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Lecture 12: Flow Graphs

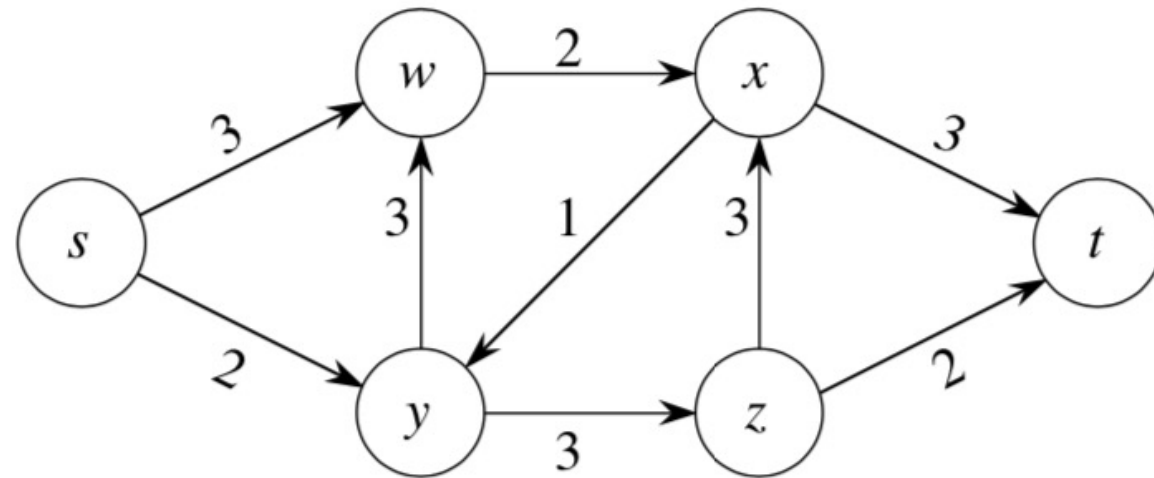
CSCI 3070U: Design and Analysis of Algorithms

Learning Outcomes

- Flow Networks:
 - Concepts and Foundations
 - Algorithms

Flow Network

- A **flow network** $G=(V, E)$ is a directed graph in which each edge $(u, v) \in E$ has a nonnegative **capacity** $c(u, v) \geq 0$



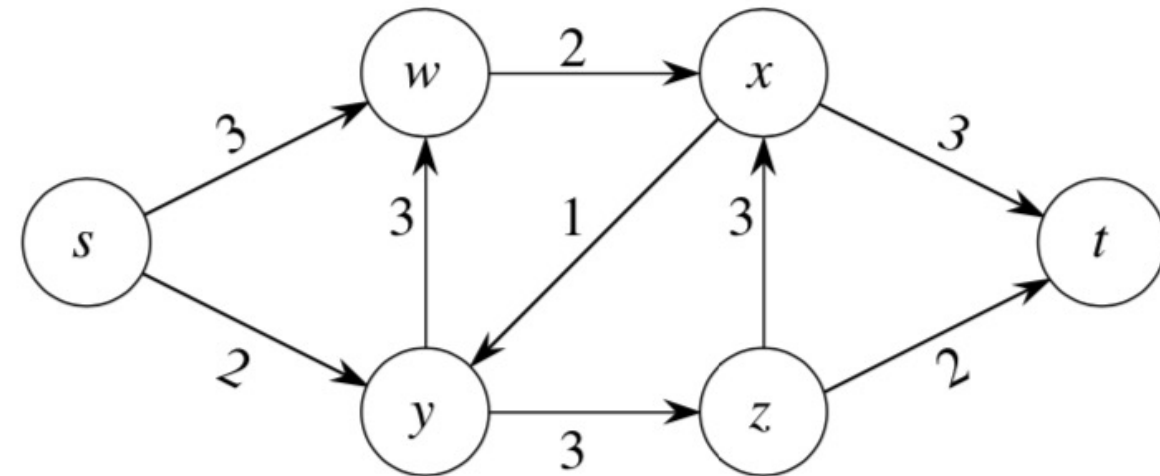
Flow Network

- We distinguish two vertices in a flow network:
 - Source s
 - Sink t

- We assume that each vertex lies on a path from source to sink

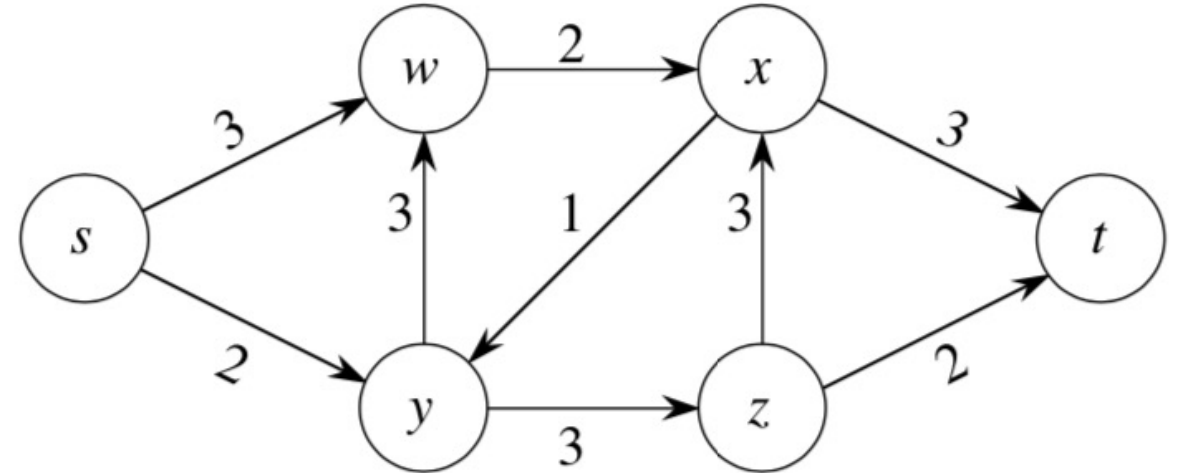
- For all $v \in V$, we have a path

$$s \rightsquigarrow v \rightsquigarrow t$$



Flow Networks Properties

- A **flow network** $G=(V, E)$ is a directed graph in which each edge $u \in E$ has a nonnegative **capacity** $c(u, v) \geq 0$
- If $(u, v) \in E$, then $(v, u) \notin E$
- If $(v, u) \notin E$, then $c(v, u) = 0$



