

CMP(N)302: Design and Analysis of Algorithms



Lecture 09: Flow Networks

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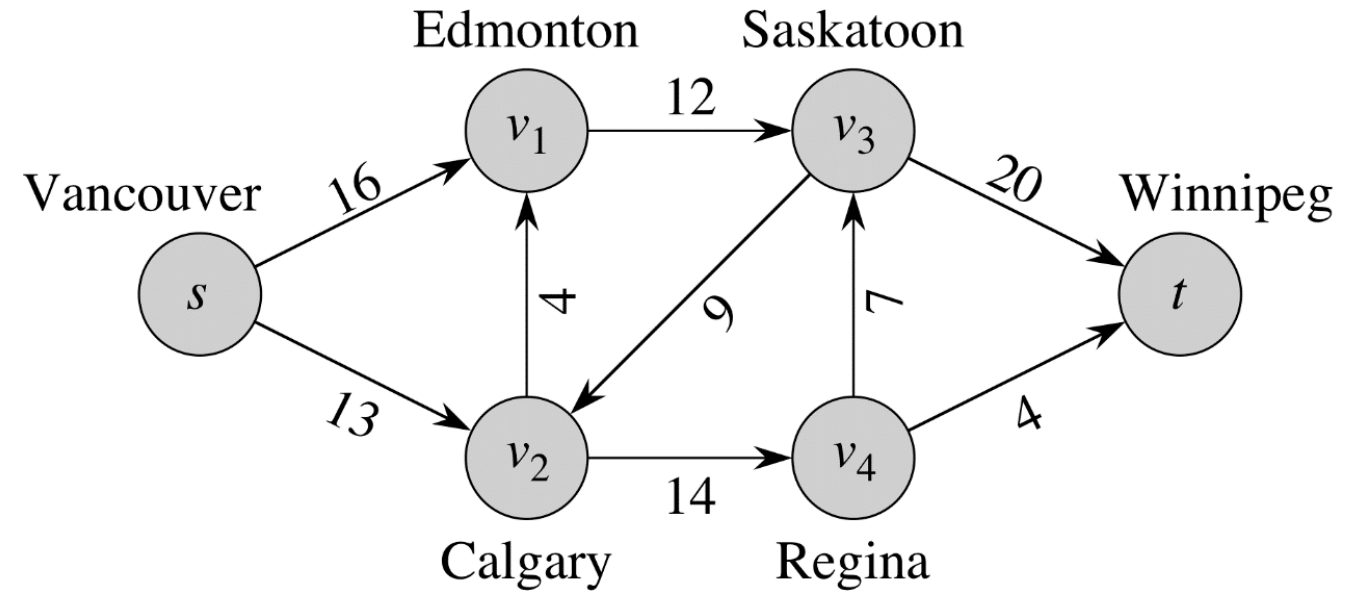
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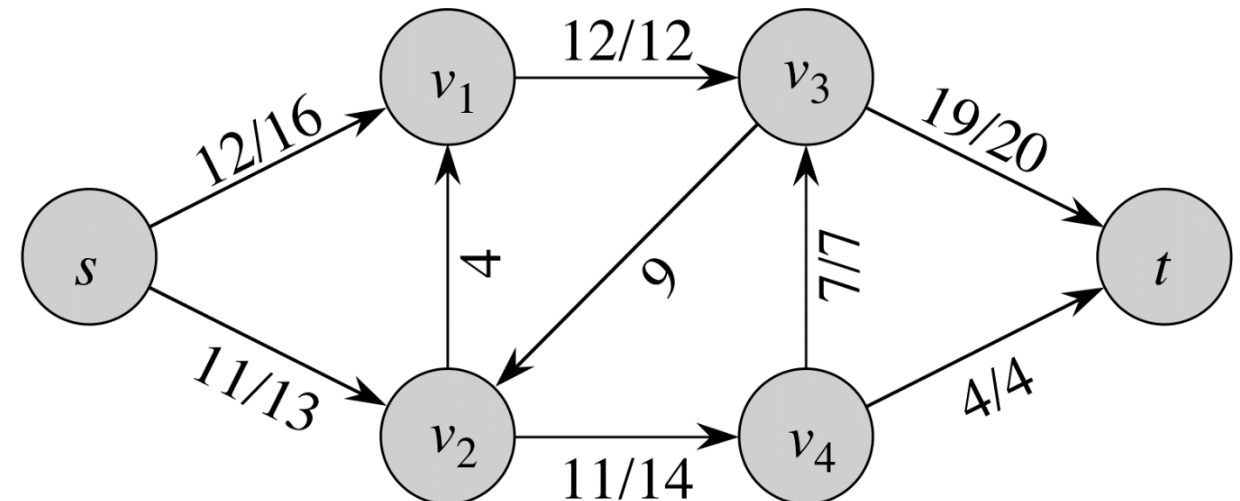
Fall 2019

Real life application

- Find max water rate flowing from **source** (Vancouver) to **sink** (Winnipeg) in the shown pipe network based on the shown capacities.
- Can you guess an upper-bound for what we can get from just looking?
- Iteratively how do you come up with the best solution?
- Between iterations, how to simplify and prepare the network for the next iteration?

