

# Analysis of Algorithms, I

## CSOR W4231.002

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Network flows

- 1 Flow networks
  - Applications
- 2 The residual graph and augmenting paths
- 3 The Ford-Fulkerson algorithm for max flow
- 4 Correctness of the Ford-Fulkerson algorithm
- 5 Application: max bipartite matching

# Today

## 1 Flow networks

- Applications

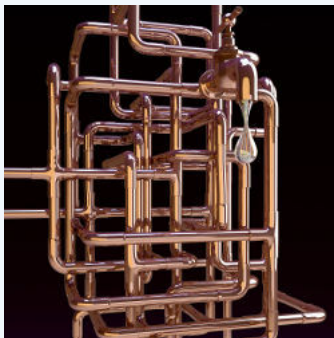
## 2 The residual graph and augmenting paths

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## 5 Application: max bipartite matching

# Modeling transportation networks



Source: Communications of the ACM, Vol. 57, No. 8

Can model a fluid network or a highway system by a **graph**:  
edges carry *traffic*, nodes are *switches* where traffic gets diverted.

# Flow networks

A flow network  $G = (V, E)$  is a directed graph such that

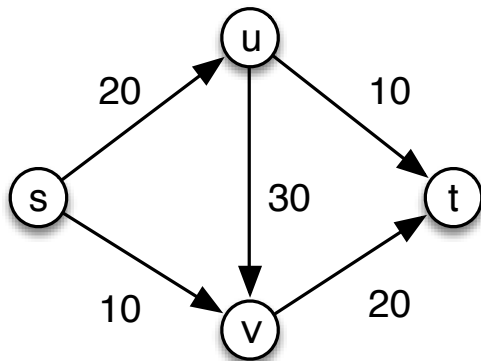
1. Every edge has a capacity  $c_e \geq 0$ .     *A1: integer capacities*
2. There is a single source  $s \in V$ .     *A2: no edge enters  $s$*
3. There is a single sink  $t \in V$ .     *A3: no edge leaves  $t$*

Two more assumptions for the purposes of the analysis

- ▶ *A4: If  $(u, v) \in E$  then  $(v, u) \notin E$ .*
- ▶ *A5: Every  $v \in V - \{s, t\}$  is on some  $s$ - $t$  path.*

Hence  $G$  has  $m \geq n - 1$  edges.

## An example flow network



Given a flow network  $G$ , an  $s$ - $t$  flow  $f$  in  $G$  is a function

$$f : E \rightarrow R^+.$$

Intuitively, the flow  $f(e)$  on edge  $e$  is the amount of *traffic* that edge  $e$  carries.

## Two kinds of constraints that every flow must satisfy

1. **Capacity constraints:** for all  $e \in E$ ,  $0 \leq f(e) \leq c_e$ .
2. **Flow conservation:** for all  $v \in V - \{s, t\}$ ,

$$\sum_{(u,v) \in E} f(u,v) = \sum_{(v,w) \in E} f(v,w) \quad (1)$$

In words, the flow **into** node  $v$  equals the flow **out of**  $v$ , or

$$\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$$



# A cleaner equation for flow conservation constraints

Define

$$1. f^{\text{out}}(v) = \sum_{e \text{ out of } v} f(e)$$

$$2. f^{\text{in}}(v) = \sum_{e \text{ into } v} f(e)$$

So we can rewrite equation (1) as: for all  $v \in V - \{s, t\}$

$$f^{\text{in}}(v) = f^{\text{out}}(v) \tag{2}$$

# The value of a flow

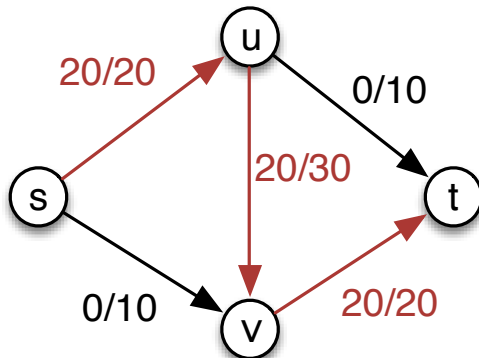
## Definition 1.

