

CSE 421

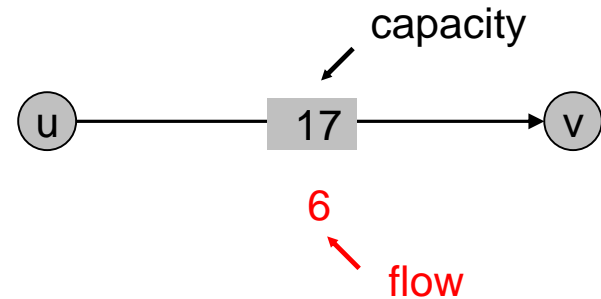
Max Flow Min Cut, Bipartite Matching

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Residual Graph

Original edge: $e = (u, v) \in E$.

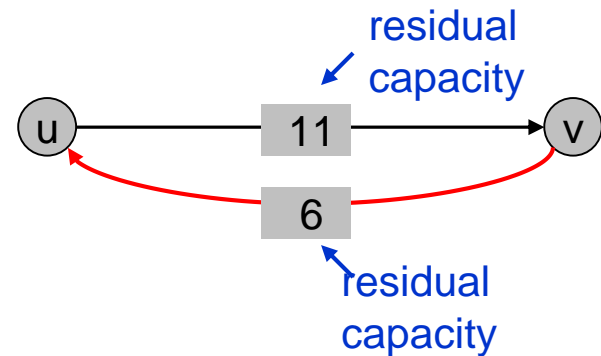
- Flow $f(e)$, capacity $c(e)$.



Residual edge.

- "Undo" flow sent.
- $e = (u, v)$ and $e^R = (v, u)$.
- Residual capacity:

$$c_f(e) = \begin{cases} c(e) - f(e) & \text{if } e \in E \\ f(e) & \text{if } e^R \in E \end{cases}$$



Residual graph: $G_f = (V, E_f)$.

- Residual edges with positive residual capacity.
- $E_f = \{e : f(e) < c(e)\} \cup \{e : f(e^R) > 0\}$.

Augmenting Path Algorithm

```
Augment(f, c, P) {  
  b ← bottleneck(P) ← Smallest capacity edge on P  
  foreach e ∈ P {  
    if (e ∈ E) f(e) ← f(e) + b ← Forward edge  
    else      f(eR) ← f(e) - b ← Reverse edge  
  }  
  return f  
}
```

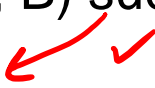

```
Ford-Fulkerson(G, s, t, c) {  
  foreach e ∈ E f(e) ← 0  
  Gf ← residual graph  
  
  while (there exists augmenting path P) {  
    f ← Augment(f, c, P)  
    update Gf  
  }  
  return f  
}
```

Max Flow Min Cut Theorem

Augmenting path theorem. Flow f is a max flow iff there are no augmenting paths.

Max-flow min-cut theorem. [Ford-Fulkerson 1956] The value of the max s - t flow is equal to the value of the min s - t cut.

Proof strategy. We prove both simultaneously by showing the TFAE:

- (i) There exists a cut (A, B) such that $v(f) = \text{cap}(A, B)$.
 - (ii) Flow f is a max flow. 
 - (iii) There is no augmenting path relative to f .
- 

(i) \Rightarrow (ii) This was the corollary to weak duality lemma.

(ii) \Rightarrow (iii) We show contrapositive.

Let f be a flow. If there exists an augmenting path, then we can improve f by sending flow along that path.

Pf of Max Flow Min Cut Theorem

(iii) \Rightarrow (i)

No augmenting path for $f \Rightarrow$ there is a cut (A,B) : $v(f)=\text{cap}(A,B)$

- Let f be a flow with no augmenting paths.
- Let A be set of vertices reachable from s in residual graph.
- By definition of A , $s \in A$.
- By definition of f , $t \notin A$.

$$\begin{aligned}v(f) &= \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e) \\ &= \sum_{e \text{ out of } A} c(e) \\ &= \text{cap}(A, B)\end{aligned}$$

Running Time

Assumption. All capacities are integers between 1 and C .

Invariant. Every flow value $f(e)$ and every residual capacities $c_f(e)$ remains an **integer** throughout the algorithm.

Theorem. The algorithm terminates in at most $v(f^*) \leq nC$ iterations, if f^* is optimal flow.

Pf. Each augmentation increase value by at least 1.

Corollary. If $C = 1$, Ford-Fulkerson runs in $O(mn)$ time.

