



# Algorithms and Data Structures

## Lecture 22 Flow Network

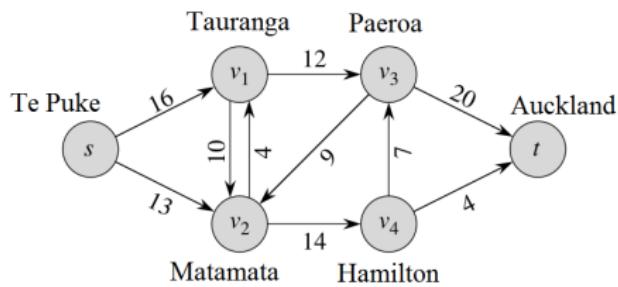
Jiamou Liu  
The University of Auckland

# Lecture 22 Flow Network

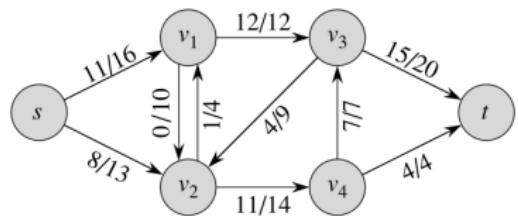
## Part I: Flow Network – Basic Terminology

# A Shipping Problem

A farmer in plans to lease trucks to ship his fruits to Auckland. Due to road and traffic conditions, the trucks can ship at most  $c(u, v)$  cartons per day between each pair of towns  $u$  and  $v$ . What is the maximum amount of fruits that he can ship per day?

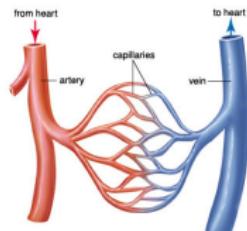
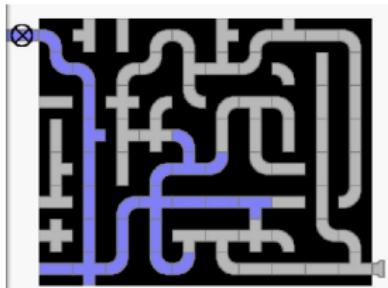


A possible shipping arrangement



# Flow Network

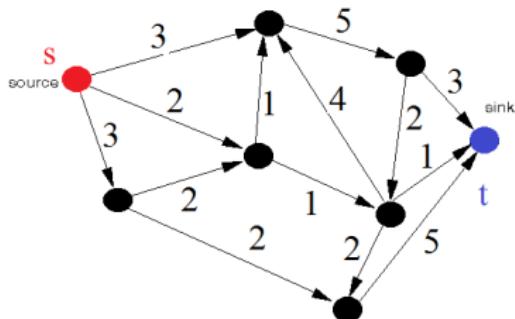
- Liquid (pipeline/blood vessel) networks
- Electricity networks
- Traffic networks
- Evacuation networks
- Telecommunication networks



# Flow Network

## Definition [Flow Network]

- A **source node** in a directed graph is a node with no incoming edges.
- A **sink node** in a directed graph is a node with no outgoing edges.
- A **flow network** is a directed graph  $(V, E, c, s, t)$  where  $(V, E, c)$  forms a weighted graph with positive edge weights,  $s \in V$  is a source node and  $t \in V$  is a sink node.

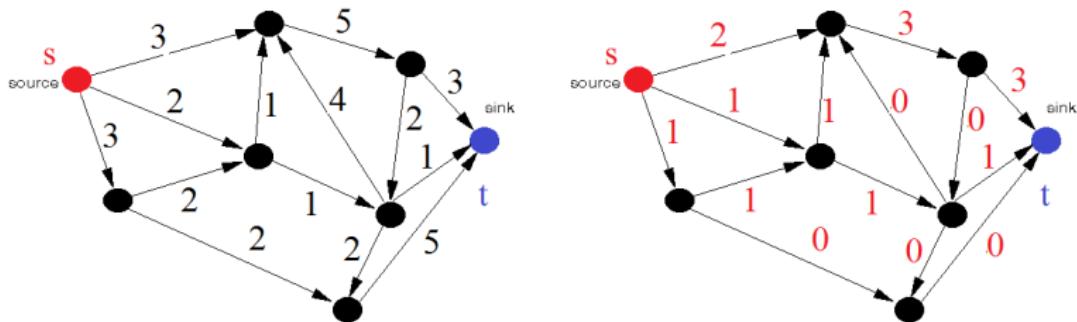


## Definition [Flow]

A **flow** in a flow network  $(V, E, c, s, t)$  is a function  $f : E \rightarrow \mathbb{N}$  such that  $f(u, v) \leq c(u, v)$  for every  $(u, v) \in E$  and

for any node  $u$  that is not  $s$  nor  $t$ , we have  $\sum_{(u,x) \in E} f(u, x) = \sum_{(y,u) \in E} f(y, u)$ .

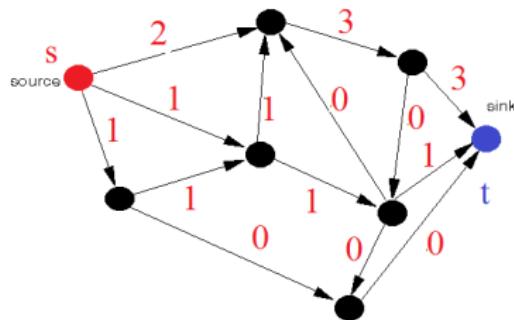
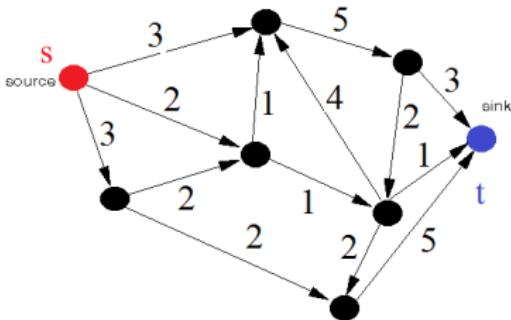
Intuitively, a flow defines a way to send oil from  $s$  to  $t$  without exceeding the pipeline capacity, and without any leak on the way.



## Definition [Size of a Flow]

The **size** of a flow  $f$  in a flow network is  $\text{size}(f) = \sum\{f(s, u) \mid (s, u) \in E\}$ . In other words, it is the total quantity of stuff sent from  $s$  to  $t$ .

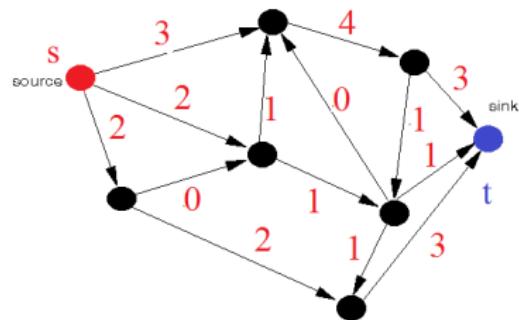
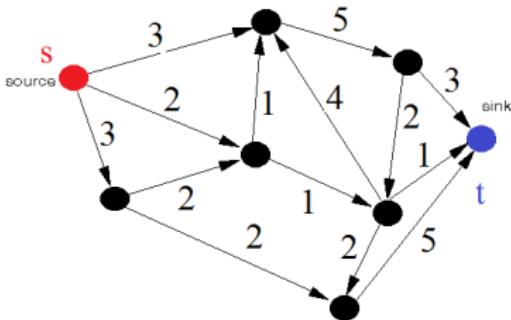
**Example:** Size of the following flow = 4



## Definition [Size of a Flow]

The **size** of a flow  $f$  in a flow network is  $\text{size}(f) = \sum\{f(s, u) \mid (s, u) \in E\}$ . In other words, it is the total quantity of stuff sent from  $s$  to  $t$ .

