - 0 \\ &= f(\{s\}, V) + f(S - \{s\}, V) \end{aligned}$$

Due to **flow conservation**, $\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v)$ for $u \in V - \{s, t\}$.

So, $f(S - \{s\}, V) = 0 \implies f(S, T) = f(\{s\}, V) = |f|$ □

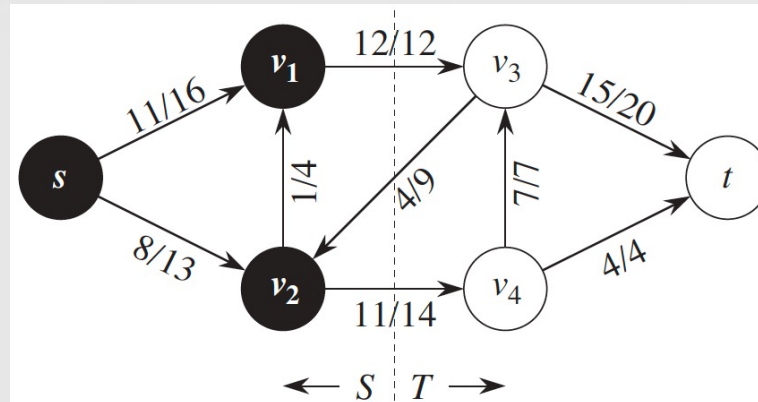
**See Lemma 26.4 on p.721-722 of the textbook for a complete proof.*

Capacity of A Cut

- The **capacity** of cut (S, T) is

$$c(S, T) = \sum_{u \in S} \sum_{v \in T} c(u, v).$$

- **Example:** $c(S, T) = ?$



A: $c(S, T) = 26$

Corollary: The value of any **flow** \leq capacity of any **cut**.

Proof. $|f| = f(S, T) = \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{u \in S} \sum_{v \in T} f(v, u)$
 $\leq \sum_{u \in S} \sum_{v \in T} c(u, v) = c(S, T).$ □

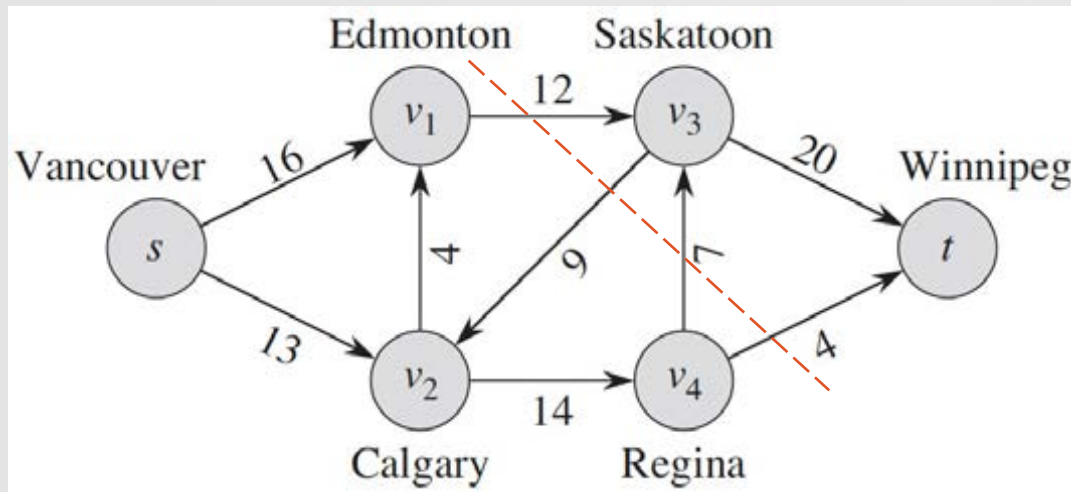
Minimum Cut of A Flow Network

Input: A flow network $G = (V, E)$ with source $s \in V$, sink $t \in V$, and capacity $c(u, v) \geq 0$ for each edge $(u, v) \in E$.

Output: A s - t cut (S, T) of G with **minimum capacity**:

$$c(S, T) = \sum_{u \in S} \sum_{v \in T} c(u, v).$$

- **Example:**



- A minimum s - t cut:
 $S = \{s, v_1, v_2, v_4\}$
 $T = \{v_3, t\}$
- $c(S, T) = ?$



The Ford-Fulkerson Method

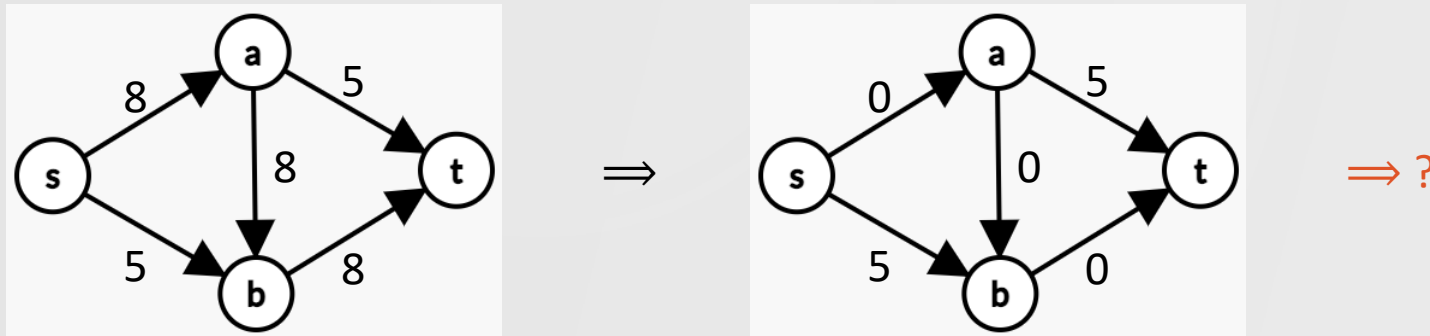
The Ford-Fulkerson Method

Ford-Fulkerson-Method(G, s, t)

- 1 initialize flow f to 0
- 2 **while** there exists an *augmenting path* p in the *residual network* G_f
- 3 augment flow f along p
- 4 **return** f

- The *residual network* consists of edges with capacities that represent how we can change the flow on edges of G .
- An *augmenting path* p is a *simple path* from s to t in the *residual network*.

• Example:





The Ford-Fulkerson Method

Residual Network

Residual Network

- Given a flow f in network $G = (V, E)$, the **residual network** consists of edges with capacities that represent how we can change the flow in G .
- That's the **residual capacity**:

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

- Then, the **residual network** is $G_f = (V, E_f)$, where

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

Residual Network Examples ^(1/2)

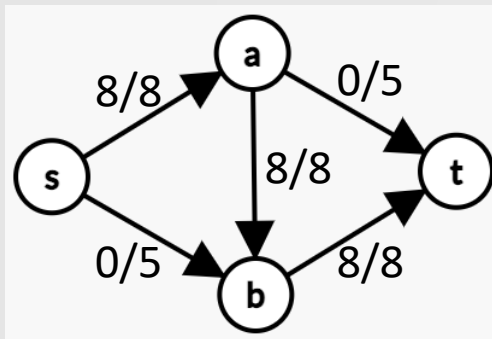
- The **residual network** is $G_f = (V, E_f)$, where

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

- The **residual capacity**:

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

- **Example 1:** Given a flow f in network G as follows, what is G_f ?

