




# Intro to Flow Problems

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## WHAT YOU SHOULD EXPECT FROM THIS LECTURE

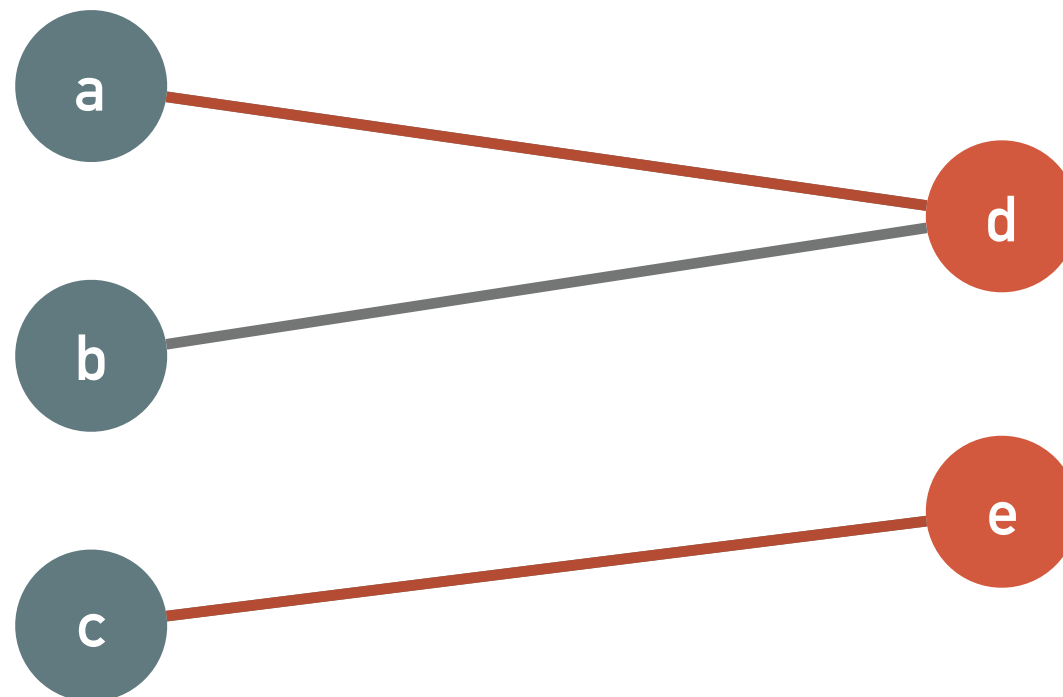
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- First of all, know that this is an advanced topic and don't get frustrated if you can't understand it immediately.
- Know Bipartite Matching
- Know what a Flow Network is
- Know how to model a problem as a Flow Network
- Understand two basic algorithms: Ford-Fulkerson & Edmonds-Karp
- Know that more advanced algorithms exist

# BIPARTITE MATCHING

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- Given a bipartite graph, with  $n$  nodes on one side and  $m$  nodes on the other side
- Find a maximum set of edges such that no two edges in the set share a common node





# BIPARTITE MATCHING – AUGMENTING PATH

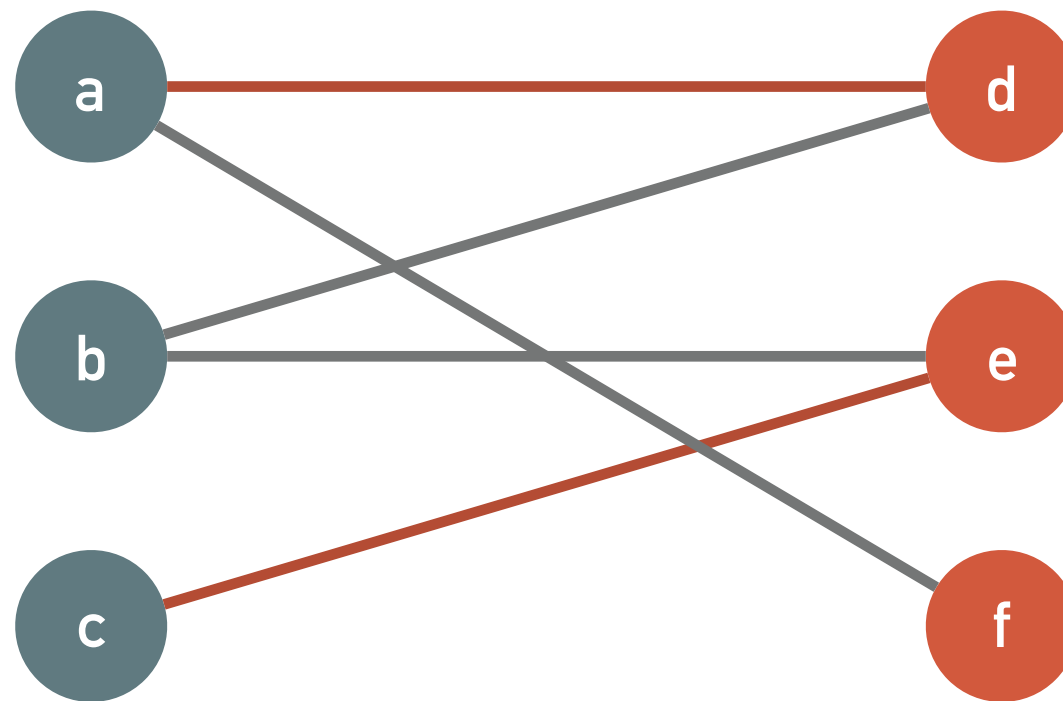
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- In order to compute bipartite matching algorithmically, we need to use the idea of “Augmenting Path”
- Let’s say we have a set  $S$  which contains the edges we have currently selected (initially empty)
- An augmenting path is a path such that
  - The length is odd
  - The 1st, 3rd, 5th, ... edges are currently NOT in  $S$
  - The 2nd, 4th, 6th, ... edges are currently in  $S$