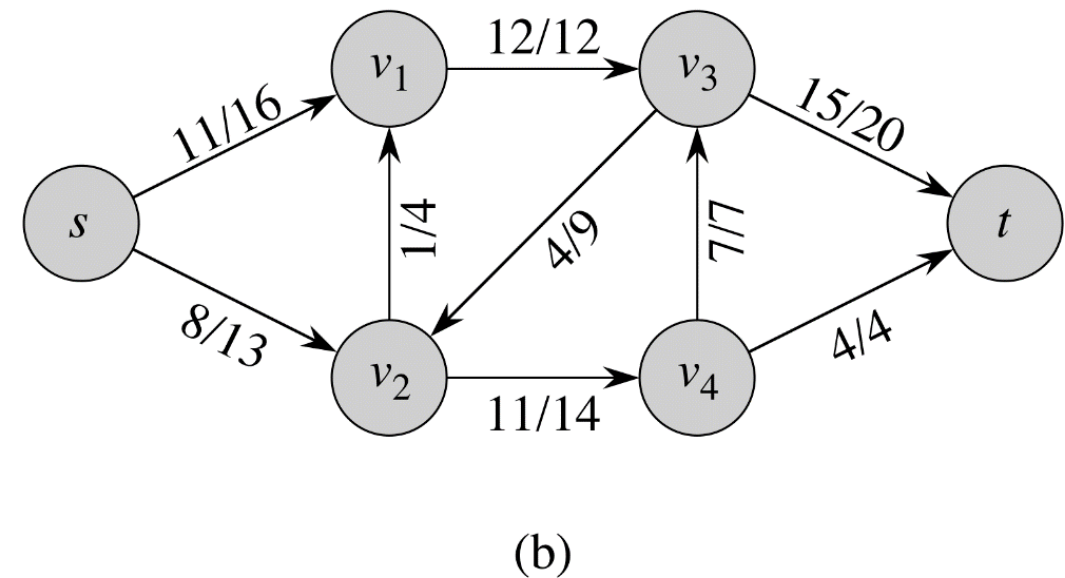
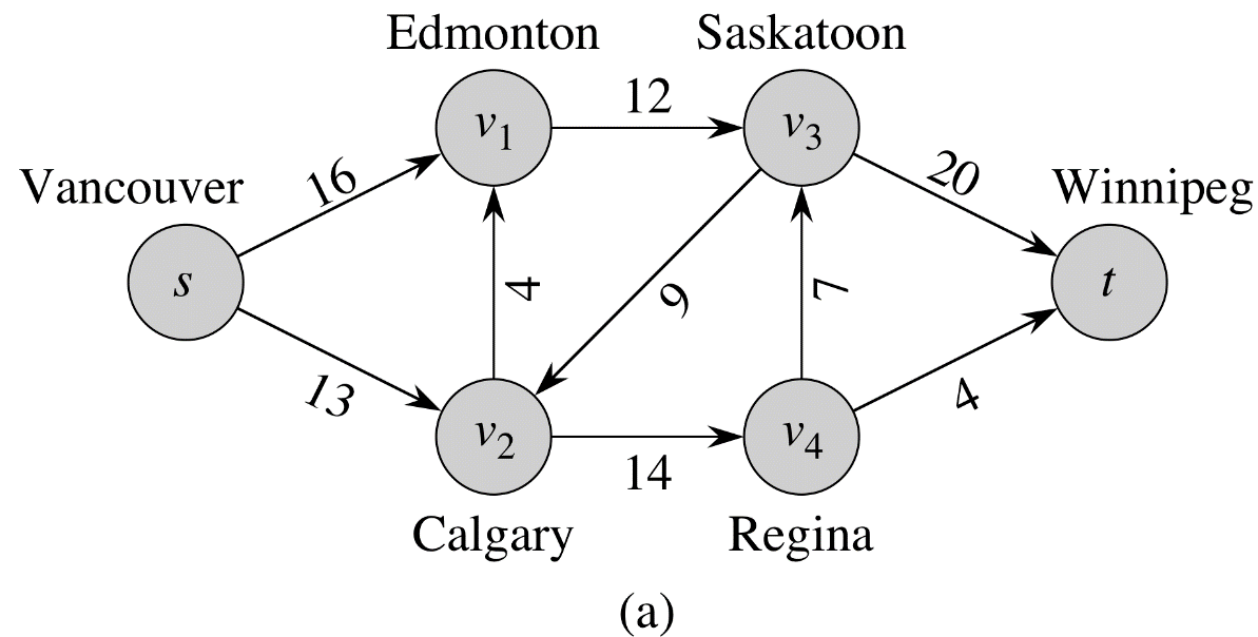


(a)



What is a flow network?

