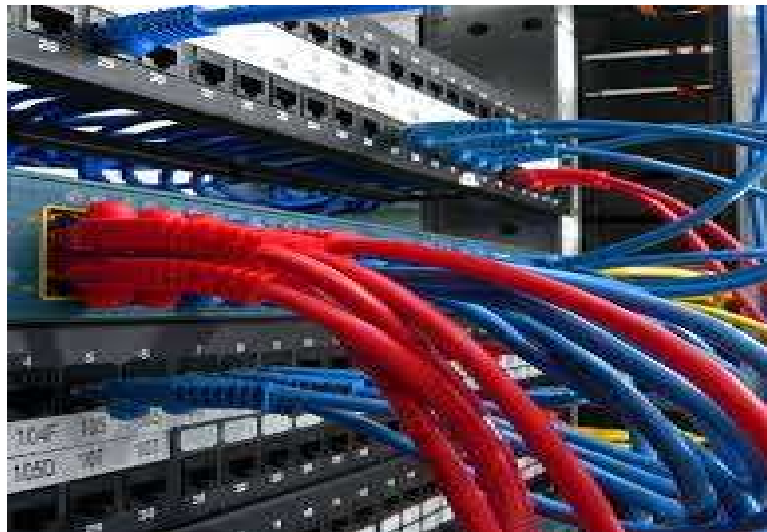


12: Maximum Flow

- Maximum Flow Problem
- The Ford-Fulkerson method
- Maximum bipartite matching



Flow networks:

- A **flow network** $G=(V,E)$: a directed graph, where each edge $(u,v) \in E$ has a nonnegative **capacity** $c(u,v) \geq 0$.
- If $(u,v) \notin E$, we assume that $c(u,v)=0$.
- two distinct nodes: a **source** s and a **sink** t .

SOURCE:

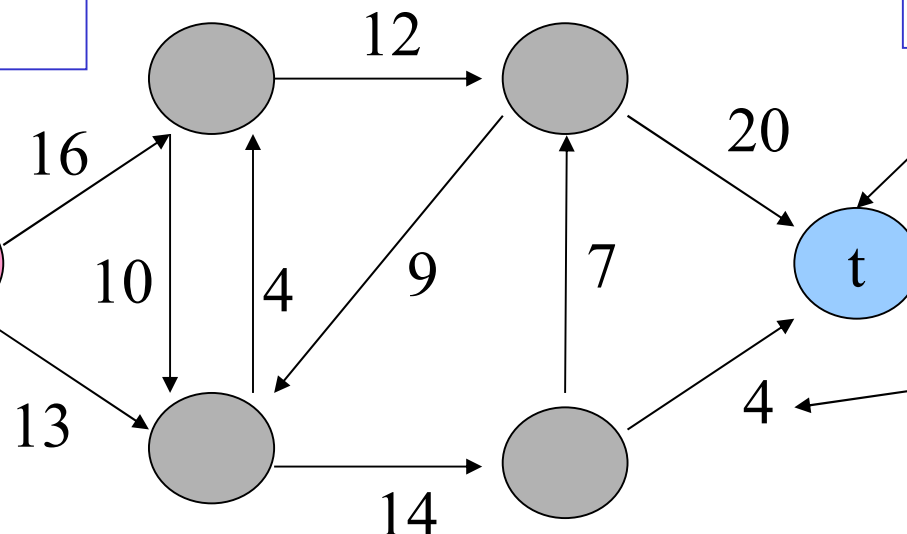
*Node with net outflow;
Production point*

SINK:

*Node with net inflow;
Consumption point*

CAPACITY:

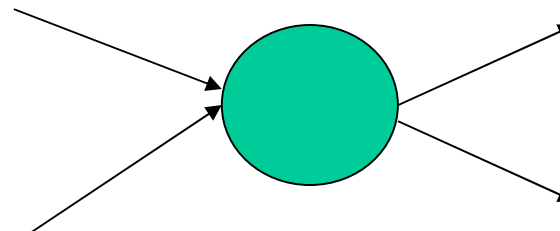
*Maximum flow
on an edge*



Flow:

- Given: $G=(V, E)$: a flow network with capacity function c .
s---the source and t--- the sink.
- **A flow in G**: a real-valued function $f: E \rightarrow \mathbb{R}$ satisfying the following two properties:
- **Capacity constraint**: For all $u,v \in V$, we require
$$f(u,v) \leq c(u,v).$$
- **Flow conservation**: For all $v \in V - \{s,t\}$, we require

$$\sum_{e \in \text{in}.v} f(e) = \sum_{e \in \text{out}.v} f(e)$$

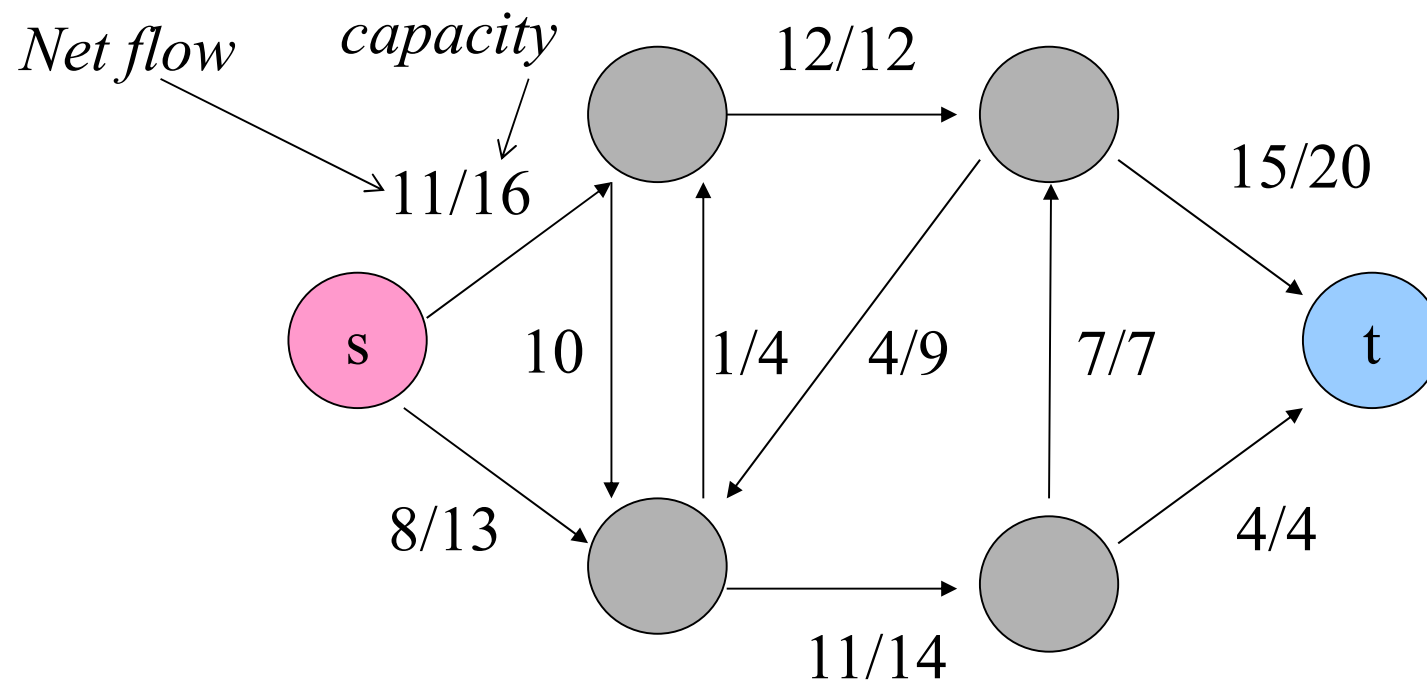


Net flow and value of a flow f :

- The quantity $f(u,v)$ is called the **net flow** from vertex u to vertex v .
- The **value** of a flow is defined as

$$|f| = \sum_{v \in V} f(s, v)$$

- The total flow from the **source** (**s**) to any other vertices.
- The same as the total flow from any vertices to **the sink** (**t**).



A flow f in G with value $|f| = 19$.

Maximum-flow problem:

- Given a flow network G with source s and sink t
- Find a flow of maximum value from s to t .
- How to solve it efficiently?

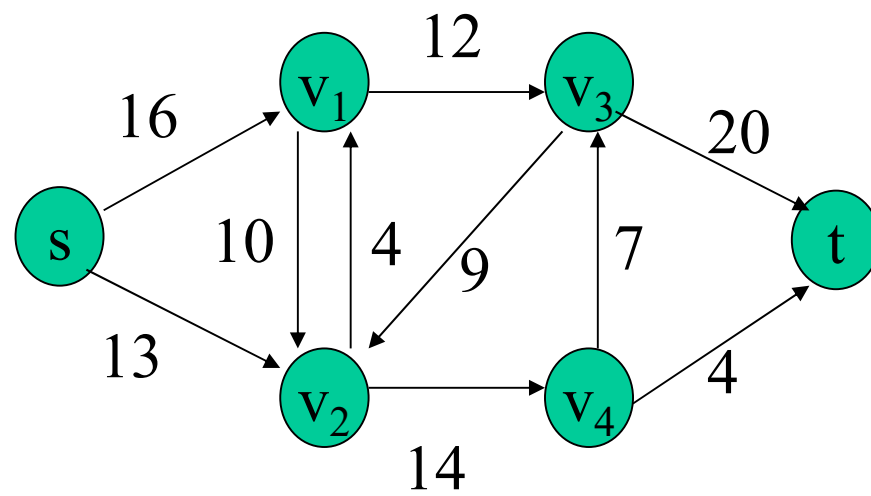


The Ford-Fulkerson method:

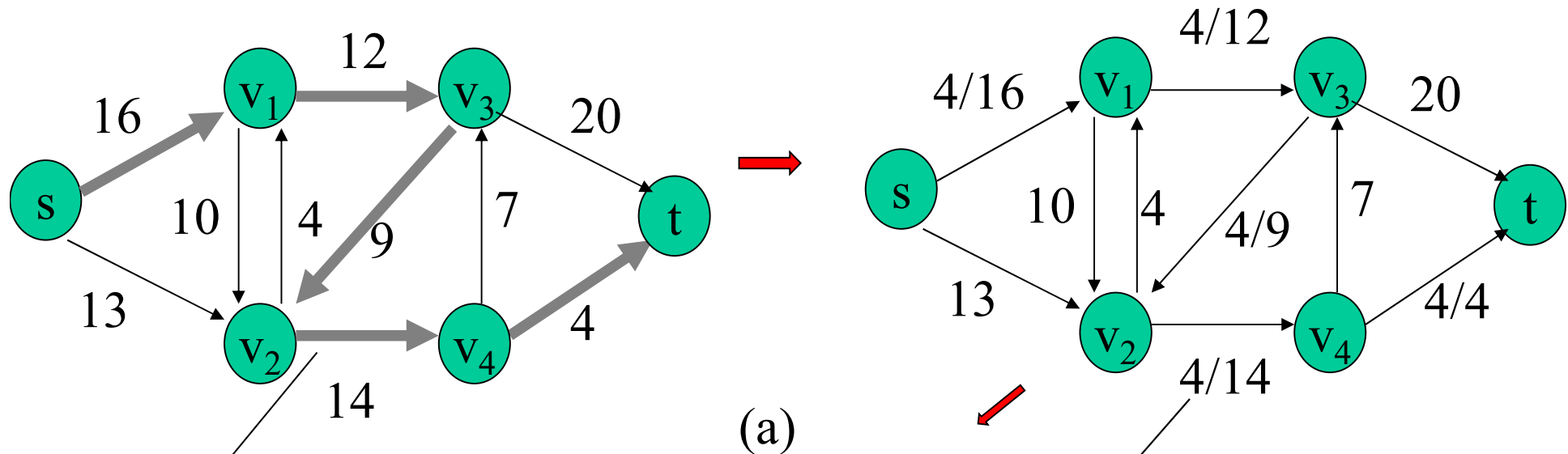
- Ford-Fulkerson method
 - it is a “method” rather than an “algorithm” because it encompasses several implementations with different running times.
 - The Ford-Fulkerson method depends on three important ideas:
 - residual networks
 - augmenting paths, and
 - cuts.