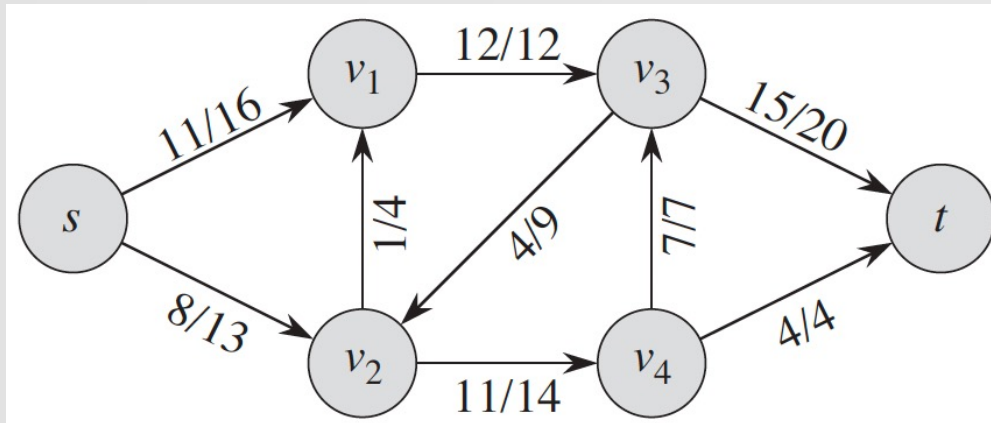


Residual Network Examples (2/2)

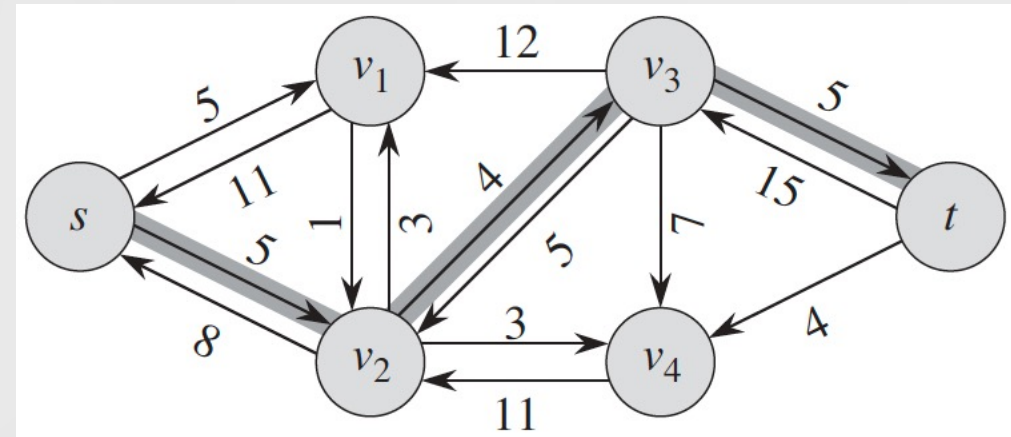
- The *residual capacity*:

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

- Example 2:**



G with f



G_f



The Ford-Fulkerson Method

Augmenting Path

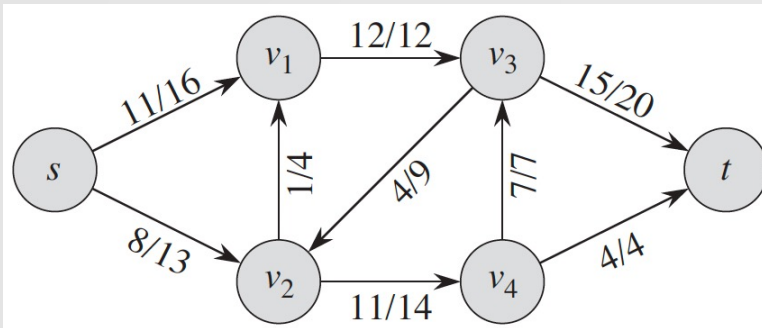
Augmenting Path

- An **augmenting path** p is a simple path from s to t in G_f .
- The flow value can be increased along an **augmenting path** p by

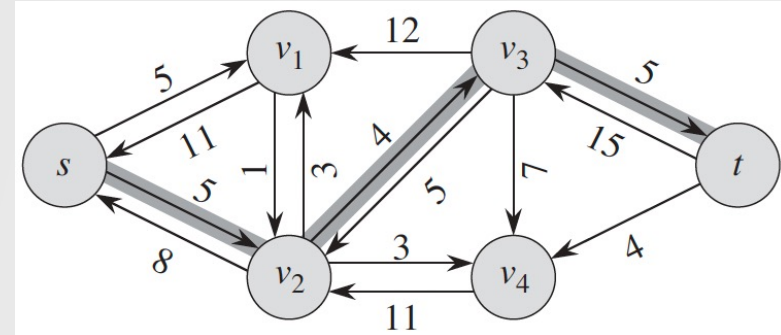
$$c_f(p) = \min_{(u,v) \in p} \{c_f(u,v)\}.$$

• Example:

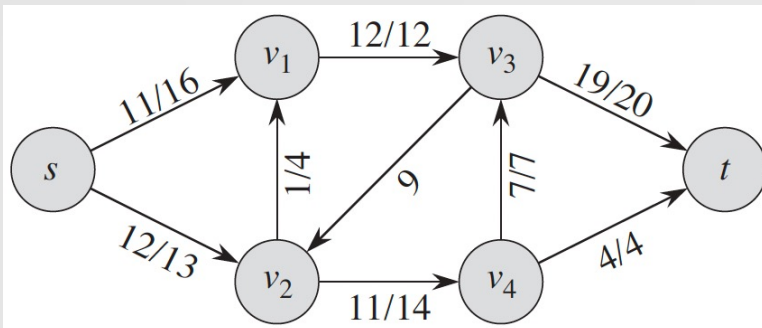
G with f_1 :



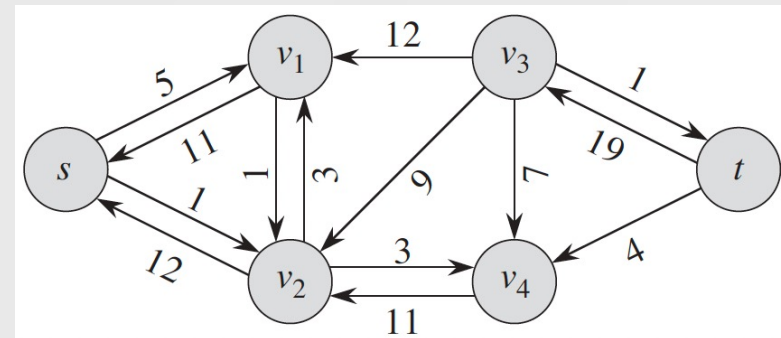
G_{f_1} :



G with f_2 :



G_{f_2} :