**Example:** Size of the following flow = 7

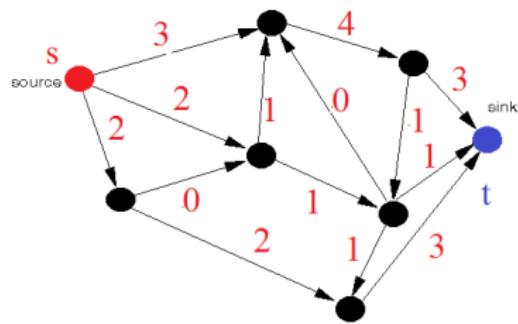
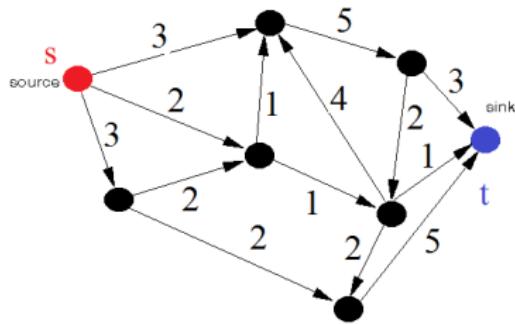


# Maximal Flows in a Flow Network

## Definition [Maximal Flow]

A flow in a flow network is **maximal** if it has maximal size.

**Example:** the following is a maximal flow.

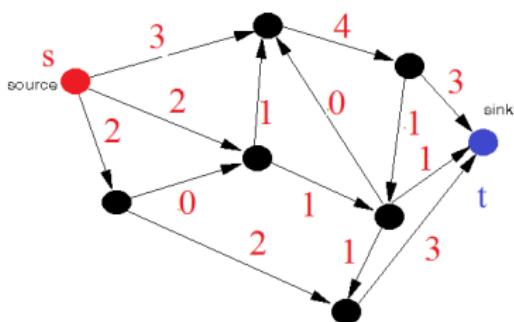
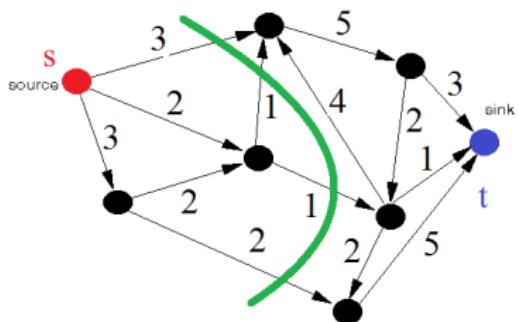


# Cuts in a Flow Network

## Definition [Cut]

A **cut** in a graph is a set  $C$  of edges such that removing them would disconnect the graph into  $\text{left}(C)$  (nodes reachable from  $s$ ) and the right part  $\text{right}(C)$  (nodes that may reach  $t$ ).

The green line shows a cut:



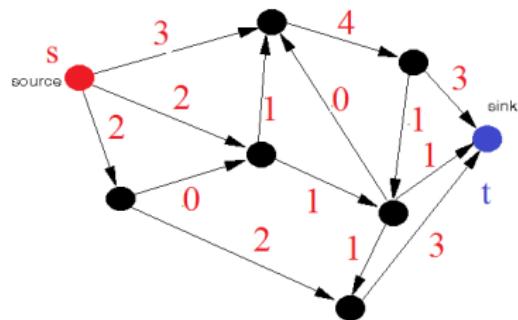
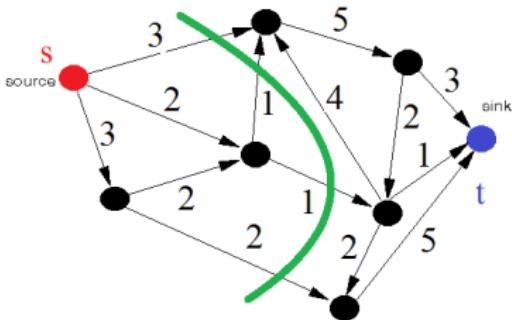
## Definition [Capacity of a Cut]

The **capacity** of a cut  $C$  in a flow network is the sum of the capacity of edges that goes from left to right:

$$\text{capacity}(C) = \sum \{c(u, v) \mid (u, v) \in E, u \in \text{left}(C), v \in \text{right}(C)\}$$

A **minimal cut** is a cut with minimal capacity.

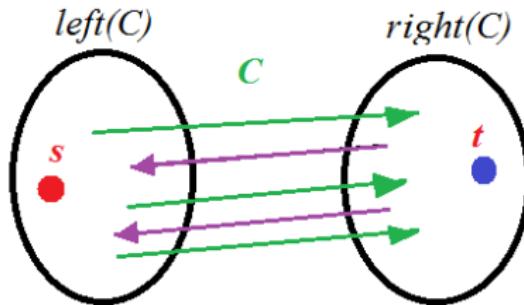
For example, the green line shows a minimal cut of capacity 7.



## Cuts in a Flow

Let  $C$  be a cut in a network  $G$  and  $f$  be a flow. The **net flow** of  $f$  over  $C$

$$f(C) = \sum \{f(u, v) \mid (u, v) \in E, u \in \text{left}(C), v \in \text{right}(C)\} - \sum \{f(v, u) \mid (v, u) \in E, u \in \text{left}(C), v \in \text{right}(C)\}$$

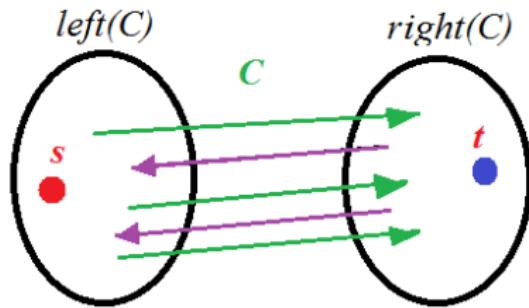


# Flow and Cut

## Fact 1

For any cut  $C$  and flow  $f$ ,

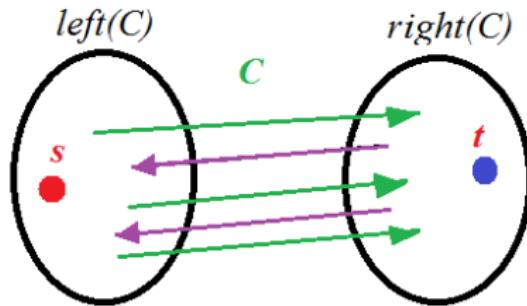
$$\text{size}(f) = f(C)$$



## Fact 2

For any flow network G,

the size of max-flow  $\leq$  the capacity of min-cut



# Two Problems

## Max-Flow Problem

Given a flow network, find a maximal flow in the network?

In other words, **what is the maximal quantity** we can send from  $s$  to  $t$ ?

## Min-Cut Problem

Given a flow network, find a minimal cut in the network?

In other words, **what is the bottleneck** across the entire network?

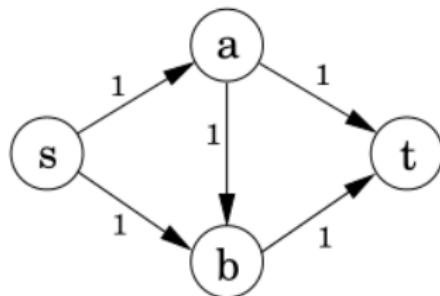
Later we will show that these two are the same problem.

