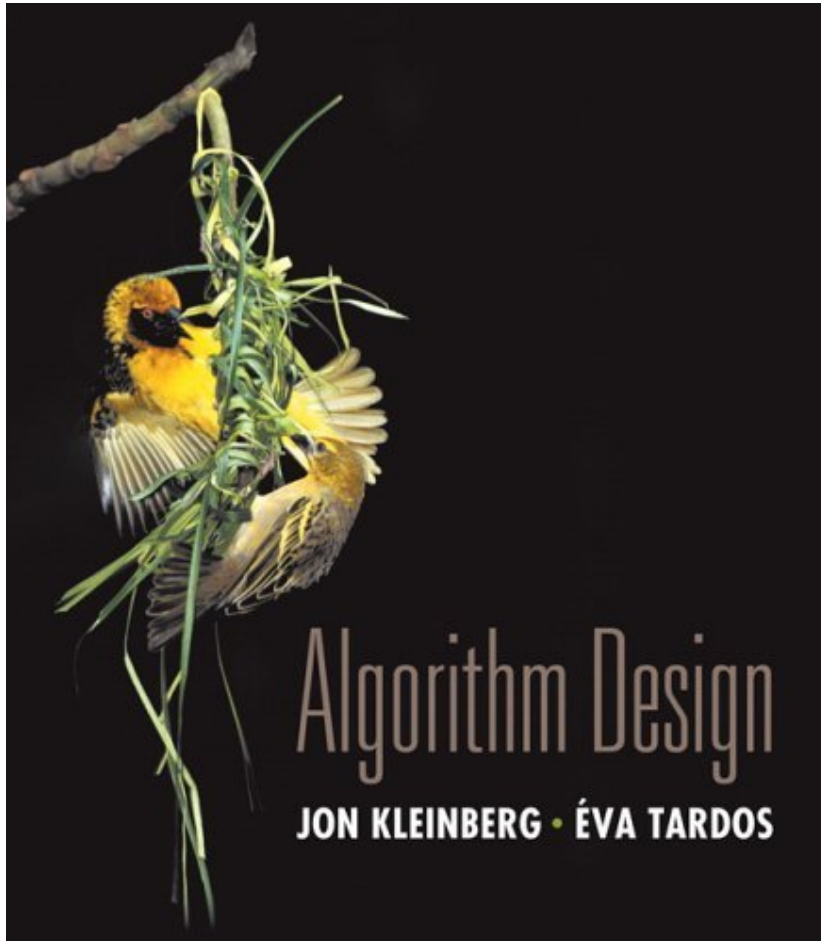


# CS 381 – FALL 2019

Week 12.1, Monday, Nov 4

Homework 6: Planning to Release Soon



# Network Flow



Slides by Kevin Wayne.  
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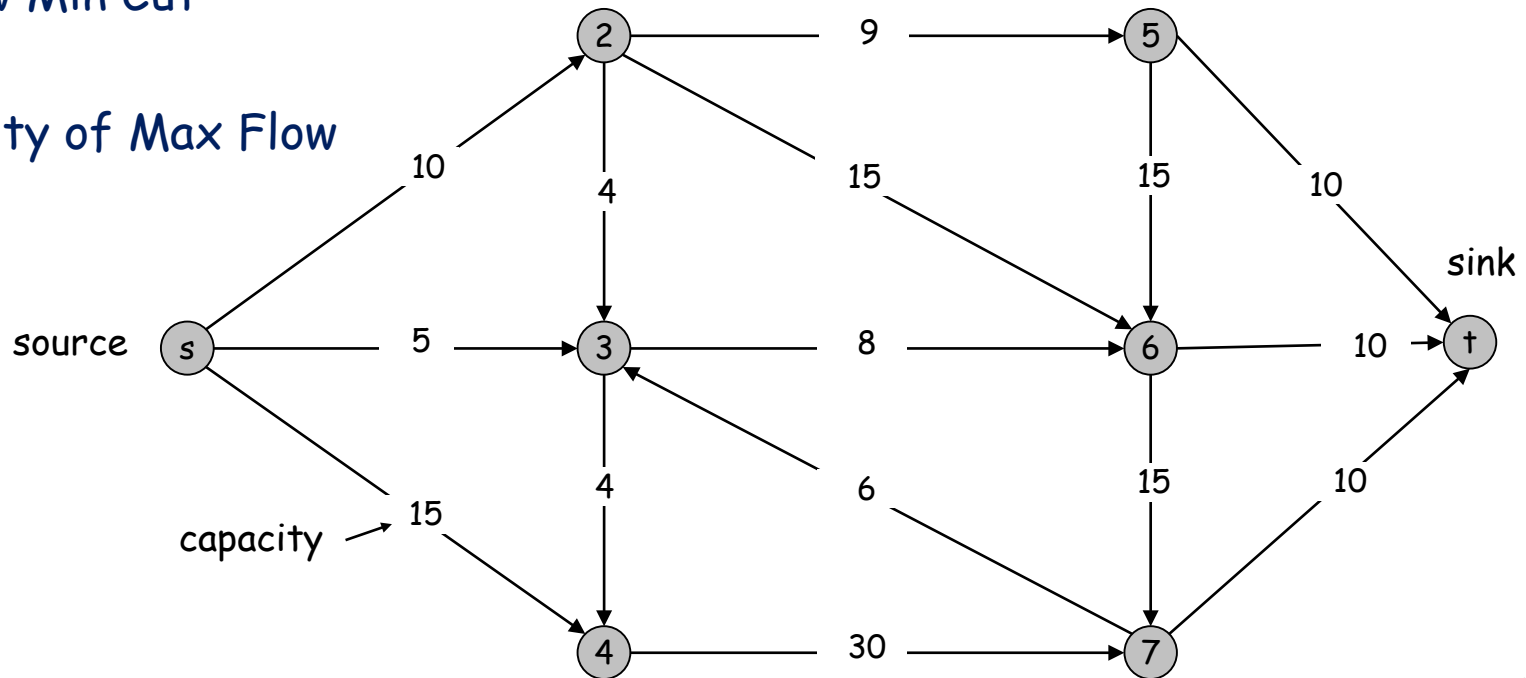
# Recap

## Residual Graph $G_f$

- Augmenting Path
- Ford-Fulkerson Algorithm
  - While the residual graph contains an augmenting path
    -  Increase Flow (Augment)
    -  Update Residual Graph

## Max Flow Min Cut

## Integrality of Max Flow



## Running time

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**Assumption.** Capacities are integers between 1 and  $C$ .