The **value** of a flow  $f$ , denoted by  $|f|$ , is

$$|f| = \sum_{e \text{ out of } s} f(e) = f^{\text{out}}(s)$$

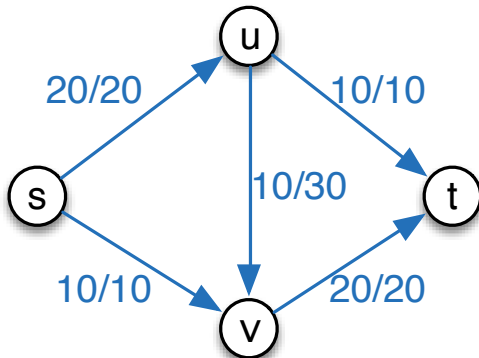
Exercise: show that  $|f| = f^{\text{in}}(t)$ .

## An example flow of value 20



A flow  $f$  of value 20.

## A flow of value 30



A (max) flow of value 30.

# Max flow problem

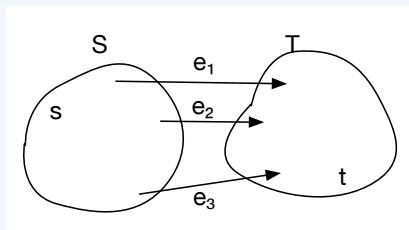
**Input:**  $(G, s, t, c)$  such that

- ▶  $G = (V, E)$  is a flow network;
- ▶  $s, t \in V$  are the source and sink respectively;
- ▶  $c$  is the (integer-valued) capacity function.

**Output:** a flow of maximum possible value

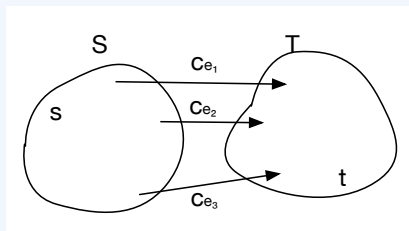
## Definition 2.

An  $s$ - $t$  cut  $(S, T)$  in  $G$  is a bipartition of the vertices into two sets  $S$  and  $T$ , such that  $s \in S$  and  $t \in T$ .



# A natural upper bound for the max value of a flow

- ▶ Flow  $f$  **must cross**  $(S, T)$  to go from source  $s$  to sink  $t$ .
- ▶ So it uses some (at most all) of the capacity of the edges crossing from  $S$  to  $T$ .



- ▶ So, intuitively, the value of the flow cannot exceed

$$\sum_{e \text{ out of } S} c_e$$

## Definition 3.

The capacity  $c(S, T)$  of an  $s$ - $t$  cut  $(S, T)$  is defined as

$$c(S, T) = \sum_{e \text{ out of } S} c_e.$$

△ Note asymmetry in the definition of  $c(S, T)$ !

So, *intuitively*, the value of the max flow is upper bounded by the capacity of *every* cut in the flow network, that is,

$$\max_f |f| \leq \min_{(S, T) \text{ cut in } G} c(S, T) \quad (3)$$



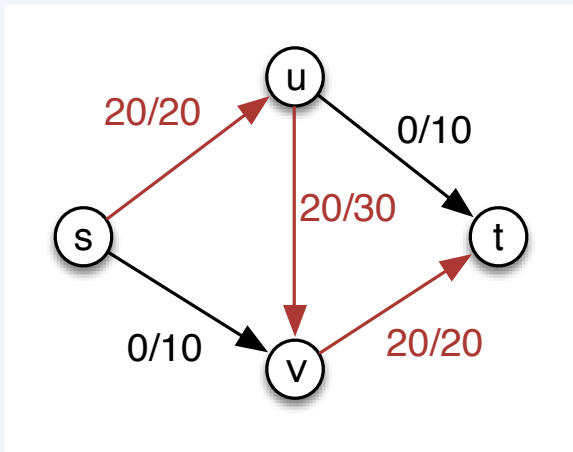
# Applications of max-flow and min-cut

- ▶ Find a set of edges of smallest capacity whose deletion disconnects the network (min cut)
- ▶ Bipartite matching (max flow) —*coming up*
- ▶ Airline scheduling (max flow)
- ▶ Baseball elimination (max flow)
- ▶ Distribution of goods to cities (max flow)
- ▶ Image segmentation (min cut)
- ▶ Survey design (max flow)
- ▶ ...

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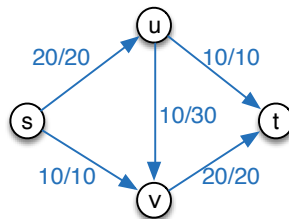
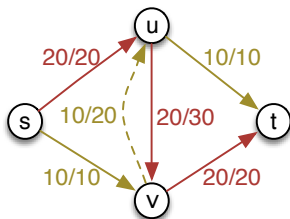
## “Undoing” flow



A flow  $f$  of value 20.