# Lecture 22 Flow Network

Part II: Max-Flow Min-Cut

# A Network Example

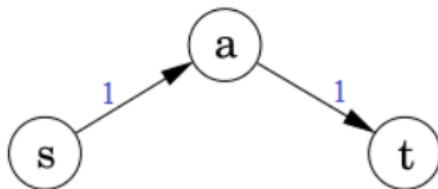
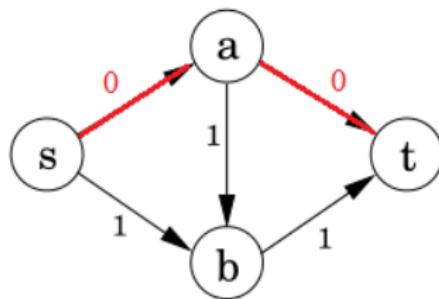
Example:



- ① Start with zero flow
- ② Repeat: Choose a path from  $s$  to  $t$  and maximize flow along this path.

# A Network Example

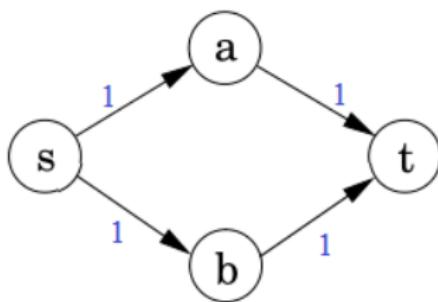
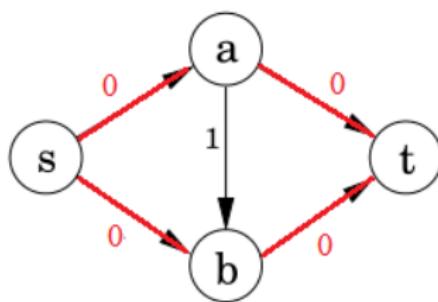
Find a path from  $s$  to  $t$  with positive capacity:



- ① Start with zero flow
- ② Repeat: Choose a path from  $s$  to  $t$  and maximize flow along this path.

# A Network Example

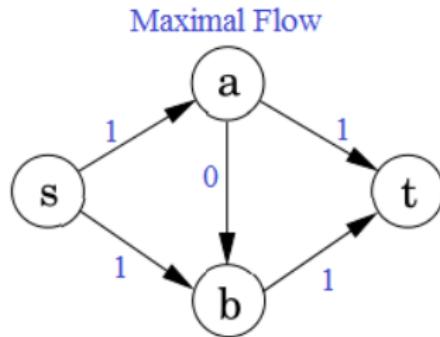
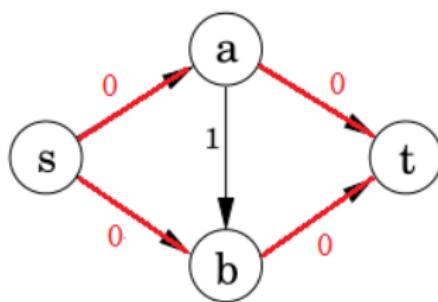
Find a remaining path from  $s$  to  $t$  with positive capacity:



- ① Start with zero flow
- ② Repeat: Choose a path from  $s$  to  $t$  and maximize flow along this path.

# A Network Example

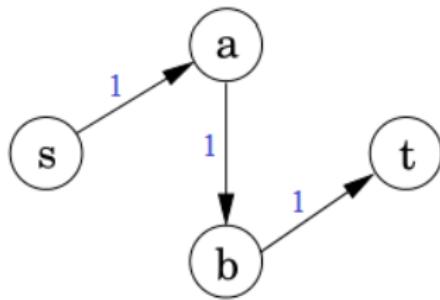
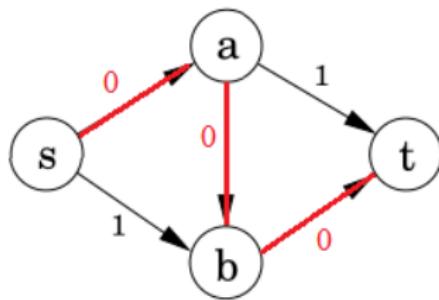
The final flow



- ① Start with zero flow
- ② Repeat: Choose a path from  $s$  to  $t$  and maximize flow along this path.

# A Network Example

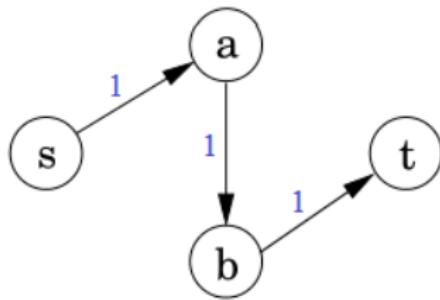
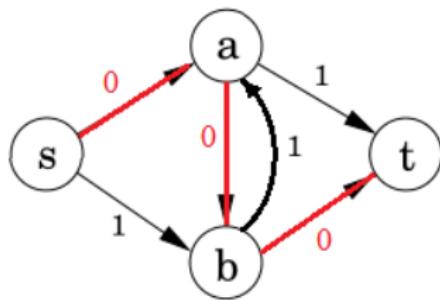
Potential Problem:



The network on the **left** represents the amount of additional flow that could be pushed along the edges.  
Formally, this is captured using **residual networks**.

# A Network Example

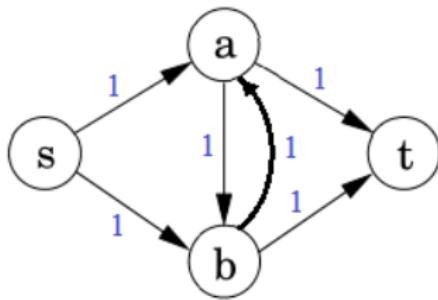
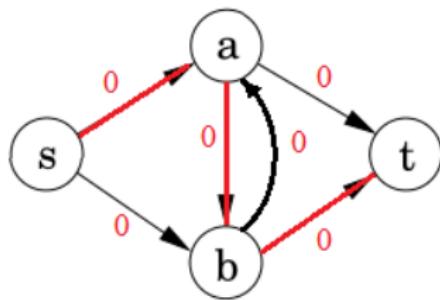
Add a reversed edge for cancellation:



The network on the **left** represents the amount of additional flow that could be pushed along the edges.  
Formally, this is captured using **residual networks**.

# A Network Example

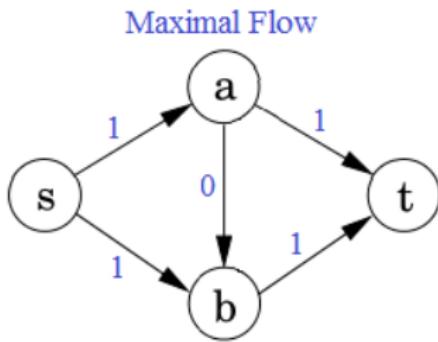
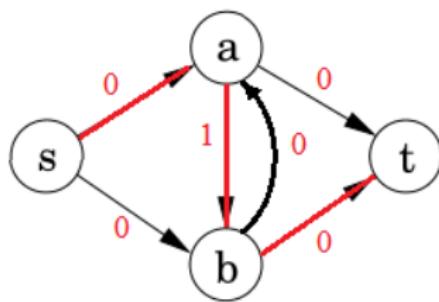
Find a path with reversed edge:



The network on the **left** represents the amount of additional flow that could be pushed along the edges.  
Formally, this is captured using **residual networks**.

# A Network Example

Cancel the two edges:



The network on the **left** represents the amount of additional flow that could be pushed along the edges.  
Formally, this is captured using **residual networks**.