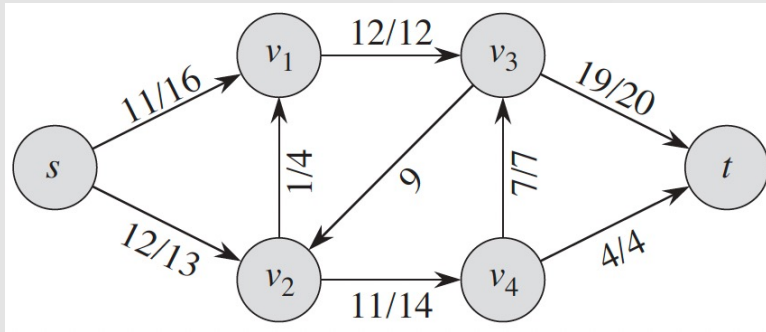


Augmented Flow Network

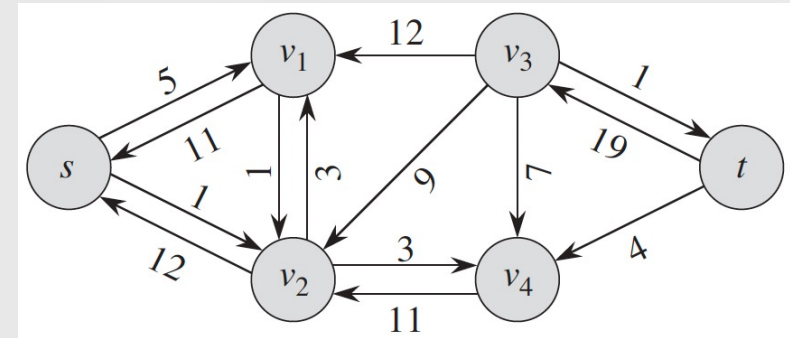
- When there is no augmenting path in G_f , f is a maximum flow in G .

- **Example:**

G with f_2 :



G_{f_2} :



- There is no augmenting path in G_{f_2} .
 - Consider cut (S, T) in G_{f_2} : $S = \{s, v_1, v_2, v_4\}$ and $T = \{v_3, t\}$
 - No edges cross the cut in the direction from S to T .
- Thus, the maximum flow in G is f_2 and $|f_2| = 23$.



The Ford-Fulkerson Method

Max-Flow Min-Cut Theorem

Max-Flow Min-Cut Theorem (1/3)

Max-Flow Min-Cut Theorem* If f is a flow in a flow network $G = (V, E)$ with source s and sink t , the the following conditions are equivalent:

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

Proof. (1) \Rightarrow (2): (prove by contraposition)

- Assume there is an augmenting path in G_f .
- Then the flow value could be increased.

(3) \Rightarrow (1):

- $|f| \leq c(S, T)$ for any cut (S, T) . - According to Corollary (on slide 12)

* This is Theorem 26.6 on p.723 of the textbook.

Max-Flow Min-Cut Theorem (2/3)

Max-Flow Min-Cut Theorem If f is a flow in a flow network $G = (V, E)$ with source s and sink t , the the following conditions are equivalent:

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

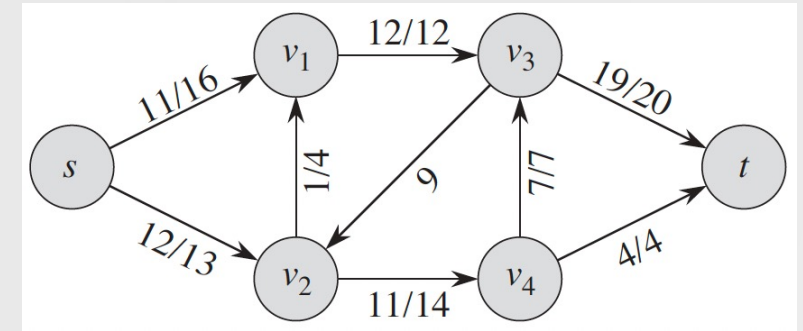
Proof. (cont'd) (2) \Rightarrow (3): When G_f contains no augmenting paths, define a cut (S, T) s.t.

- $S = \{v \in V : \exists s \rightsquigarrow v \text{ path in } G_f\}$ ($s \in S$)
- $T = V - S$ ($t \in T$)

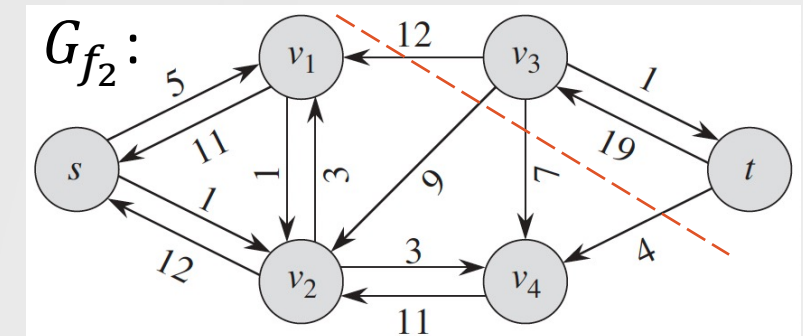
For any $u \in S$ and $v \in T$, we must have $c_f(u, v) = 0$.

By Lemma (on slide 11), we have $|f| = f(S, T)$.

Need to show: $f(S, T) = c(S, T)$.



G with f_2



Max-Flow Min-Cut Theorem (3/3)

Max-Flow Min-Cut Theorem If f is a flow in a flow network $G = (V, E)$ with source s and sink t , the the following conditions are equivalent:

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

Proof. (cont'd) (2) \Rightarrow (3): Show that $f(S, T) = c(S, T)$.

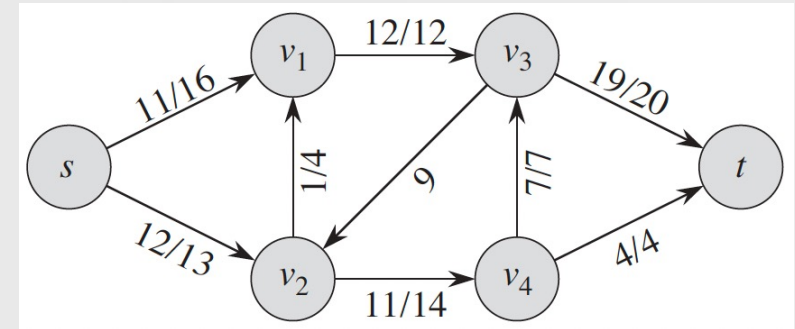
For any $u \in S$ and $v \in T$, we must have $c_f(u, v) = 0$.

Based on definitions of residual capacity,

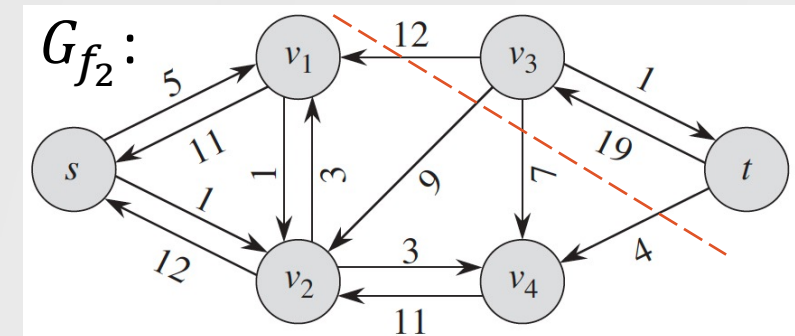
- If $(u, v) \in E$, then $c(u, v) = f(u, v)$
- If $(v, u) \in E$, then $f(v, u) = 0$

$$\begin{aligned} f(S, T) &= \sum_{u \in S} \sum_{v \in T} f(u, v) - \sum_{u \in S} \sum_{v \in T} f(v, u) \\ &= \sum_{u \in S} \sum_{v \in T} c(u, v) = c(S, T) \end{aligned}$$