Continue:

- FORD-FULKERSON-METHOD(G, s, t)
- initialize flow f to 0
- **while** there exists an *augmenting* path p
- **do** *augment* flow f along p
- return f



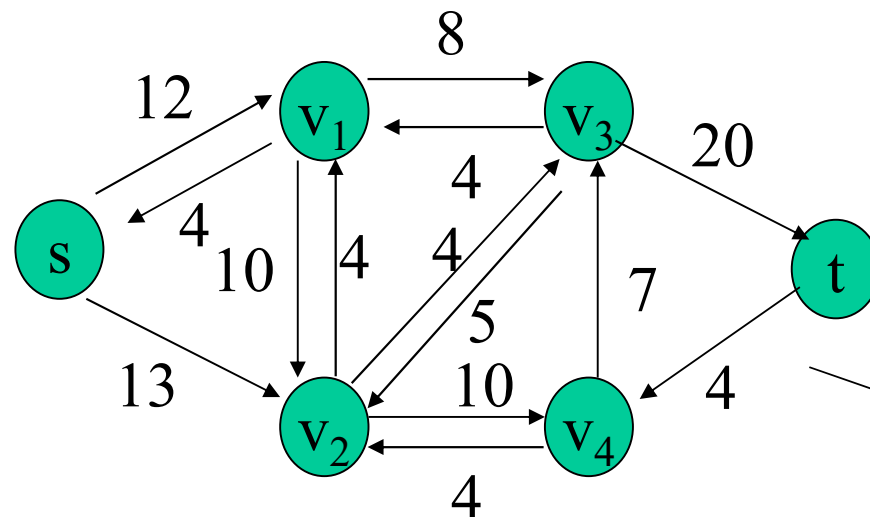
Example



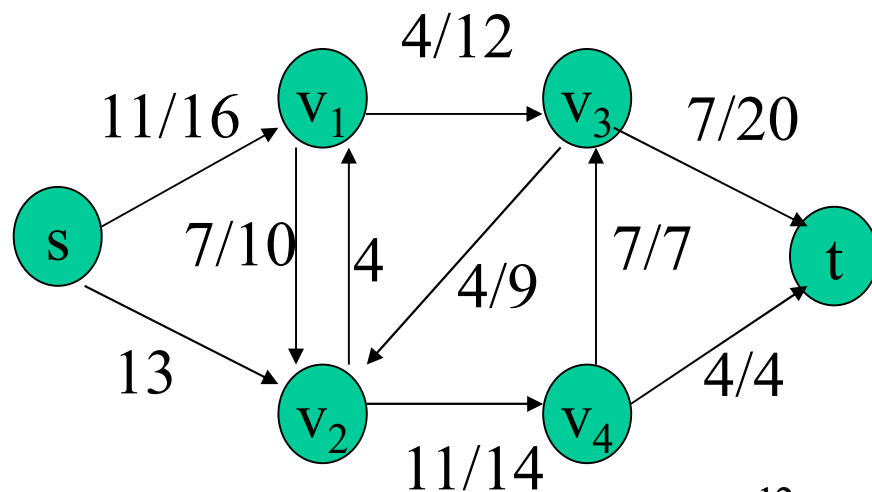
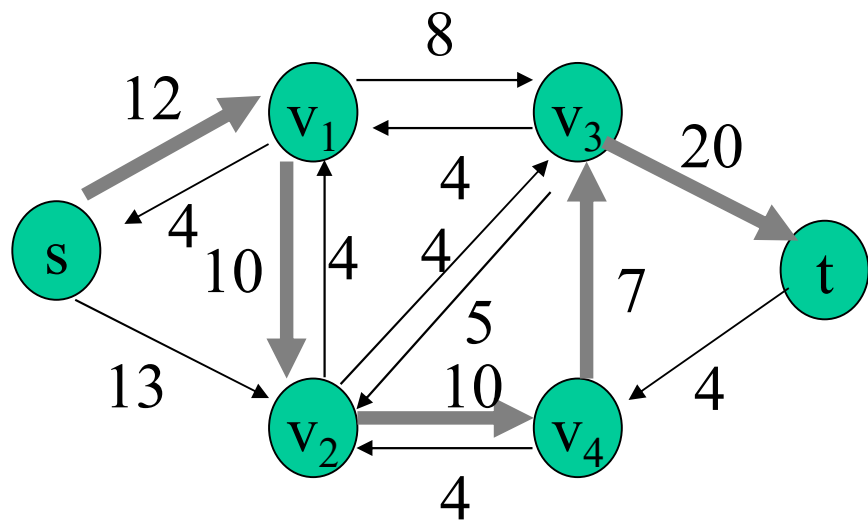
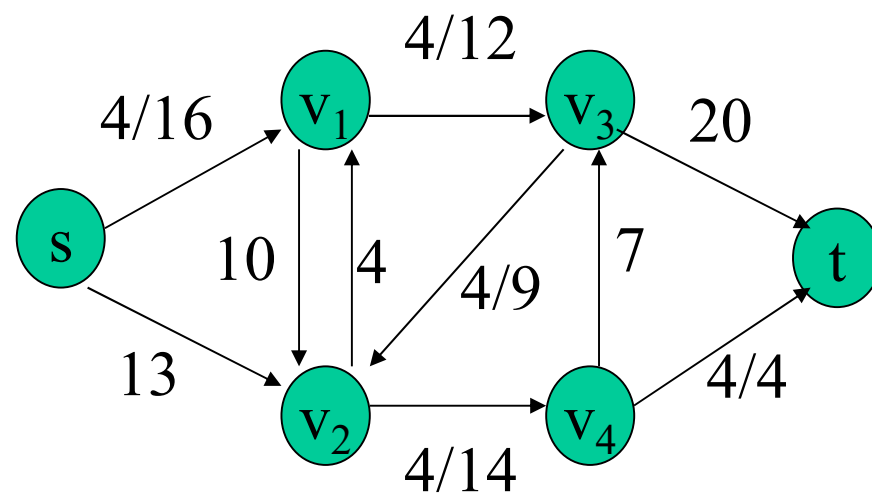
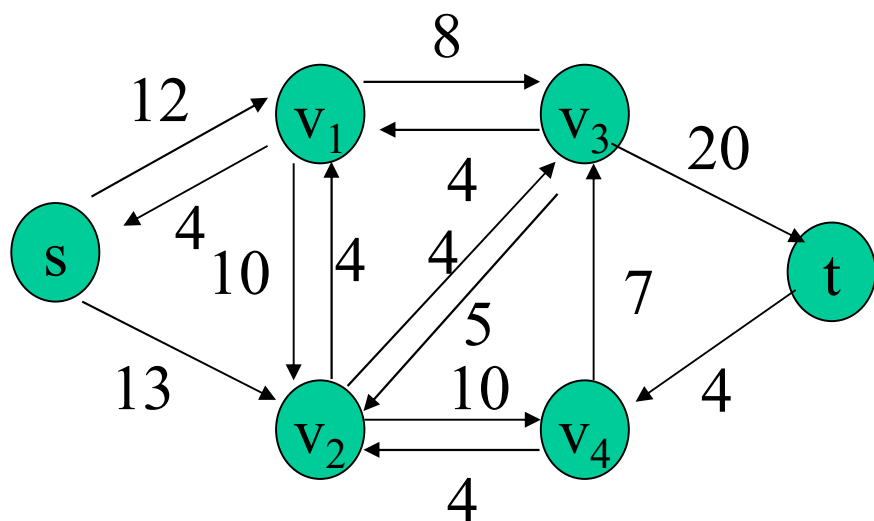
Bold path: augment path

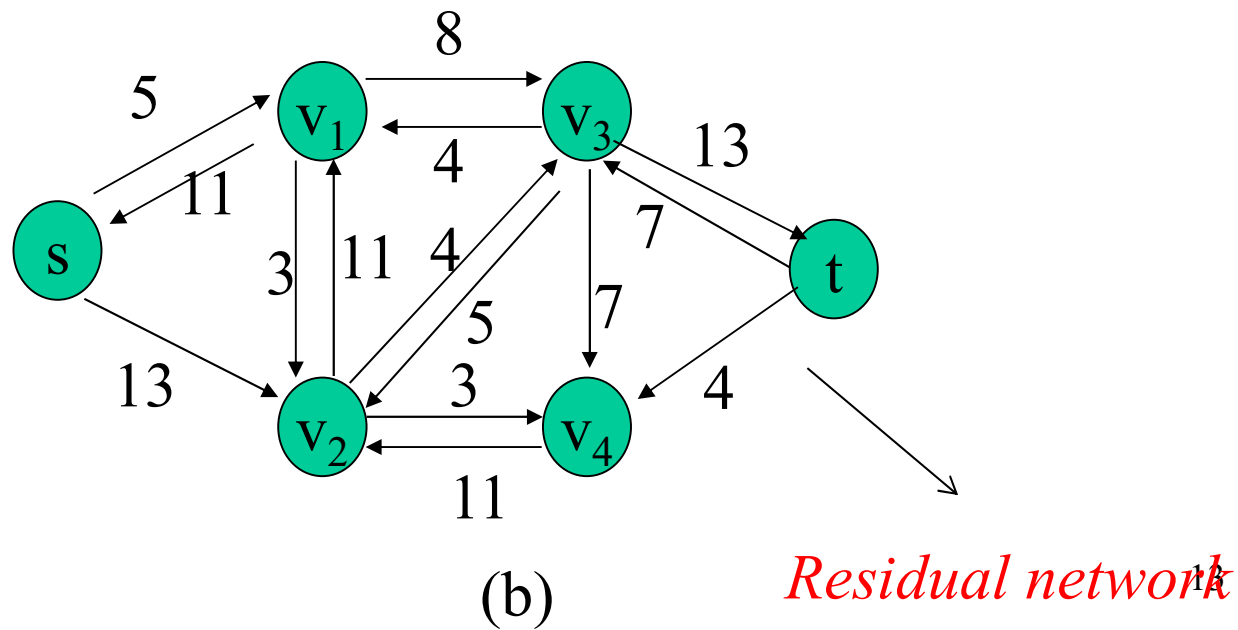
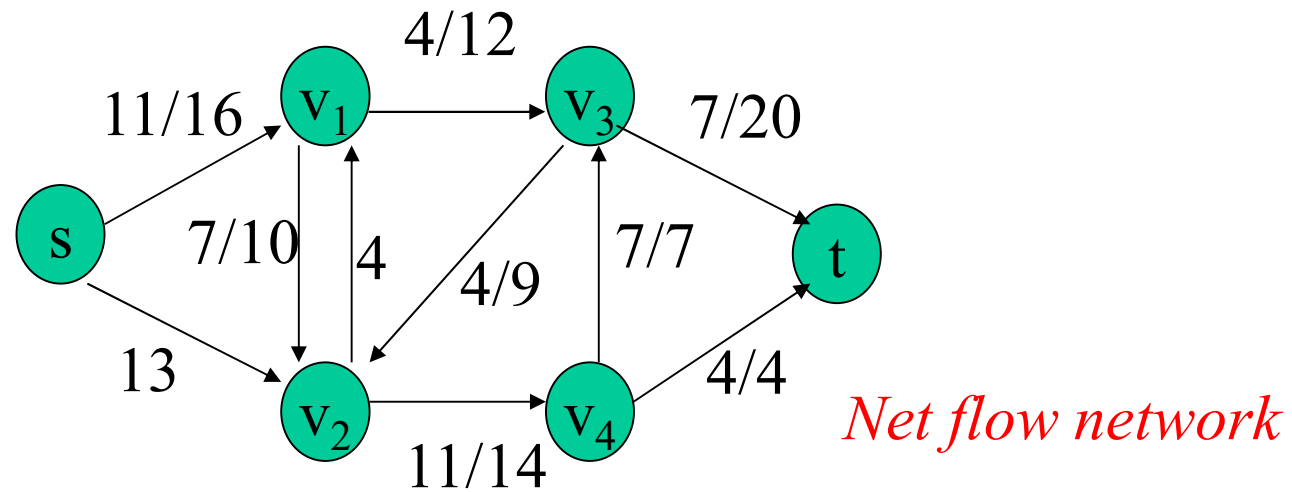
Net flow

Flow



Residual network





Residual networks:

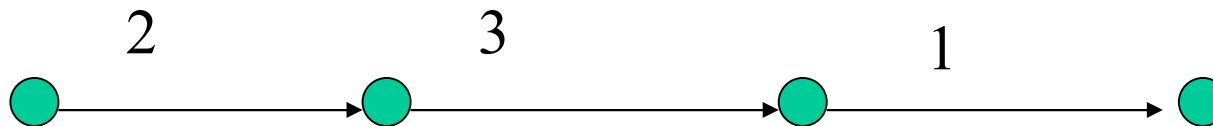
- Given a flow network and **a flow**, the **residual network** consists of edges that can admit more net flow.
- f : a flow in G .
- The **residual capacity** of (u,v) , given by:
 - $c_f(u,v) = c(u,v) - f(u,v)$
 - *in the other direction*
 - $c_f(v,u) = c(v,u) + f(u,v)$.