**Integrality invariant.** Throughout the algorithm, the flow values  $f(e)$  and the residual capacities  $c_f(e)$  are integers.

**Theorem.** The algorithm terminates in at most  $val(f^*) \leq nC$  iterations.

**Pf.** Each augmentation increases the value by at least 1.

**Corollary.** The running time of Ford-Fulkerson is  $O(mnC)$ .

**Corollary.** If  $C = 1$ , the running time of Ford-Fulkerson is  $O(mn)$ .

**Integrality theorem.** Then exists a max-flow  $f^*$  for which every flow value  $f^*(e)$  is an integer.

**Pf.** Since algorithm terminates, theorem follows from invariant. ■

Pseudo-polynomial

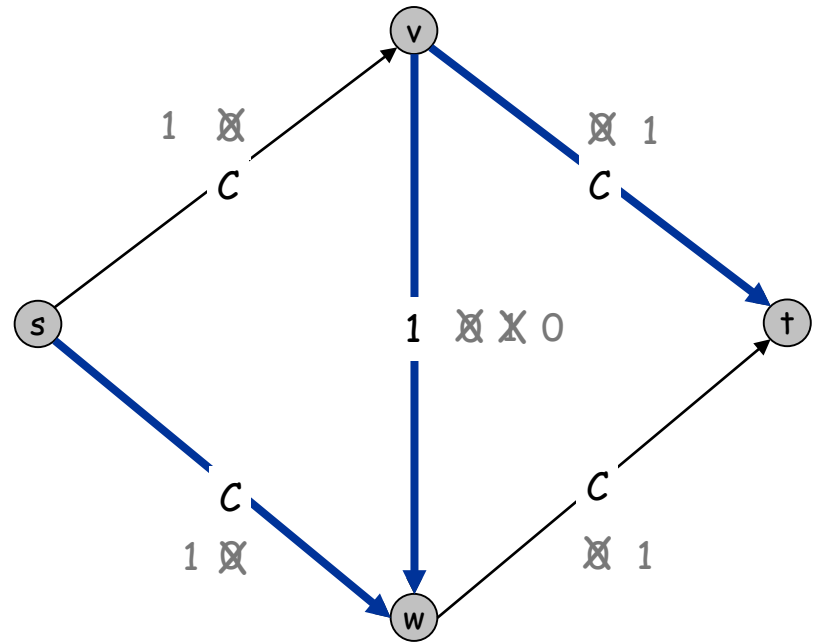
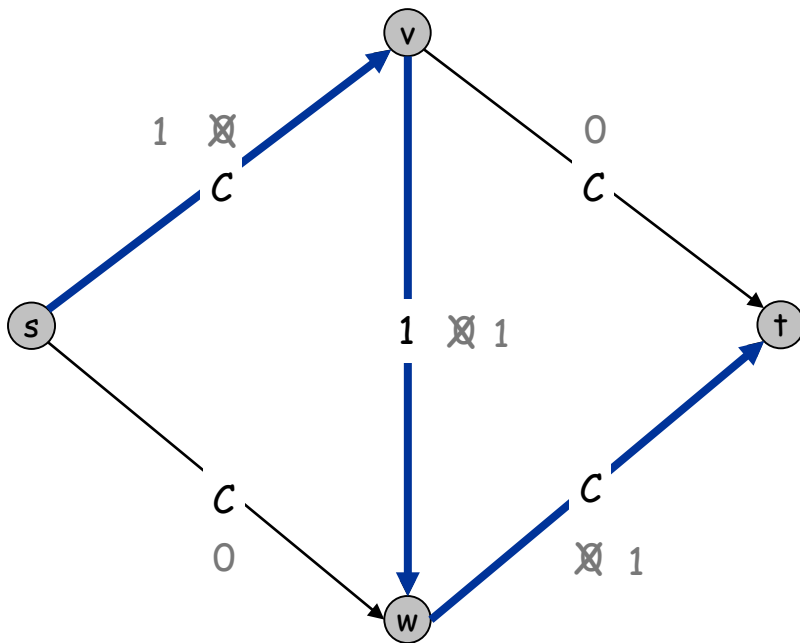


# Ford-Fulkerson: Exponential Number of Augmentations

Q. Is generic Ford-Fulkerson algorithm polynomial in input size?

$m, n,$  and  $\log C$  ↗

A. No. If max capacity is  $C$ , then algorithm can take  $C$  iterations.



# Bad case for Ford-Fulkerson

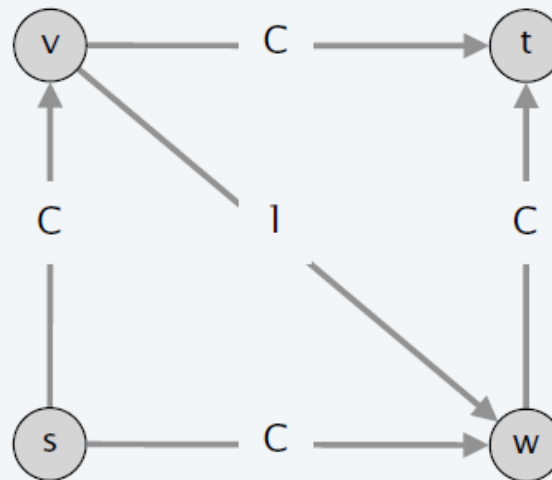
Q. Is generic Ford-Fulkerson algorithm poly-time in input size?