□



G with f_2





The Ford-Fulkerson Method

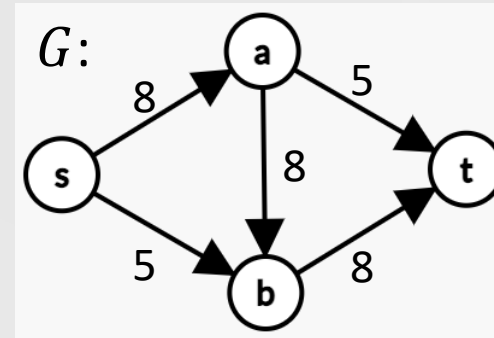
The Basic Ford-Fulkerson Algorithm

The Basic Ford-Fulkerson Algorithm

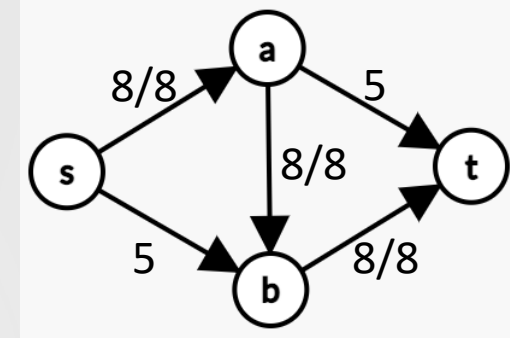
FORD-FULKERSON(G, s, t)

- 1 for each edge $(u, v) \in G.E$
- 2 $(u, v).f = 0$
- 3 build the residual network G_f
- 4 **while** there exists a path p from s to t in G_f
- 5 augment f by $c_f(p)$
- 6 update G_f

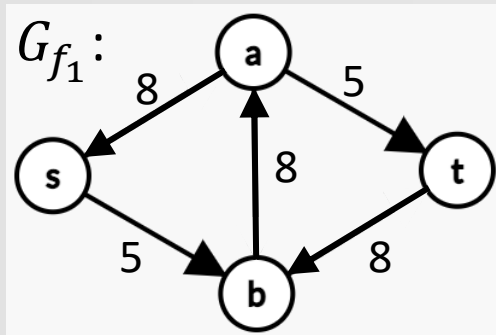
Example 1:



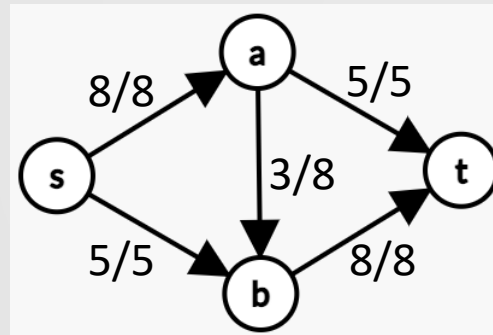
$$c_{f_0}(p_1) = 8$$



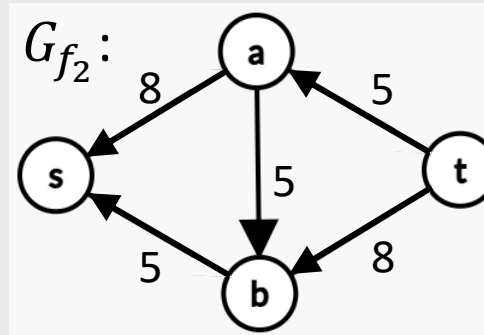
G with f_1



$$c_{f_1}(p_2) = 5$$



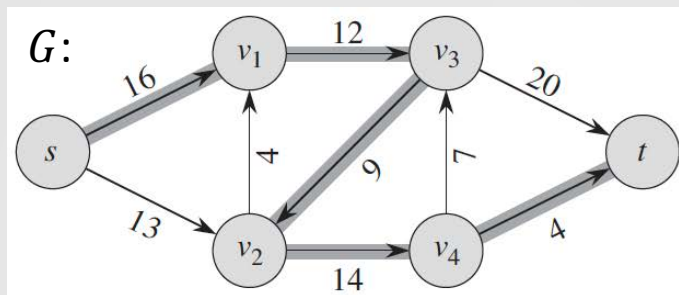
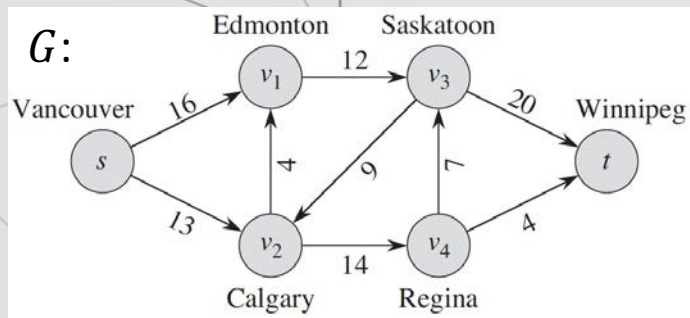
G with f_2



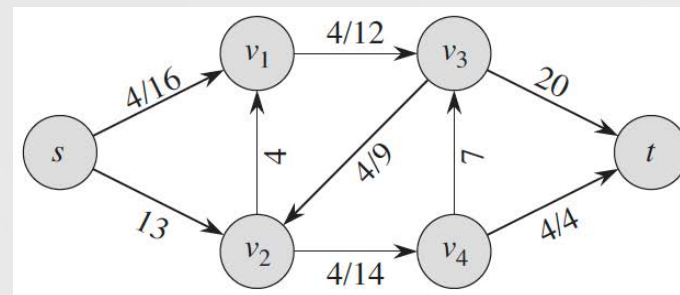
No augmenting path

- f_2 is a max flow in G .
- $|f_2| = 13$
- A min-cut of G :
 $S = \{s\}$
 $T = \{a, b, t\}$

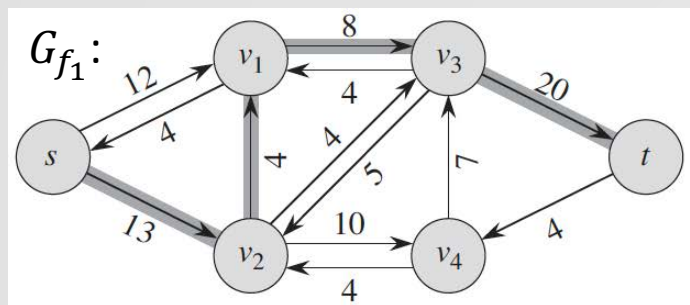
The Trucking Problem Example (1/2)