Integrality theorem. If all capacities are integers, then there exists a max flow f for which every flow value $f(e)$ is an integer.

Pf. Since algorithm terminates, theorem follows from invariant.

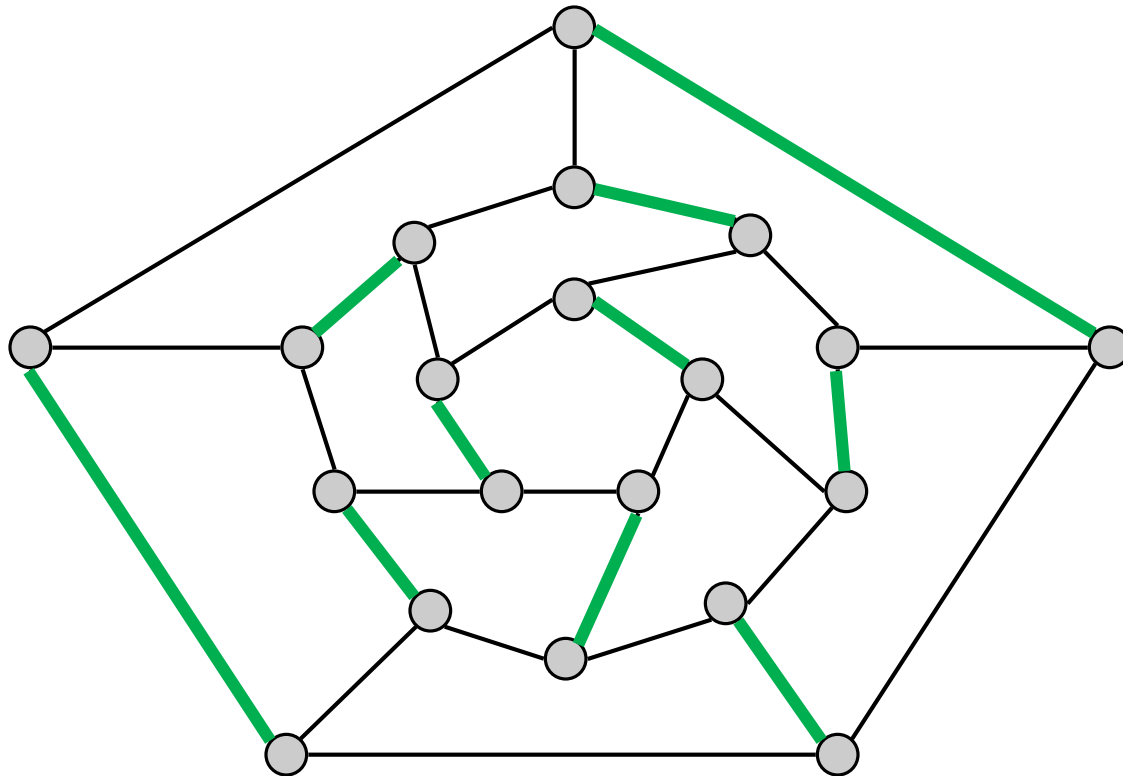
Applications of Max Flow: Bipartite Matching

Maximum Matching Problem

Given an undirected graph $G = (V, E)$.

A set $M \subseteq E$ is a **matching** if each node appears in at most one edge in M .

Goal: find a matching with largest cardinality.