**Goal:** *undo* 10 units of flow along  $(u, v)$ , divert it along  $(u, t)$ .

# Pushing flow back



- ▶ “Push back” 10 units of flow along  $(v, u)$ .
- ▶ Send 10 more units from  $s$  to  $t$  along the green path edges  $(s, v)$ ,  $(v, u)$ ,  $(u, t)$ .
- ▶ New flow  $f'$  (on the right) with value 30.

# Pushing flow forward and backward

By pushing flow back on  $(v, u)$ , we created an  $s$ - $t$  path on which we are pushing flow

- ▶ **forward**, on edges with leftover capacity (e.g.,  $(s, v)$ );
- ▶ **backward**, on edges that are already carrying flow so as to divert it to a different direction (e.g.,  $(u, v)$ ).

# The residual graph $G_f$

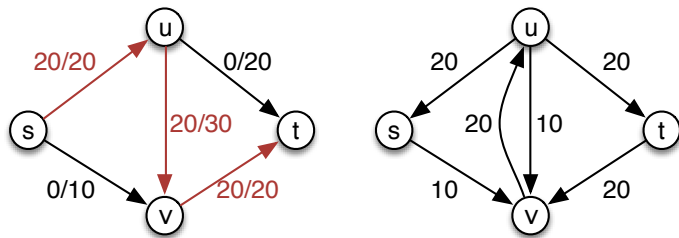
## Definition 4.

Given flow network  $G$  and flow  $f$ , the residual graph  $G_f$  has

- ▶ the **same vertices** as  $G$ ;
- ▶ for every edge  $e = (u, v) \in E$  with  $f(e) < c_e$ , an edge  $e = (u, v)$  with residual capacity  $c_f(e) = c_e - f(e)$  (**forward** edge);
- ▶ for every edge  $e = (u, v) \in E$  such that  $f(e) > 0$ , an edge  $e^r = (v, u)$  in  $G_f$  with residual capacity  $c_f(e^r) = f(e)$  (**backward** edge).

So  $G_f$  has  $\leq 2m$  edges and every  $e \in G_f$  has  $c_f(e) > 0$ .

## Example residual graph



Left: a graph  $G$  and a flow  $f$  of value 20.

Right: the residual graph  $G_f$  for flow network  $G$  and flow  $f$ .

# The residual graph $G_f$ as a roadmap for augmenting $f$

1. Let  $P$  be a **simple**  $s$ - $t$  path in  $G_f$ .
2. Augment  $f$  by pushing extra flow on  $P$ .

*Question: How much flow can we push on  $P$  without violating capacity constraints in  $G_f$ ?*



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## Definition 5.

The **bottleneck** capacity  $c(P)$  of a simple path  $P$  is the minimum residual capacity of **any** edge of  $P$ . In symbols

$$c(P) = \min_{e \in P} c_f(e).$$

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*Answer:* The max amount of flow we can safely push on **every** edge of  $P$  is  $c(P)$ .

# The augmented flow $f'$

Let  $P$  be an augmenting path in the residual graph  $G_f$ .

We obtain an **augmented flow**  $f'$  as follows:

1. For a **forward** edge  $e \in P$ , set  $f'(e) = f(e) + c(P)$
2. For a **backward** edge  $e^r = (u, v) \in P$ , let  $e = (v, u) \in G$ ;  
set  $f'(e) = f(e) - c(P)$
3. For  $e \in E$  but not in  $P$ ,  $f'(e) = f(e)$ .

**Claim 1.**

$f'$  is a flow.

## Pseudocode for subroutine Augment

**Input:** a flow  $f$ , and an augmenting path  $P$  in  $G_f$

**Output:** the augmented flow  $f'$

Augment( $f, P$ )

**for** each edge  $(u, v) \in P$  **do**

**if**  $e = (u, v)$  is a forward edge **then**

$f(e) = f(e) + c(P)$

**else**

$f(v, u) = f(v, u) - c(P)$

**end if**

**end for**

**return**  $f$

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# The Ford-Fulkerson algorithm

```
Ford-Fulkerson( $G = (V, E, c), s, t$ )  
  for all  $e \in E$  do  $f(e) = 0$   
  end for  
  while there is an  $s$ - $t$  path in  $G_f$  do  
    Let  $P$  be a simple  $s$ - $t$  path in  $G_f$   
     $f' = \text{Augment}(f, P)$   
    Set  $f = f'$   
    Set  $G_f = G_{f'}$   
  end while  
  Return  $f$ 
```

# Running time analysis

The algorithm **terminates** if the following facts are both true.