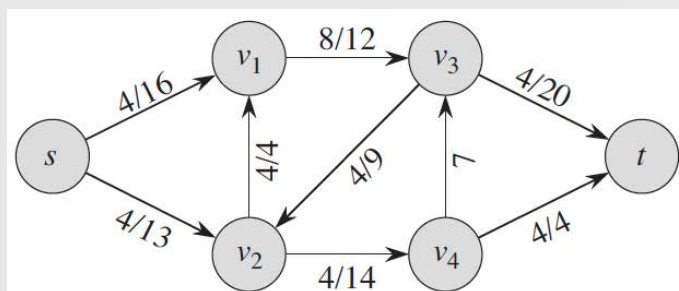
$$c_{f_0}(p_1) = 4$$



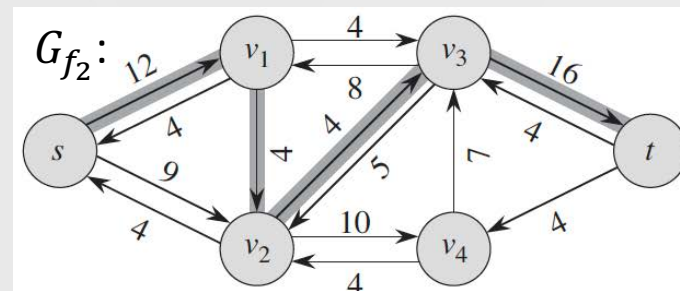
$$G \text{ with } f_1$$



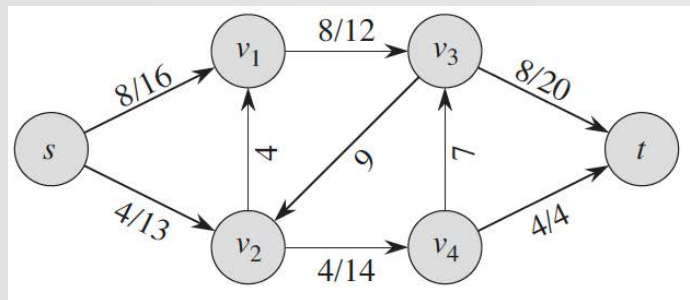
$$c_{f_1}(p_2) = 4$$



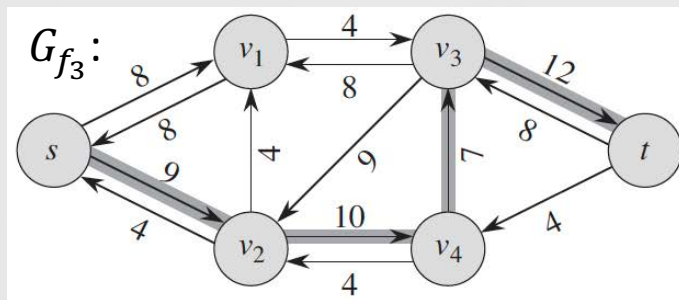
$$G \text{ with } f_2$$



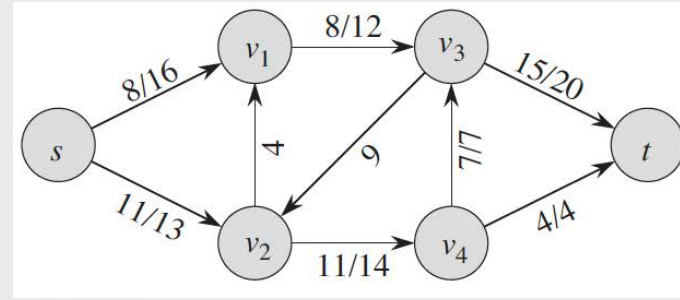
$$c_{f_2}(p_3) = 4$$



$$G \text{ with } f_3$$

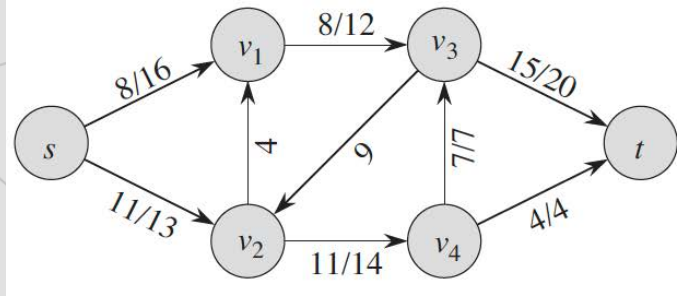


$$c_{f_3}(p_4) = 7$$

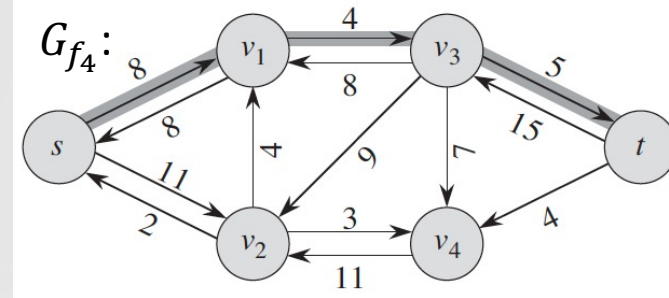


$$G \text{ with } f_4$$

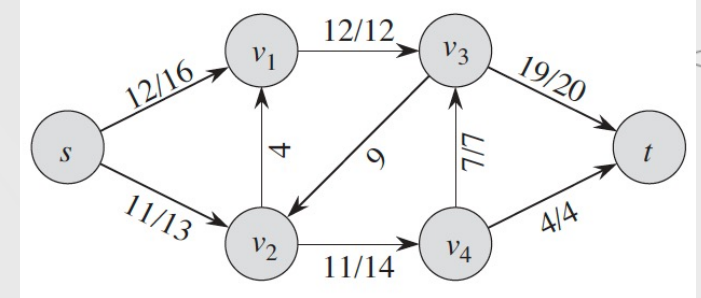
The Trucking Problem Example (2/2)



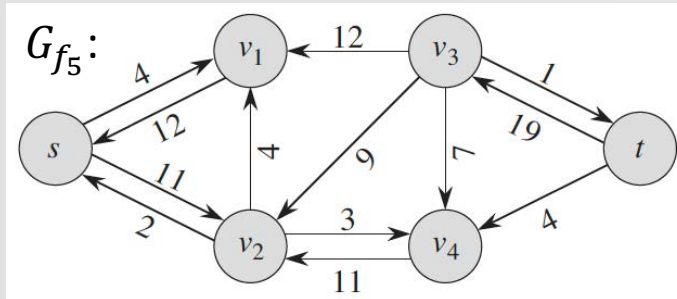
G with f_4



$$c_{f_4}(p_5) = 4$$



G with f_5



No more augmenting path.

- f_5 is a maximum flow in G .
- $|f_5| = 23$
- A min-cut of G :
 $S = \{s, v_1, v_2, v_4\}$
 $T = \{v_3, t\}$