- **Flow network** is a directed graph $G(V, E)$
- Each edge $(u, v) \in E$ has a nonnegative **capacity** $c(u, v) \geq 0$. No self-loops.
- If there is $(u, v) \in E$, then there is no edge (v, u) in reverse direction, and $c(v, u) = 0$
- Typically, there is **source** s and **sink** t



What is a flow?

- **Flow** is a real-valued function $f: V \times V \rightarrow \mathbb{R}$ that satisfies:

- Capacity constraint: For all $u, v \in E$, we require

$$0 \leq f(u, v) \leq c(u, v)$$

- For all $u, v \in V - \{s, t\}$, we require

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

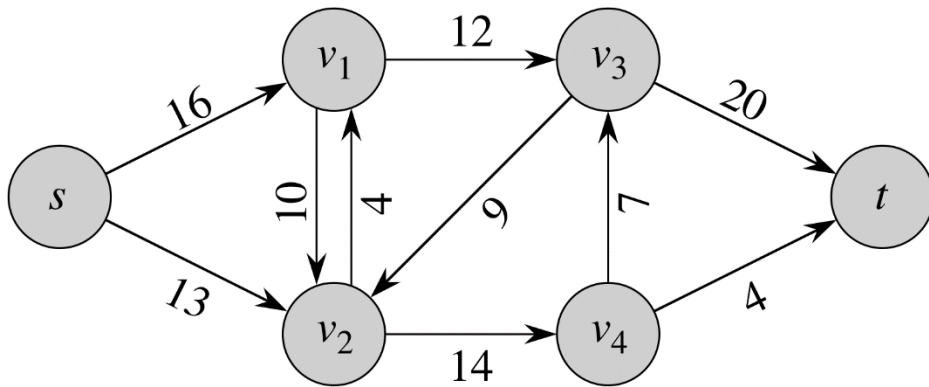
Modeling with antiparallel edges

- Antiparallel edge:

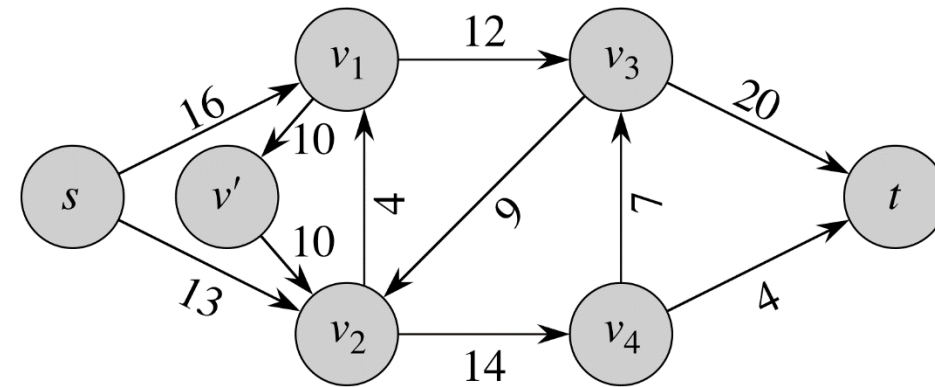
Edges (v_1, v_2) and (v_2, v_1) are called antiparallel

- Workaround:

Split one of the edges using a new vertex

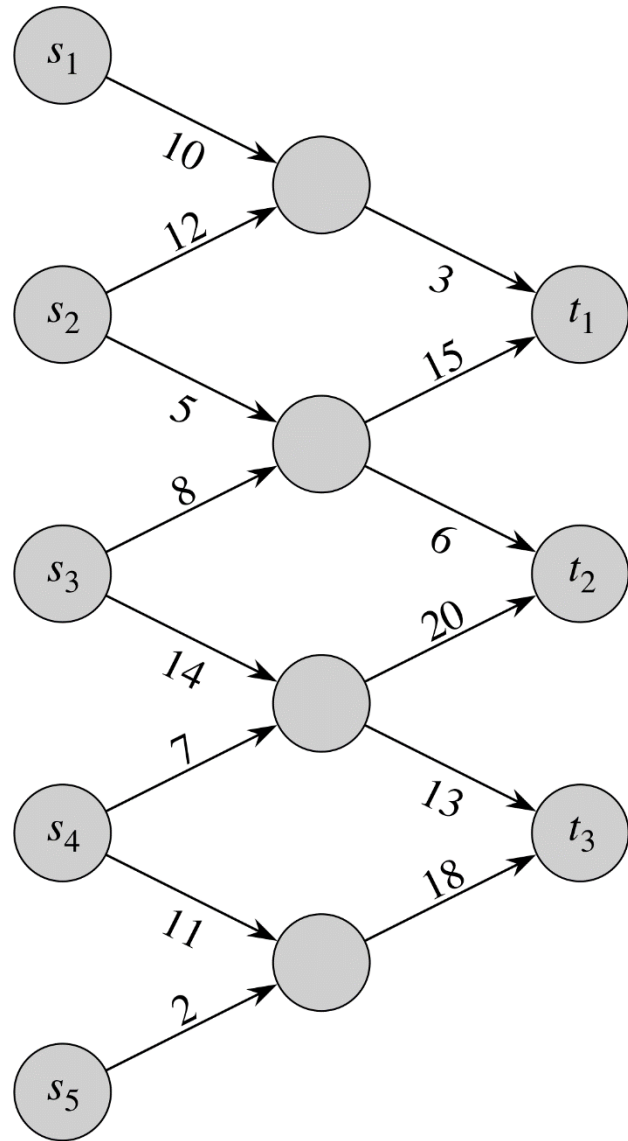


(a)