## Fact 6.

*Every iteration of the while loop returns a flow increased by an integer amount.*

## Fact 7.

*There is a finite upper bound to the flow.*

# Running time analysis

The algorithm **terminates** if the following facts are both true.

**Fact 6.**

*Every iteration of the while loop returns a flow increased by an integer amount.*

**Fact 7.**

*There is a finite upper bound to the flow.*

**Proof of Fact 7.**

Let  $U$  be the largest edge capacity, that is,  $U = \max_e c_e$ . Then

$$|f| \leq \sum_{e \text{ out of } s} c_e \leq nU.$$





$f$  increases by an integer amount after  $\text{Augment}(f, P)$

Proof of Fact 6.

It follows from the following claims.

**Claim 2.**

*During execution of the Ford-Fulkerson algorithm, the flow values  $\{f(e)\}$  and the residual capacities in  $G_f$  are all integers.*

**Claim 3.**

*Let  $f$  be a flow in  $G$  and  $P$  a simple  $s$ - $t$  path in  $G_f$  with residual capacity  $c(P) > 0$ . Then after  $\text{Augment}(f, P)$*

$$|f'| = |f| + c(P) \geq |f| + 1.$$



$f$  increases by an integer amount after  $\text{Augment}(f, P)$

### Proof of Claim 3.

Recall that  $|f| = f^{\text{out}}(s)$ .

1. Since  $P$  is an  $s$ - $t$  path, it contains some edge out of  $s$ , say  $(s, u)$ .
2. Since  $P$  is simple, it does not contain any edge entering  $s$  ( $P$  is in  $G_f$ , where there could be edges entering  $s$ !).
3. Since no edge enters  $s$  in  $G$ ,  $(s, u)$  is a forward edge in  $G_f$ , thus its augmented flow is  $f(s, u) + c(P) \geq f(s, u) + 1$ .
4. Since no other edge going out of  $s$  is updated, it follows that the value of  $f'$  is

$$|f'| = |f| + c(P) \geq |f| + 1.$$



# Running time of Ford-Fulkerson

1. Claim 3 guarantees at most  $nU$  iterations.
2. The running time of each iteration is bounded as follows:
  - ▶  $O(m + n)$  to create  $G_f$  using adjacency list representation.
  - ▶  $O(m + n)$  to run BFS or DFS to find the augmenting path.
  - ▶  $O(n)$  for  $\text{Augment}(f, P)$  since  $P$  has at most  $n - 1$  edges. $\Rightarrow$  Hence one iteration requires  $O(m)$  time.

The running time of Ford-Fulkerson is  $O(mnU)$ .

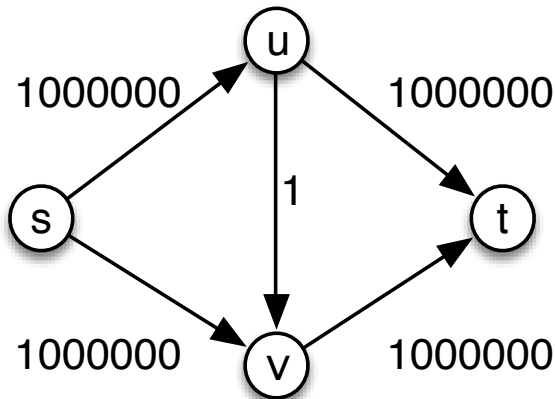
## Definition 8 (Pseudo-polynomial algorithms).

An algorithm is pseudo-polynomial if it is polynomial in the size of the input when the **numeric** part of the input is encoded in **unary**.

## Remark 1.

*Ford-Fulkerson is a **pseudo-polynomial** time algorithm.*

## Problems with pseudo-polynomial running times



# Improved algorithms

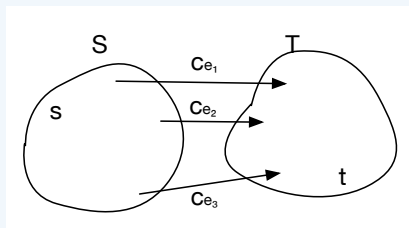
- ▶ FF can be made polynomial: use BFS instead of DFS
  - ▶ Edmonds-Karp:  $O(nm^2)$ , Dinitz:  $O(n^2m)$ ,  
other improvements:  $O(nm \log n)$ ,  $O(n^3)$
- ▶ **Unit** capacities:  $O(\min\{m^{3/2}, mn^{2/3}\})$  [EvenTarjan1975]
  - ▶ **Improved for sparse graphs:**  $\tilde{O}(m^{10/7})$  [Madry2013]
- ▶ **Integral** capacities:  $O(\min\{m^{3/2}, mn^{2/3}\} \log(n^2/m) \log U)$  [GoldbergRao1998]
  - ▶ **Improved:**  $\tilde{O}(m\sqrt{n} \log^2 U)$  [LeeSidfort2014];  
*also improves for dense graphs with unit capacities*
- ▶ **Real** capacities:  $O(nm \log(n^2/m))$ 
  - ▶ **Improved:**  $O(nm)$  [Orlin2013]