Flow Definition

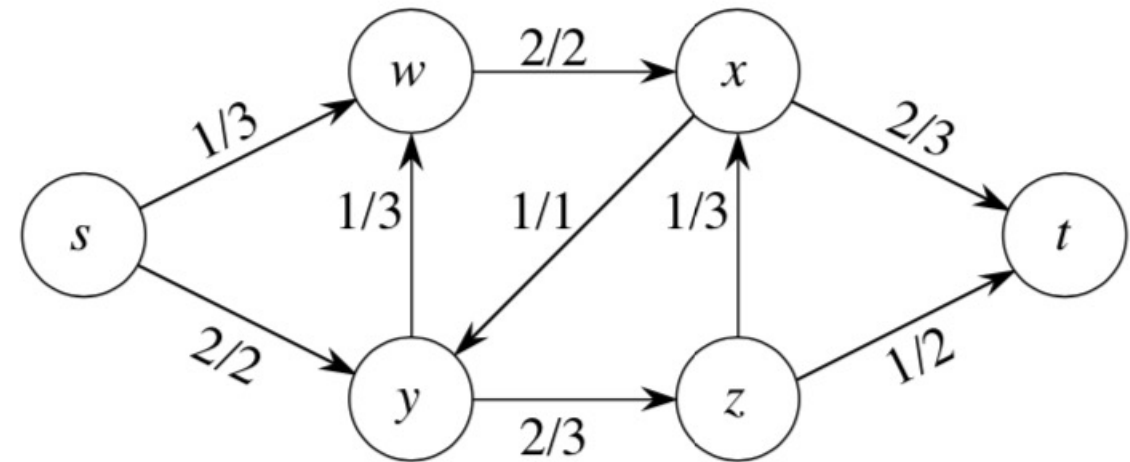
- Flow is a function $f : V \times V \rightarrow \mathbb{R}$ satisfying:

- Capacity Constraint:

For all $u, v \in V, 0 \leq f(u, v) \leq c(u, v)$.

- Flow Conservation:

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



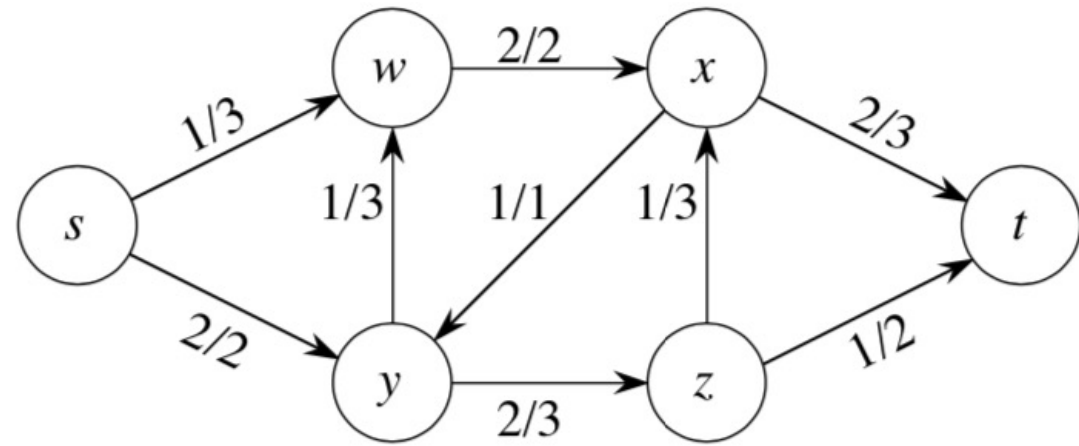
Value of Flow

- Value of flow f is defined as follows:

$$\begin{aligned}\text{Value of flow } f &= |f| \\ &= \sum_{v \in V} f(s, v) - \sum_{v \in V} f(v, s) \\ &= \text{flow out of source} - \text{flow into source}\end{aligned}$$

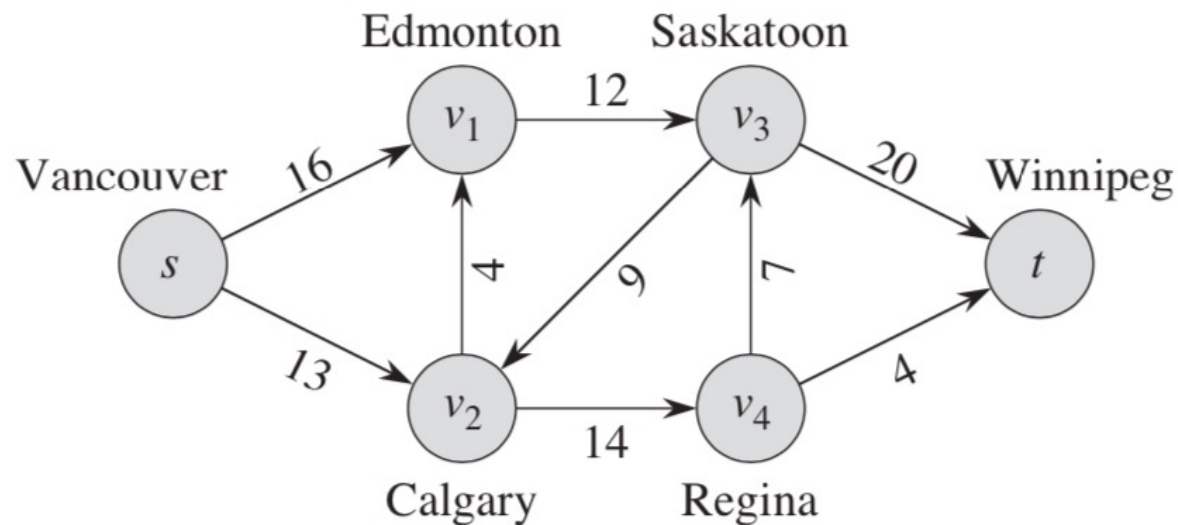
- Example:
 $|f| = ?$

Answer = 3



Maximum Flow Problem

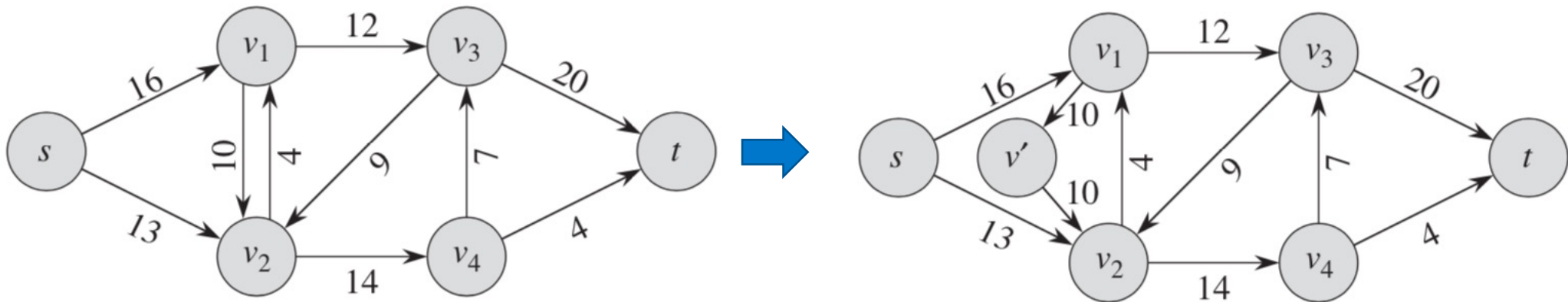
- Given G , s , t , and c , find a flow whose value is **maximum**