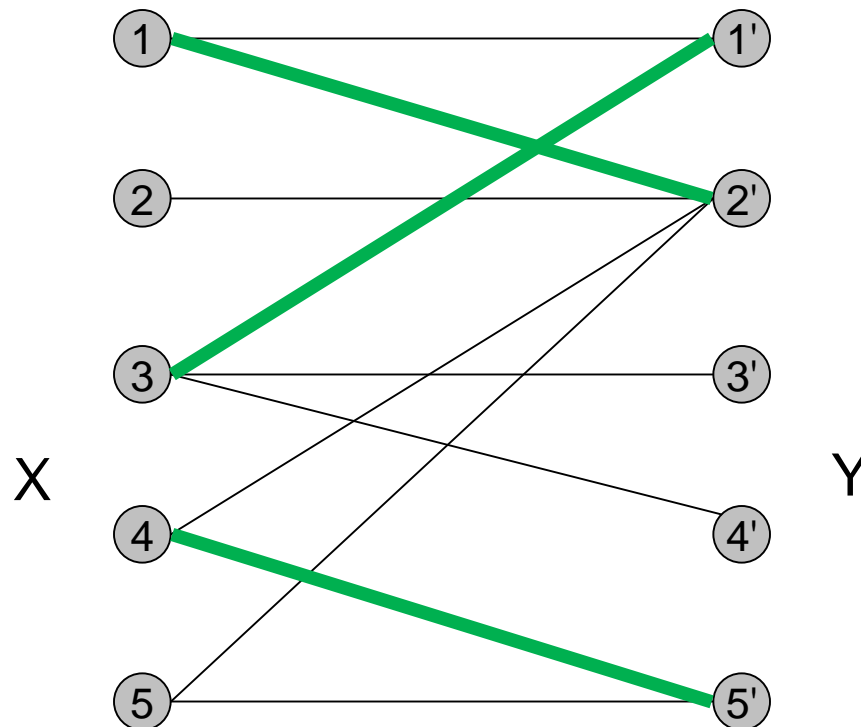


Bipartite Matching Problem

Given an undirected bipartite graph $G = (X \cup Y, E)$

A set $M \subseteq E$ is a **matching** if each node appears in at most one edge in M .

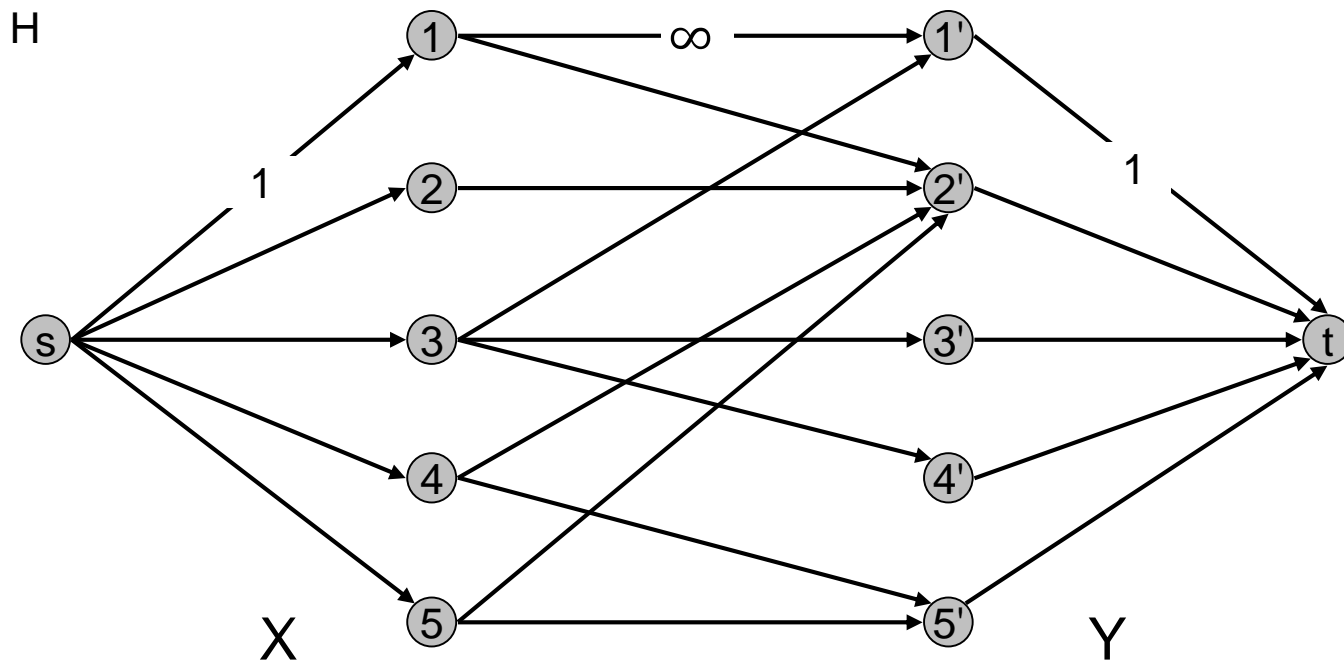
Goal: find a matching with largest cardinality.