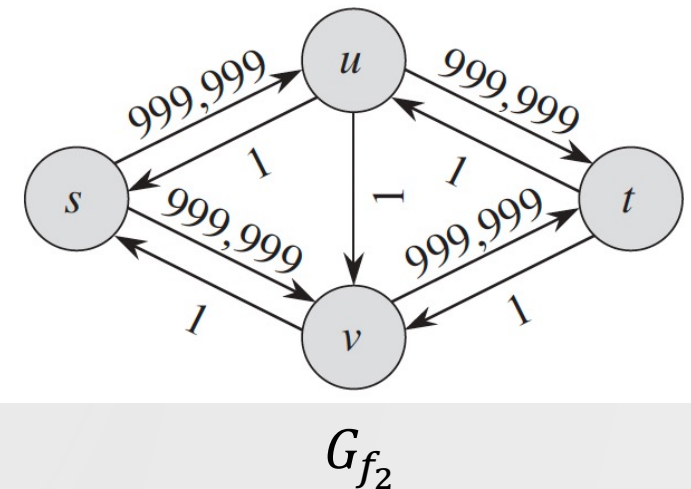
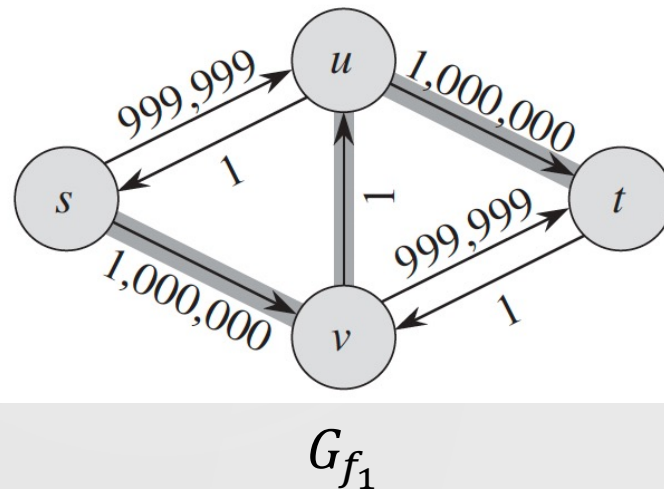
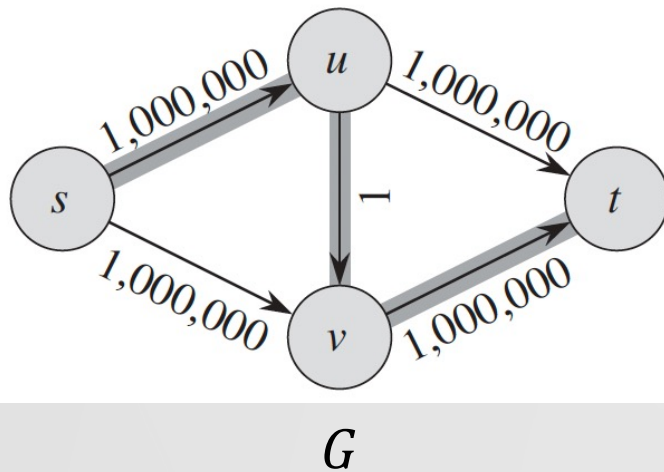
S contains all the vertices that can be reached from s in G_{f_5} .

Time Complexity of Ford-Fulkerson Algorithm

FORD-FULKERSON(G, s, t)

```
1 for each edge  $(u, v) \in G.E$ 
2    $(u, v).f = 0$ 
3 build the residual network  $G_f$ 
4 while there exists a path  $p$  from  $s$  to  $t$  in  $G_f$ 
5   augment  $f$  by  $c_f(p)$ 
6   update  $G_f$ 
```

- The time complexity depends on how we find the augmenting path in line 4.
- It can be very slow.
- **Example:**
It takes 2×10^6 iterations.



Analysis of Ford-Fulkerson Algorithm

FORD-FULKERSON(G, s, t)

```
1 for each edge  $(u, v) \in G.E$ 
2    $(u, v).f = 0$ 
3 build the residual network  $G_f$ 
4 while there exists a path  $p$  from  $s$  to  $t$  in  $G_f$ 
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6   update  $G_f$ 
```

- Lines 1-2: $O(m)$
- Line 3: $O(n + m)$
- Lines 4-6: in each iteration,
 - Line 4: $O(n + m)$ – by DFS or BFS
 - Line 5: $O(n)$
 - Line 6: $O(n)$

- **Assumption:** Capacities are all integers.
- Then, each augmenting path increases $|f|$ by at least 1. – Why?
- If max flow is f^* , then there will be at most $|f^*|$ iterations.
- Total running time: $O(m|f^*|)$ – NOT polynomial in input size!



The Ford-Fulkerson Method

The Edmonds-Karp algorithm

Edmonds-Karp Algorithm

- It follows the basic **Ford-Fulkerson** algorithm.
- Computes augmenting paths by using **BFS** in G_f to find the shortest path from s to t with all edge weights being 1. Call it the **BFS path**.
- Time complexity of **Edmonds-Karp algorithm**: $O(nm^2)$
 - NO assumption on values of capacities
- That is, the number of iterations of the **while** loop is in $O(nm)$.
 - In each iteration, G_f is updated with ≥ 1 edge deleted.
 - We will prove: Each edge in G_f can be deleted and reinserted back later for at most $n/2$ times.
 - Then, with $O(m)$ edges, there are $O(nm)$ iterations in total.

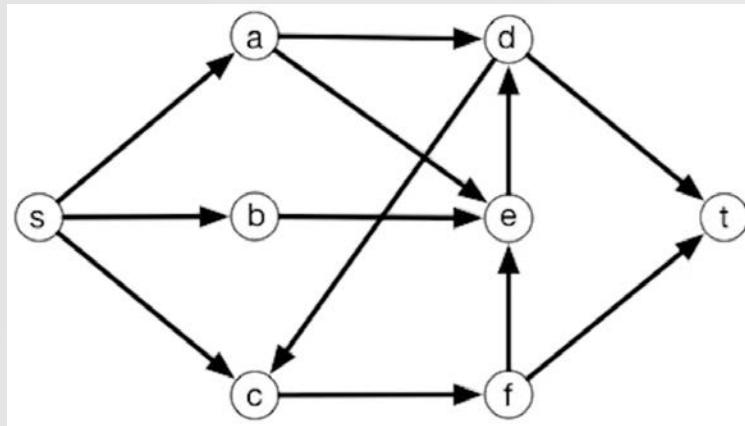
** This is Theorem 26.8 on p.729 of the textbook.*

Analysis of Edmonds-Karp Algorithm (1/5)

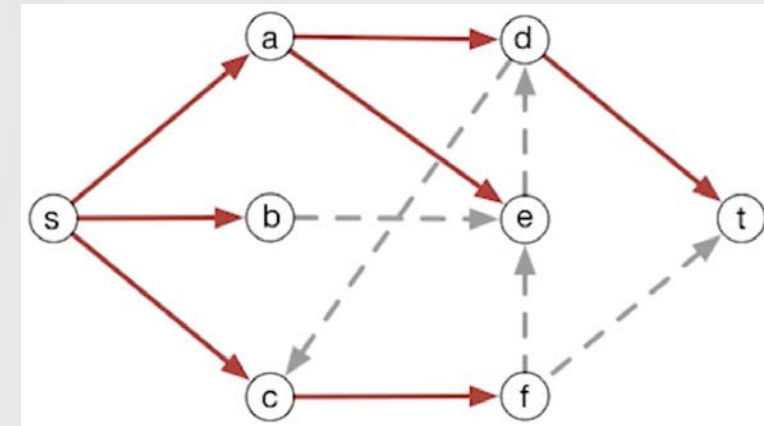
Lemma: Each edge (u, v) in G_f can be deleted and reinserted back later for at most $n/2$ times.

- *Recall:* BFS takes a graph G_f with starting vertex s . It can compute $v.d$ for every $v \in V$, where $v.d$ = distance (smallest # of edges) from s to v .
- For any edge (u, v) on a BFS path, $v.d = u.d + 1$.

Example:



G_f



Breadth-first tree of G_f

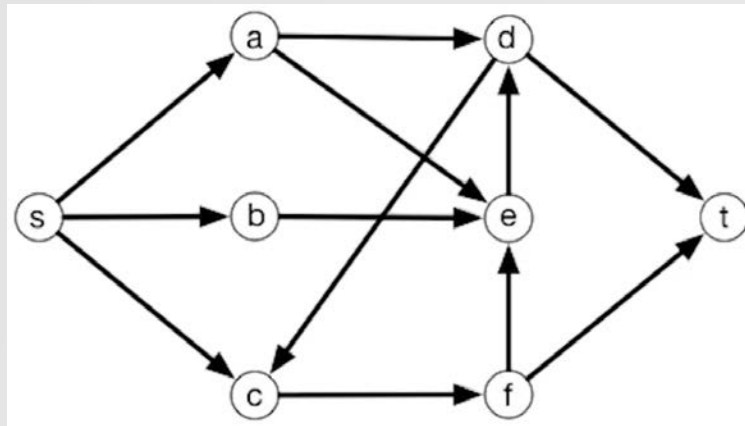
Analysis of Edmonds-Karp Algorithm (2/5)

Q: How does $v.d$ change as G_f changes?