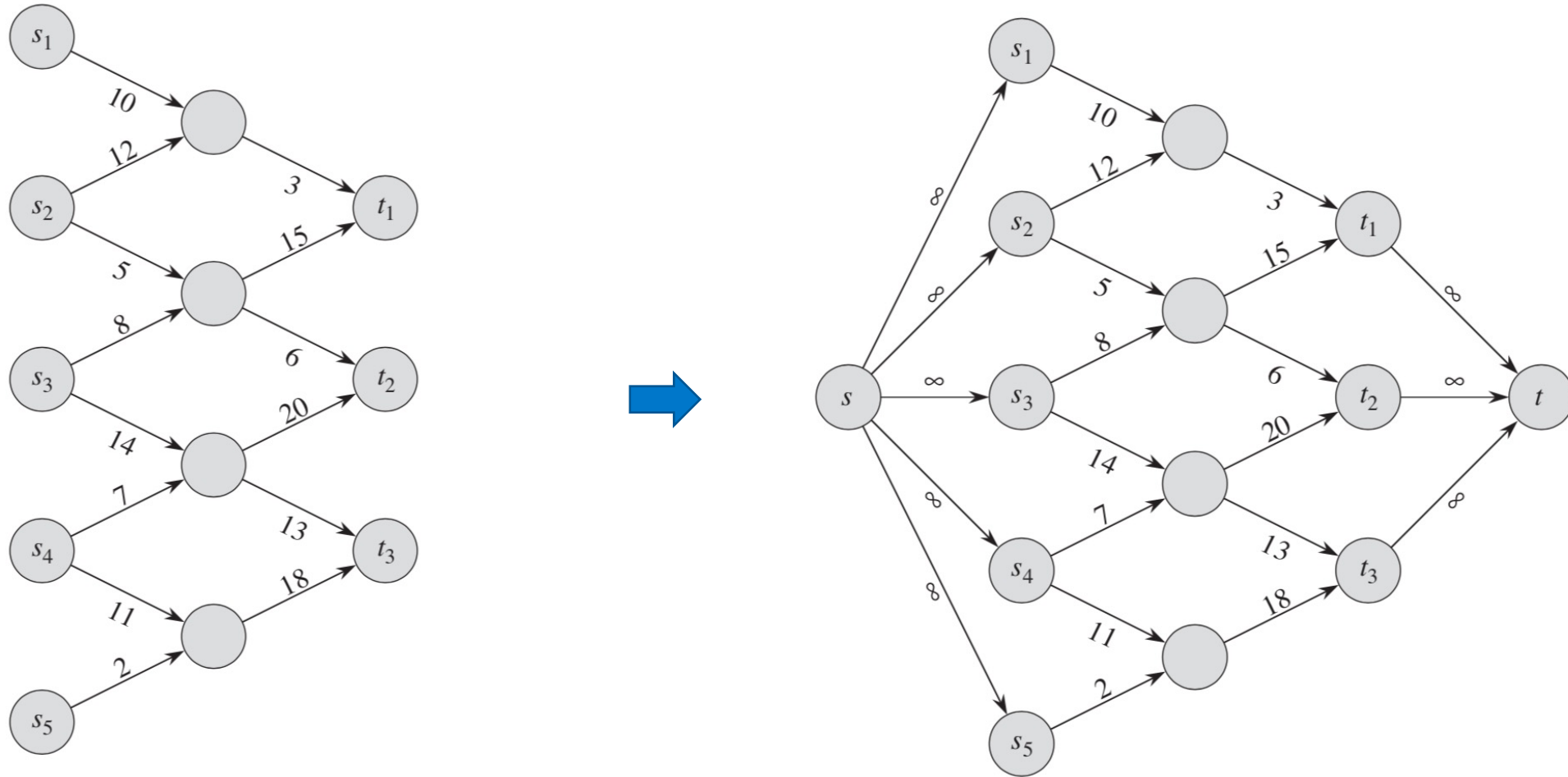
- Maximum rate of shipping product from Vancouver to Winnipeg through intermediate cities

Antiparallel Edges

- Definition of flow network does not allow both (u, v) and (v, u) to be edges. These edges would be **antiparallel**
- What if we really need antiparallel edges?



Networks with Multiple Sources and Sinks

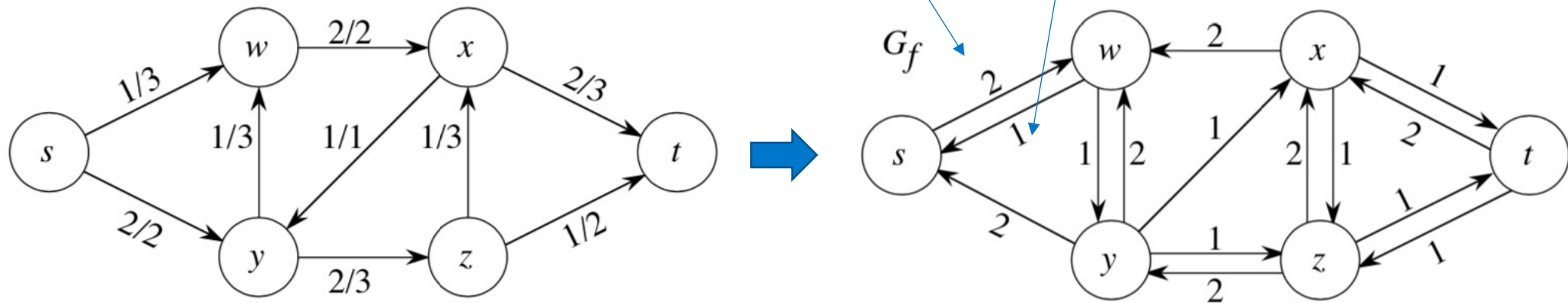


Residual Network

- Given a flow f in network $G = (V, E)$, the residual network of G is G_f

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

$$c_f(v, u) = f(u, v)$$



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

Augmenting Paths

- Given a flow network $G = (V, E)$ and a flow f , an **augmenting path** p is a simple path from s to t in **the residual network** G_f .
 - A simple path is a path without cycle.