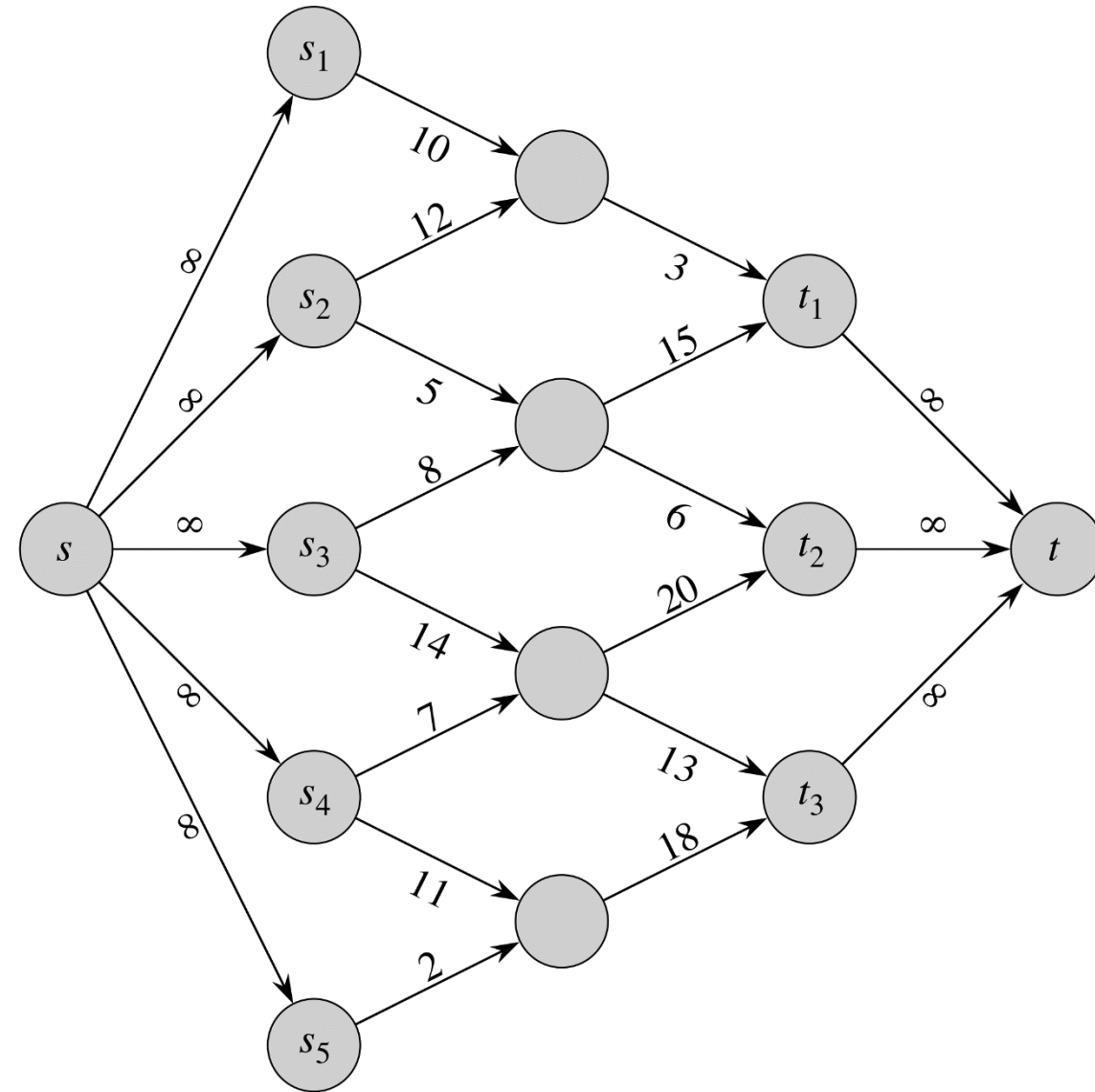


(b)

Modeling with multiple sources/sinks



(a)



(b)

Ford-Fulkerson method

- Method not algorithm because it has several implementations with different complexities
 - Greedy, greedy, greedy!!!
 - Main ideas
 - Residual networks
 - Augmenting paths
 - Cuts
- 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