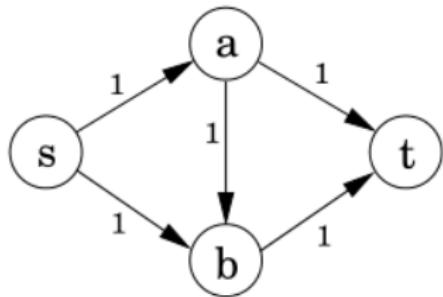
# Residual Network

## Residual Network

Let  $G = (V, E, c, s, t)$  be a flow network and  $f$  be a flow in  $G$ . The **residual network**  $G_f$  of  $G$  and  $F$  is another flow network  $(V, E_f, c_f, s, t)$ :

- (a) if  $(u, v) \in E$ , and  $f(u, v) < c(u, v)$ , put  $(u, v) \in E_f$  and set  $c_f(u, v) = c(u, v) - f(u, v)$ .
- (b) if  $f(u, v) > 0$ , put  $(v, u) \in E_f$  and set  $c_f(v, u) = f(u, v)$ .

Flow Network  $G$



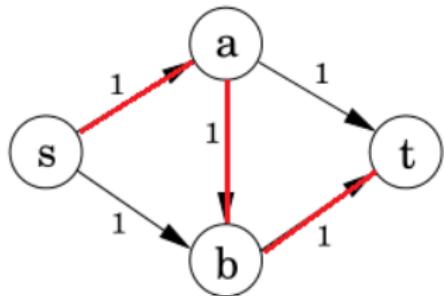
# Residual Network

## Residual Network

Let  $G = (V, E, c, s, t)$  be a flow network and  $f$  be a flow in  $G$ . The **residual network**  $G_f$  of  $G$  and  $F$  is another flow network  $(V, E_f, c_f, s, t)$ :

- (a) if  $(u, v) \in E$ , and  $f(u, v) < c(u, v)$ , put  $(u, v) \in E_f$  and set  $c_f(u, v) = c(u, v) - f(u, v)$ .
- (b) if  $f(u, v) > 0$ , put  $(v, u) \in E_f$  and set  $c_f(v, u) = f(u, v)$ .

Flow Network  $G$

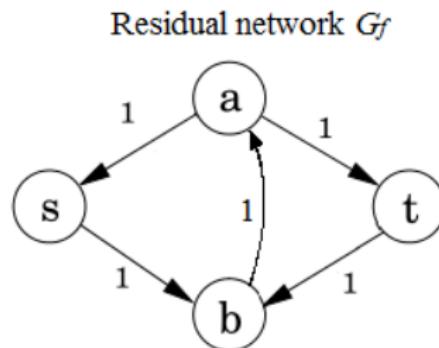
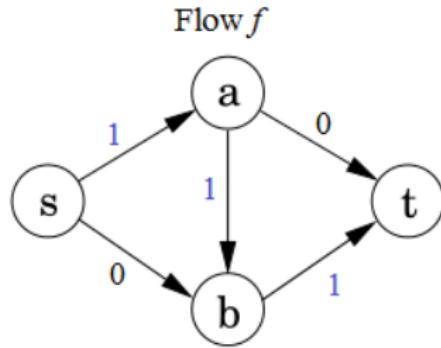


# Residual Network

## Residual Network

Let  $G = (V, E, c, s, t)$  be a flow network and  $f$  be a flow in  $G$ . The **residual network**  $G_f$  of  $G$  and  $F$  is another flow network  $(V, E_f, c_f, s, t)$ :

- if  $(u, v) \in E$ , and  $f(u, v) < c(u, v)$ , put  $(u, v) \in E_f$  and set  $c_f(u, v) = c(u, v) - f(u, v)$ .
- if  $f(u, v) > 0$ , put  $(v, u) \in E_f$  and set  $c_f(v, u) = f(u, v)$ .

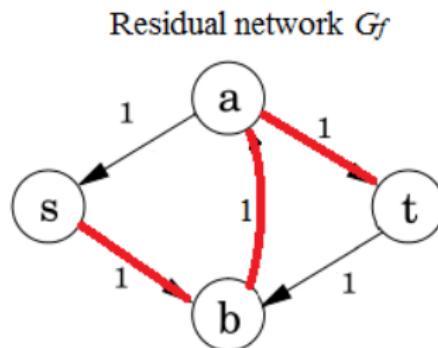
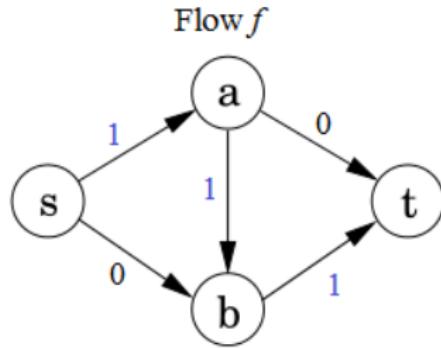


# Residual Network

## Residual Network

Let  $G = (V, E, c, s, t)$  be a flow network and  $f$  be a flow in  $G$ . The **residual network**  $G_f$  of  $G$  and  $F$  is another flow network  $(V, E_f, c_f, s, t)$ :

- if  $(u, v) \in E$ , and  $f(u, v) < c(u, v)$ , put  $(u, v) \in E_f$  and set  $c_f(u, v) = c(u, v) - f(u, v)$ .
- if  $f(u, v) > 0$ , put  $(v, u) \in E_f$  and set  $c_f(v, u) = f(u, v)$ .



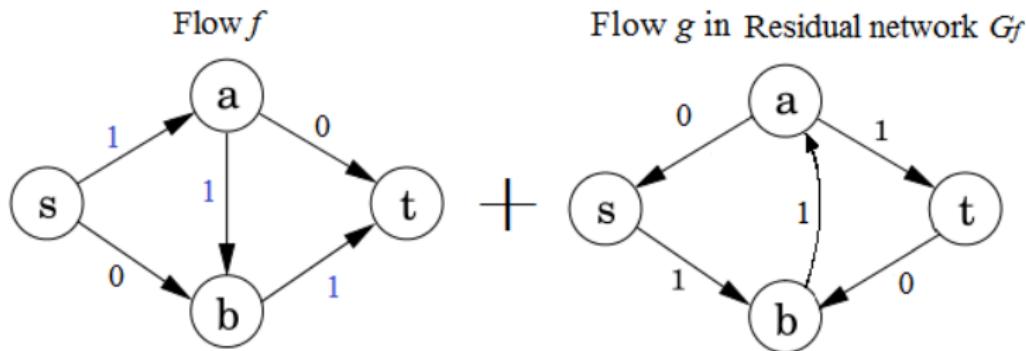
# Combining Flows

## Combining Flows

Let  $f$  be a flow in the network  $G$ . Let  $g$  be a flow in the residual network  $G_f$ . Then we can obtain another flow  $h$  in  $G$  by **adding**  $f$  and  $g$ : For all  $(u, v) \in E$ , let

$$h(u, v) = f(u, v) + g(u, v) - g(v, u)$$

**Note:** The flow  $h$  is larger than  $f$ .



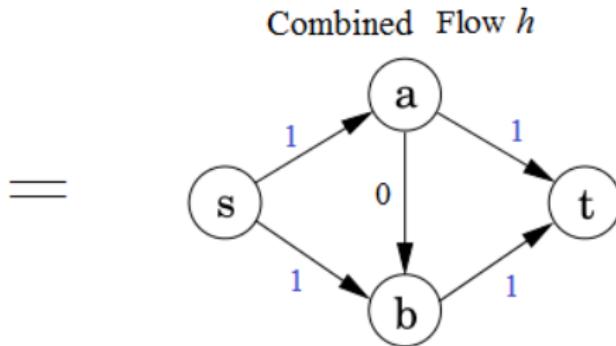
# Combining Flows

## Combining Flows

Let  $f$  be a flow in the network  $G$ . Let  $g$  be a flow in the residual network  $G_f$ . Then we can obtain another flow  $h$  in  $G$  by **adding**  $f$  and  $g$ : For all  $(u, v) \in E$ , let