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# A natural upper bound for the max value of a flow

- ▶ An  $s$ - $t$  cut  $(S, T)$  in  $G$  is a **bipartition** of the vertices into two sets  $S$  and  $T$ , such that  $s \in S$  and  $t \in T$ .



- ▶ The **capacity**  $c(S, T)$  of  $s$ - $t$  cut  $(S, T)$  is  $\sum_{e \text{ out of } S} c_e$ .
- ▶ Then intuitively

$$\max_f |f| \leq \min_{(S, T) \text{ cut in } G} c(S, T) \quad (4)$$

# Roadmap for proving optimality of Ford-Fulkerson

Recall that  $|f| = f^{\text{out}}(s)$ . We will prove the following.

1. For any flow  $f$ , the value of the flow  $|f|$  cannot exceed the capacity of **any** cut in  $G$ .
2. Let  $f$  be the flow *upon termination* of the Ford-Fulkerson algorithm. We will exhibit a specific cut  $(S^*, T^*)$  such that the value of  $f$  equals the capacity of  $(S^*, T^*)$ . In symbols,

$$|f| = c(S^*, T^*)$$

From 1., 2., we can conclude that  $f$  is a maximum flow.

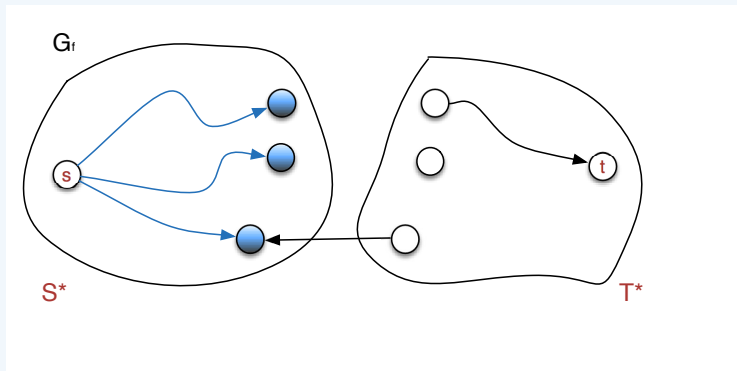
- And  $(S^*, T^*)$  is a cut of minimum capacity.



# Ford-Fulkerson terminates when $\nexists$ $s$ - $t$ path in $G_f$

Consider the residual graph  $G_f$  upon *termination* of the algorithm. Let  $(S^*, T^*)$  be the cut in  $G_f$  where

- ▶  $S^*$  is the set of nodes **reachable from the source  $s$** ;
- ▶  $T^*$  contains every other node.



*Is  $(S^*, T^*)$  also a cut in  $G$ ?*

## On the cut $(S^*, T^*)$

1.  $(S^*, T^*)$  is an  $s$ - $t$  cut: that is,  $s \in S^*$ ,  $t \in T^*$ . *Why?*
2. In  $G_f$ , no edge crosses from  $S^*$  to  $T^*$ . *Why?*
3. Hence, if  $e = (x, y) \in E$  with  $x \in S^*$  and  $y \in T^*$ , then  $f(e) = c_e$  (thus  $e \notin E_f$ ).
4. Similarly, if  $e' = (u, v) \in E$  with  $u \in T^*$  and  $v \in S^*$ , then  $f(e') = 0$ . *Why?*

## On the cut $(S^*, T^*)$

1.  $(S^*, T^*)$  is an  $s$ - $t$  cut: that is,  $s \in S^*$ ,  $t \in T^*$ . *Why?*

Because there is no  $s$ - $t$  path in  $G_f$ .