Residual network

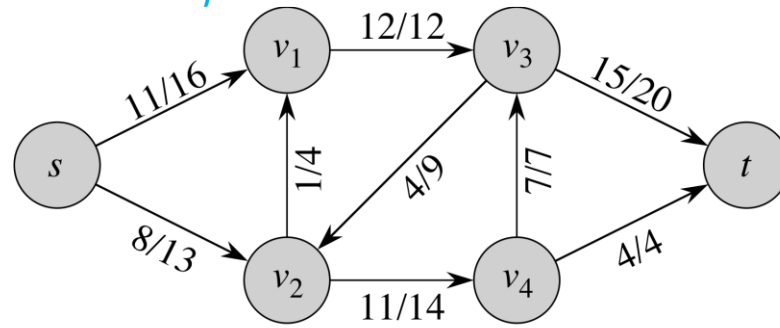
- Residual network G_f contains the residual capacities from G
- It may contain extra edges
 - To allow for decreasing previously-allocated flows
 - Reversed edges for previously-allocated flows
- Kind of similar to a flow network, except it allows for reversed edges
- Now, why there is no reversed edges in flow networks??

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

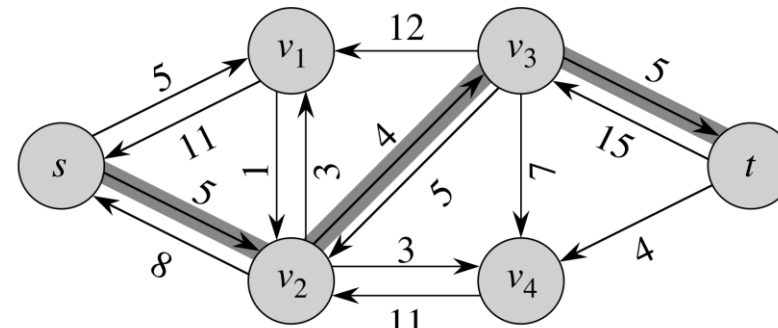
Augmenting paths

- Augmenting path p is a simple path from s to t in the residual network G_f

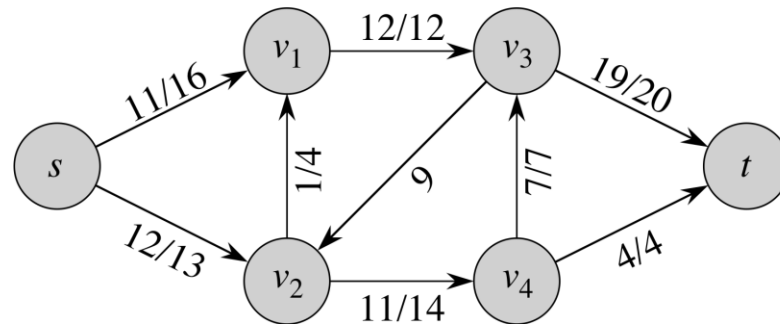
A path is simple if all the vertices on the path are distinct.



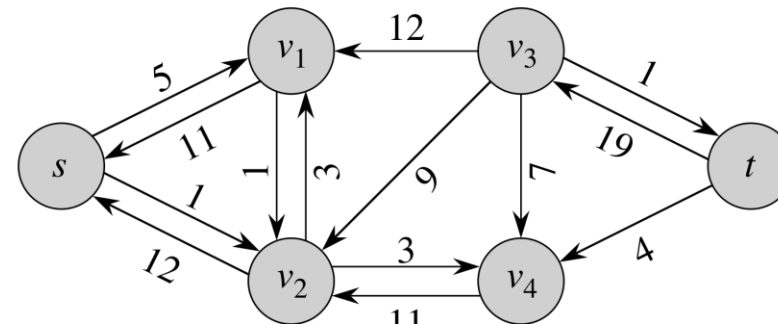
(a)



(b)



(c)



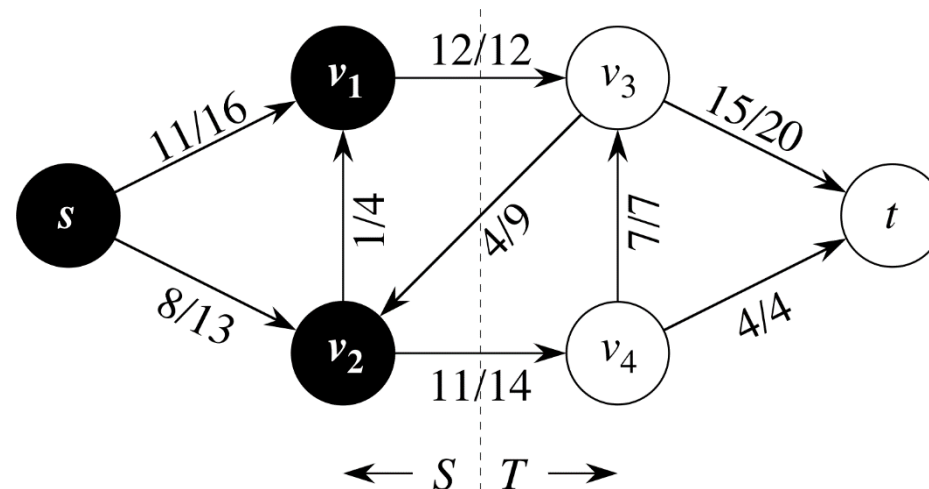
(d)