$$h(u, v) = f(u, v) + g(u, v) - g(v, u)$$

**Note:** The flow  $h$  is larger than  $f$ .



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

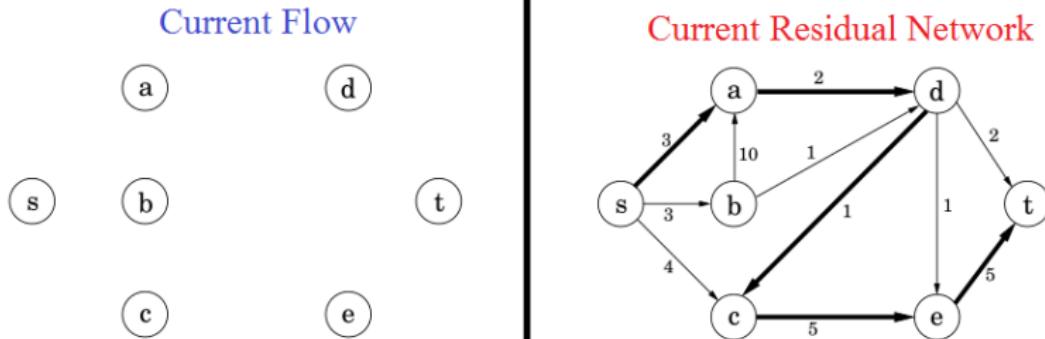
- Start with empty flow, and **full** residual network.

- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

- Stop when we cannot find any single-path flow

Iteration 0:



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

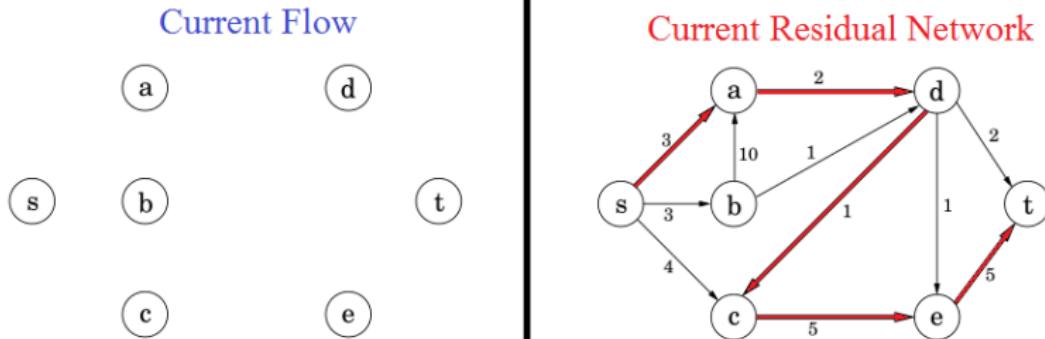
**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

- Stop when we cannot find any single-path flow

Iteration 1: Pick



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

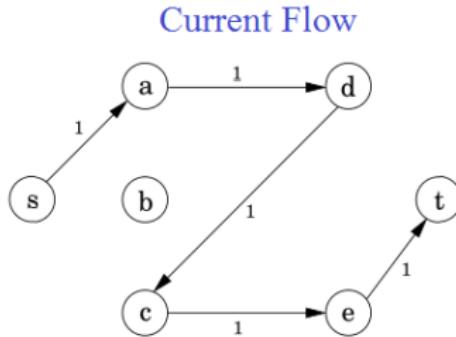
- Start with empty flow, and **full** residual network.

- Repeat:

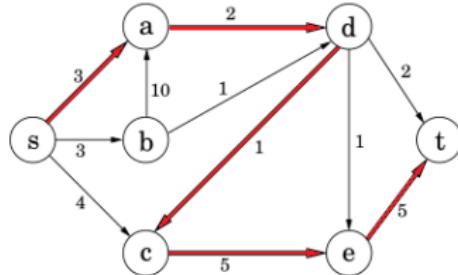
- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

- Stop when we cannot find any single-path flow

### Iteration 1: Combine



### Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

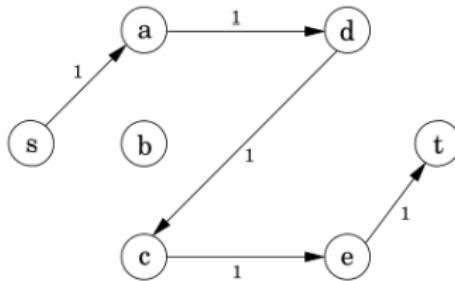
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

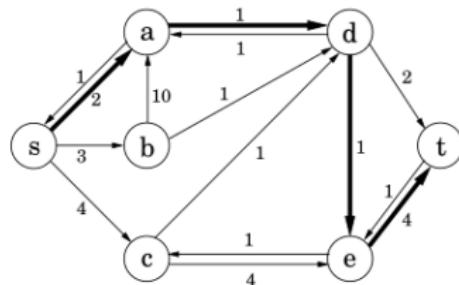
- Stop when we cannot find any single-path flow

Iteration 1: Update

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

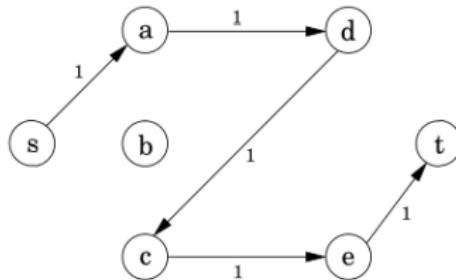
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

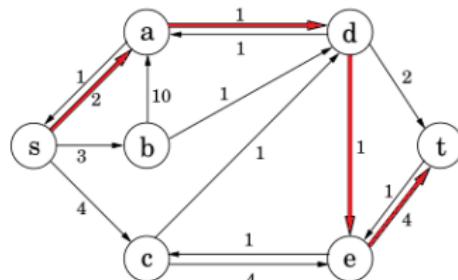
- Stop when we cannot find any single-path flow

Iteration 2: Pick

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

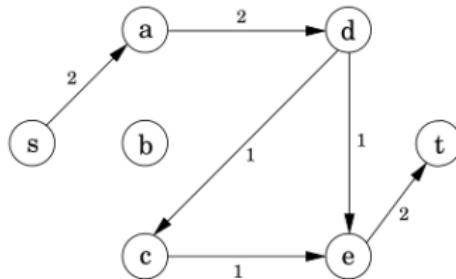
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

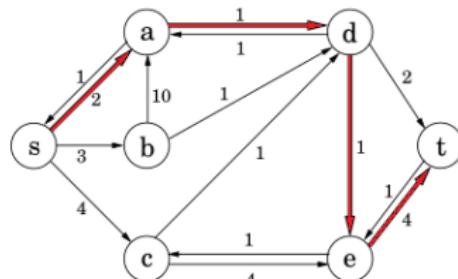
- Stop when we cannot find any single-path flow

### Iteration 2: Combine

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

- Repeat:

- 1) **Pick** a single-path flow from the residual network.