# BIPARTITE MATCHING – AUGMENTING PATH

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- Here the red edges have been selected

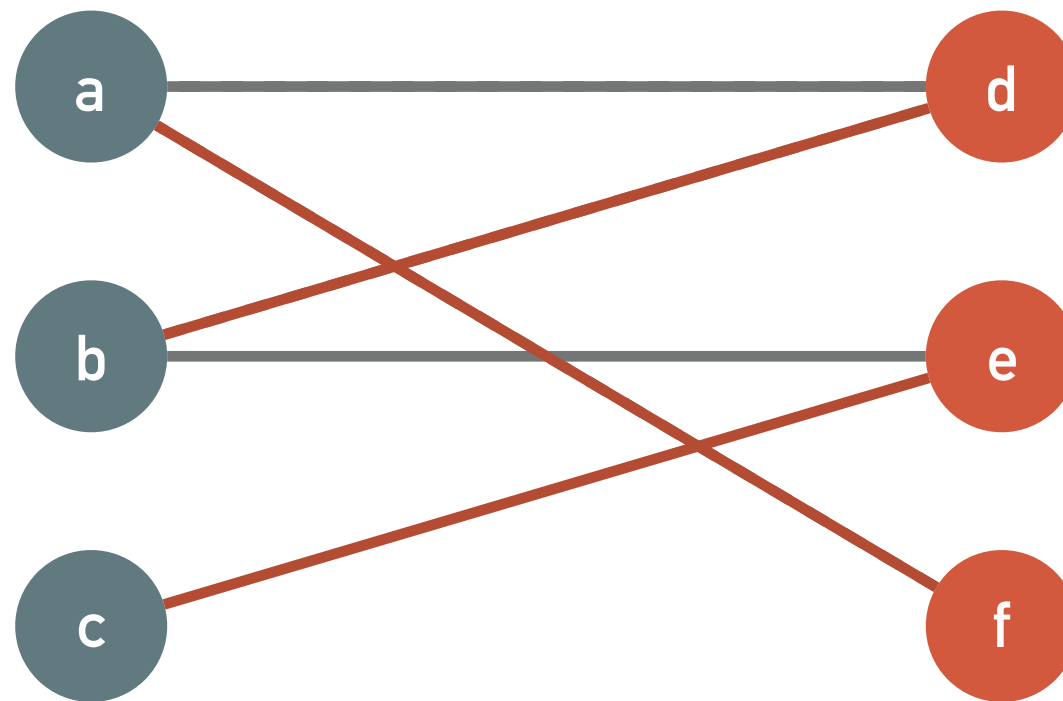


- Can you find an augmenting path?

# BIPARTITE MATCHING – AUGMENTING PATH

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- If we \*invert\* the augmenting path we just found



- We increase the size of our matching by 1

# BIPARTITE MATCHING – HUNGARIAN ALGORITHM

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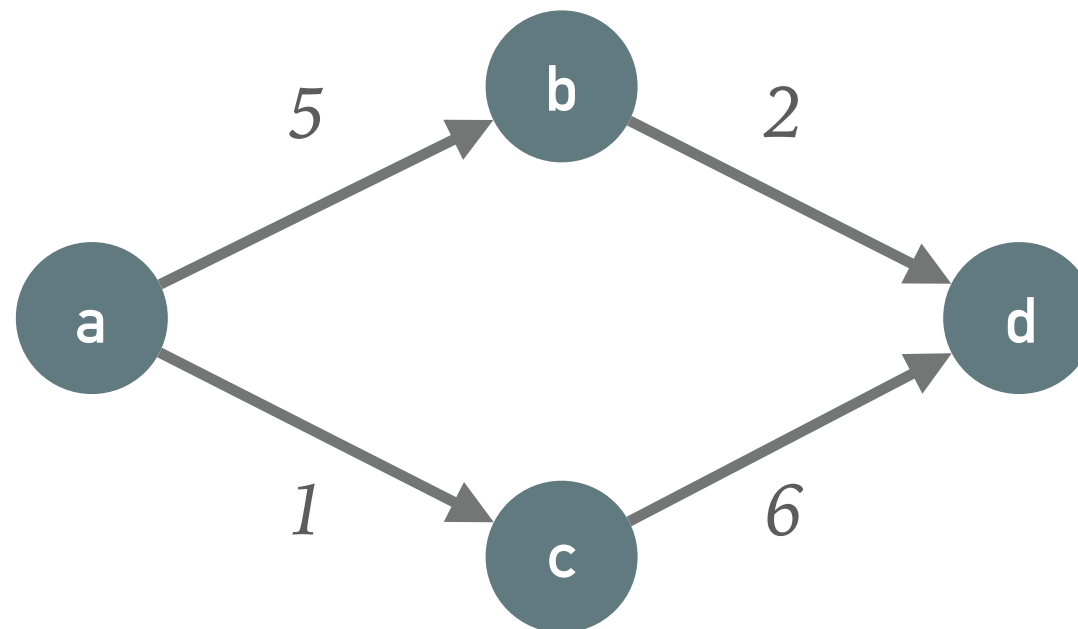
- Berge's Lemma: a matching in a graph is maximum if and only there is no augmenting path
- In other words, we can just keep finding augmenting paths (using bfs/dfs) and inverting them, and eventually we will get a maximum matching
- This idea is formalized as the Hungarian algorithm

```
for each node v on the left:  
    find an augmenting path starting with v  
    if found:  
        invert the path  
  
    // if not found, do not terminate the process  
    // move on to the next node  
    // this is necessary since the graph may be disconnected
```