Figure 26.4 (a) The flow network G and flow f of Figure 26.1(b). (b) The residual network G_f with augmenting path p shaded; its residual capacity is $c_f(p) = c_f(v_2, v_3) = 4$. Edges with residual capacity equal to 0, such as (v_1, v_3) , are not shown, a convention we follow in the remainder of this section. (c) The flow in G that results from augmenting along path p by its residual capacity 4. Edges carrying no flow, such as (v_3, v_2) , are labeled only by their capacity, another convention we follow throughout. (d) The residual network induced by the flow in (c).

Cuts of flow networks

- How to know when the algorithm terminates??
- A cut (S, T) of flow network $G(V, E)$ is a partition of V into S and $T = V - S$
- **Net flow** $f(S, T)$ across the cut (S, T) is defined as

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



Cuts of flow networks

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

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

- A *minimum cut* of a network is a cut whose capacity is minimum over all cuts of the network

Max-flow min-cut

Theorem 26.6 (Max-flow min-cut theorem)

If f is a flow in a flow network $G = (V, E)$ with source s and sink t , then 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 .

Basic 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  $(v, u).f = (v, u).f - c_f(p)$ 
```

- Path-finding is performed using either BFS or DFS, thus $O(V + E) = O(E)$
- The while loop is executed $|f^*|$ (if every iteration just adds on unit value), where f^* is the maximum flow
- Overall complexity is $O(E |f^*|)$

FFA

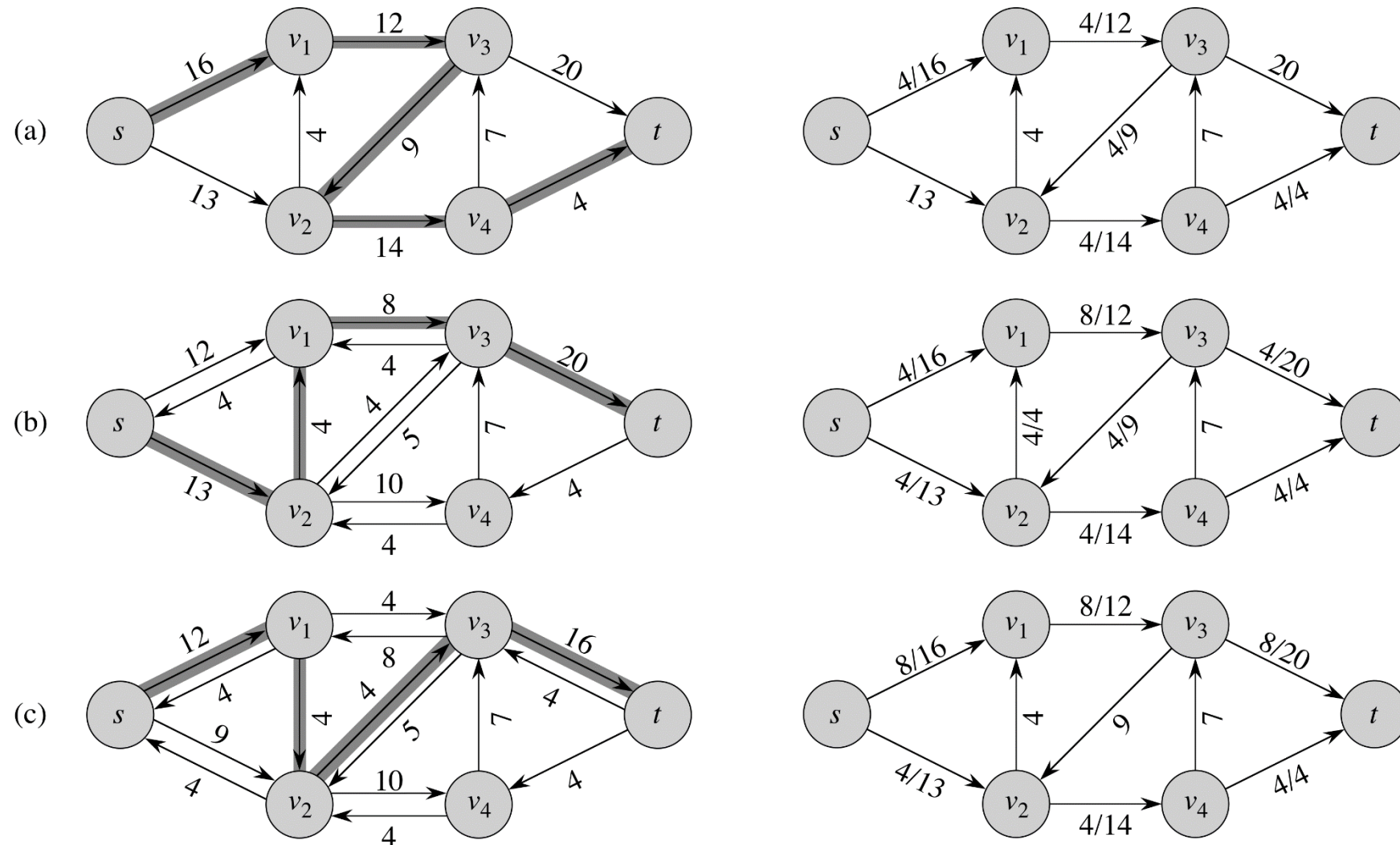
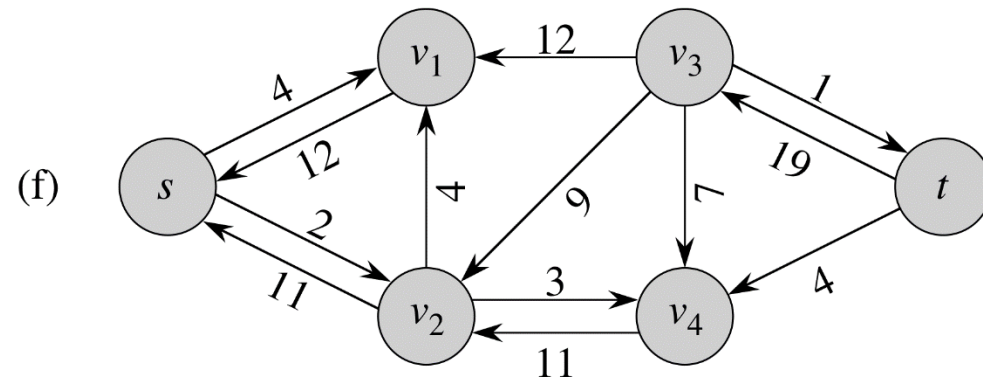
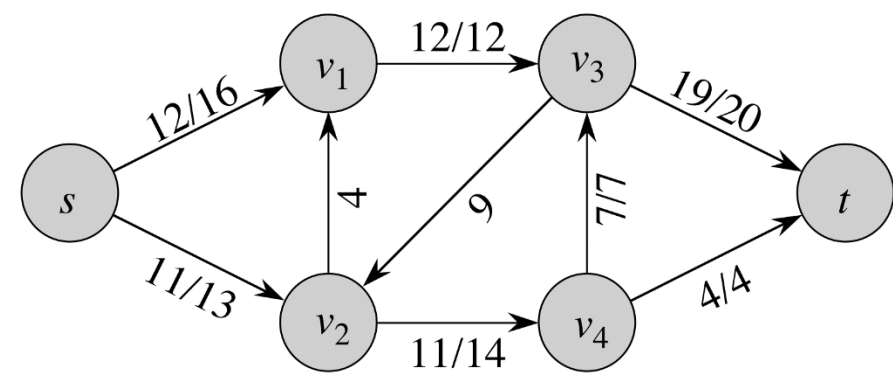
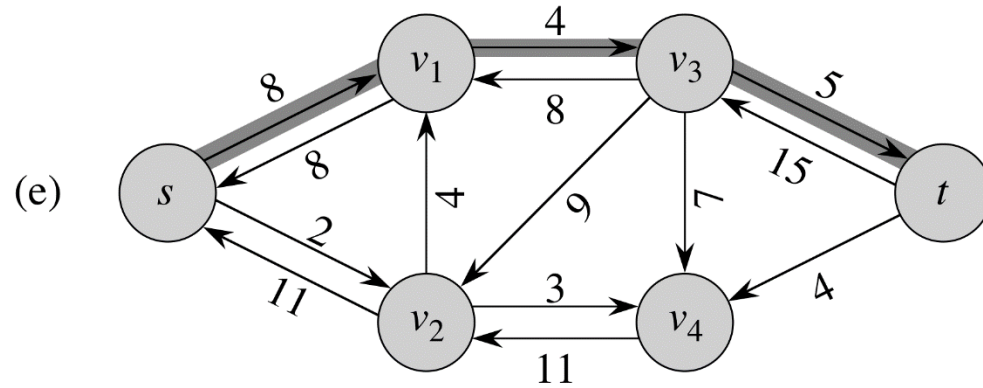