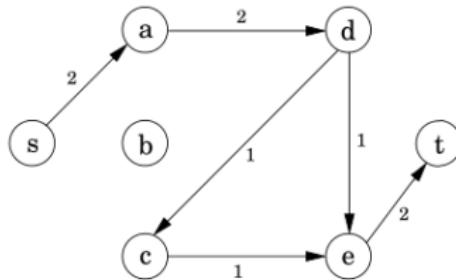
- 2) **Combine** this flow with the current flow

- 3) **Update** the residual network

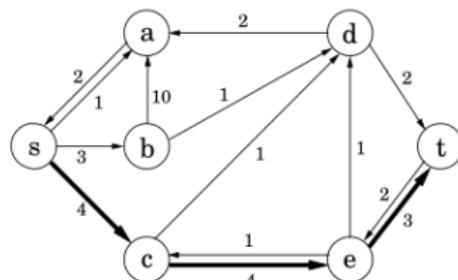
- Stop when we cannot find any single-path flow

### Iteration 2: Update

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

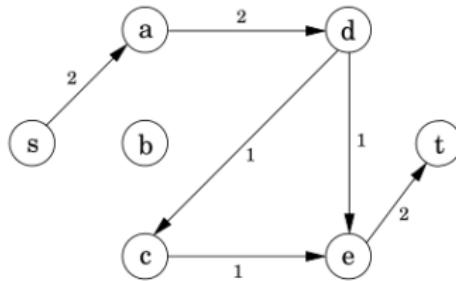
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

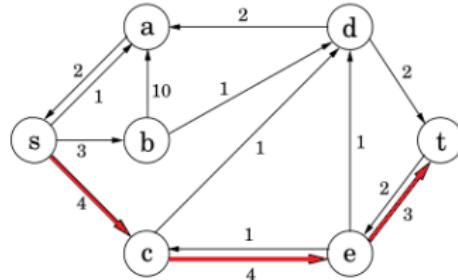
- Stop when we cannot find any single-path flow

Iteration 3: Pick

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

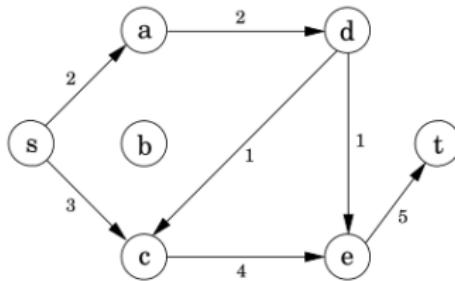
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

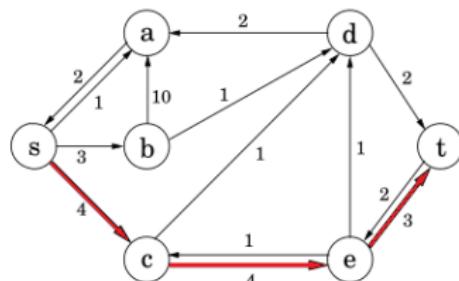
- Stop when we cannot find any single-path flow

### Iteration 3: Combine

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

- Repeat:

- 1) **Pick** a single-path flow from the residual network.

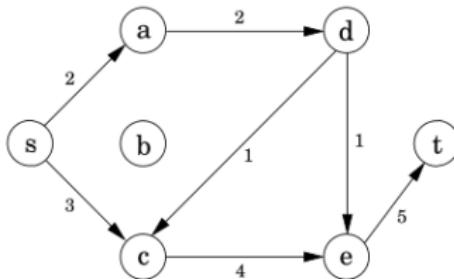
- 2) **Combine** this flow with the current flow

- 3) **Update** the residual network

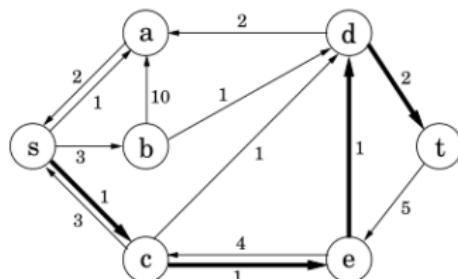
- Stop when we cannot find any single-path flow

### Iteration 3: Update

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

- Repeat:

- 1) **Pick** a single-path flow from the residual network.

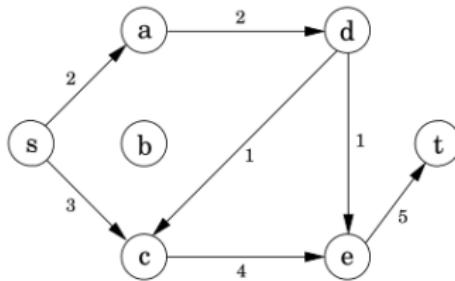
- 2) **Combine** this flow with the current flow

- 3) **Update** the residual network

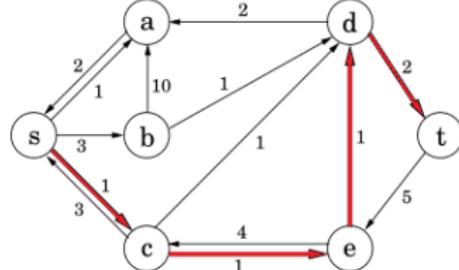
- Stop when we cannot find any single-path flow

Iteration 4: Pick

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

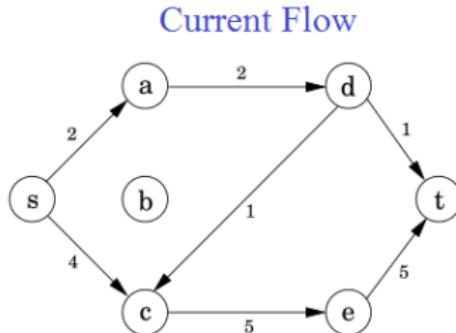
- Start with empty flow, and **full** residual network.

- Repeat:

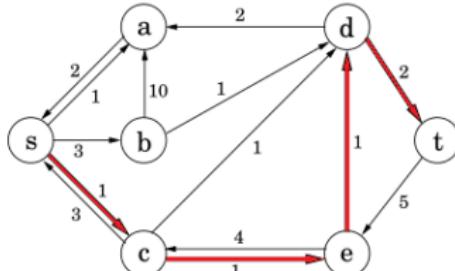
- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

- Stop when we cannot find any single-path flow

### Iteration 4: Combine



### Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

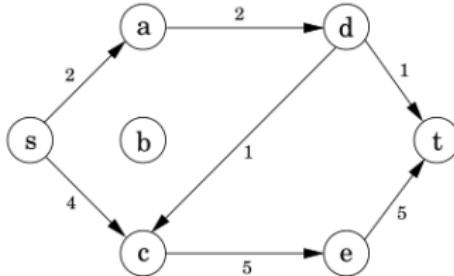
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

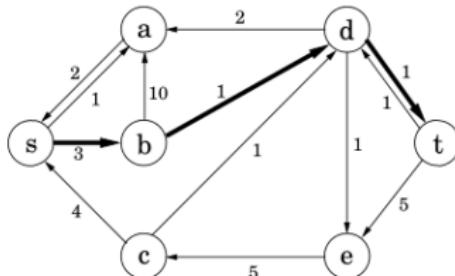
- Stop when we cannot find any single-path flow

### Iteration 4: Update

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

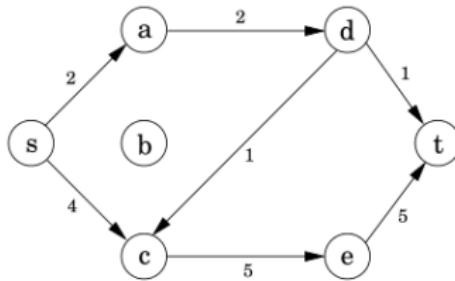
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

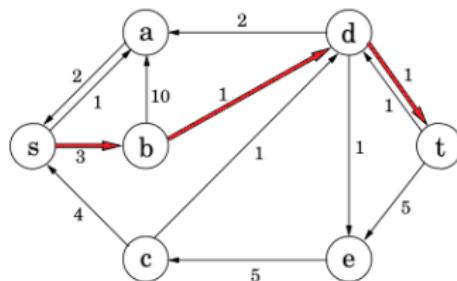
- Stop when we cannot find any single-path flow

Iteration 5: Pick

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

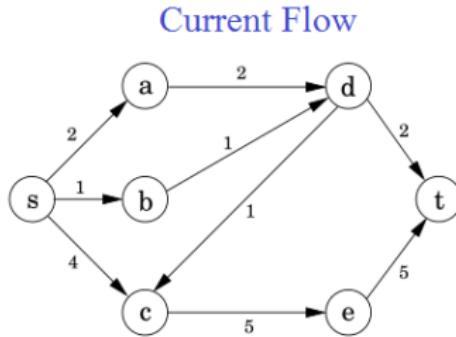
- Start with empty flow, and **full** residual network.