Bipartite Matching using Max Flow

Create digraph H as follows:

- Orient all edges from X to Y, 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

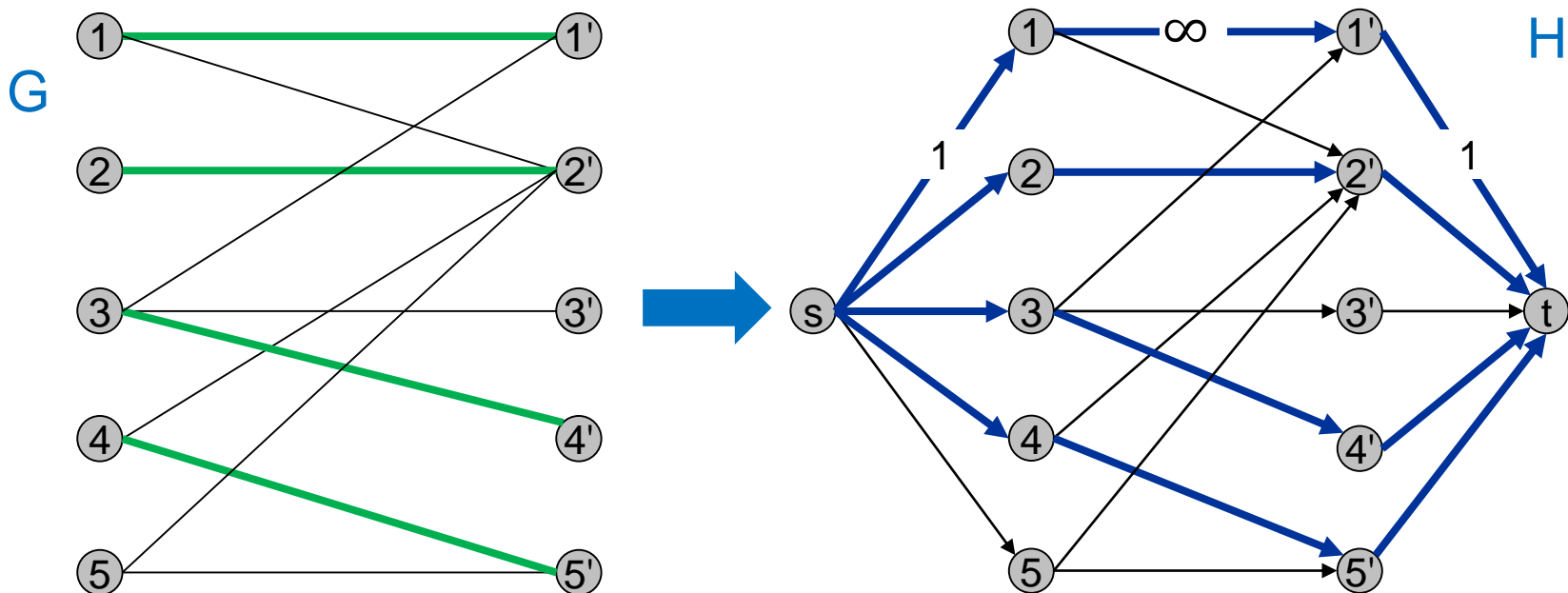
Thm. Max cardinality matching in G = value of max flow in H .

Pf. \leq

Given max matching M of cardinality k .

Consider flow f that sends 1 unit along each of k edges of M .