Fact 1:

- Let $G=(V,E)$ be a flow network with source s and sink t , and let f be a flow in G .
- Let G_f be the residual network of G induced by f , and let f' be a flow in G_f . Then, the flow sum $f+f'$ is a flow in G with value \cdot .
- $f+f'$: the flow in the same direction will be added.
the flow in different directions will be cancelled.

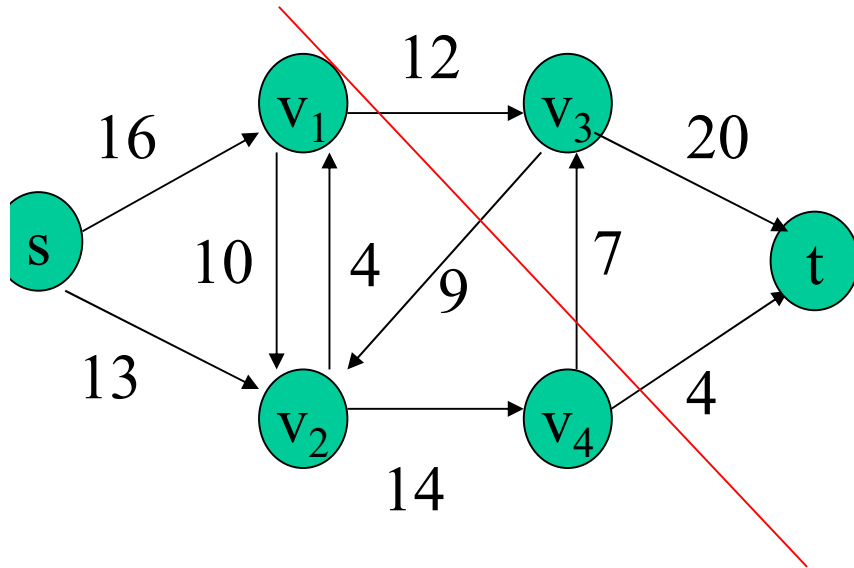
Augment paths:

- Given a flow network $G=(V,E)$ and a flow f , an **augment path** is a simple path from s to t in the residual network G_f .
- **Residual capacity** of p : the maximum amount of net flow that we can ship along p , i.e.,
$$c_f(p) = \min \{c_f(u,v) : (u,v) \text{ is on } p\}.$$

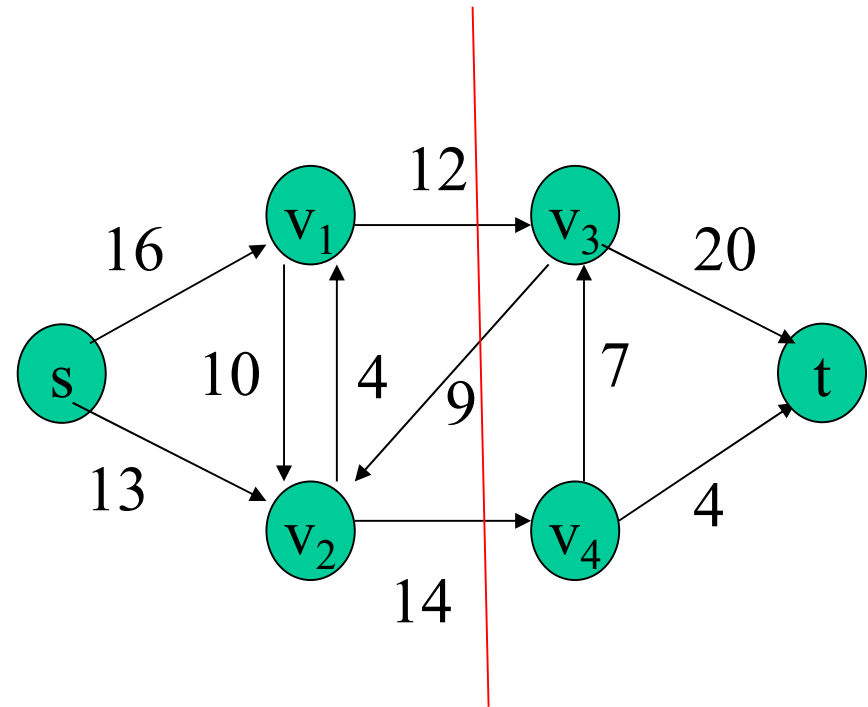


The residual capacity is 1.

Cut and flow



$$|f| \leq 12 + 7 + 4,$$



$$|f| \leq 12 + 14,$$

Theorem $\max |f| = \min |cut|$

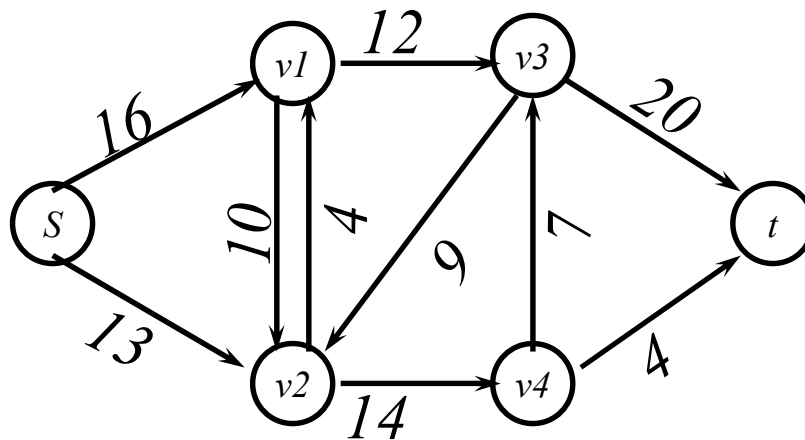
The basic Ford-Fulkerson algorithm:

- 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 a path p from s to t in the residual network G_f
- **do** $c_f(p) \leftarrow \min \{c_f(u, v) : (u, v) \text{ is in } p\}$
- **for** each edge (u, v) in p
- **do** $f[u, v] \leftarrow f[u, v] + c_f(p)$
-

The basic Ford Fulkerson algorithm

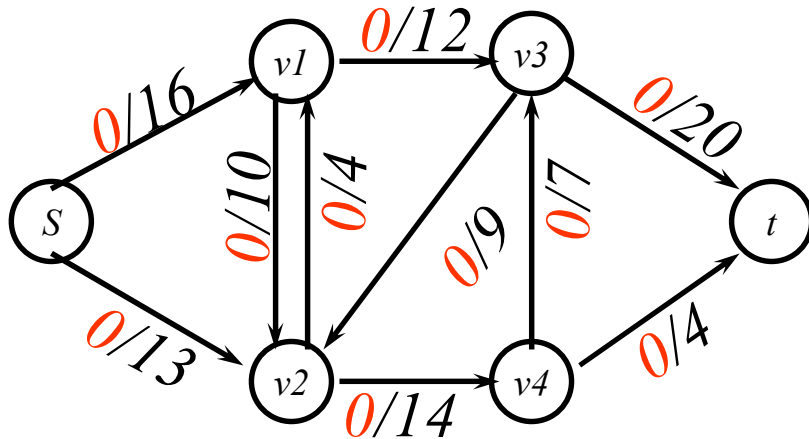
example of an execution

Network



The basic Ford Fulkerson algorithm

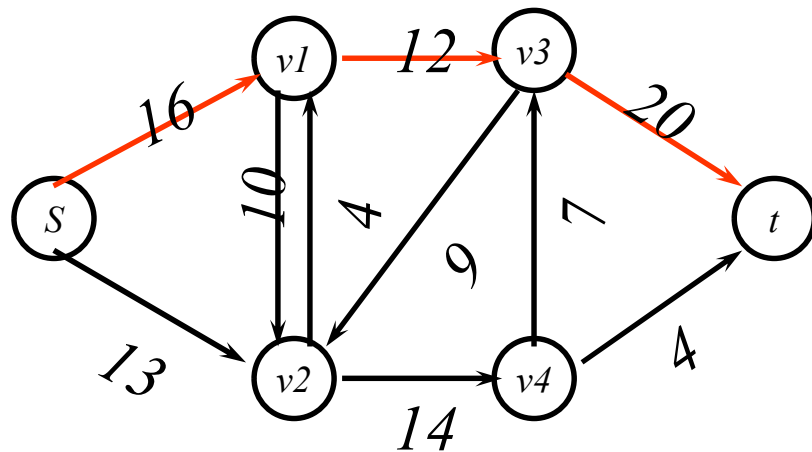
example of an execution



```
1  for each edge  $(u, v) \in E [G]$ 
2    do  $f[u, v] = 0$ 
3     $f[v, u] = 0$ 
4  while there exists a path  $p$  from  $s$  to  $t$ 
    in the residual network  $G_f$ 
5    do  $c_f(p) = \min \{c_f(u, v) \mid (u, v) \in p\}$ 
6    for each edge  $(u, v)$  in  $p$ 
7      do  $f[u, v] = f[u, v] + c_f(p)$ 
8
```

The basic Ford Fulkerson algorithm

example of an execution



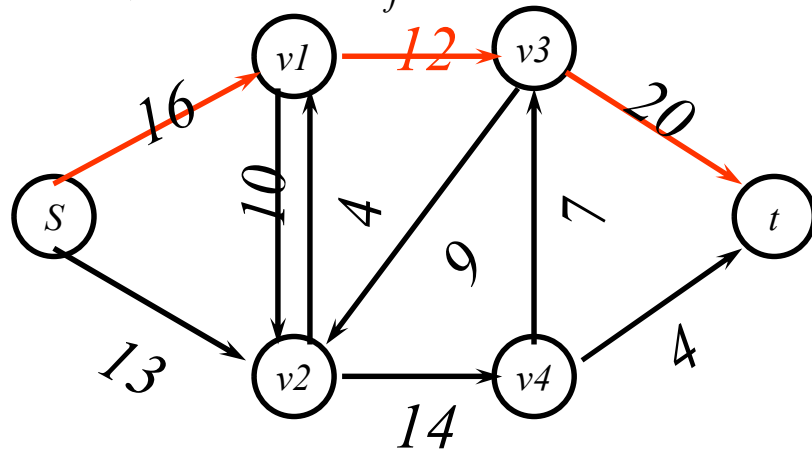
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7          do  $f[u, v] = f[u, v] + c_f(p)$ 
    
```

The basic Ford Fulkerson algorithm

example of an execution

(residual) network G_f



```

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```

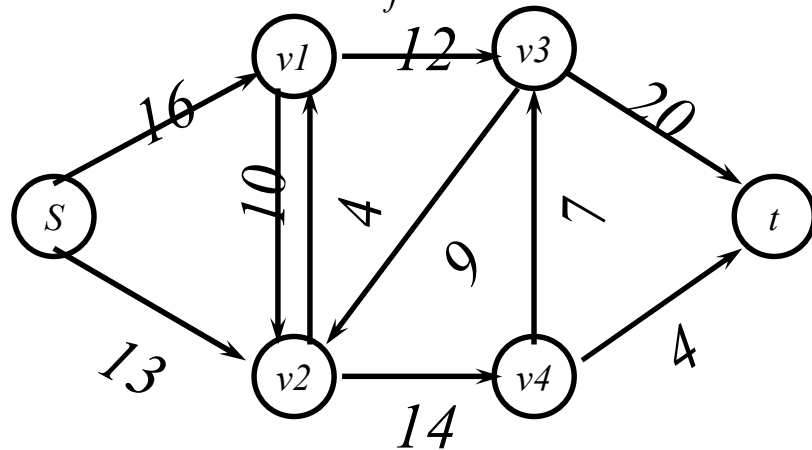
temporary variable:

$$c_f(p) = 12$$

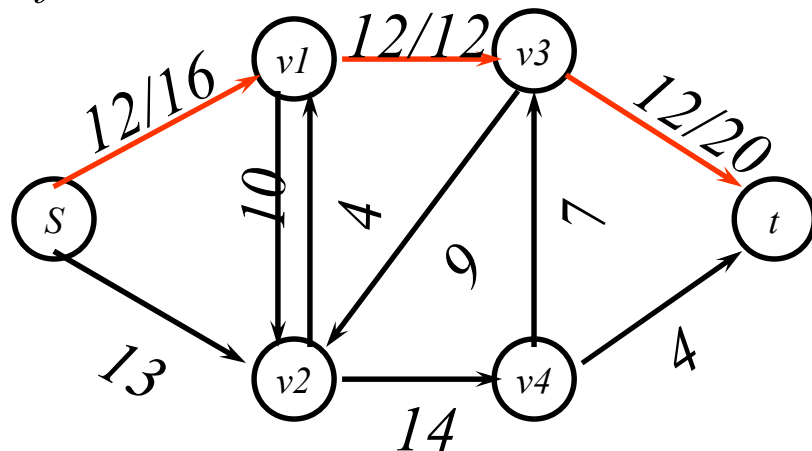
The basic Ford Fulkerson algorithm

example of an execution

(residual) network G_f



new flow network G



```

1  for each edge  $(u, v) \in E [G]$ 
2      do  $f[u, v] = 0$ 
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4  while there exists a path  $p$  from  $s$  to  $t$ 
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```

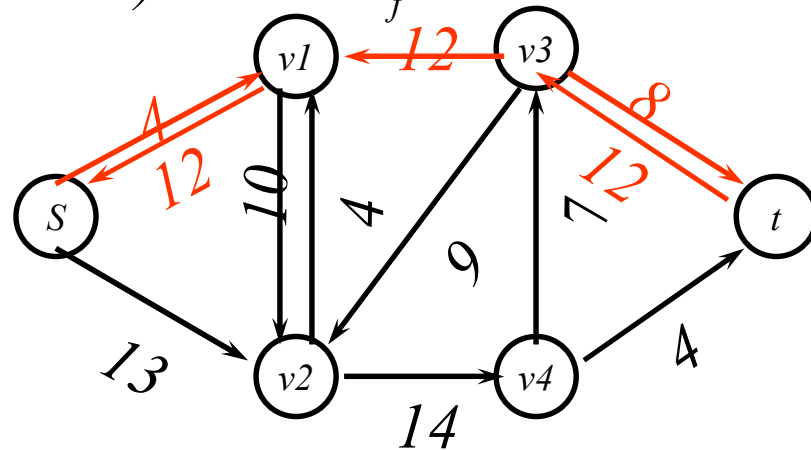
temporary variable:

$$c_f(p) = 12$$

The basic Ford Fulkerson algorithm

example of an execution

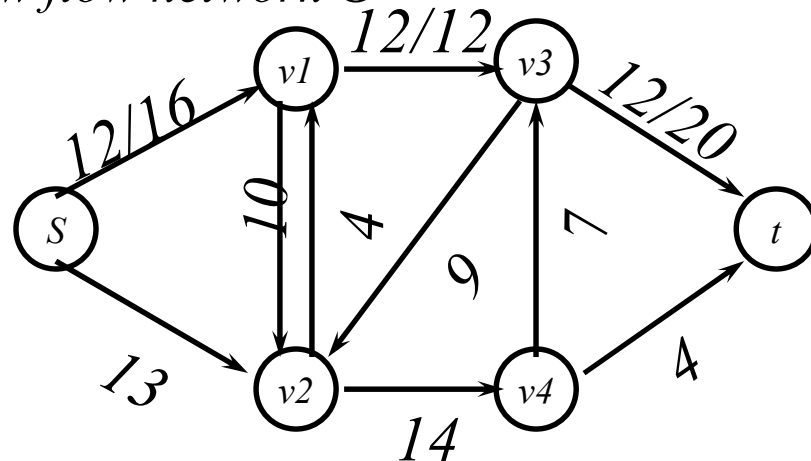
(residual) network G_f



```

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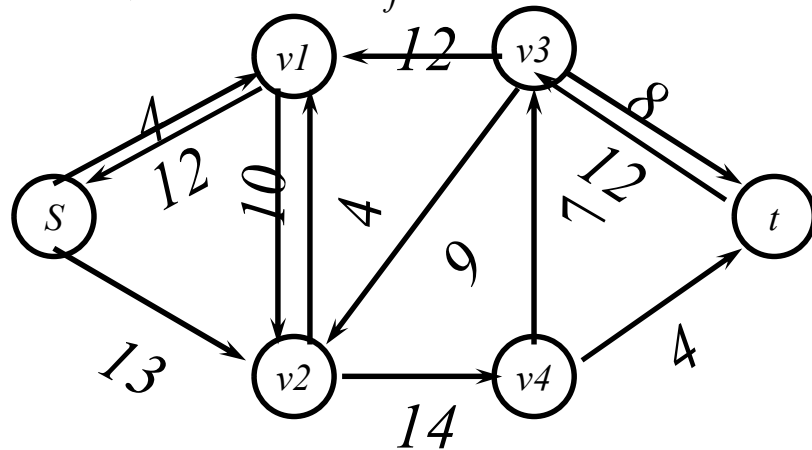
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

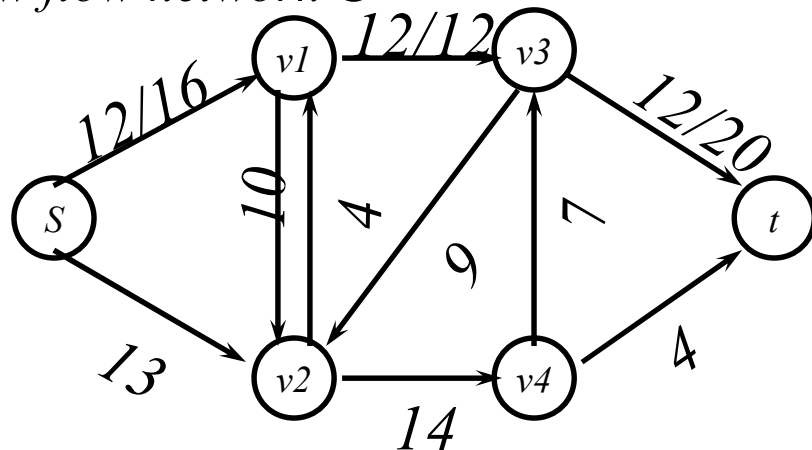
(residual) network G_f



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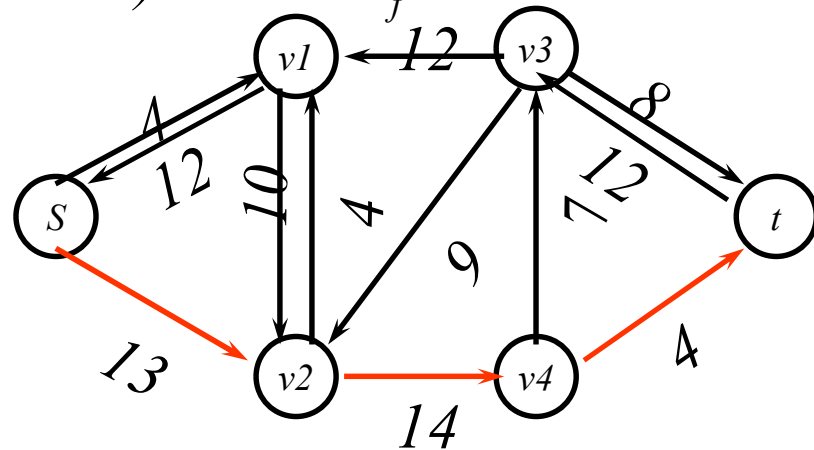
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

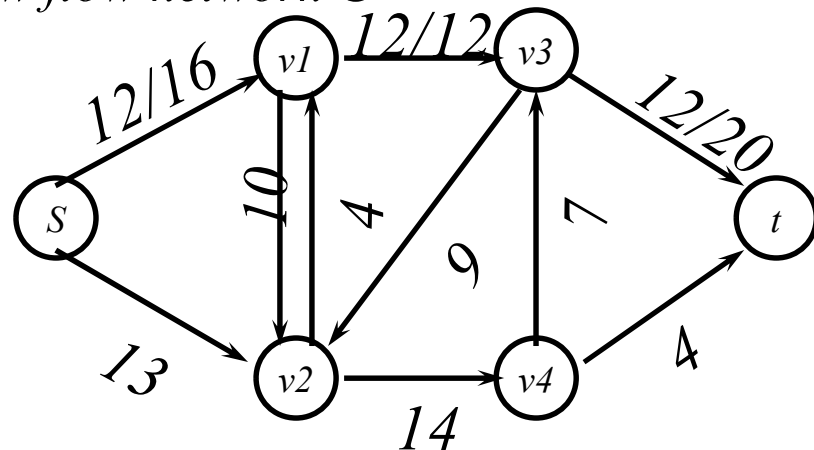
(residual) network G_f



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```

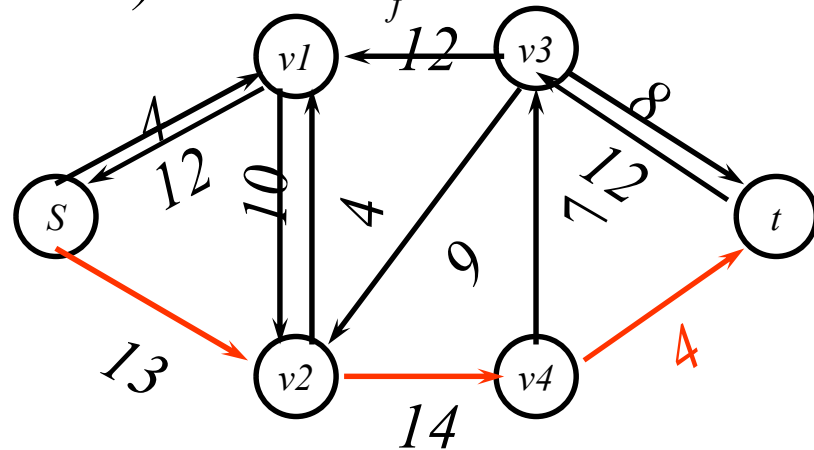
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

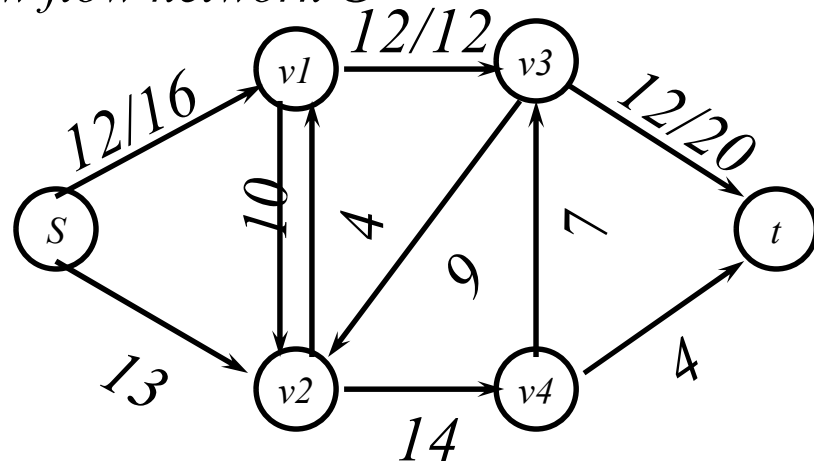
(residual) network G_f



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```

new flow network G



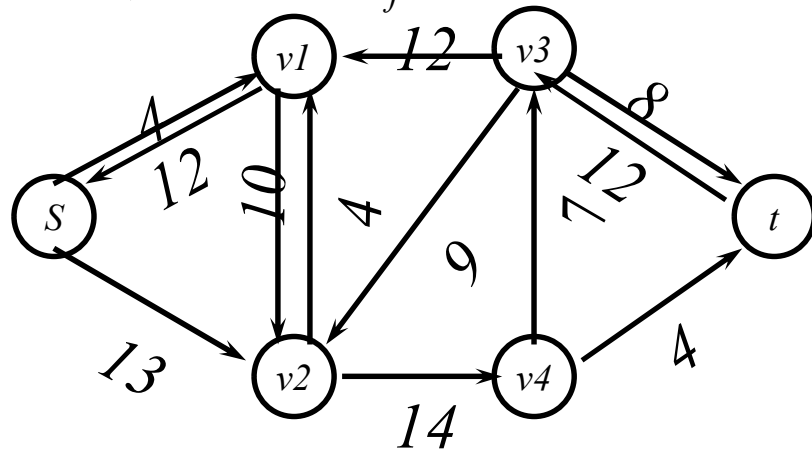
temporary variable:

$$c_f(p) = 4$$

The basic Ford Fulkerson algorithm

example of an execution

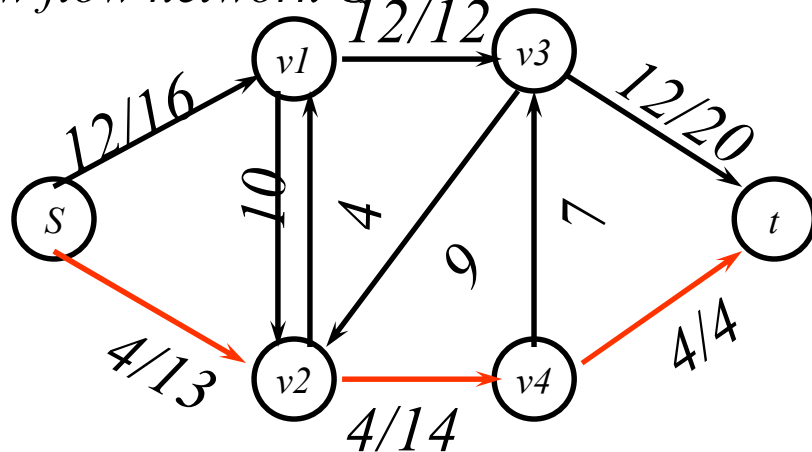
(residual) network G_f



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```

new flow network G



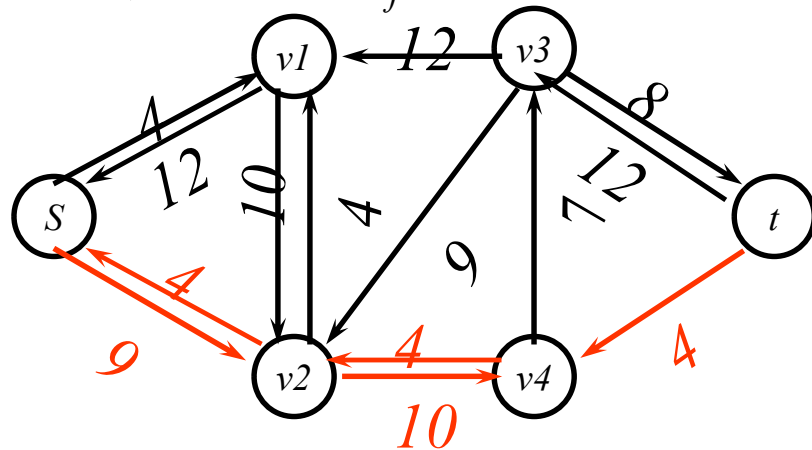
temporary variable:

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The basic Ford Fulkerson algorithm

example of an execution

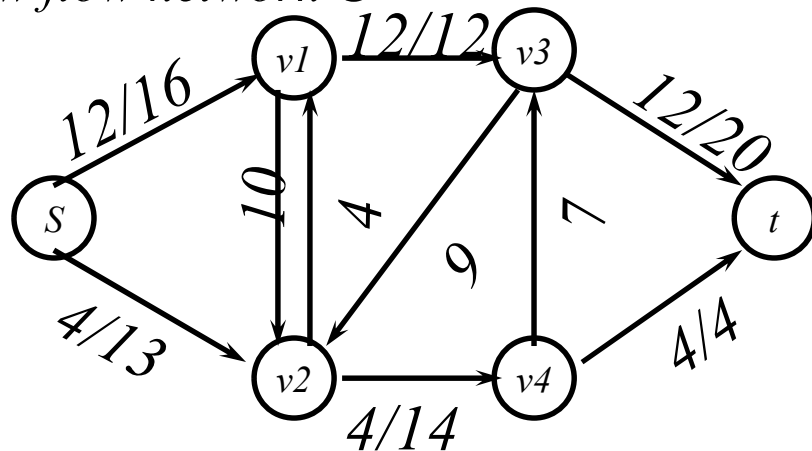
(residual) network G_f



```

1  for each edge  $(u, v) \in E [G]$ 
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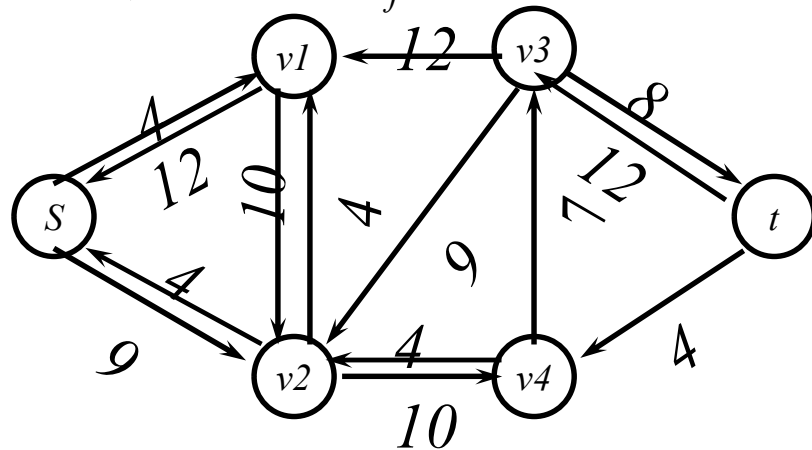
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

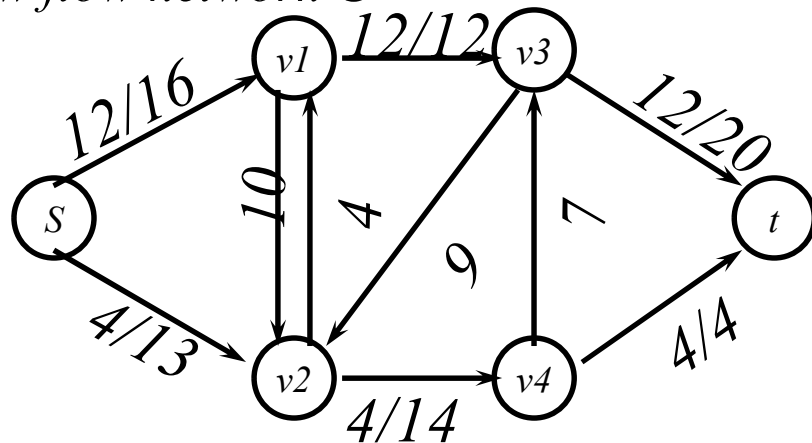
(residual) network G_f



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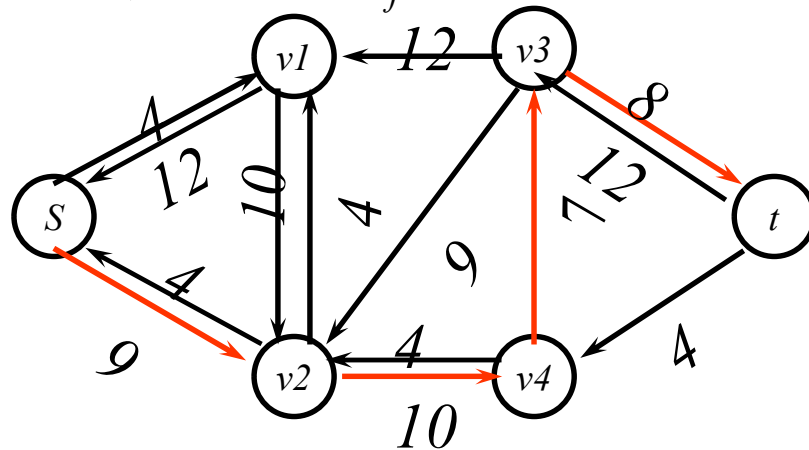
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

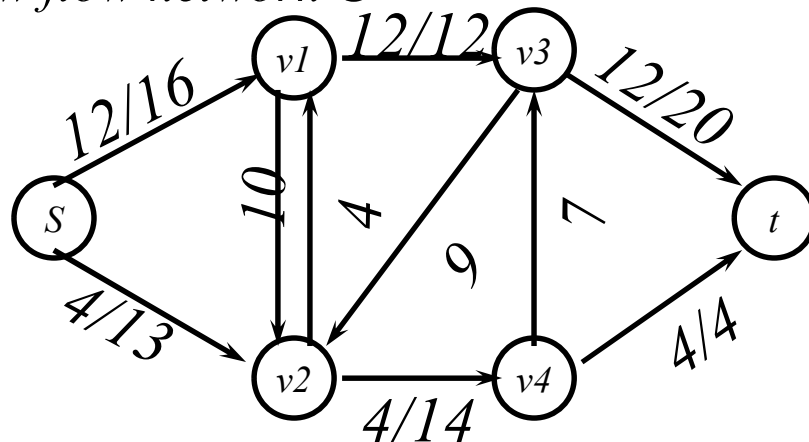
(residual) network G_f



```

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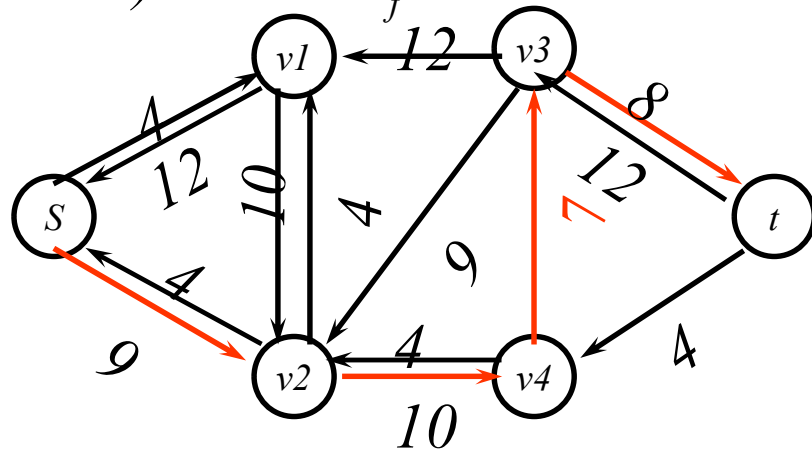
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

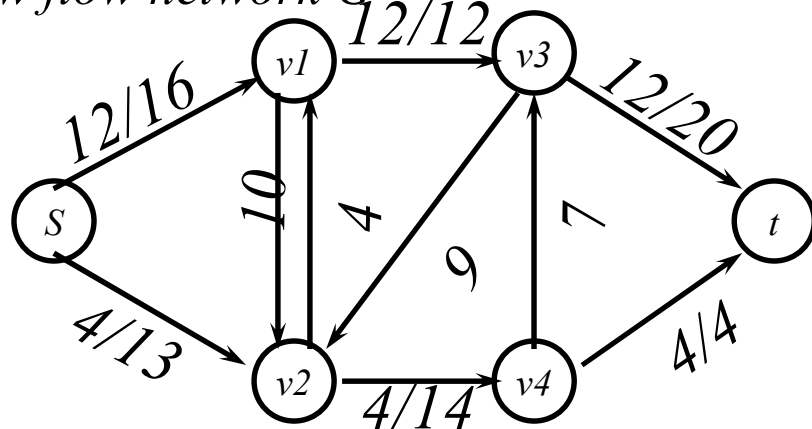
(residual) network G_f



```

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```

new flow network G



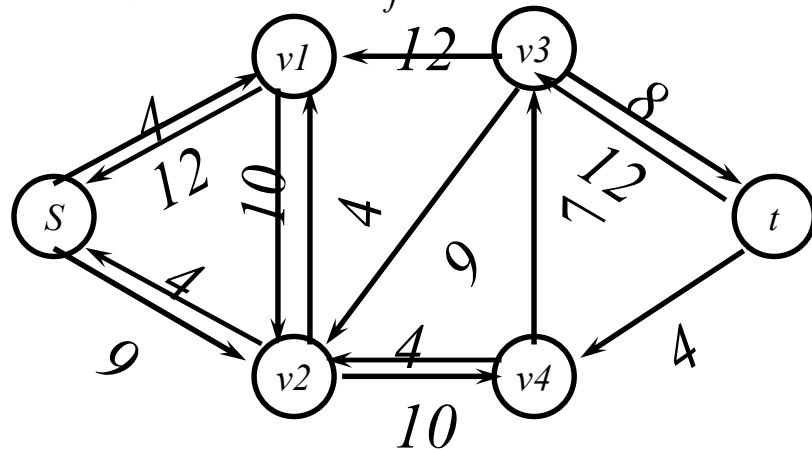
temporary variable:

$$c_f(p) = 7$$

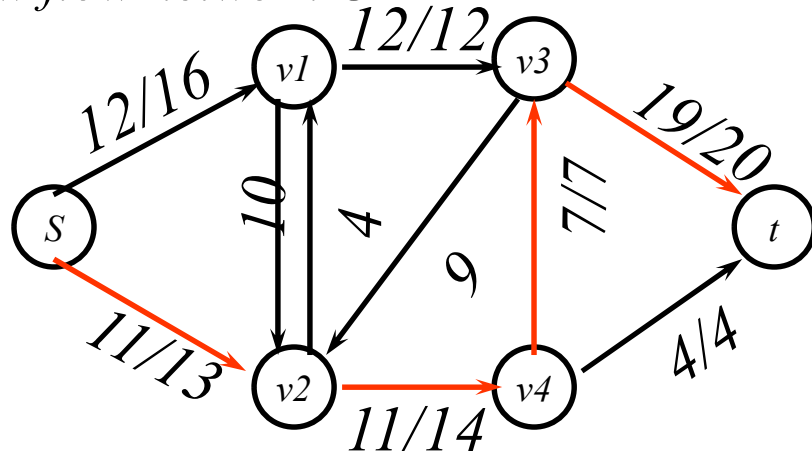
The basic Ford Fulkerson algorithm

example of an execution

(residual) network G_f



new flow network G