$$s \rightsquigarrow t$$

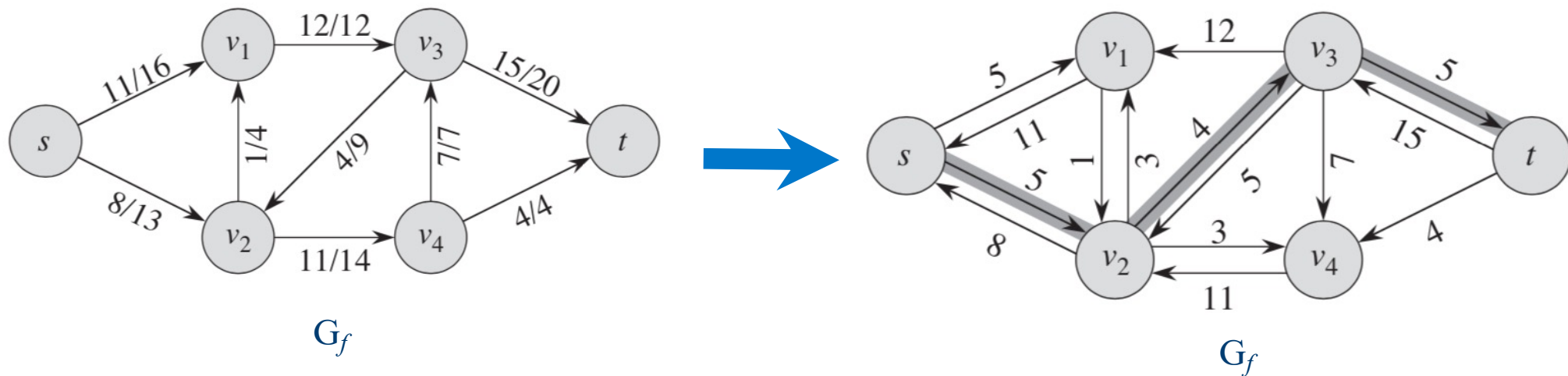
- How much more flow can we push from s to t along augmenting path p ?

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

Residual Capacity

Augmenting Paths

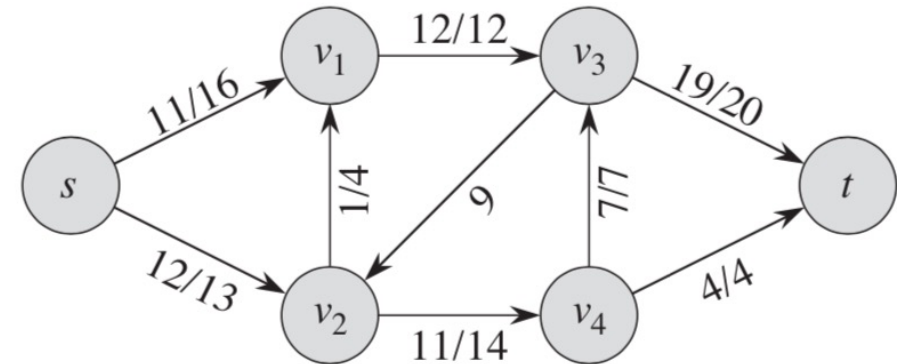
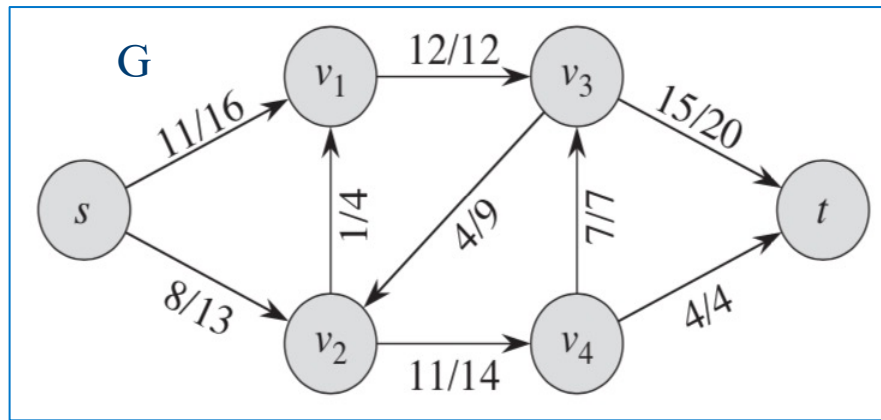
- The shaded path is an augmenting path



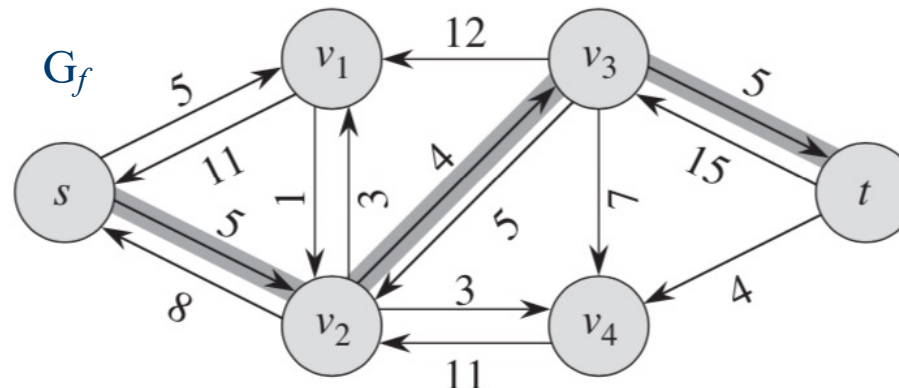
$$c_f(p) = \min \{c_f(u, v) : (u, v) \text{ is on } p\} = \min\{5, 4, 5\} = 4$$

Augmenting Paths

- Insight:** We can increase the flow through each edge of this path by up to 4 units without violating the capacity constraint:



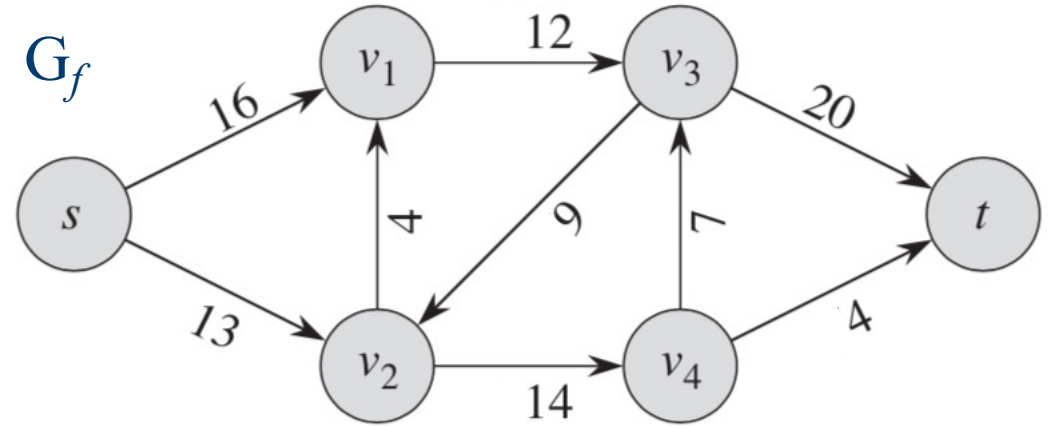
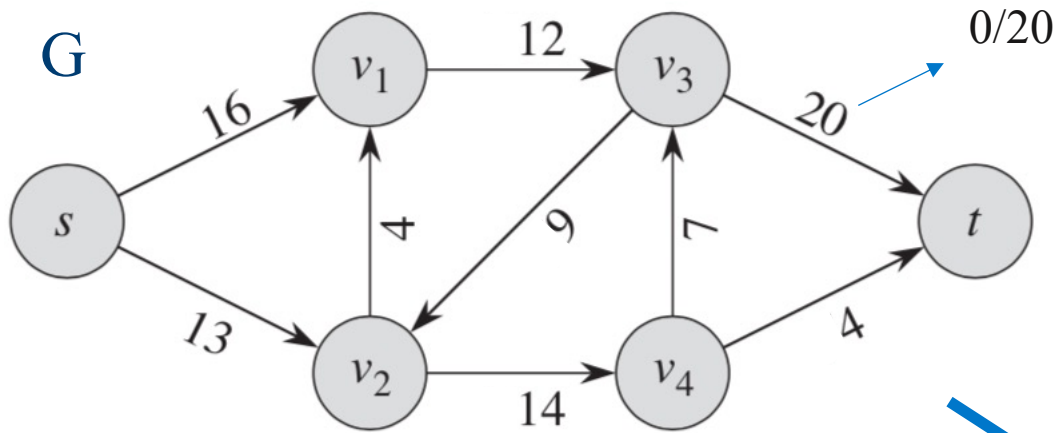
G with augmented flow
based on path p



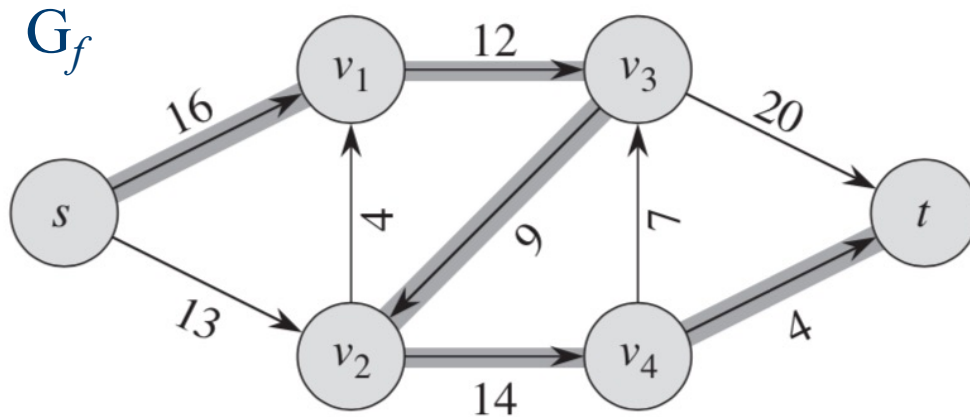
Ford-Fulkerson Algorithm

- Given G , s , t , and c , it finds a flow whose value is **maximum**
- General ideas:
 - In each iteration of the Ford-Fulkerson method, we find **some augmenting path p** and use p to modify the flow f
 - That is, adding flow when the residual edge in p is an original edge and subtracting it otherwise
 - When no augmenting paths exist, the flow f is a maximum flow.