$m, n,$  and  $\log C$

A. No. If max capacity is  $C$ , then algorithm can take  $\geq C$  iterations.

- $s \rightarrow v \rightarrow w \rightarrow t$
- $s \rightarrow w \rightarrow v \rightarrow t$
- $s \rightarrow v \rightarrow w \rightarrow t$
- $s \rightarrow w \rightarrow v \rightarrow t$
- ...
- $s \rightarrow v \rightarrow w \rightarrow t$
- $s \rightarrow w \rightarrow v \rightarrow t$

each augmenting path  
sends only 1 unit of flow  
(# augmenting paths =  $2C$ )



## 7.3 Choosing Good Augmenting Paths

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# Choosing Good Augmenting Paths

Use care when selecting augmenting paths.

- Some choices lead to exponential algorithms.
- Clever choices lead to polynomial algorithms.
- If capacities are irrational, algorithm not guaranteed to terminate!

Goal: choose augmenting paths so that:

- Can find augmenting paths efficiently.
- Few iterations.

Choose augmenting paths with: [Edmonds-Karp 1972, Dinitz 1970]

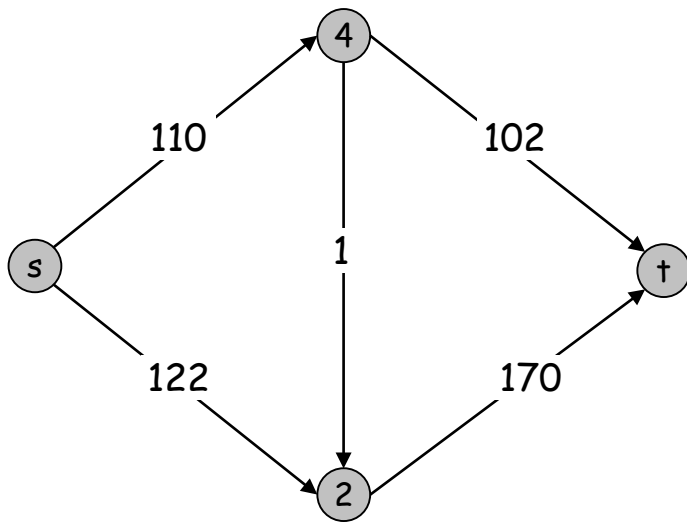
- Max bottleneck capacity.
- Sufficiently large bottleneck capacity.
- Fewest number of edges.



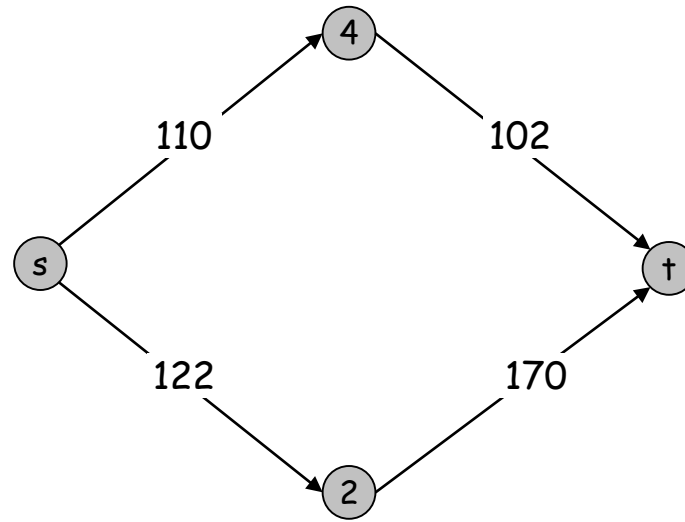
# Capacity Scaling

**Intuition.** Choosing path with highest bottleneck capacity increases flow by max possible amount.

- Don't worry about finding exact highest bottleneck path.
- Maintain scaling parameter  $\Delta$ .
- Let  $G_f(\Delta)$  be the subgraph of the residual graph consisting of only arcs with capacity at least  $\Delta$ .



$G_f$



$G_f(100)$

# Capacity Scaling

```
Scaling-Max-Flow( $G, s, t, c$ ) {  
  foreach  $e \in E$   $f(e) \leftarrow 0$   
   $\Delta \leftarrow$  smallest power of 2 greater than or equal to  $C$  //max capacity  
   $G_f \leftarrow$  residual graph  
  
  while ( $\Delta \geq 1$ ) {  
     $G_f(\Delta) \leftarrow \Delta$ -residual graph  
    while (there exists augmenting path  $P$  in  $G_f(\Delta)$ ) {  
       $f \leftarrow$  augment( $f, c, P$ )  
      update  $G_f(\Delta)$   
    }  
     $\Delta \leftarrow \Delta / 2$   
  }  
  return  $f$   
}
```

# Capacity Scaling: Correctness

**Assumption.** All edge capacities are integers between 1 and  $C$ .

**Integrality invariant.** All flow and residual capacity values are integral.

**Correctness.** If the algorithm terminates, then  $f$  is a max flow.

**Pf.**

- By integrality invariant, when  $\Delta = 1 \Rightarrow G_f(\Delta) = G_f$ .
- Upon termination of  $\Delta = 1$  phase, there are no augmenting paths. ▪