- Repeat:

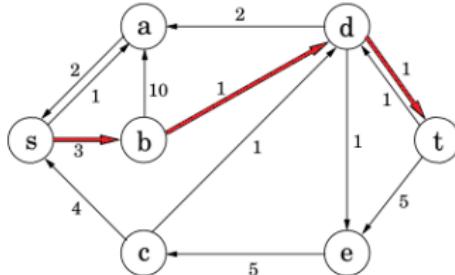
- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

- Stop when we cannot find any single-path flow

### Iteration 5: Combine



### Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

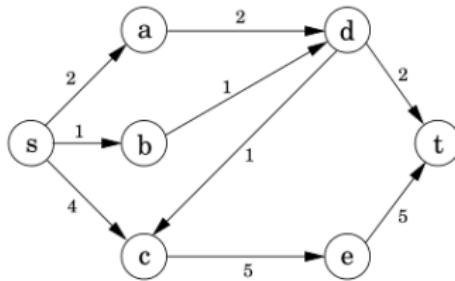
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

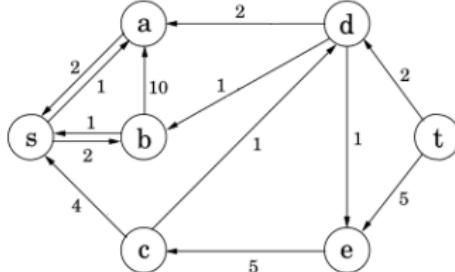
- Stop when we cannot find any single-path flow

### Iteration 5: Update

Current Flow



Current Residual Network



# Solving Max-Flow Problem

## Ford-Fulkerson Algorithm

**INPUT:** A flow network  $G = (V, E, c, s, t)$

**OUTPUT:** A maximal flow in  $G$

- Start with empty flow, and **full** residual network.

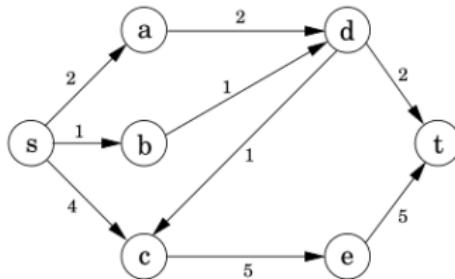
- Repeat:

- 1) **Pick** a single-path flow from the residual network.
- 2) **Combine** this flow with the current flow
- 3) **Update** the residual network

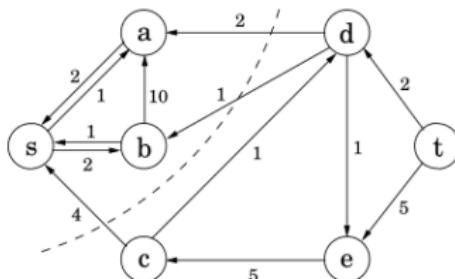
- Stop when we cannot find any single-path flow

Stop p

Current Flow



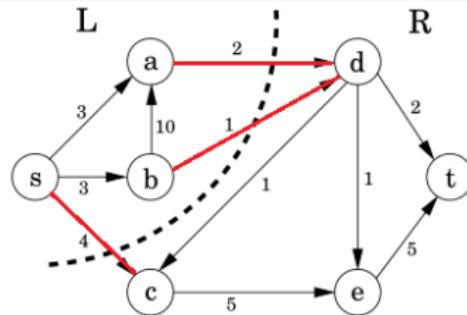
Current Residual Network



# Ford-Fulkerson Algorithm: Correctness

Why does Ford-Fulkerson algorithm find a maximal flow?

Because this flow actually matched the capacity of a cut!



## Fact 3

Let  $f$  be the flow produced by Ford-Fulkerson. Consider the residual network  $G_f$ . Let  $L$  be the nodes that are reachable from  $s$  in  $G_f$ . Let  $R = V \setminus L$ . Let  $C$  be the cut between  $L$  and  $R$ . Then

$$\text{size}(f) = \text{capacity}(C)$$

# Max-Flow, Min-Cut

## Max-Flow Min-Cut Theorem

In any flow network, the size of the maximal flow equals the capacity of the minimal cut.

## Furthermore

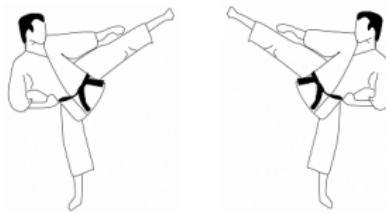
Ford-Fulkerson algorithm finds a maximal flow, as well as a minimal cut in a network.

# Lecture 7 Flow Network

## Part III: Further Applications

# Application 1: Community Detection

**Zachary's Karate Club**<sup>1</sup> A social network of a karate club was studied by Wayne W. Zachary, with 34 members and 78 links between members who interacted outside the club. During the study a conflict arose between the administrator "John A" and instructor "Mr. Hi", which led to the split of the club into two. Half of the members formed a new club around Mr. Hi, members from the other part found a new instructor or gave up karate. Basing on collected data Zachary assigned correctly all but one member of the club to the groups they actually joined after the split.



---

<sup>1</sup>"An Information Flow Model for Conflict and Fission in Small Groups", Wayne W. Zachary (1977)

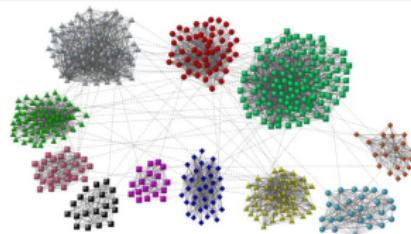
# Application 1: Community Detection

**Question:** How did Zachary come up with the partition?

## Community Detection Problem

Partition a given network into subgraphs that capture “realistic communities” in real-life networks:

- Circles of friends
- Political bloggers with similar opinions
- Consumers with similar purchasing behaviours
- Functional groups of proteins (in protein-protein network)