# FLOW NETWORK – NON-RIGOROUS DEFINITION

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- A **Flow Network** is essentially a directed graph with a **capacity** assigned to each edge
- Think of it as the roads in a city, or a system of interconnected water pipes

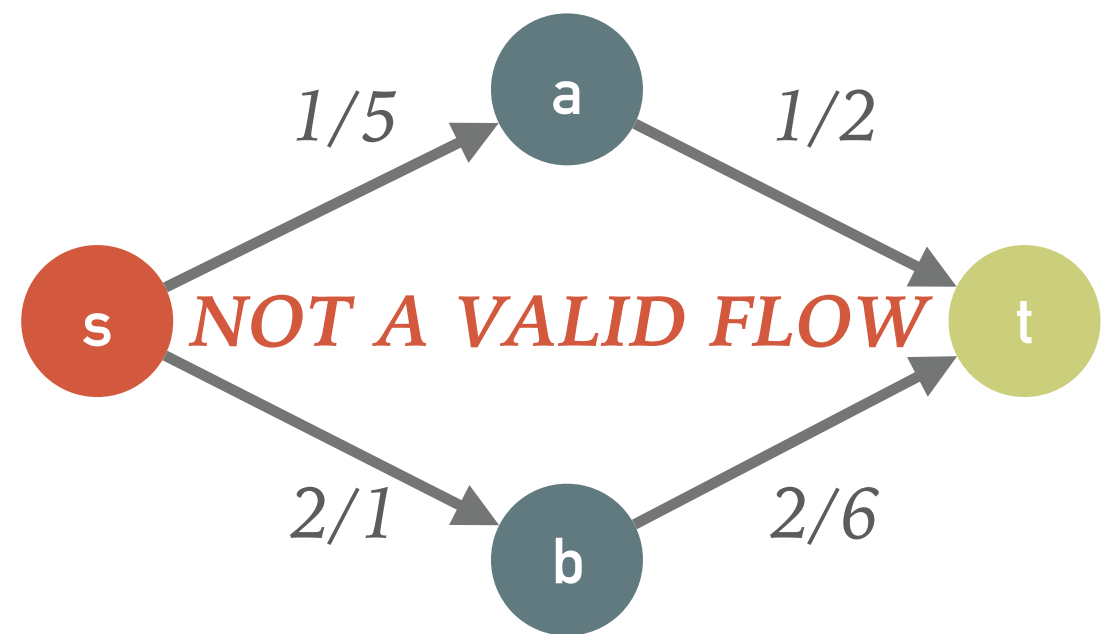
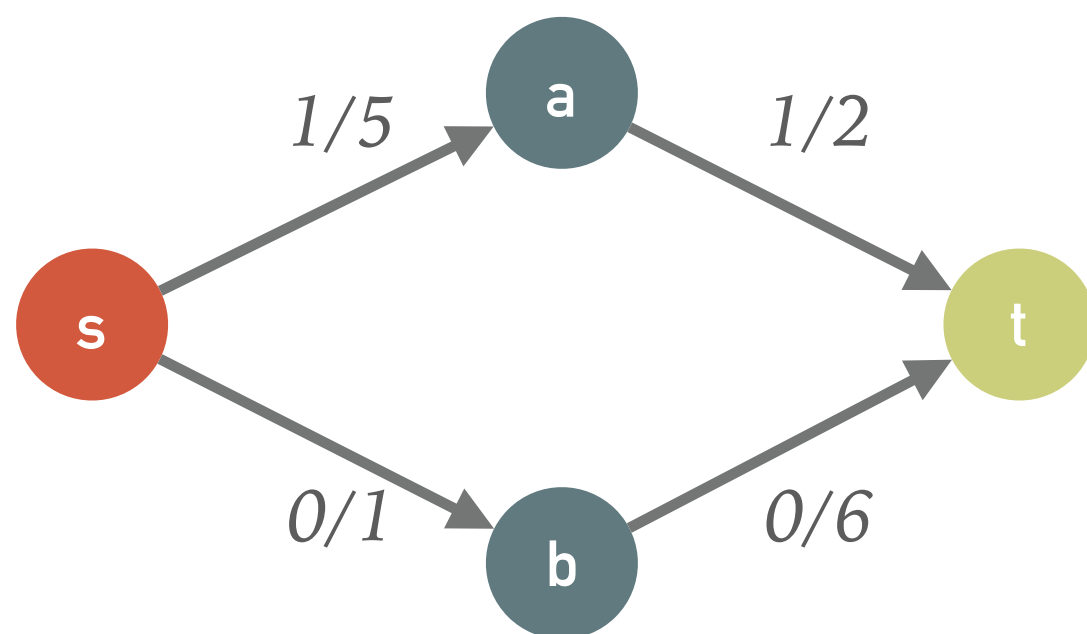