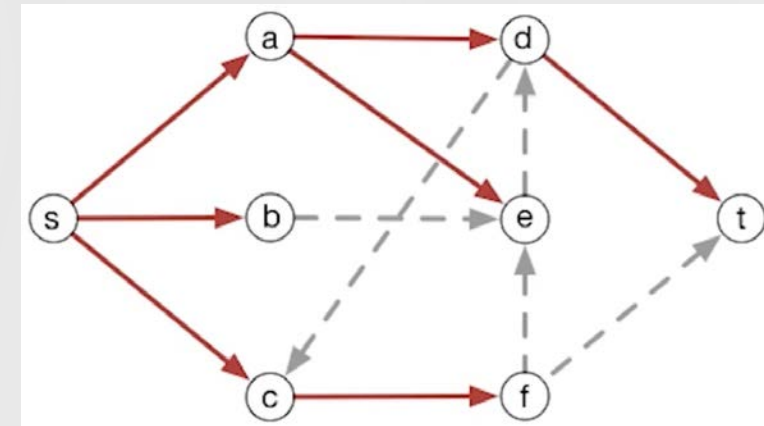
Claim: For every $v \in V$, $v.d$ never decreases.

Proof. When an edge in G_f is **deleted**, $v.d$ will not decrease. – **Why?**

- Only need to discuss the cases when an edge is **added** to G_f .
- **Q:** How does G_f change in each iteration?



G_f



Breadth-first tree of G_f

Analysis of Edmonds-Karp Algorithm (3/5)

Let f be the flow before the current iteration and f' be the flow updated with augmenting path (BFS path) p in the current iteration.

- **Recall:** Residual capacity is defined as:

For any edge (u, v) ,

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

- If (u, v) is **added** to G_f :
 - **Case 1:** $f(u, v) = c(u, v)$ and $f'(u, v) < c(u, v) \Rightarrow (v, u)$ is on p .
 - **Case 2:** $f(v, u) = 0$ and $f'(v, u) > 0 \Rightarrow (v, u)$ is on p .
- If (u, v) is **deleted** from G_f :
 - **Case 1:** $f(u, v) < c(u, v)$ and $f'(u, v) = c(u, v) \Rightarrow (u, v)$ is on p .
 - **Case 2:** $f'(v, u) = 0$ and $f(v, u) > 0 \Rightarrow (u, v)$ is on p .

Analysis of Edmonds-Karp Algorithm (4/5)

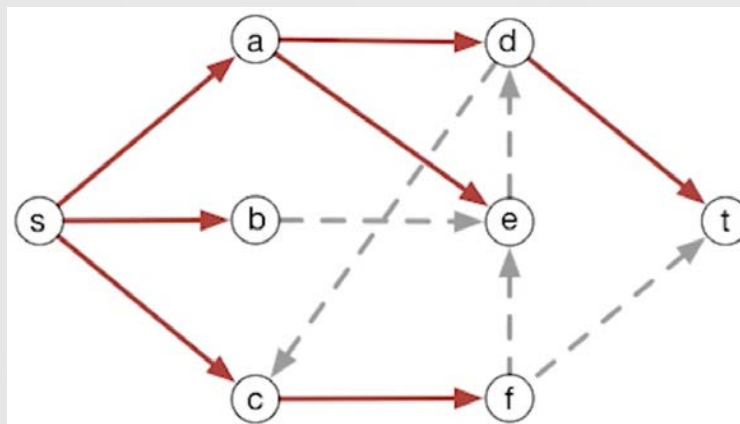
Claim: For every $v \in V$, $v.d$ never decreases.

Proof. (cont'd) Discuss cases when an edge (u, v) is **added** to G_f .

We've shown: if an edge (u, v) is **added** to G_f , then (v, u) is on p .

- Suppose $v.d = k$ before edge (u, v) is **added** to G_f .
- As p is a BFS path, for (v, u) , $u.d = v.d + 1 = k + 1$.
- So, adding (u, v) will not decrease $v.d$. – **Why?**

□



Analysis of Edmonds-Karp Algorithm (5/5)

Lemma: Each edge (u, v) in G_f can be deleted and reinserted back later for at most $n/2$ times.

Proof. Suppose at some point, $u.d = k$.

- When edge (u, v) is **deleted** from G_f , (u, v) is on the BFS path p .
So, $v.d = u.d + 1 \geq k + 1$.
- When edge (u, v) is later **added** back into $G_{f'}$, (v, u) is on the BFS path p' .
So, $u.d' = v.d + 1 \geq k + 2$.

That is, $u.d$ will increase by at least 2.

As $0 \leq u.d \leq n$, lemma is proved. □

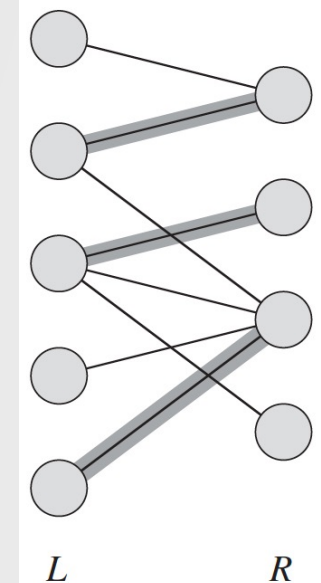
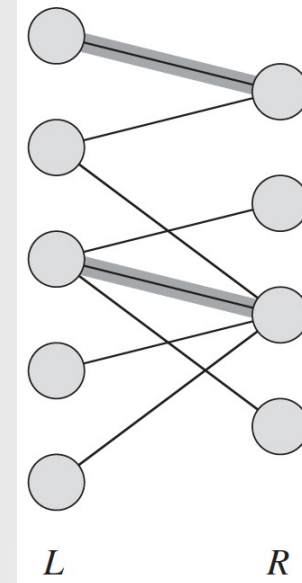


Maximum Bipartite Matching

The Maximum-Bipartite-Matching Problem

- A graph $G = (V, E)$ is **bipartite** if we can **partition** $V = L \cup R$ ($L \cap R = \emptyset$) such that all edges in E go between L and R .
- In an undirected graph $G = (V, E)$, a **matching** is a subset of edges $M \subseteq E$ s.t. for all $v \in V$, at most one edge of M is incident on v .
 - Vertex v is **matched** by M if an edge of M is incident on it; otherwise v is **unmatched**.
 - A **maximum matching** is a matching of maximum cardinality, that is, a matching M s.t. $|M| \geq |M'|$ for any matching M' .
- **The Maximum-Bipartite-Matching Problem:**
 - **Given:** A bipartite graph $G = (L \cup R, E)$
 - **Goal:** Find a maximum matching.