f is a flow, and has cardinality k . \blacksquare



Bipartite Matching: Proof of Correctness

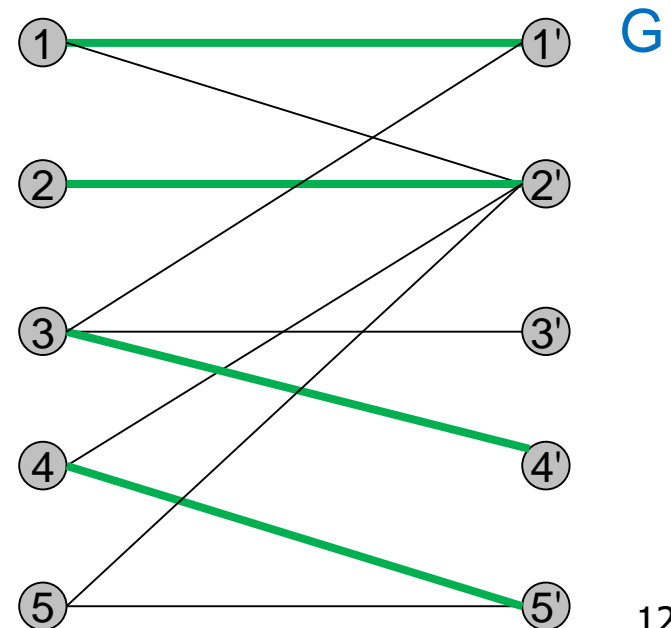
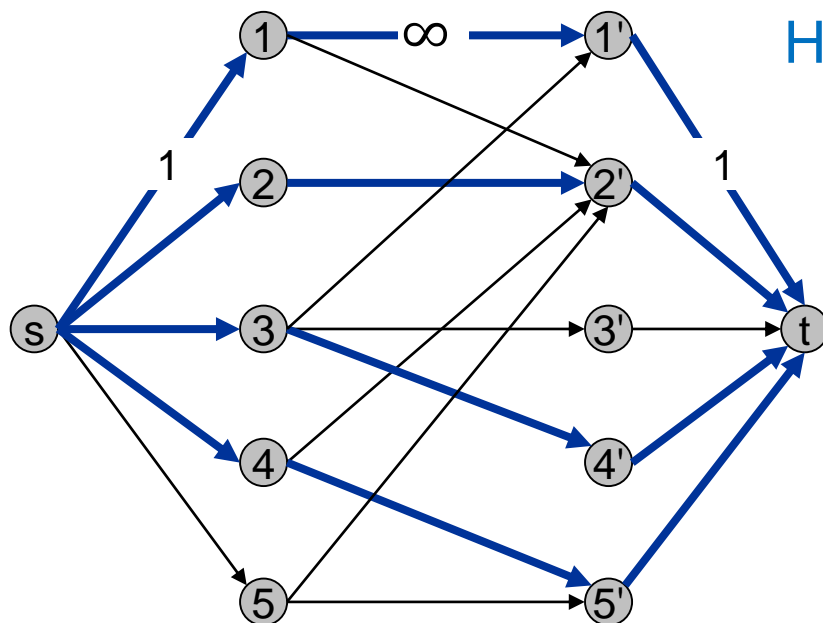
Thm. Max cardinality matching in G = value of max flow in H .

Pf. (of \geq) Let f be a max flow in H of value k .

Integrality theorem \Rightarrow k is integral and we can assume f is 0-1.

Consider M = set of edges from X to Y with $f(e) = 1$.

- each node in X and Y participates in at most one edge in M
- $|M| = k$: consider s-t cut $(s \cup X, t \cup Y)$



Perfect Bipartite Matching

Perfect Bipartite 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:

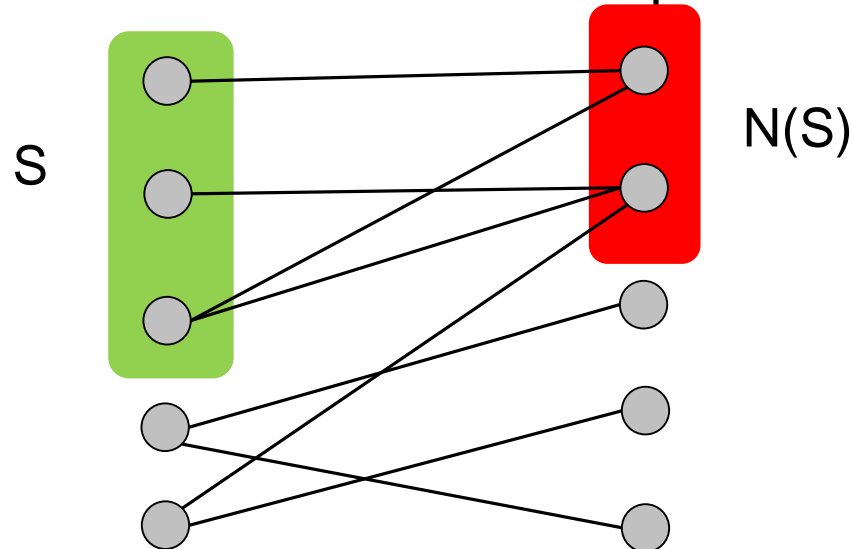
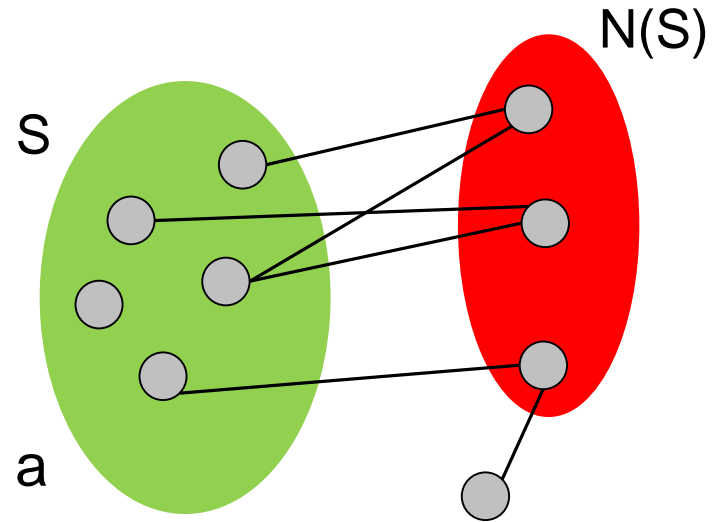
- Clearly we must have $|X| = |Y|$.
- What other conditions are necessary?
- What conditions are sufficient?