# FLOW NETWORK – NON-RIGOROUS DEFINITION

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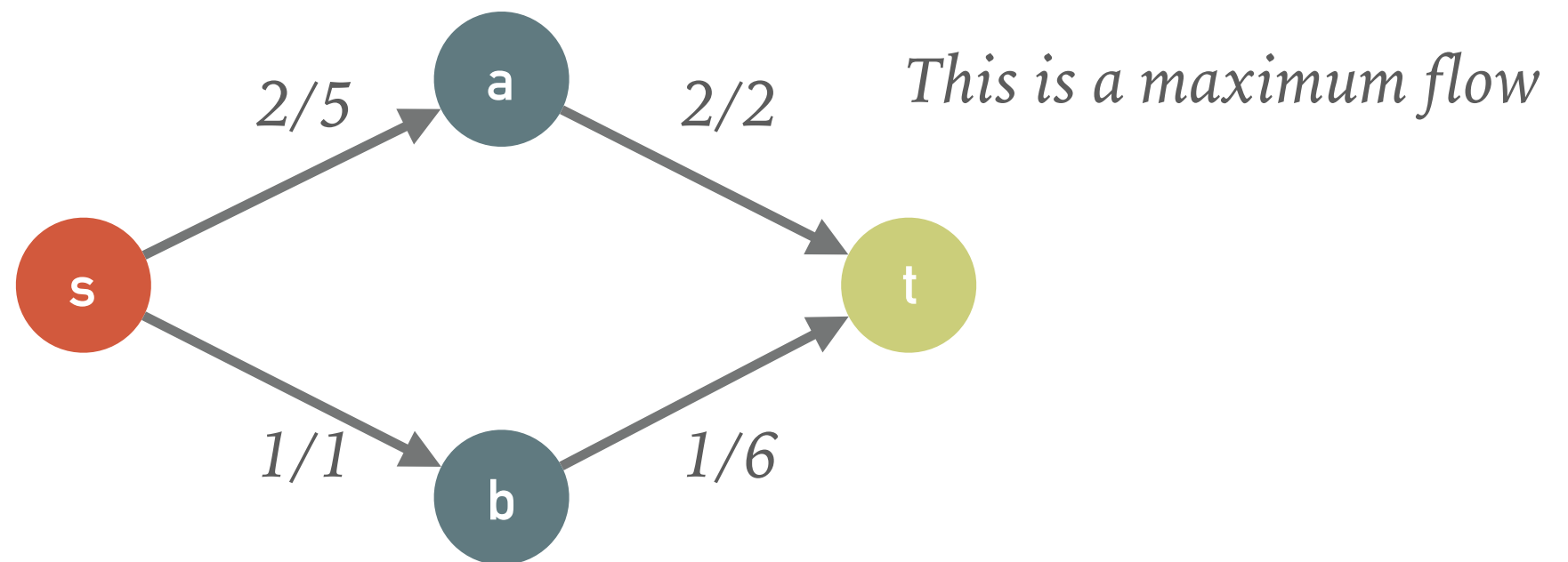
- A (feasible) **Flow** in a network, is a way to assign a non-negative **value** to each edge such that
  - The **value** is less than or equal to the **capacity** for each edge
  - Except two nodes that are specified as **source** and **sink**, the amount of flow going to and leaving each node should be equal (in other words, the net flow should be 0)



# FLOW NETWORK – MAXIMUM FLOW

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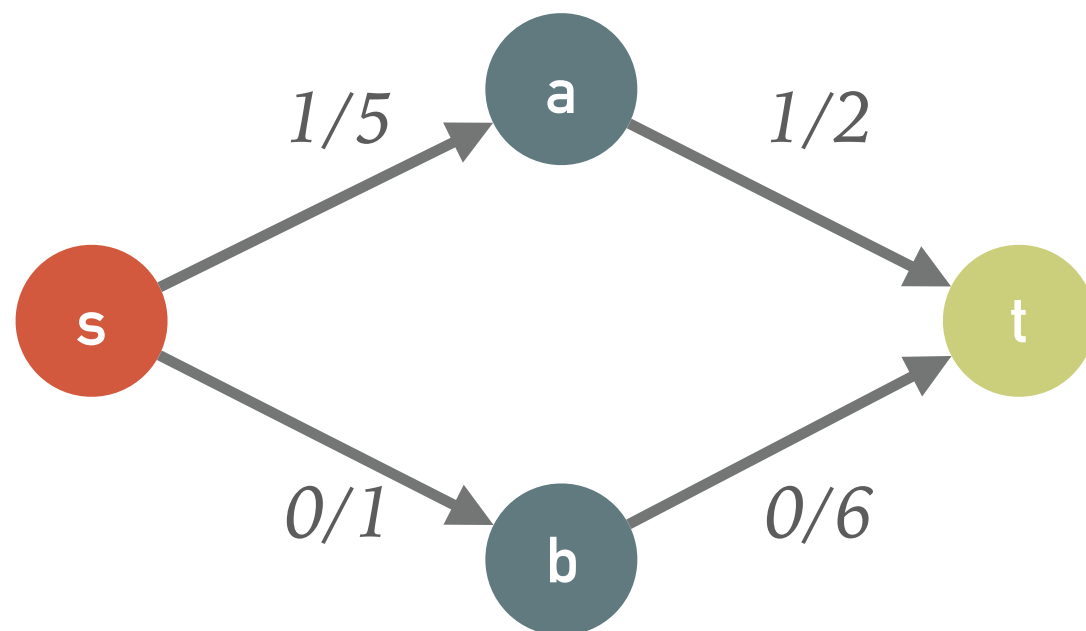
- The maximum flow problem asks for the maximum possible amount of **flow** going from the **source** to the **sink** in a given **flow network**
- A real example: in average, how many cars can go from one location to another per minute, knowing the capacity of each road in the city? (bandwidth)