**Fact:** The algorithm terminates in polynomial time in  $n$ ,  $m$  and  $\log(C)$

**Proof:** Homework 6! (We provide the hints you provide the proof)

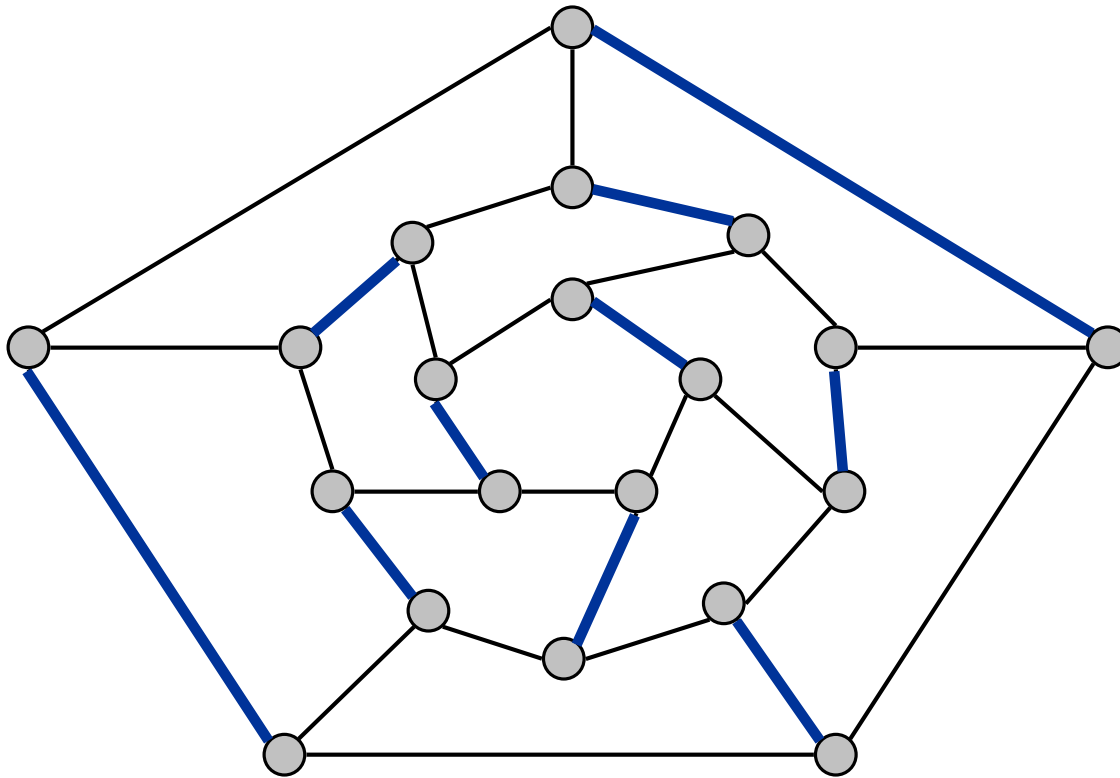
## 7.5 Bipartite Matching

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# Matching

## Matching.

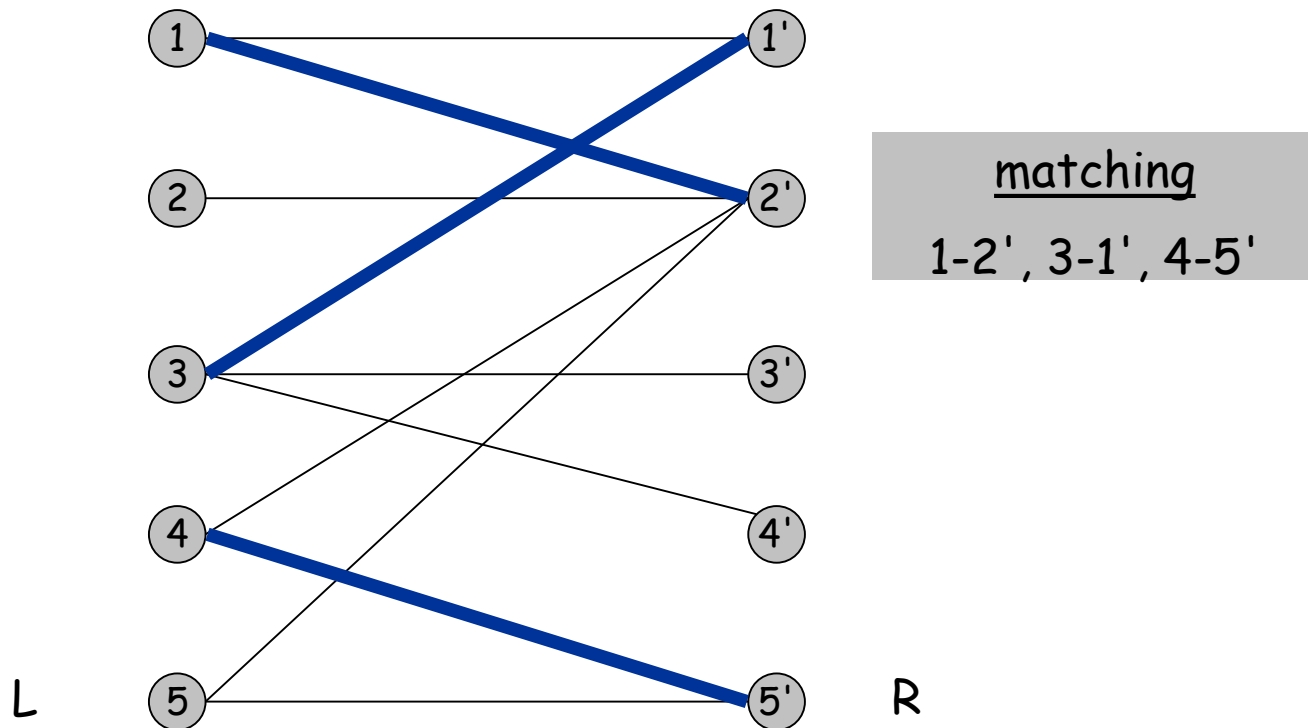
- Input: undirected graph  $G = (V, E)$ .
- $M \subseteq E$  is a **matching** if each node appears in at most edge in  $M$ .
- Max matching: find a max cardinality matching.