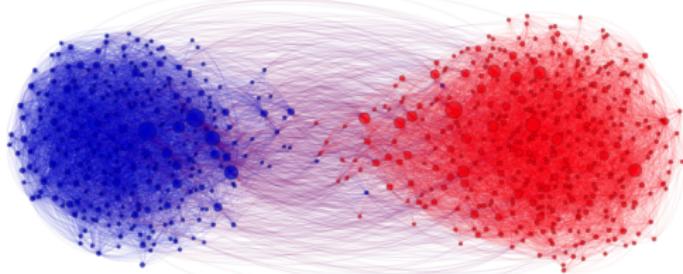


# Application 1: Community Detection

## Community Detection using Minimum Cut Method

**Polarisation:** Divide the network into two parts, between which there exist the smallest number of links.

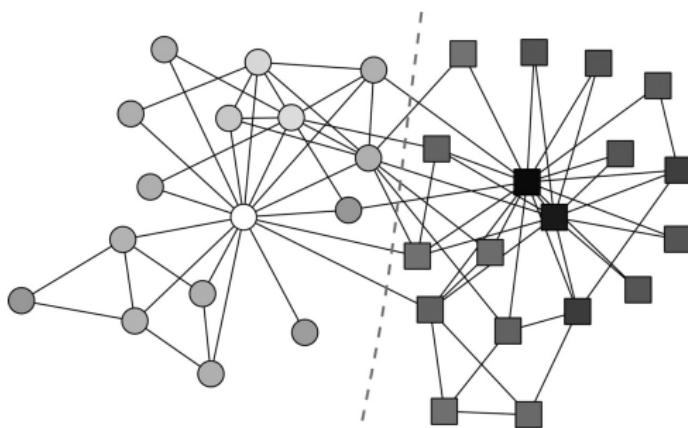
Useful in analysing e.g. political campaign



**Note:** This is essentially the minimal cut problem.

# Application 1: Community Detection

- Ford-Fulkerson's algorithm is the earliest algorithm for community detection
- Zachary run Ford-Fulkerson's algorithm on the social interaction graph of the karate club, with  $s, t$  being the administrator and the instructor, respectively, to obtain the following partition:



# Application 2: Bipartite Matching

## A Problem of Match Making

There are  $m$  male and  $n$  female users in an online dating website. By analysing their profiles, we represent possible matchings as a bipartite graph. Now it needs to find a way for matching men with women so that there are as many matched users as possible.



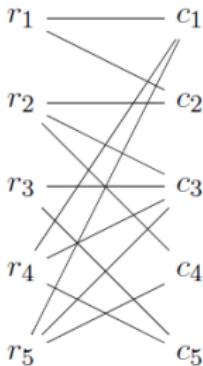
# Application 2: Bipartite Matching

## Bipartite Matching

Let  $G = (X \cup Y, E)$  be a bipartite graph. A **matching** is a set  $M \subseteq E$  of edges that are **independent**, that is, they do not share any nodes in common.

A node  $u \in V = X \cup Y$  is called **matched** if there is  $v$  such that  $\{u, v\} \in E$ .

Example.



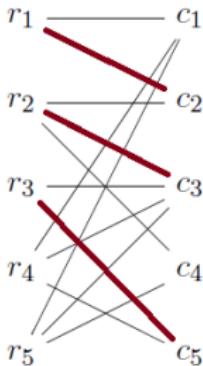
# Application 2: Bipartite Matching

## Bipartite Matching

Let  $G = (X \cup Y, E)$  be a bipartite graph. A **matching** is a set  $M \subseteq E$  of edges that are **independent**, that is, they do not share any nodes in common.

A node  $u \in V = X \cup Y$  is called **matched** if there is  $v$  such that  $\{u, v\} \in E$ .

Example.



# Application 2: Bipartite Matching

## Bipartite Matching

A matching  $M$  of  $G$  is **maximal** if it is not a proper subset of another matching of  $G$ .

A matching of  $G$  is **perfect** if every node in  $G$  is matched.

**Maximal Bipartite Matching Problem:** Given a bipartite graph, compute a maximal matching

Applications of Matching:

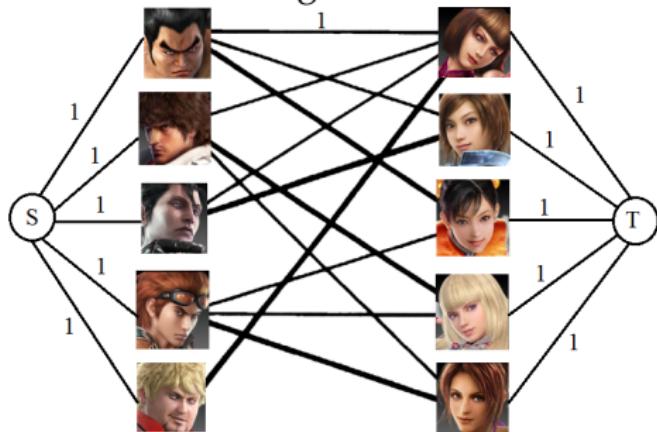
- Consumer v.s. products
- People with required skill set v.s. Tasks

# Application 2: Bipartite Matching

We can reduce the bipartite matching problem to maximal flow problem: Given a bipartite graph  $(X \cup Y, E)$

- ① Add a source node  $s$  and a sink node  $t$
- ② Add edges between  $s$  and all nodes in  $X$
- ③ Add edges between  $t$  and all nodes in  $Y$
- ④ Set capacity of each edge to 1

Then a maximal flow in the resulting network corresponds to a maximal matching.



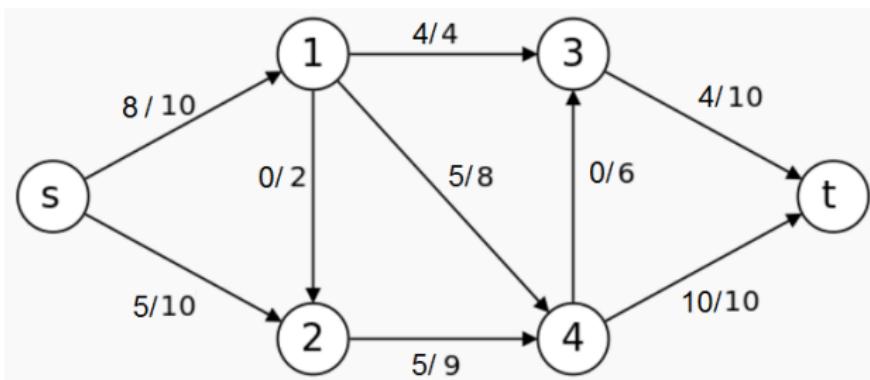
# Summary

The following are the topics covered in this lecture:

- Maximal flow  $\equiv$  Minimal cut
- Ford and Fulkerson algorithm
- Application 1: Community detection (polarisation)
- Application 2: Bipartite matching

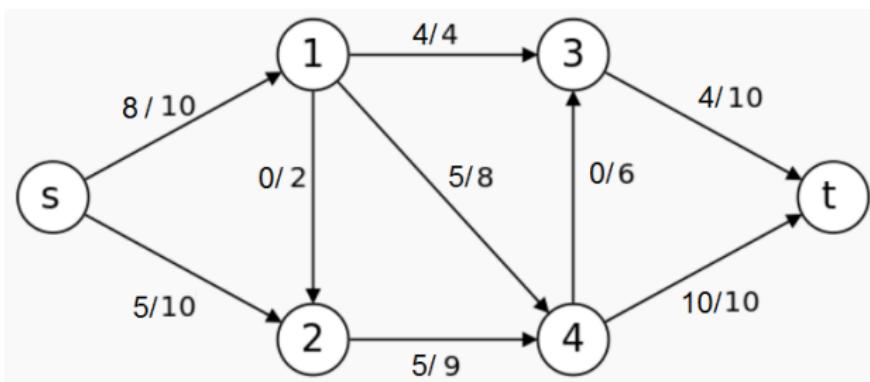
# Exercise

**Question 1.** Consider the following flow in the flow network. Write down its residual network.



# Exercise

**Question 2.** Identify a path flow in the residual network you found for the question above, and combine it with the flow given.



# Exercise

**Question 3.** Consider the flow network below. Apply Ford-Fulkerson's algorithm to find a maximal flow and also identify a minimal cut in the network.