# FLOW NETWORK – AUGMENTING PATH

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- In order to compute the maximum flow, again we need to use the idea of **augmenting path** (modified) and **residual network**
- Definition: an **augmenting path** is an s-t path with positive remaining capacity on each edge



*Can you find the augmenting paths?*

$s \rightarrow a \rightarrow t$

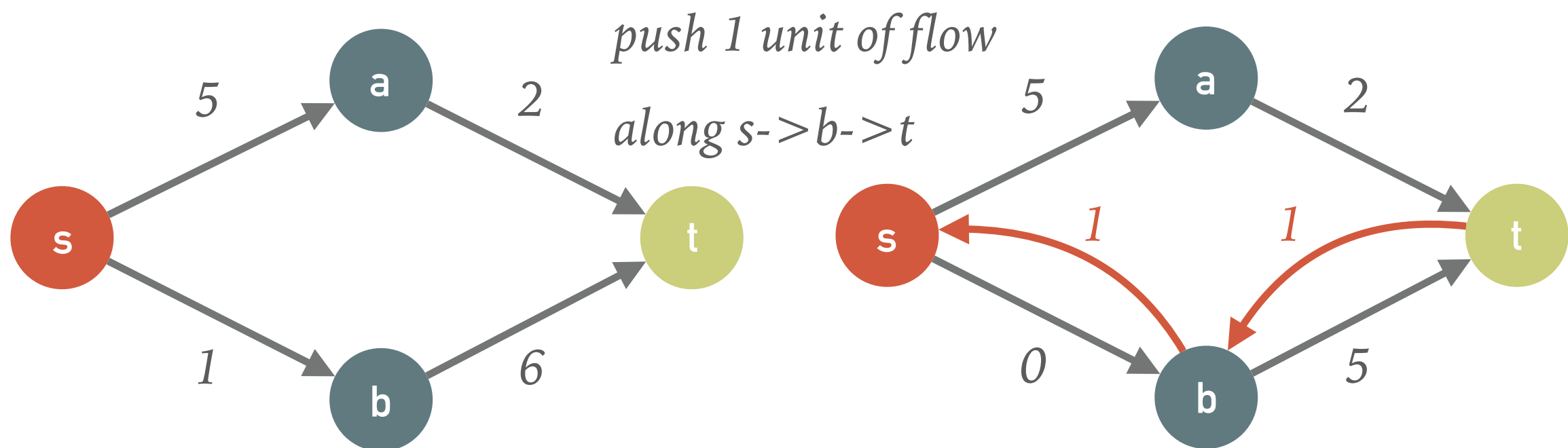
$s \rightarrow b \rightarrow t$



# FLOW NETWORK – RESIDUAL NETWORK

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- Let's say we have found an augmenting path in a network  $G$ , and we push  $f$  amount of flow along the path
- The **residual network** is the network obtained by
  - Decreasing the capacity of each edge along the path by  $f$
  - Increasing the capacity of each backward edge along the path by  $f$ , allowing the flow to come back