Example:

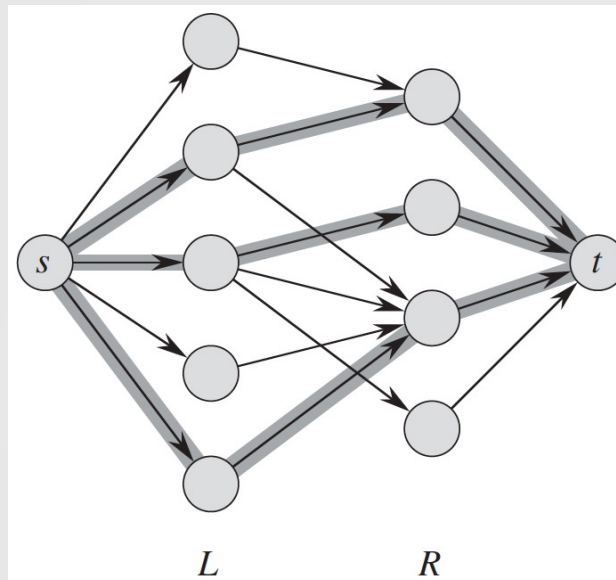


Maximum-Bipartite-Matching as Maximum-Flow

Given a **bipartite graph** $G = (L \cup R, E)$, we define the corresponding **flow network** $G' = (V', E')$ as follows:

- $V' = V \cup \{s, t\}$,
- $E' = \{(s, u) : u \in L\} \cup \{(u, v) : (u, v) \in E\} \cup \{(v, t) : v \in R\}$,
- $c(u, v) = 1$ for all $(u, v) \in E'$.

Example:



Integrality Theorem

Integrality Theorem^{*}. If the capacities of all edges in a flow network are integers, then

1. The value of the maximum flow f produced by the **Ford-Fulkerson method**, $|f|$, is also an integer.
2. For every edge (u, v) , the value of $f(u, v)$ is an integer.

Lemma^{**}: Let $G = (L \cup R, E)$ be a bipartite graph and $G' = (V', E')$ be its corresponding flow network. Then, a matching M in G corresponds to a flow f in G' , with $|M| = |f|$.

Proof. See next slide.

^{*} This is Theorem 26.10 on p.734 of the textbook.

^{**} See Lemma 26.9 on p.733-734 of the textbook for a complete proof.

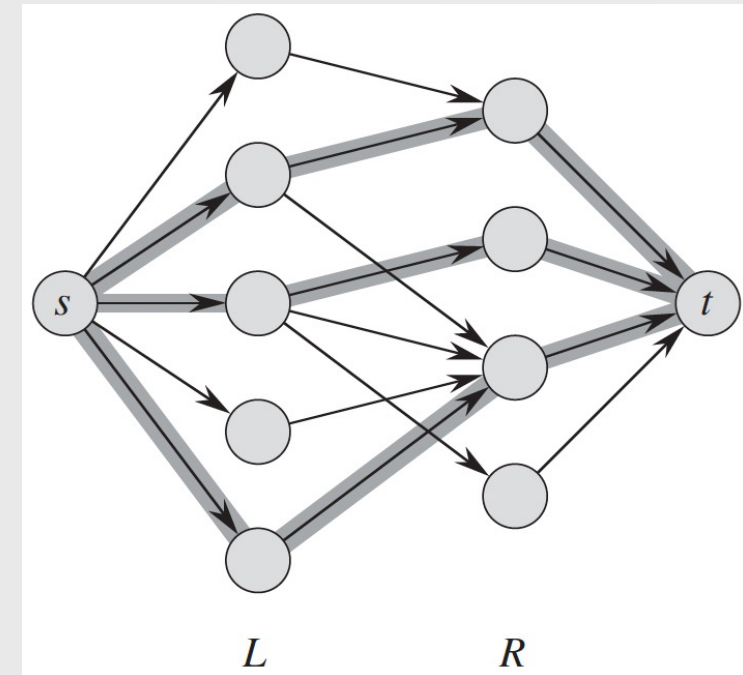
Correspondence Between Matchings and Flows (1/2)

Lemma: Let $G = (L \cup R, E)$ be a bipartite graph and $G' = (V', E')$ be its corresponding flow network. Then, a matching M in G corresponds to a flow f in G' , with $|M| = |f|$.

Proof.

- “ \Rightarrow ”: Show $|M| = k \Rightarrow \exists f$ s.t. $|f| = k$.
 - For each edge $(u, v) \in M$, with $u \in L, v \in R$, $f(u, v) = 1$, $f(s, u) = 1$, and $f(v, t) = 1$.
 - $|f| = \sum_{v \in V} f(s, v) = k$.
- “ \Leftarrow ”: Show $|f| = k \Rightarrow \exists M$ s.t. $|M| = k$.
 - If $f(u, v) > 0$, then $f(u, v) = 1$
 - $M = \{(u, v) : u \in L, v \in R, f(u, v) = 1\}$
 - M is a matching. – Why?
 - $|M| = f(L \cup \{s\}, R \cup \{t\}) = |f| = k$. □

Example:



Correspondence Between Matchings and Flows (2/2)

Integrality Theorem. If the capacities of all edges in a flow network are integers, then

1. The value of the maximum flow f produced by the **Ford-Fulkerson method**, $|f|$, is also an integer.
2. For every edge (u, v) , the value of $f(u, v)$ is an integer.

Lemma: Let $G = (L \cup R, E)$ be a bipartite graph and $G' = (V', E')$ be its corresponding flow network. Then, a matching M in G corresponds to a flow f in G' , with $|M| = |f|$.

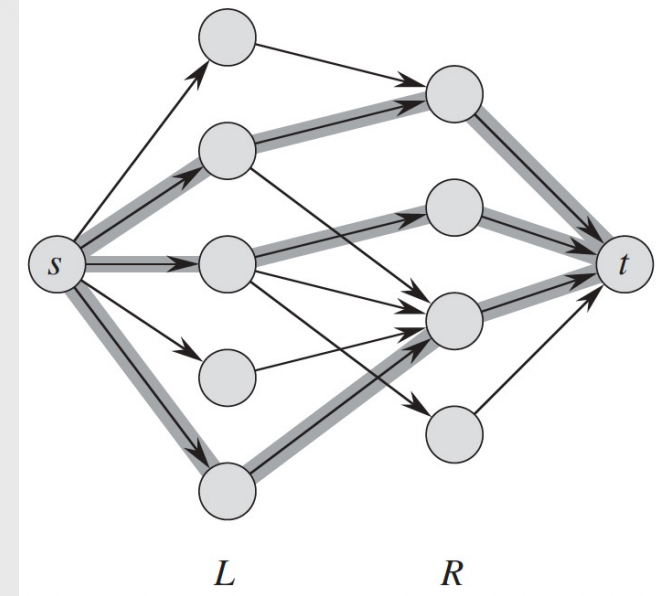
The **Integrality Theorem** and the **Lemma** lead to the following conclusion.

Corollary (26.11): The cardinality of a maximum matching in a bipartite graph = the value of a maximum flow in its corresponding flow network.

Finding a Maximum Bipartite Matching

Given a bipartite graph $G = (L \cup R, E)$, with $n = |L \cup R|$ and $m = |E|$,

1. Create flow network $G' = (V', E')$
 - $|V'| = n + 2$ and $|E'| = m + n \in \Theta(m)$;
 - $c(u, v) = 1$ for all $(u, v) \in E'$.
 - Running time: $\Theta(m + n)$
2. Apply the Ford-Fulkerson method on G'
 - Let M be max matching in G ,
 - $|M| \leq \min\{|L|, |R|\} \in O(n)$.
 - If f' is max flow in G' , then $|f'| = |M| \in O(n)$.
 - Running time: $O(mn)$



Thank you!
Questions?