# Bipartite Matching

## Bipartite matching.

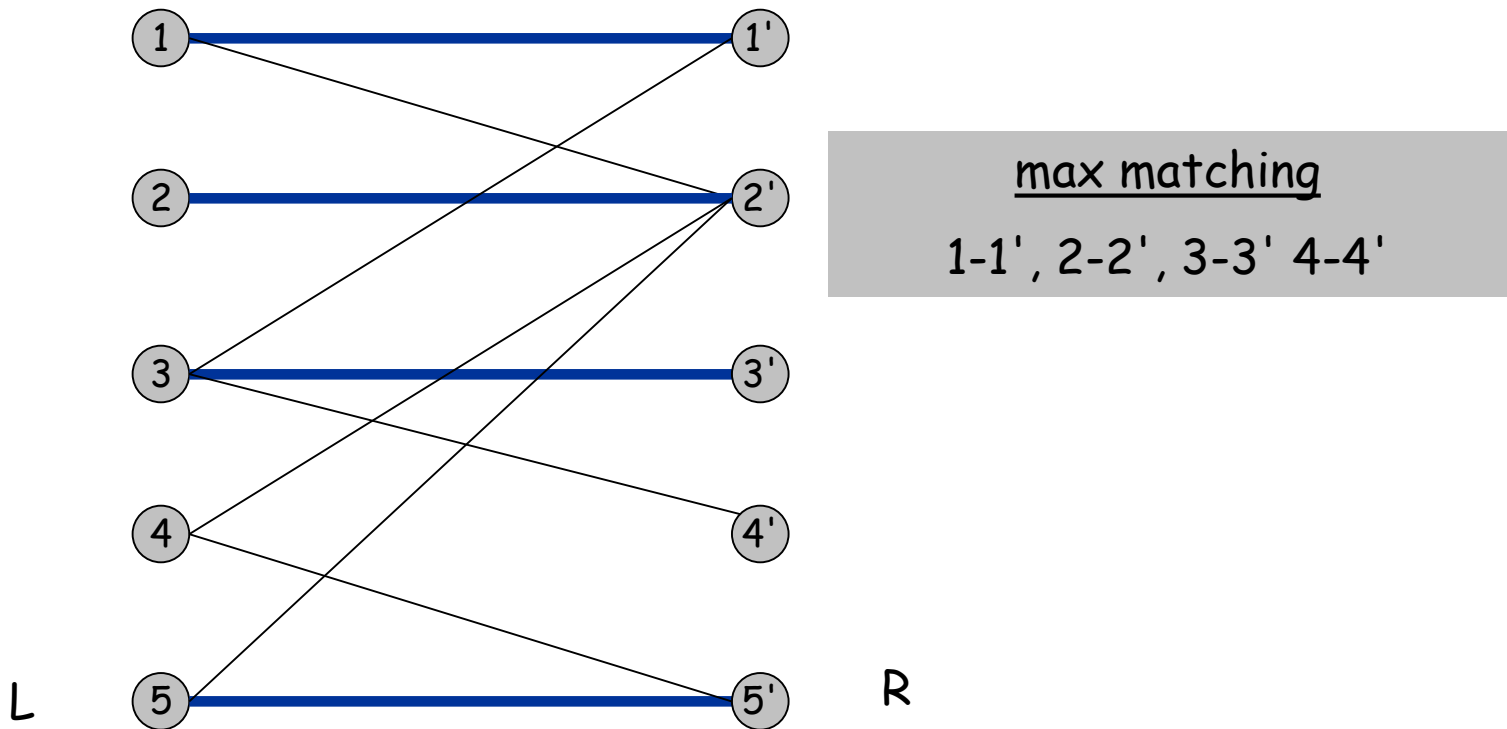
- Input: undirected, **bipartite** graph  $G = (L \cup R, E)$ .
- $M \subseteq E$  is a **matching** if each node appears in at most edge in  $M$ .
- Max matching: find a max cardinality matching.