# FLOW NETWORK – FORD-FULKERSON

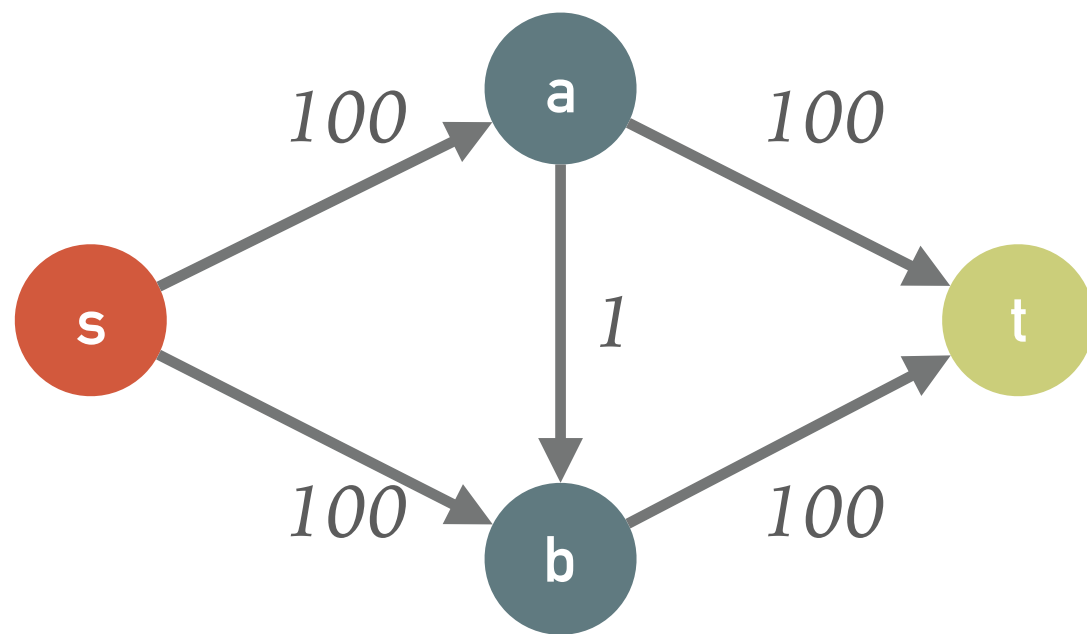
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- A flow we obtained is maximum if and only if we can not find any augmenting path in the network
- Similar to bipartite matching
  - Find an arbitrary augmenting path if there is one
  - Push as much flow as you can along the path
  - Repeat the process in the **residual network**
- This is called the **Ford-Fulkerson Algorithm**

# FLOW NETWORK – FORD-FULKERSON

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- Ford-Fulkerson has a time complexity of  $O(Ef)$ , where  $f$  is the value of the maximum flow
- It is very inefficient in a case like this:



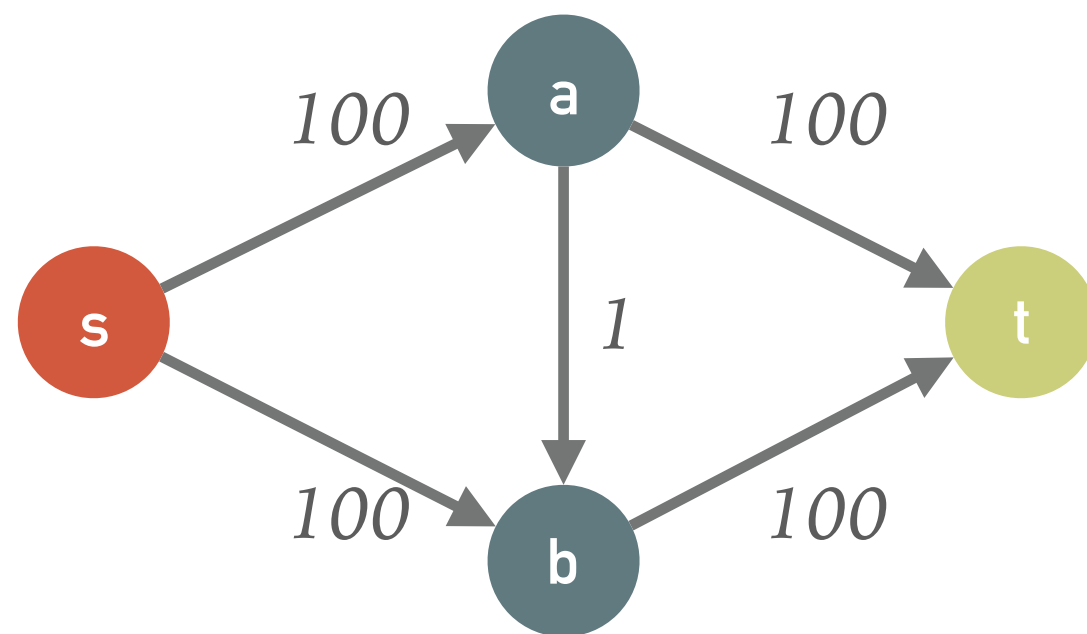
*try to push flow along  $s \rightarrow a \rightarrow b \rightarrow t$   
then try to push flow along  $s \rightarrow b \rightarrow a \rightarrow t$*



# FLOW NETWORK – EDMONDS-KARP

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- **Edmonds-Karp** is essentially the same as Ford-Fulkerson except that instead of choose an arbitrary augmenting path at a time, it always picks the shortest path
- Note “shortest” means the least number of edges



*In this case, Edmonds-Karp will pick  $s \rightarrow a \rightarrow t$  or  $s \rightarrow b \rightarrow t$*

- Time Complexity:  $O(VE^2)$  (no longer depends on  $f$ !)

# FLOW NETWORK – ADVANCED ALGORITHMS

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- There are some faster (and more complicated) algorithms to compute the maximum flow
- Here is a list from Prof. Jeff Erickson's notes:

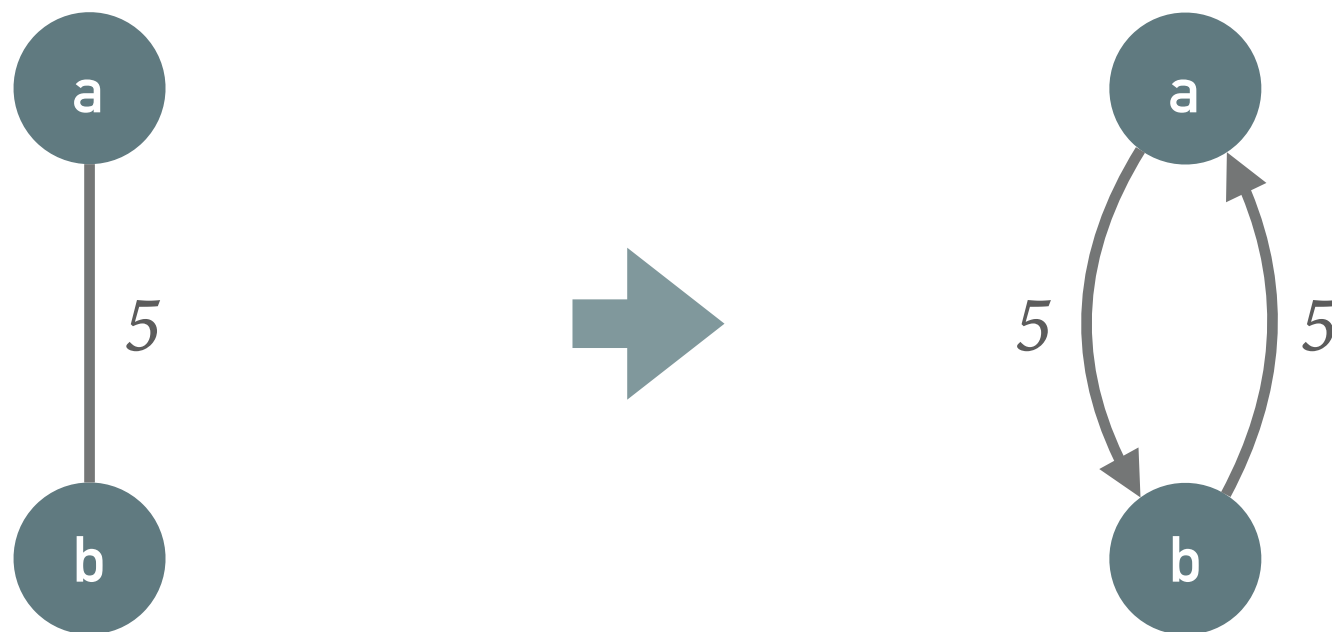
Technique	Direct	With dynamic trees	Sources
Blocking flow	$O(V^2E)$	$O(VE \log V)$	[Dinitz; Sleator and Tarjan]
Network simplex	$O(V^2E)$	$O(VE \log V)$	[Dantzig; Goldfarb and Hao; Goldberg, Grigoriadis, and Tarjan]
Push-relabel (generic)	$O(V^2E)$	—	[Goldberg and Tarjan]
Push-relabel (FIFO)	$O(V^3)$	$O(V^2 \log(V^2/E))$	[Goldberg and Tarjan]
Push-relabel (highest label)	$O(V^2\sqrt{E})$	—	[Cheriyani and Maheshwari; Tunçel]
Pseudoflow	$O(V^2E)$	$O(VE \log V)$	[Hochbaum]
Compact abundance graphs		$O(VE)$	[Orlin 2012]

Several purely combinatorial maximum-flow algorithms and their running times.

# FLOW NETWORK – A NOTE ON UNDIRECTED GRAPHS

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- If you are given a undirected graph, convert it into a directed one by replacing each undirected edge with two directed edges





# FLOW NETWORK – MAX-FLOW MIN-CUT THEOREM (OPTIONAL)

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- What is a **Cut**?
  - A cut is a partition of the nodes of a graph into two disjoint subsets  $S$ ,  $T$
  - A minimum cut is a cut such that the sum of the weights of the edges going from  $S$  to  $T$  is minimized
- **Max-Flow Min-Cut Theorem:**
  - The maximum amount of flow passing from the source to the sink is equal to the total weight of the edges in the minimum cut.

# FLOW NETWORK – A NOTE ON MIN COST MAX FLOW

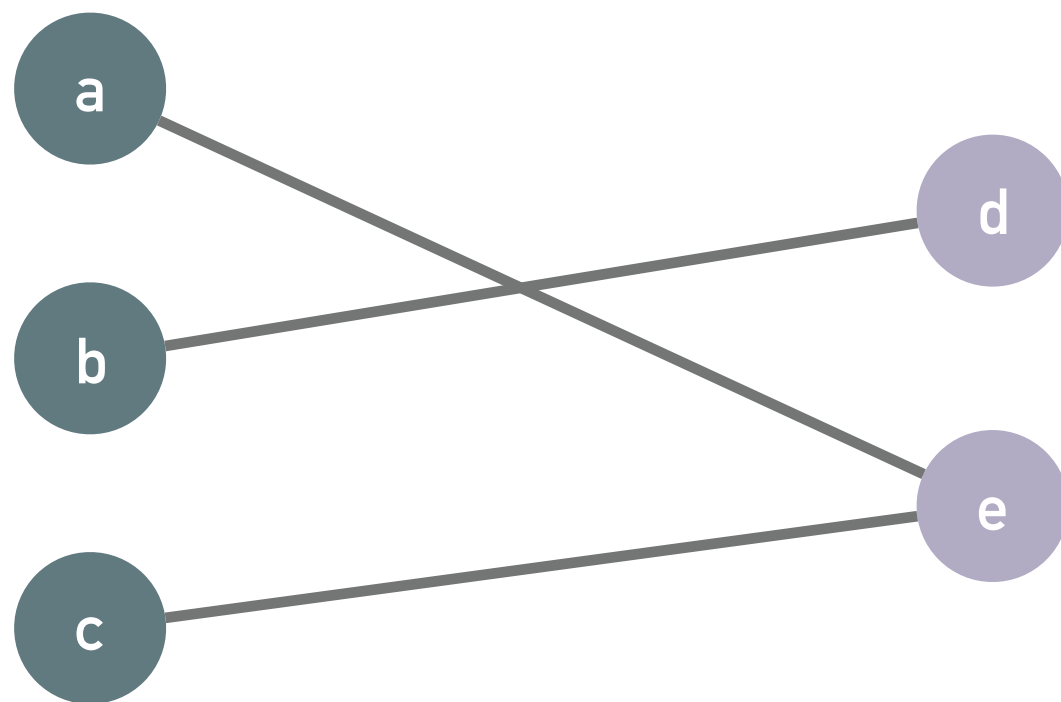
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- A variation of the maximum flow problem is the minimum cost maximum flow problem
- Each edge not only has a capacity, but also a unit cost for each unit of flow to pass by
- You need to not only find a maximum flow but also one with minimum cost
- Basic idea:
  - Use Bellman-Ford instead of BFS
  - When reversing an edge, assign (-cost) to the backward edge