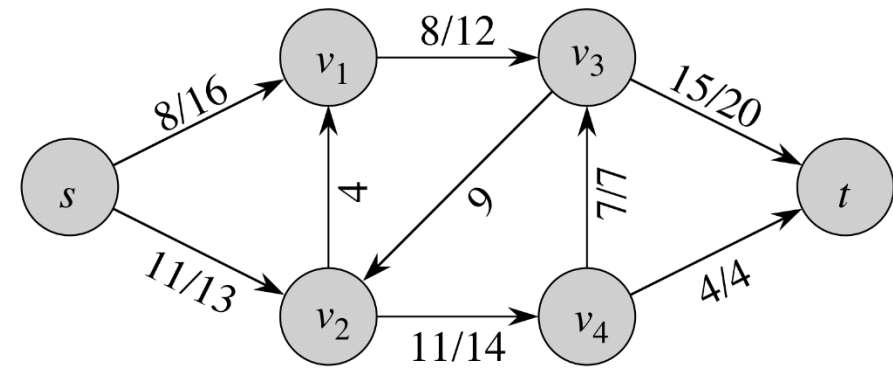
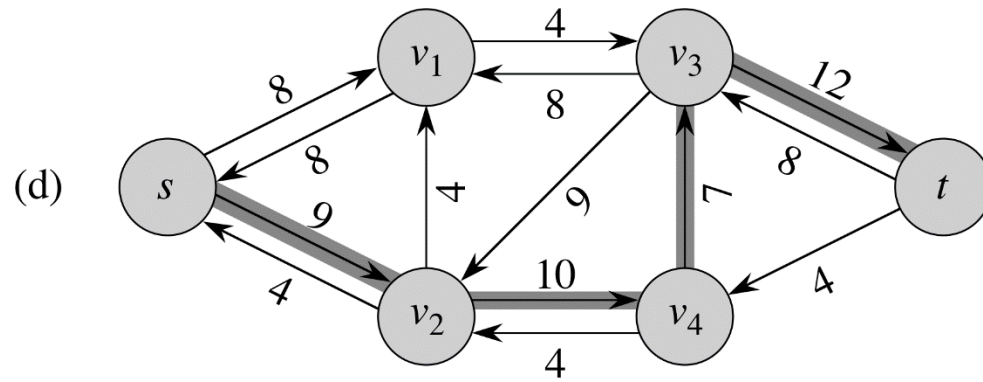


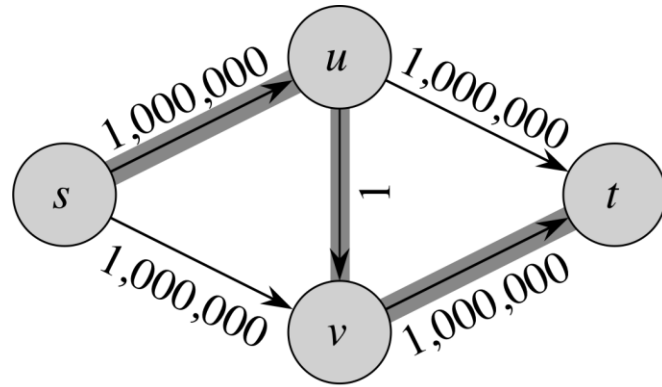
Figure 26.6 The execution of the basic Ford-Fulkerson algorithm. (a)–(e) Successive iterations of the **while** loop. The left side of each part shows the residual network G_f from line 3 with a shaded augmenting path p . The right side of each part shows the new flow f that results from augmenting f by f_p . The residual network in (a) is the input network G .

FFA

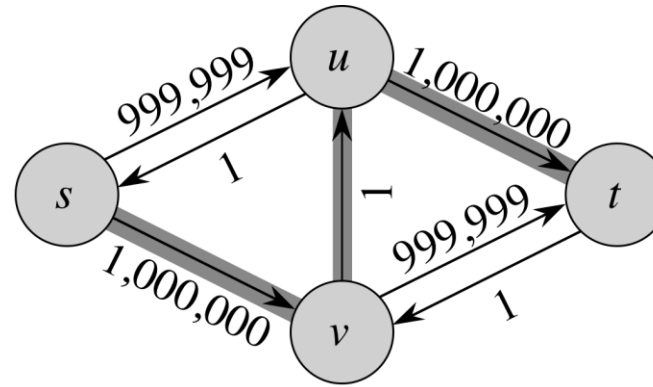


Worst case for FFA

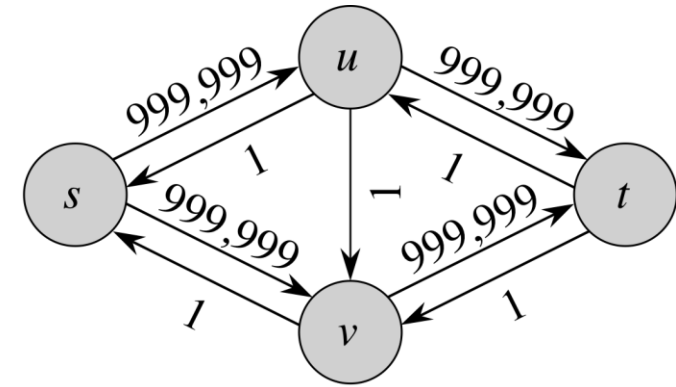
- $|f^*|$ in this case is 2,000,000



(a)



(b)



(c)

Edmonds-Karp algorithm

- Uses BFS for augmenting path (shortest path)
- Total number of flow augmentations is $O(VE)$, thus overall complexity $O(VE^2)$
- How does it perform with the worst case for FFA?