# Bipartite Matching

## Bipartite matching.

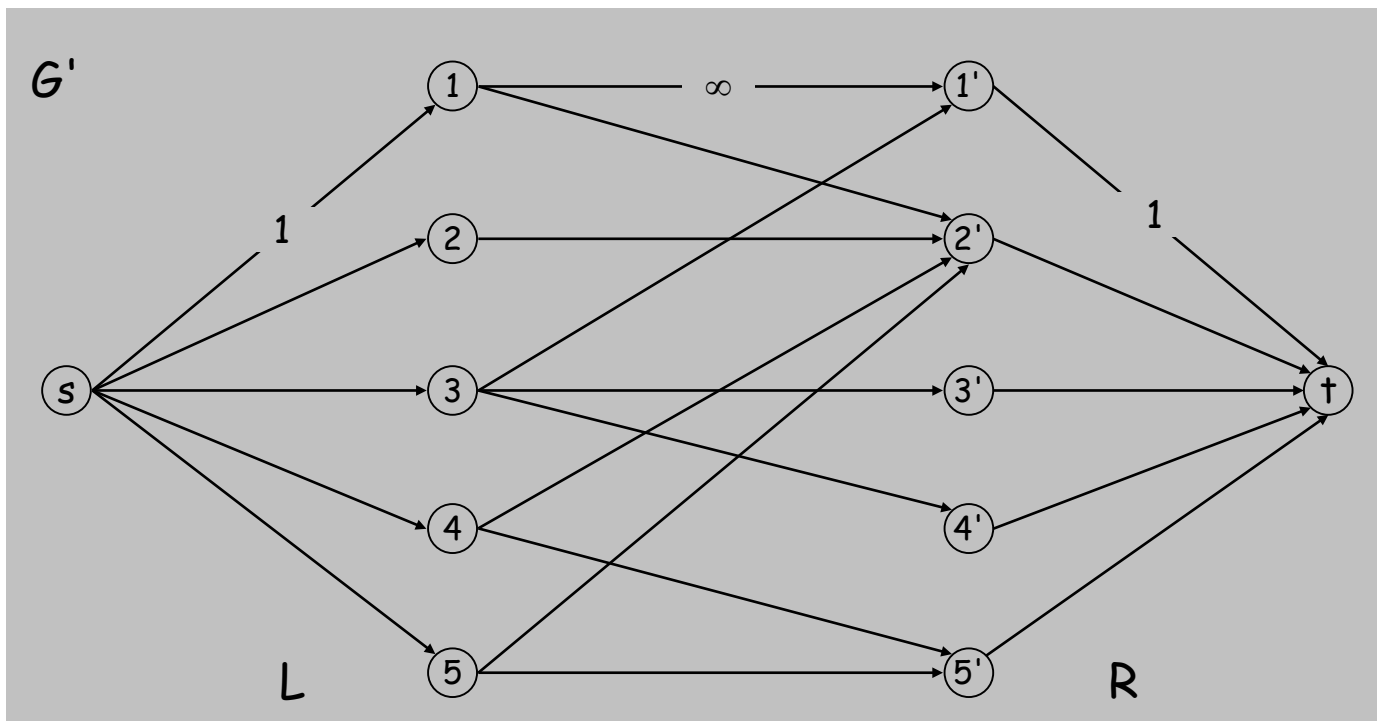
- Input: undirected, **bipartite** graph  $G = (L \cup R, E)$ .
- $M \subseteq E$  is a **matching** if each node appears in at most edge in  $M$ .
- Max matching: find a max cardinality matching.



# Bipartite Matching

## Max flow formulation.

- Create digraph  $G' = (L \cup R \cup \{s, t\}, E')$ .
- Direct all edges from  $L$  to  $R$ , and assign infinite (or unit) capacity.
- Add source  $s$ , and unit capacity edges from  $s$  to each node in  $L$ .
- Add sink  $t$ , and unit capacity edges from each node in  $R$  to  $t$ .





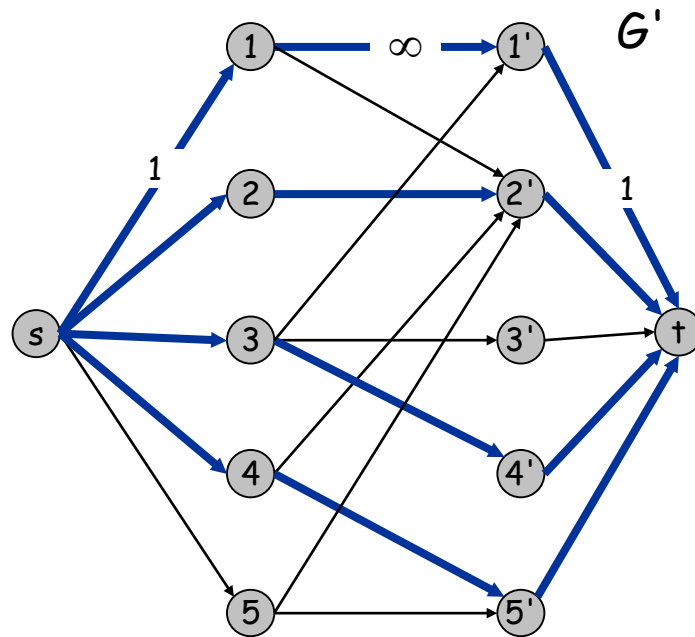
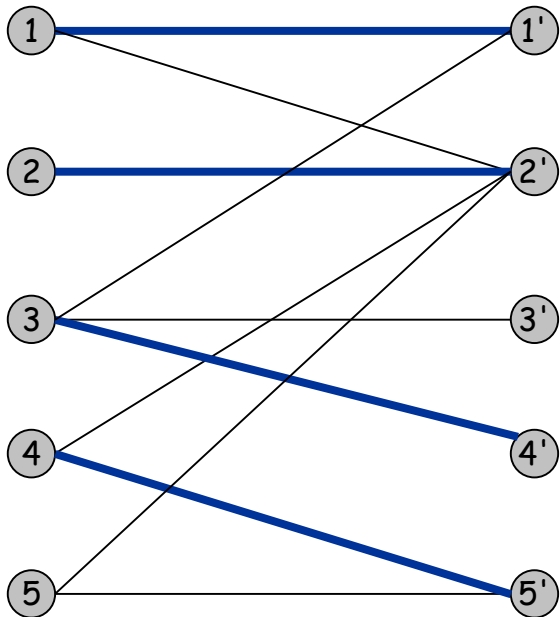
# Bipartite Matching: Proof of Correctness

**Theorem.** Max cardinality matching in  $G$  = value of max flow in  $G'$ .

**Pf.**  $\leq$

- Given max matching  $M$  of cardinality  $k$ .
- Consider flow  $f$  that sends 1 unit along each of  $k$  paths.
- $f$  is a flow, and has cardinality  $k$ . ▪