Perfect Bipartite Matching: $N(S)$

Def. 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 has a perfect matching, then $|N(S)| \geq |S|$ for all subsets $S \subseteq X$.

Pf. Each $v \in S$ has to be matched to a unique node in $N(S)$.



Marriage Theorem

Thm: [Frobenius 1917, Hall 1935] Let $G = (X \cup Y, E)$ be a bipartite graph with $|X| = |Y|$.

Then, G has a perfect matching iff $|N(S)| \geq |S|$ for all subsets $S \subseteq X$.

Pf. \Rightarrow

This was the previous observation.

If $|N(S)| < |S|$ for some S , then there is no perfect matching.

Marriage Theorem

Pf. $\exists S \subseteq X$ s.t., $|N(S)| < |S| \iff G$ does not have a perfect matching

Formulate as a max-flow and let (A, B) be the min s-t cut

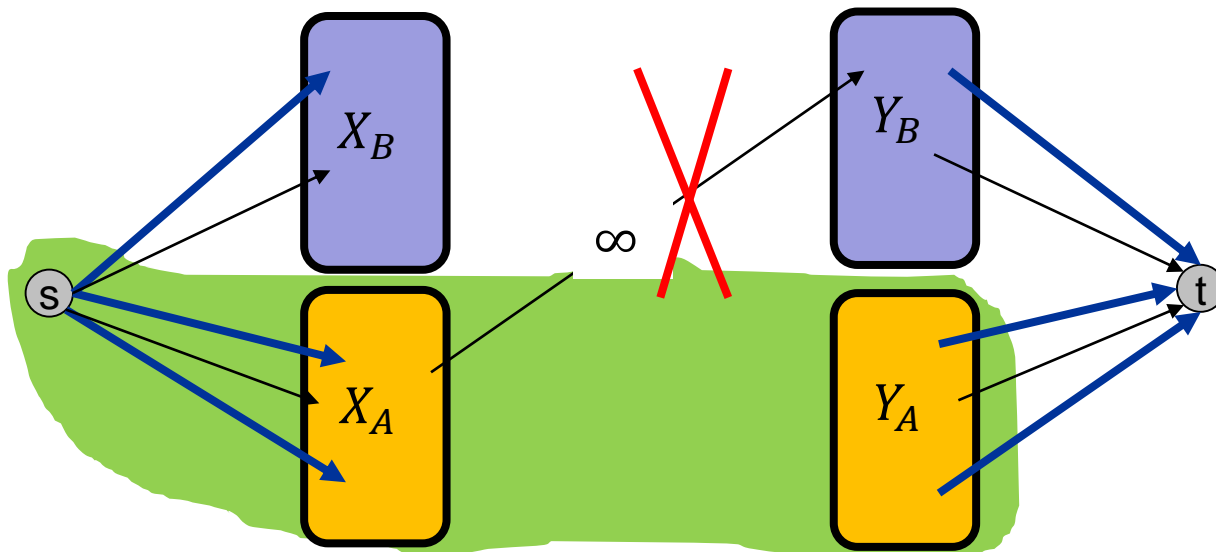
G has no perfect matching $\Rightarrow v(f^*) < |X|$. So, $cap(A, B) < |X|$

Define $X_A = X \cap A, X_B = X \cap B, Y_A = Y \cap A$

Then, $cap(A, B) \geq |X_B| + |Y_A|$

Since min-cut does not use ∞ edges, $N(X_A) \subseteq Y_A$

$|N(X_A)| \leq |Y_A| \leq cap(A, B) - |X_B| = cap(A, B) - |X| + |X_A| < |X_A|$



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]