# REDUCING BIPARTITE MATCHING TO MAXIMUM FLOW

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- Bipartite matching problem can be reduced to maximum flow problem by adding a super source and a super sink

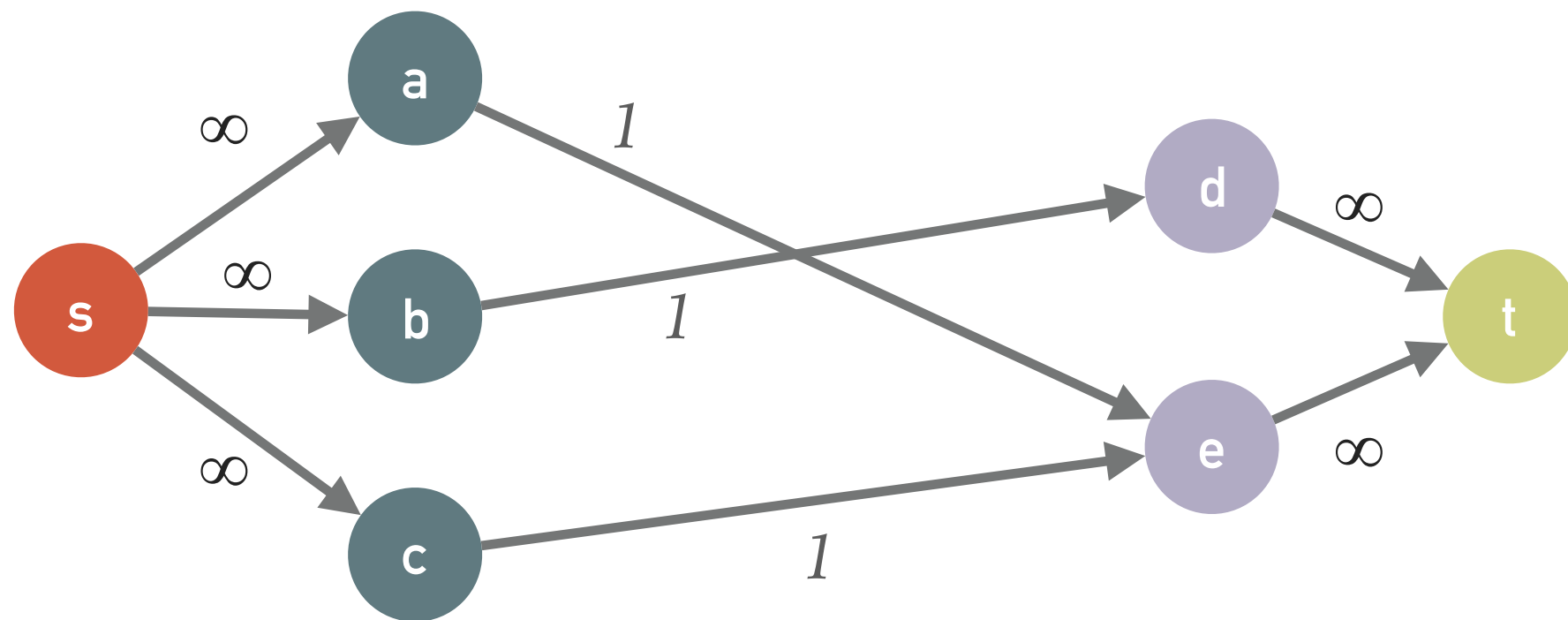




# REDUCING BIPARTITE MATCHING TO MAXIMUM FLOW

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- Each edge in the original bipartite graph is replaced by a directed one with capacity 1
- Add edges with infinite capacity from the source and to the sink



# EXAMPLE PROBLEM – DRAINAGE DITCHES

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<http://poj.org/problem?id=1273>

# EXAMPLE PROBLEM – MY T-SHIRT SUITS ME

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search “uva 11045” using your search engine



# QUESTIONS?