$G$



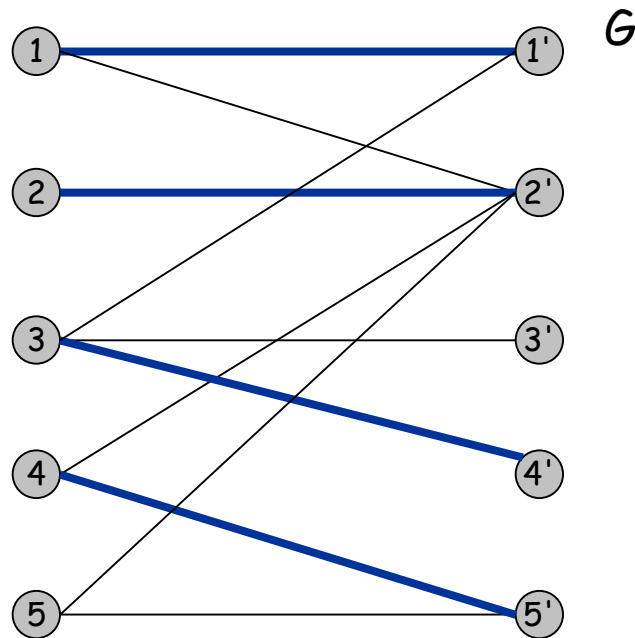
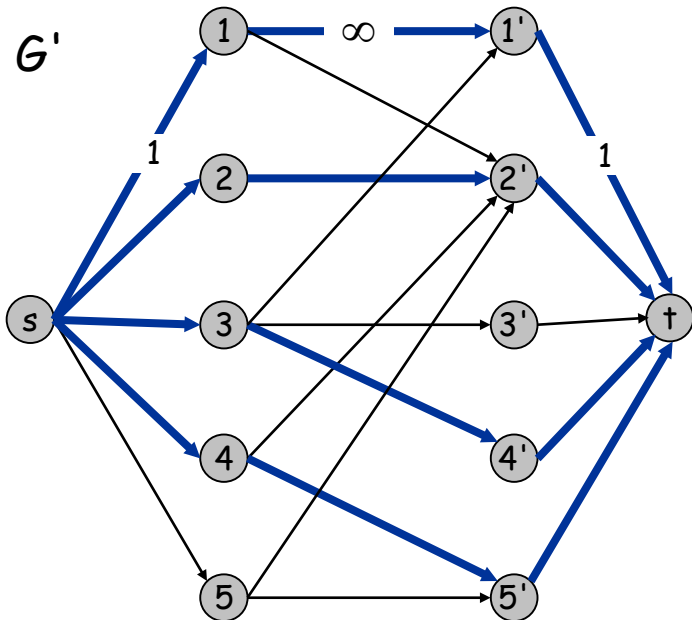
$G'$

# Bipartite Matching: Proof of Correctness

**Theorem.** Max cardinality matching in  $G$  = value of max flow in  $G'$ .

**Pf.**  $\geq$

- Let  $f$  be a max flow in  $G'$  of value  $k$ .
- Integrality theorem  $\Rightarrow$   $k$  is integral and can assume  $f$  is 0-1.
- Consider  $M$  = set of edges from  $L$  to  $R$  with  $f(e) = 1$ .
  - each node in  $L$  and  $R$  participates in at most one edge in  $M$
  - $|M| = k$ : consider *flow across the cut*  $(L \cup s, R \cup t)$  ▪



# Perfect Matching

**Def.** A matching  $M \subseteq E$  is **perfect** if each node appears in exactly one edge in  $M$ .

**Q.** When does a bipartite graph have a perfect matching?

Structure of bipartite graphs with perfect matchings.

- Clearly we must have  $|L| = |R|$ .
- What other conditions are necessary?
- What conditions are sufficient?

# Bipartite Matching: Running Time

Which max flow algorithm to use for bipartite matching?

- Generic augmenting path:  $O(m \text{ val}(f^*)) = O(mn)$ .
- Capacity scaling:  $O(m^2 \log C) = O(m^2)$ .
- Shortest augmenting path:  $O(m n^{1/2})$ .

Non-bipartite matching.

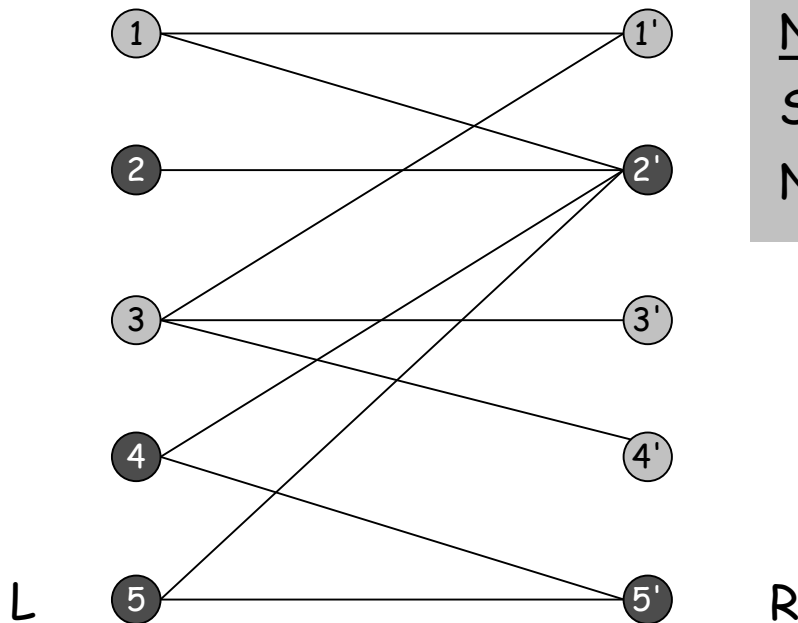
- Structure of non-bipartite graphs is more complicated, but well-understood. [Tutte-Berge, Edmonds-Galai]
- Blossom algorithm:  $O(n^4)$ . [Edmonds 1965]
- Best known:  $O(m n^{1/2})$ . [Micali-Vazirani 1980]

# Perfect Matching

**Notation.** Let  $S$  be a subset of nodes, and let  $N(S)$  be the set of nodes adjacent to nodes in  $S$ .

**Observation.** If a bipartite graph  $G = (L \cup R, E)$ , has a perfect matching, then  $|N(S)| \geq |S|$  for all subsets  $S \subseteq L$ .