```

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2      do  $f[u, v] = 0$ 
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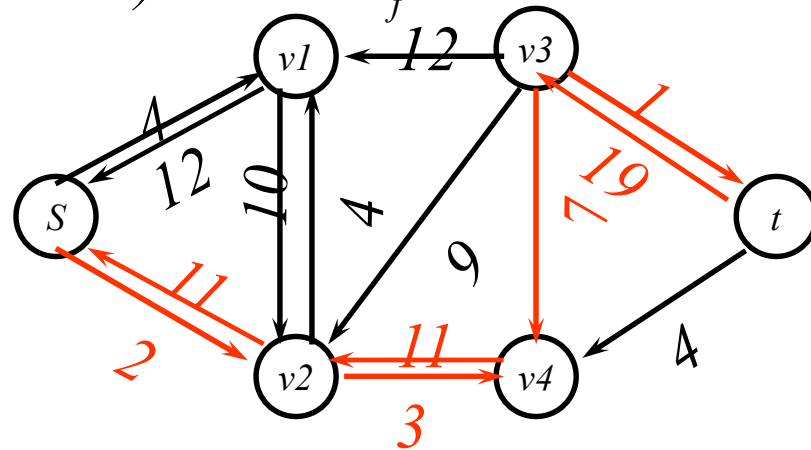
temporary variable:

$$c_f(p) = 7$$

The basic Ford Fulkerson algorithm

example of an execution

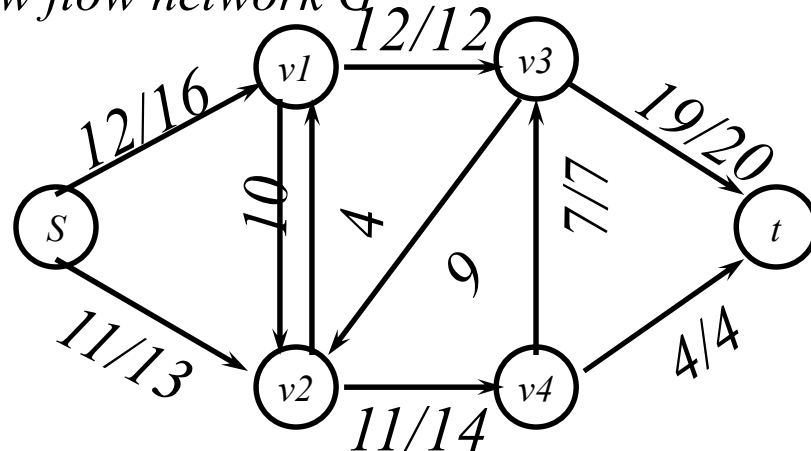
(residual) network G_f



```

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7          do  $f[u, v] = f[u, v] + c_f(p)$ 
    
```

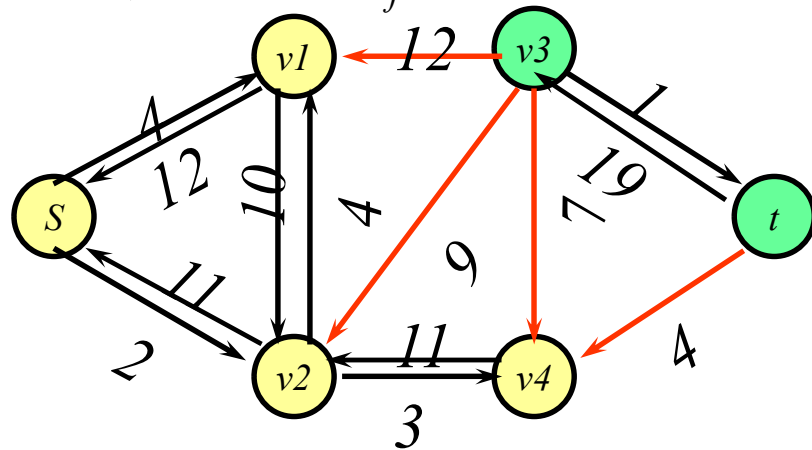
new flow network G



The basic Ford Fulkerson algorithm

example of an execution

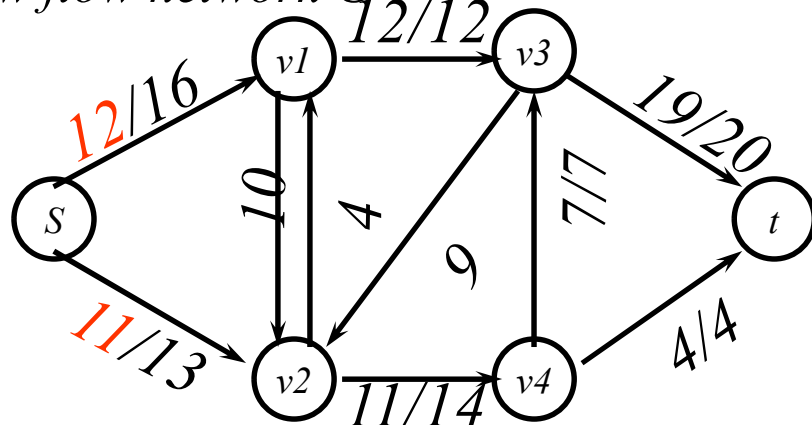
(residual) network G_f



```

1  for each edge  $(u, v) \in E [G]$ 
2      do  $f[u, v] = 0$ 
3      do  $f[v, u] = 0$ 
4  while there exists a path  $p$  from  $s$  to  $t$ 
    in the residual network  $G_f$ 
5      do  $c_f(p) = \min \{c_f(u, v) \mid (u, v) \in p\}$ 
6      for each edge  $(u, v)$  in  $p$ 
7          do  $f[u, v] = f[u, v] + c_f(p)$ 
    
```

new flow network G



Finally we have:

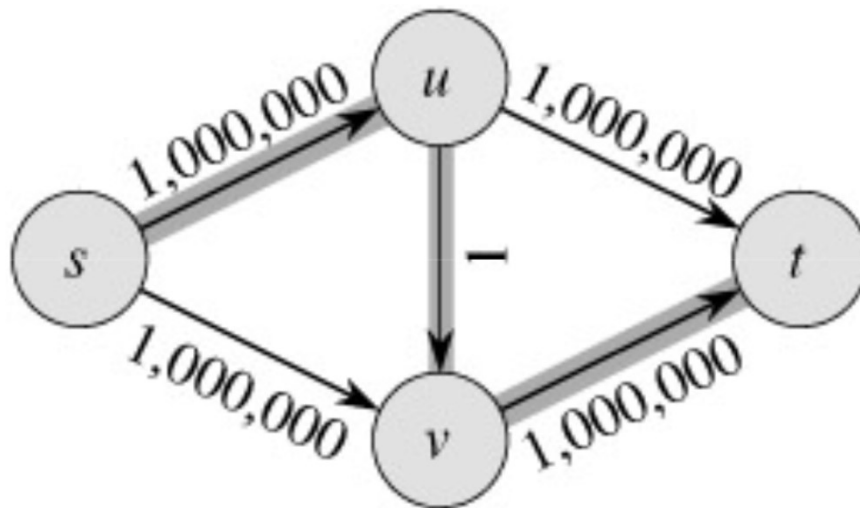
$$|f| = f(s, V) = 23$$

Time complexity:

- If each $c(e)$ is an *integer*, then time complexity is $O(|E|f^*)$, where f^* is the maximum flow.
- Reason: each time the flow is increased by at least one.
- This might not be a polynomial time algorithm since f^* can be represented by $\log(f^*)$ bits. So, the input size might be $\log(f^*)$.

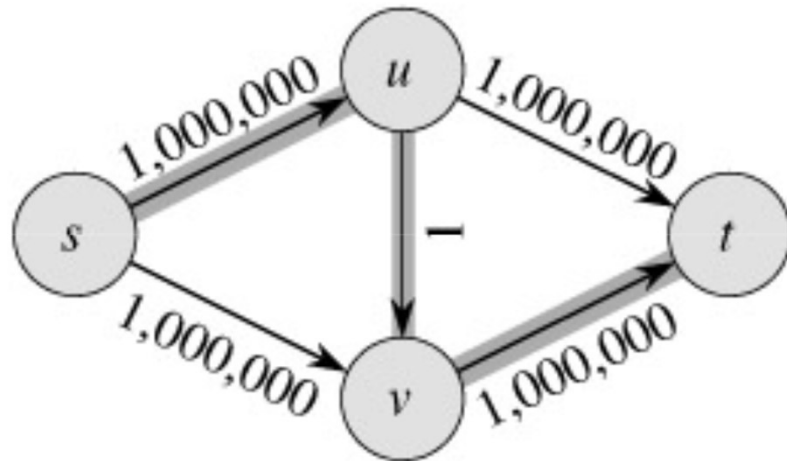
The Basic Ford-Fulkerson Algorithm

- With time $O(E |f^*|)$, the algorithm is **not** polynomial.
- Ford-Fulkerson may perform very badly