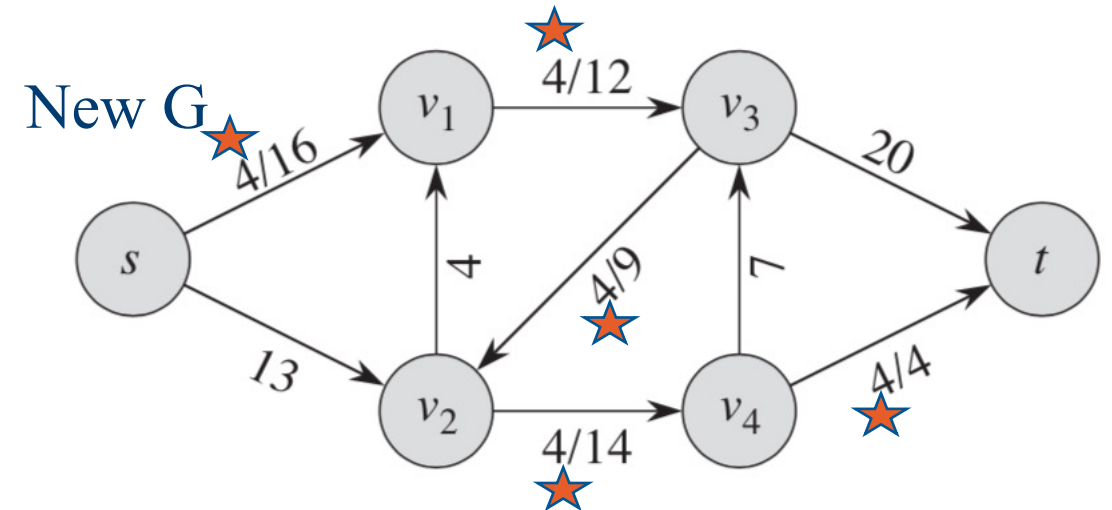
Ford-Fulkerson Algorithm Example



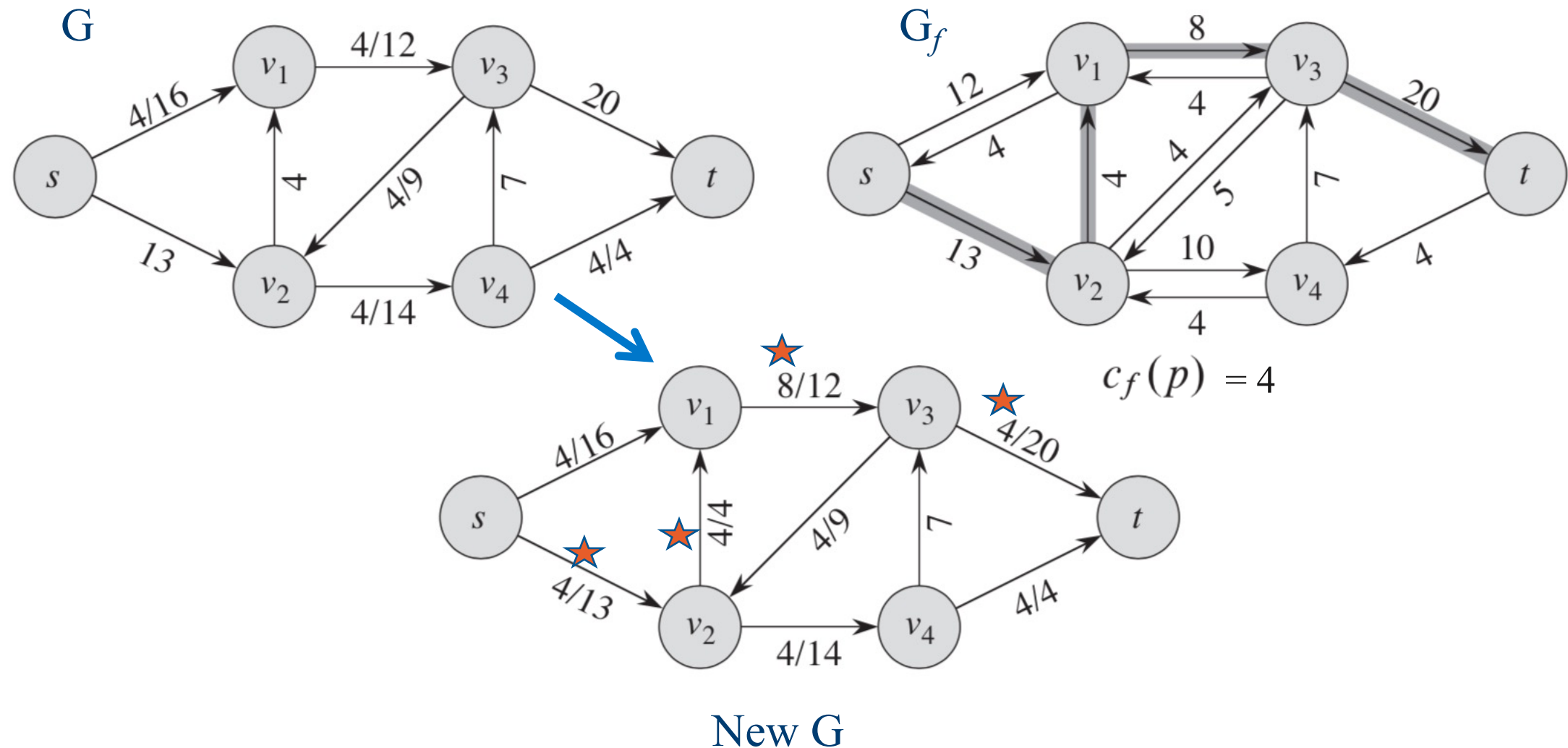
Are there any augmenting path?