**Pf.** Each node in  $S$  has to be matched to a different node in  $N(S)$ .



No perfect matching:

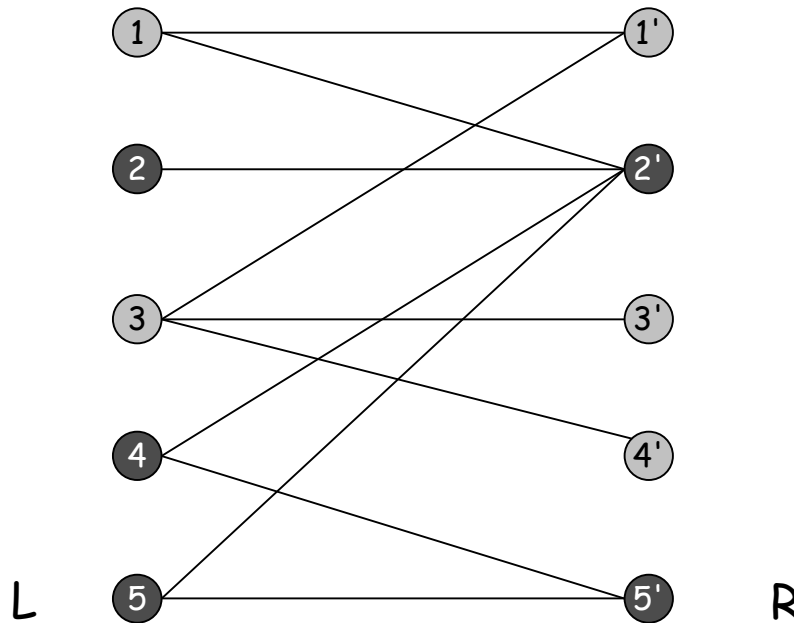
$S = \{ 2, 4, 5 \}$

$N(S) = \{ 2', 5' \}$ .

# Marriage Theorem

**Marriage Theorem.** [Frobenius 1917, Hall 1935] Let  $G = (L \cup R, E)$  be a bipartite graph with  $|L| = |R|$ . Then,  $G$  has a perfect matching iff  $|N(S)| \geq |S|$  for all subsets  $S \subseteq L$ .

**Pf.**  $\Rightarrow$  This was the previous observation.



No perfect matching:

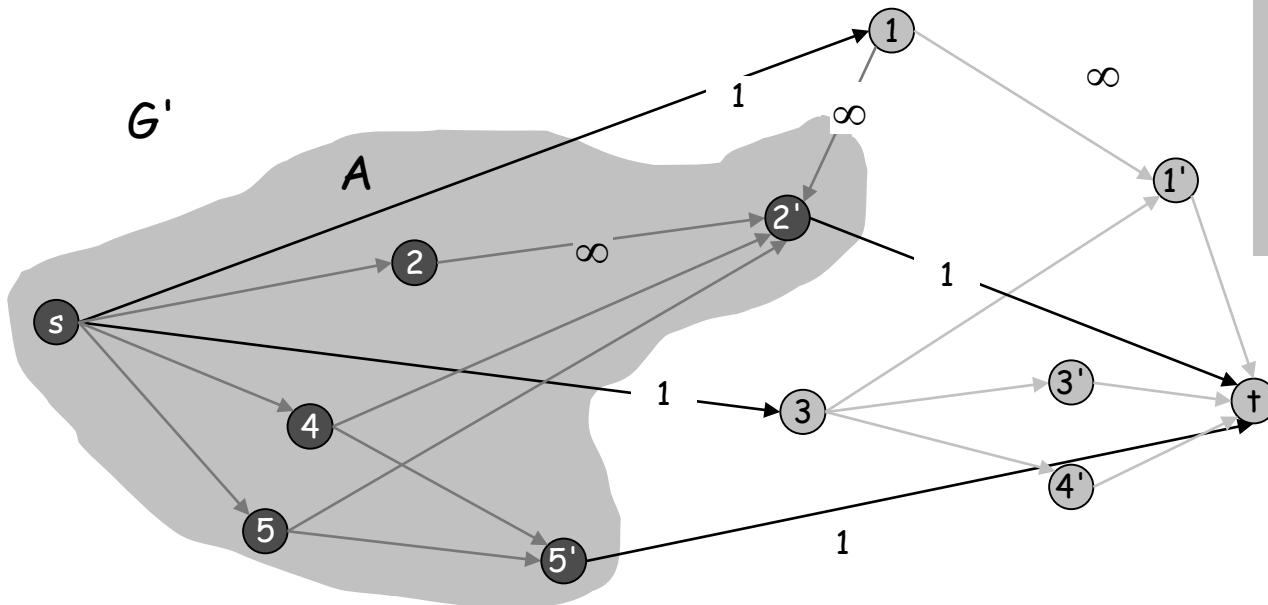
$$S = \{ 2, 4, 5 \}$$

$$N(S) = \{ 2', 5' \}.$$

# Proof of Marriage Theorem

Pf.  $\Leftarrow$  Suppose  $G$  does not have a perfect matching.

- Formulate as a max flow problem and let  $(A, B)$  be min cut in  $G'$ .
- By max-flow min-cut,  $\text{cap}(A, B) < |L|$ .
- Define  $L_A = L \cap A$ ,  $L_B = L \cap B$ ,  $R_A = R \cap A$ .
- Since min cut can't use  $\infty$  edges:  $N(L_A) \subseteq R_A$ .
- $\text{cap}(A, B) = |L_B| + |R_A|$  (again, since min cut can't use  $\infty$  edges).
- $|N(L_A)| \leq |R_A| = \text{cap}(A, B) - |L_B| < |L| - |L_B| = |L_A|$ .
- Choose  $S = L_A$ .



$L_A = \{2, 4, 5\}$   
 $L_B = \{1, 3\}$   
 $R_A = \{2', 5'\}$   
 $N(L_A) = \{2', 5'\}$