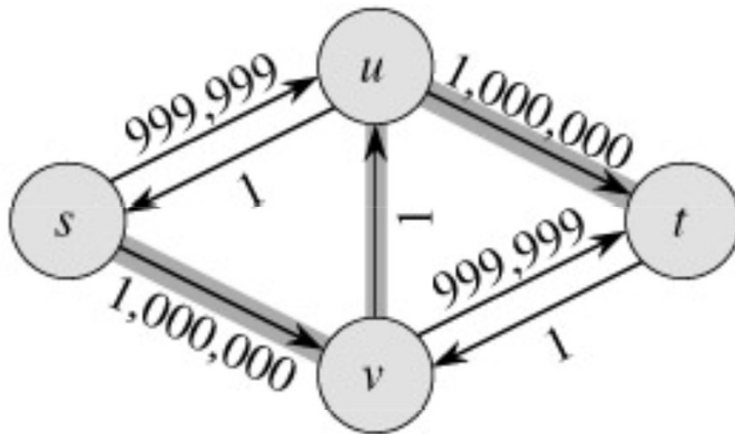


$$|f^*| = 2,000,000$$

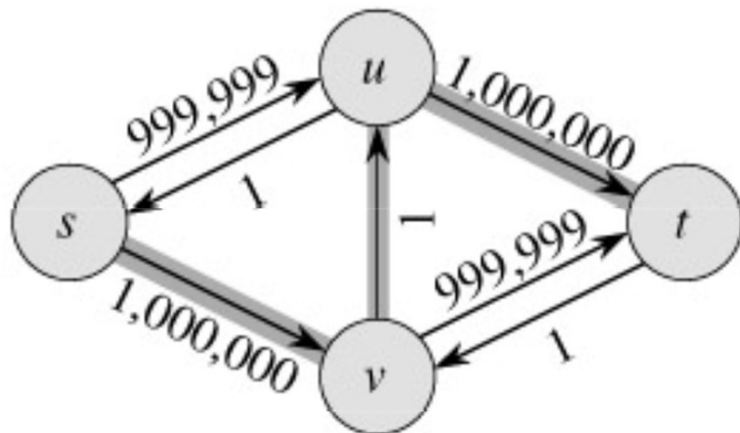
Run Ford-Fulkerson on this example



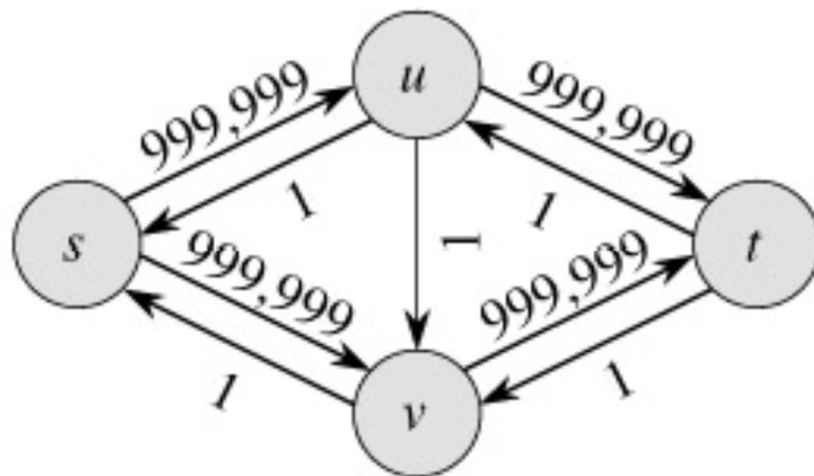
Augmenting Path



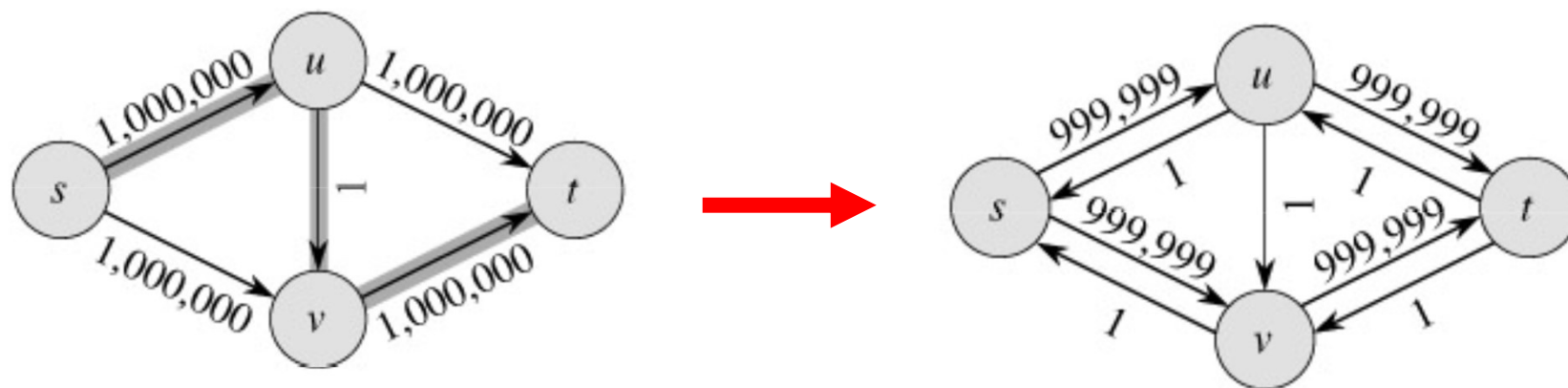
Residual Network



Augmenting Path



Residual Network

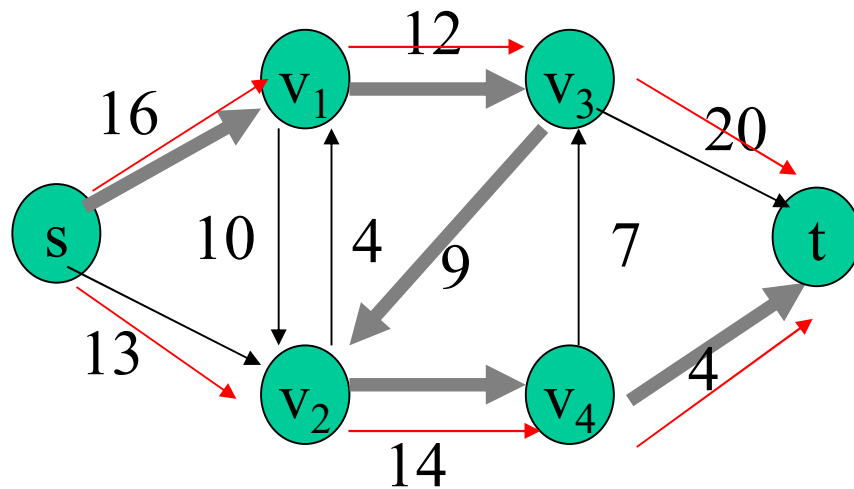


- Repeat 999,999 more times...
- Can we do better than this?

The Edmonds-Karp algorithm

- Find the augment path using breadth-first search.
- Breadth-first search gives the shortest path for graphs (Assuming the length of each edge is 1.)
- Time complexity of Edmonds-Karp algorithm is $O(V \cdot E^2)$.
- The proof is very hard and is not required here.

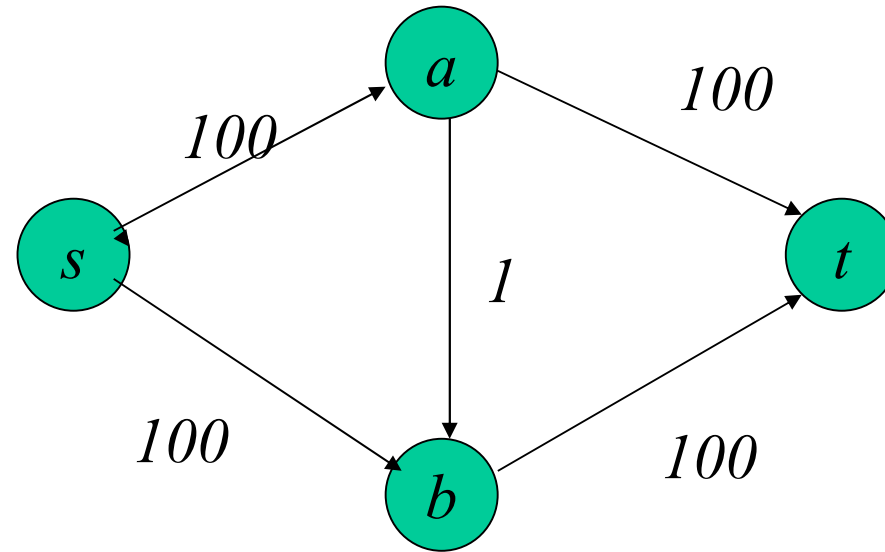
Breadth-first search



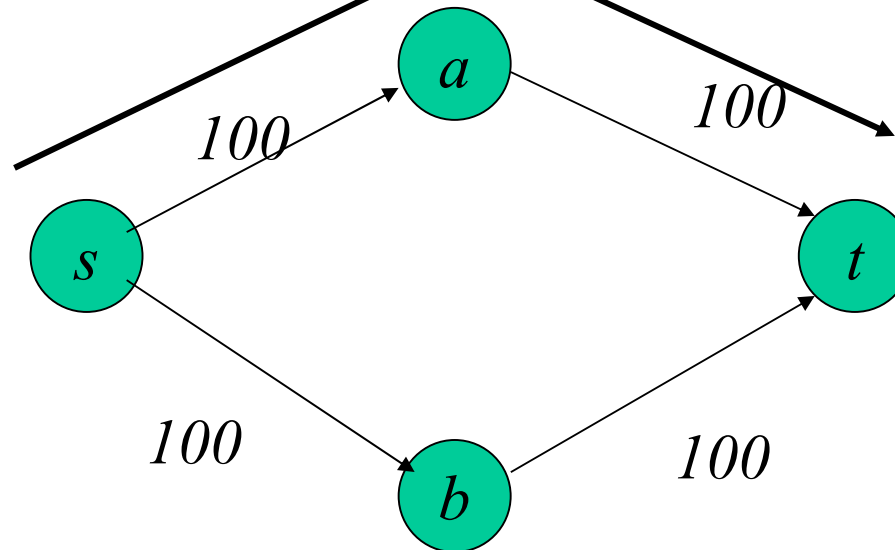
Path : $s \rightarrow v_1 \rightarrow v_3 \rightarrow t$ Path: $s \rightarrow v_2 \rightarrow v_4 \rightarrow t$.

(a)

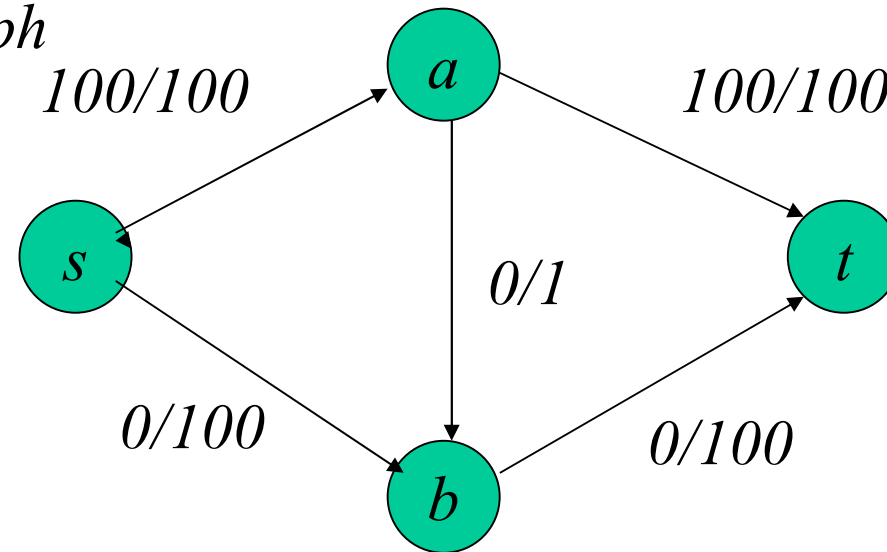
Example :



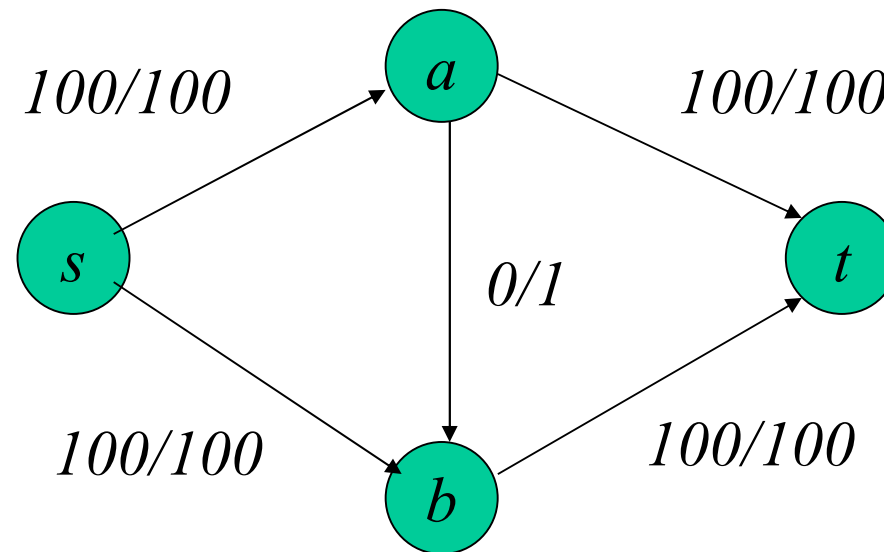
Applying breadth first search



Residual graph



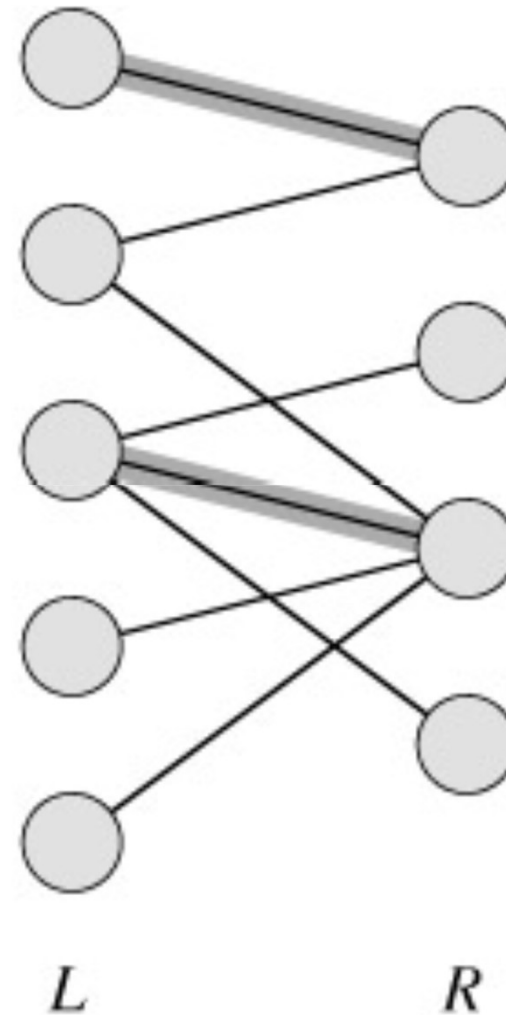
Again Run BFS So the obvious path is s - b - t