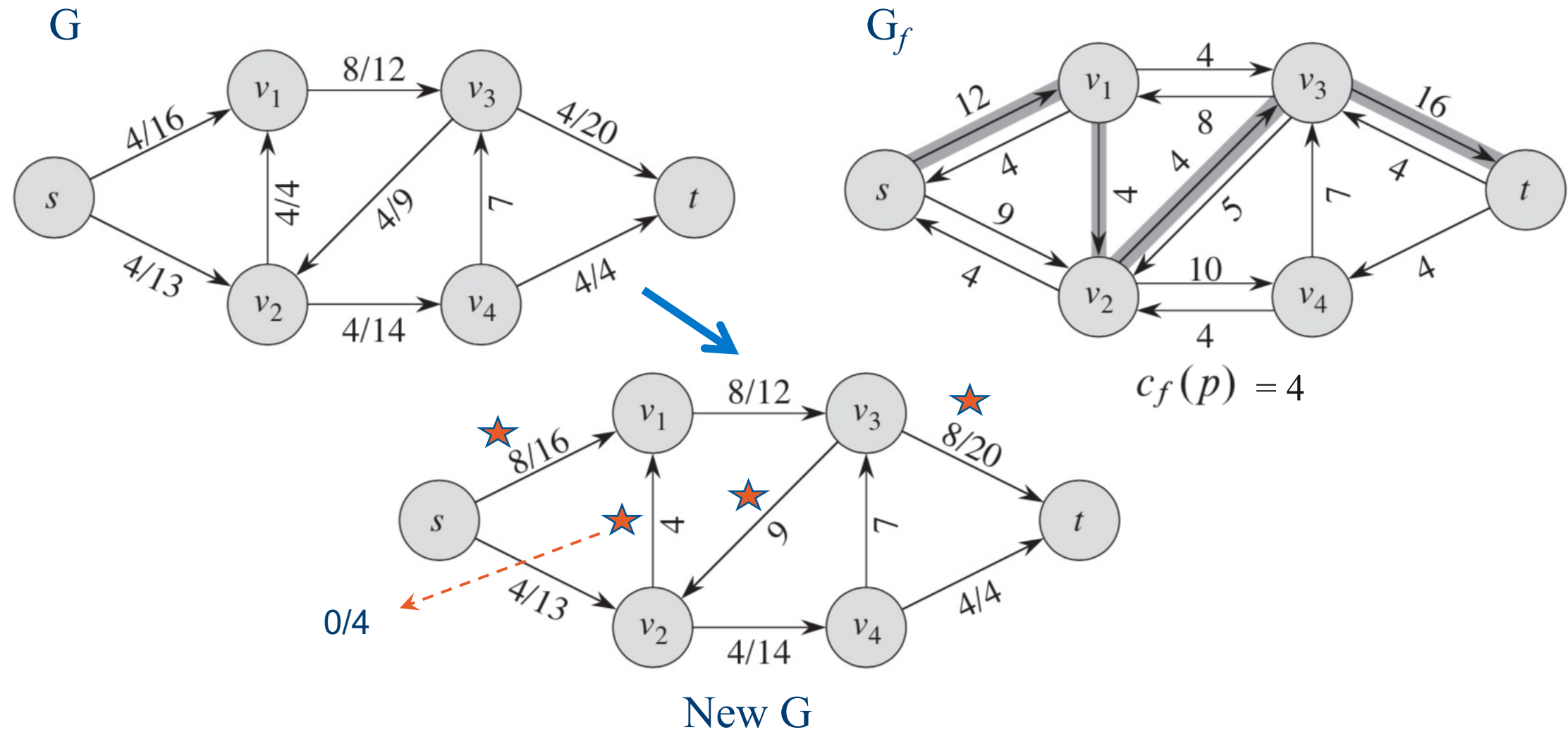
$$c_f(p) = 4$$



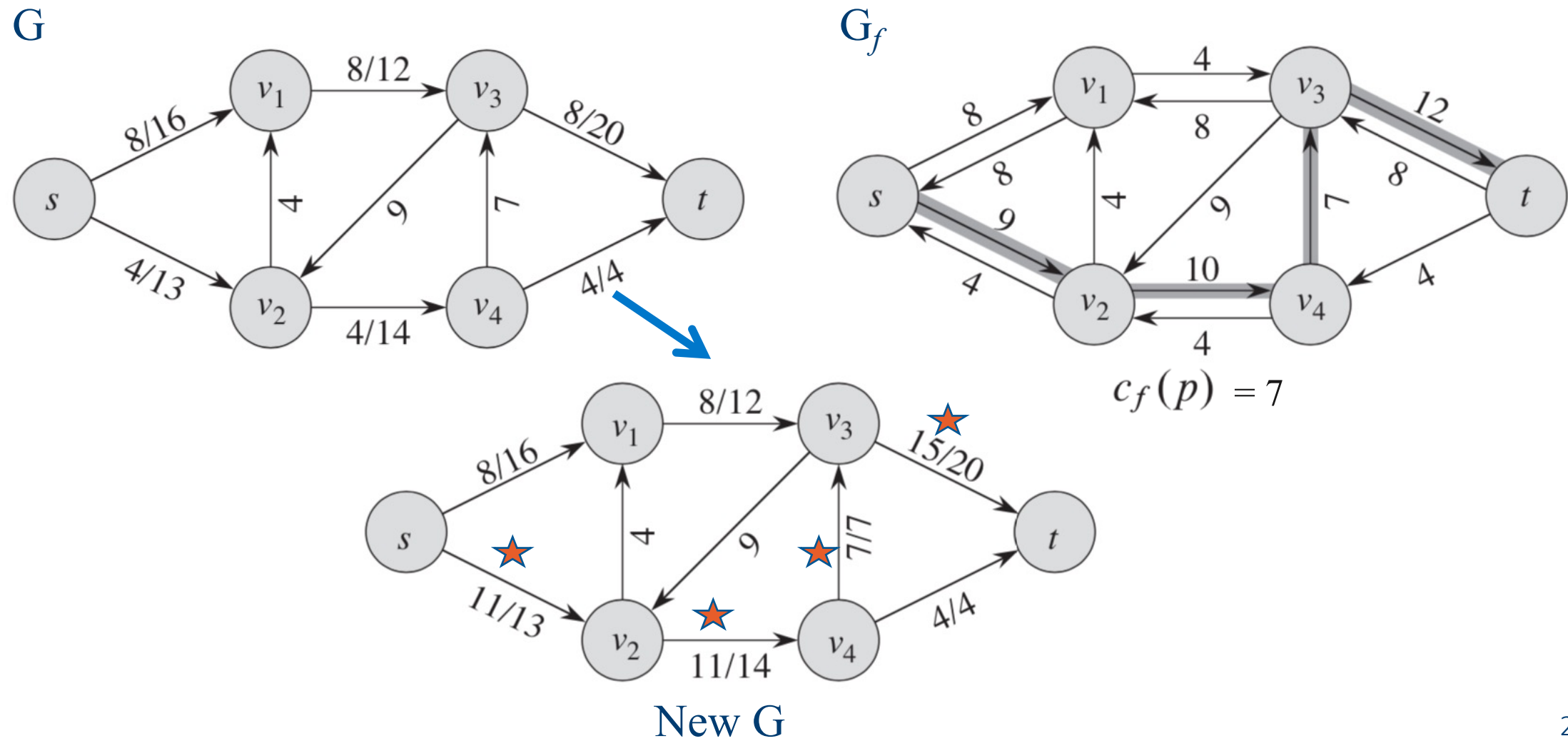
Ford-Fulkerson Algorithm Example



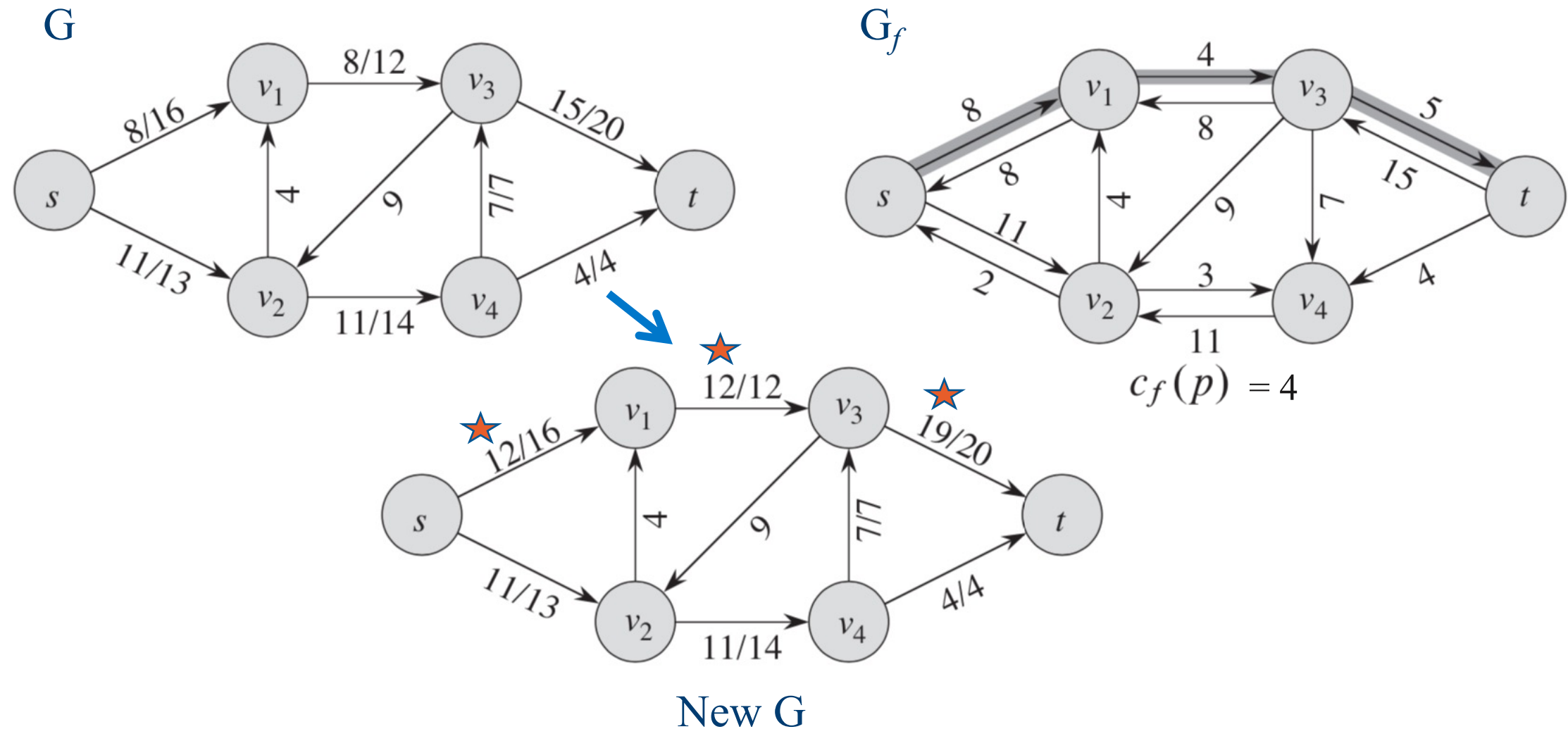
Ford-Fulkerson Algorithm Example



Ford-Fulkerson Algorithm Example

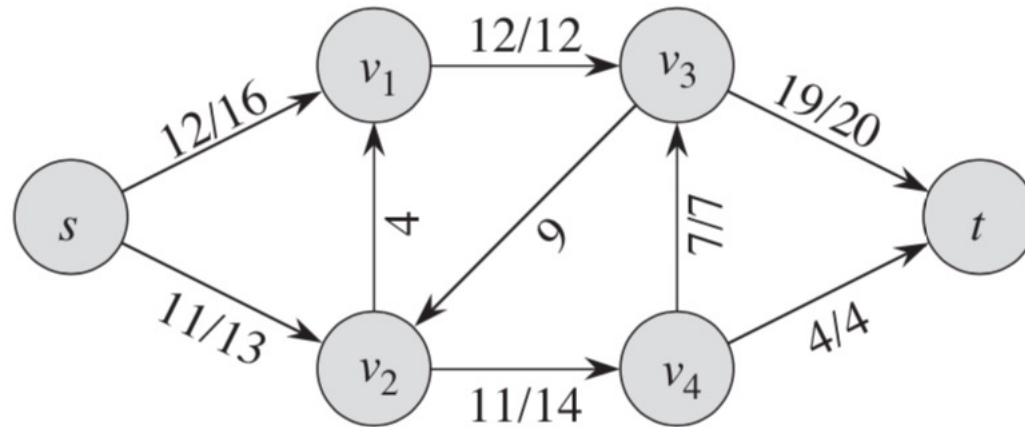


Ford-Fulkerson Algorithm Example



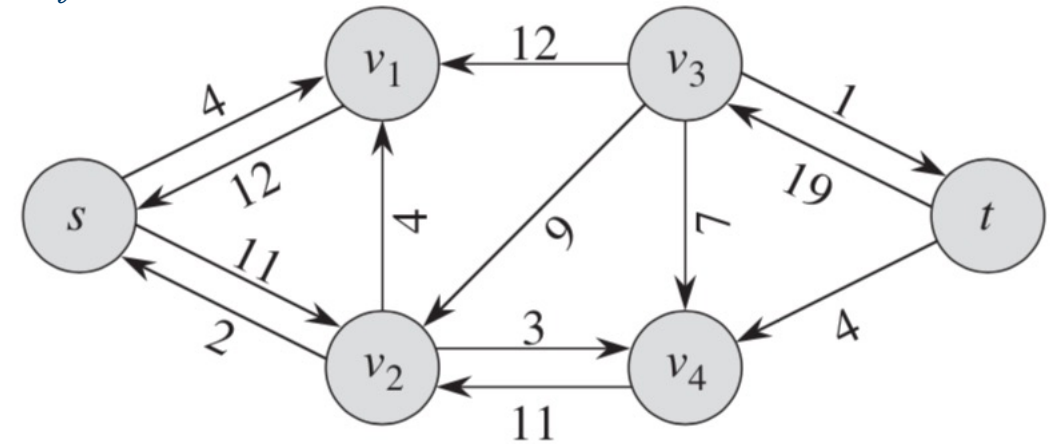
Ford-Fulkerson Algorithm Example

G



Answer = 23

G_f



No Augmenting Path !

Ford-Fulkerson Algorithm

FORD-FULKERSON(G, s, t)

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

Ford-Fulkerson Algorithm

FORD-FULKERSON(G, s, t)

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

$O(E)$

$O(|f^*|)$

$O(E|f^*|)$

Time Complexity:

$O(E|f^*|)$

$|f^*|$ denotes a maximum flow in the network

Ford-Fulkerson Algorithm

- Why we come up with maximum flow when there is no augmentation path in G_f ?
- **Theorem (Max-flow min-cut theorem):** The following are equivalent:
 - f is a maximum flow
 - G_f has no augmenting path

The Edmonds-Karp Algorithm

- We can improve the bound on Ford-Fulkerson by finding the augmenting path p in line 3 with a breadth-first search
- That is, we choose the augmenting path as *a shortest path* from s to t in the residual network, where each edge has unit distance (weight)
- The Edmonds-Karp algorithm runs in $O(VE^2)$

Wrap-up

- We learned about flow networks
 - Definitions
 - Properties
 - Max flow problem
 - Ford-Fulkerson
 - Edmonds-Karp