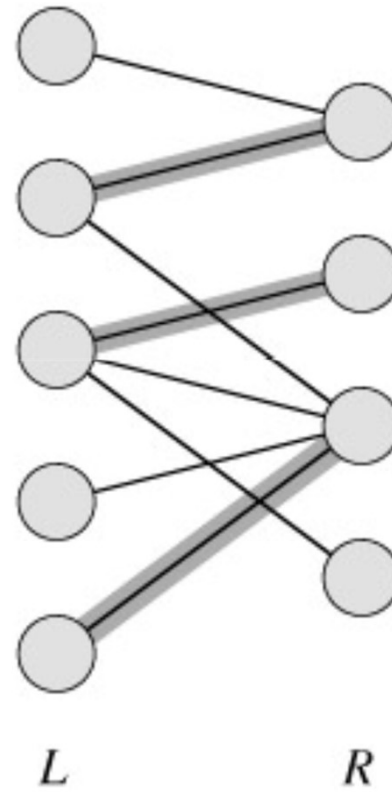
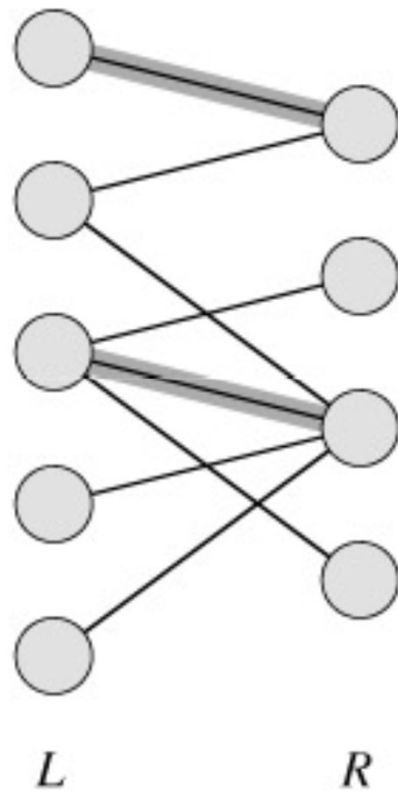
Maximum bipartite matching

Maximum Bipartite Matching

- A *bipartite graph* is a graph $G=(V,E)$ in which V can be divided into two parts L and R such that every edge in E is between a vertex in L and a vertex in R .
- e.g. vertices in L represent skilled workers and vertices in R represent jobs. An edge connects workers to jobs they can perform.

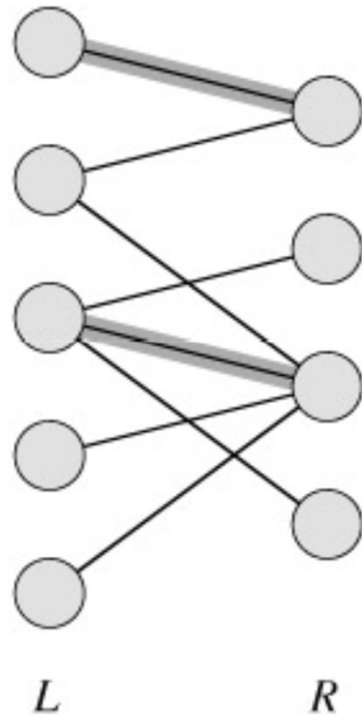


- A **matching** in a graph is a subset M of E , such that for all vertices v in V , **at most one edge of M is incident on v .**

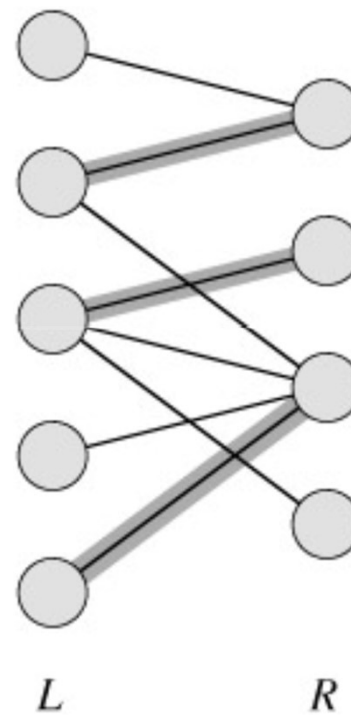


- A **maximum matching** is a matching of maximum cardinality (maximum number of edges).

not maximum

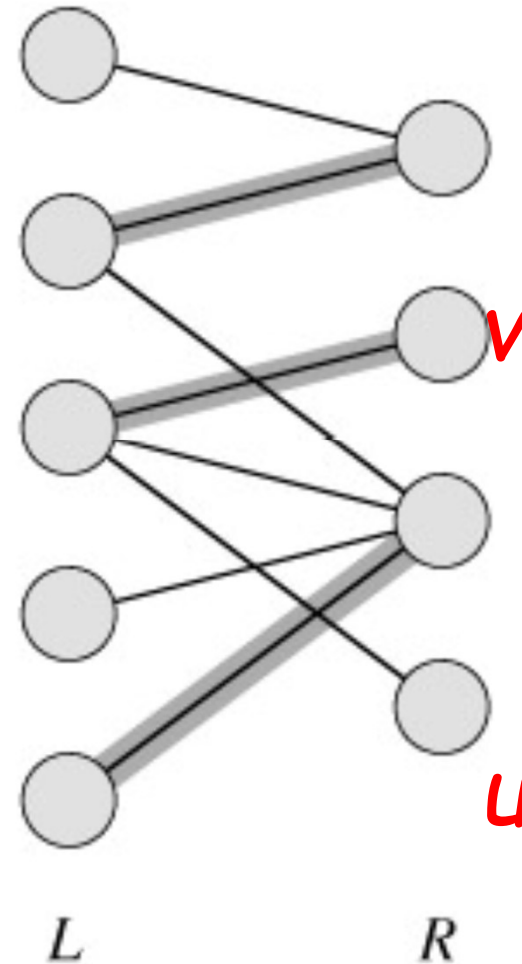


maximum



A Maximum Matching

- No matching of cardinality 4, because only one of v and u can be matched.
- In the workers-jobs example a max-matching provides work for as many people as possible.

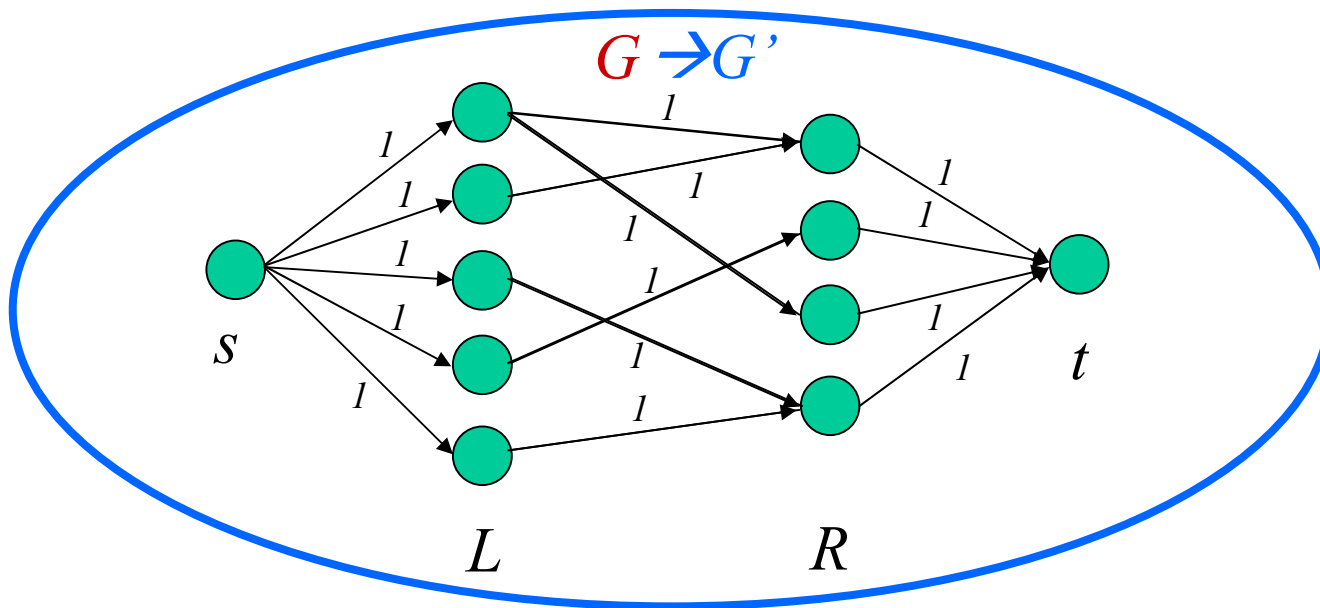


Solving the Maximum Bipartite Matching Problem

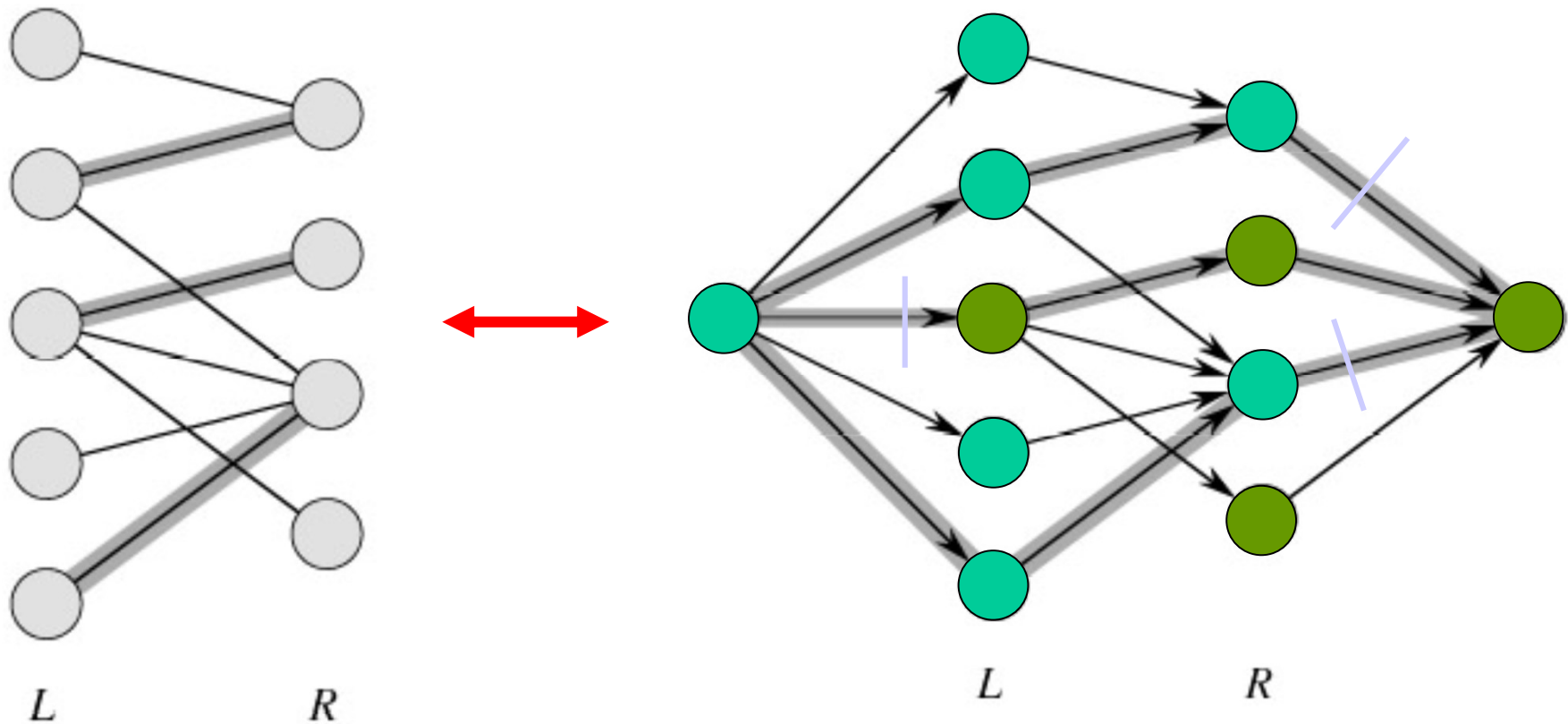
- Reduce the maximum bipartite matching problem on graph **G** to the max-flow problem on a corresponding flow network **G'**.
- Solve using Ford-Fulkerson method.

Corresponding Flow Network

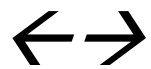
- To form the corresponding flow network G' of the bipartite graph G :
 - Add a source vertex s and edges from s to L .
 - Direct the edges in E from L to R .
 - Add a sink vertex t and edges from R to t .
 - Assign a capacity of 1 to all edges.
- Claim: max-flow in G' corresponds to a max-bipartite-matching on G .



Example



$$|M| = 3$$



$$53 \max \text{ flow} = |f| = 3$$