2. In  $G_f$ , no edge crosses from  $S^*$  to  $T^*$ . *Why?*

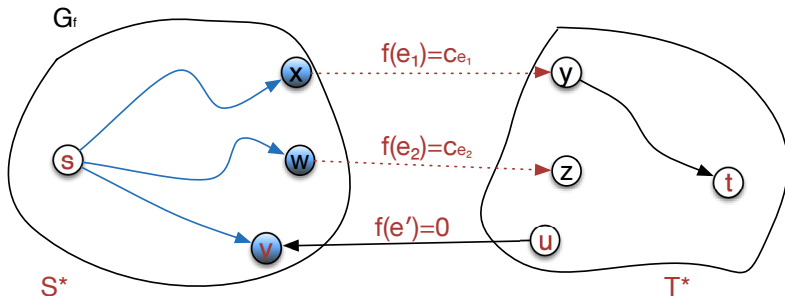
If  $(u, v)$  crosses from  $S^*$  to  $T^*$ , thus  $u \in S^*$ ,  $v \in T^*$ , then  $\exists$   $s$ - $v$  path in  $G_f$ . Hence  $v \in S^*$ ; contradiction.

3. Hence, if  $e = (x, y) \in E$  with  $x \in S^*$  and  $y \in T^*$ , then  $f(e) = c_e$  (thus  $e \notin E_f$ ).

4. Similarly, if  $e' = (u, v) \in E$  with  $u \in T^*$  and  $v \in S^*$ , then  $f(e') = 0$ . *Why?*

If  $f(e') > 0$ , then  $(v, u) \in E_f$ , with  $c_f(v, u) = f(e') > 0$ . Contradicts our second observation.

# Our observations on $(S^*, T^*)$ in a diagram



In  $G$ , every edge  $e$  crossing from  $S^*$  to  $T^*$  satisfies  $f(e)=c_e$  (of course, such  $e$  does not appear in  $G_f$ ).  
Every edge  $e'$  in  $G$  crossing from  $T^*$  to  $S^*$  satisfies  $f(e')=0$ .

# Net flow across a cut

## Definition 9.

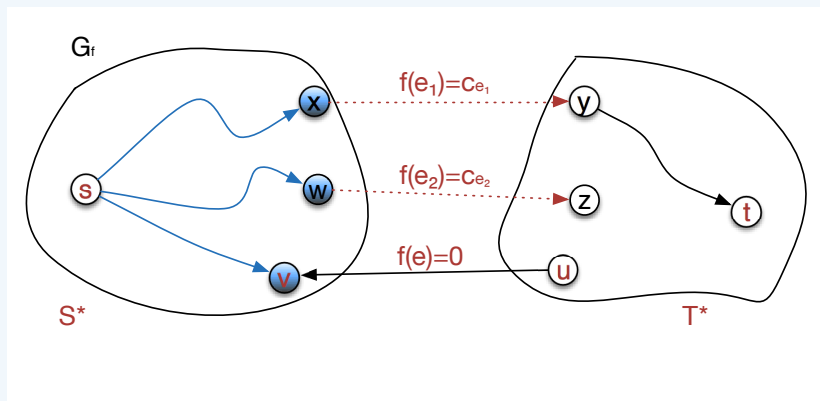
The **net flow** across an  $s$ - $t$  cut  $(S, T)$  is the amount of flow leaving the cut minus the amount of flow entering the cut

$$f^{\text{out}}(S) - f^{\text{in}}(S), \quad (5)$$

where

1.  $f^{\text{out}}(S) = \sum_{e \text{ out of } S} f(e)$
2.  $f^{\text{in}}(S) = \sum_{e \text{ into } S} f(e)$

Net flow across  $(S^*, T^*)$  equals capacity of  $(S^*, T^*)$



$$\begin{aligned}
 f^{\text{out}}(S^*) - f^{\text{in}}(S^*) &= \sum_{e \text{ out of } S^*} f(e) - \sum_{e \text{ into } S^*} f(e) \\
 &= \sum_{e \text{ out of } S^*} c_e - 0 \\
 &= c(S^*, T^*)
 \end{aligned} \tag{6}$$

# Roadmap revisited

Let  $f$  be the flow upon termination of the Ford-Fulkerson algorithm.

1. Exhibit a specific  $s$ - $t$  cut  $(S^*, T^*)$  in  $G$  such that the

$$|f| = c(S^*, T^*).$$

*Not quite there yet!*

- ▶ We exhibited  $(S^*, T^*)$  with *net flow* equal to its *capacity*.
- ▶ We need to relate the *net flow* across  $(S^*, T^*)$  to  $|f|$  (that is, the flow out of  $s$ ).
- ▶ In particular, **if we showed them equal**, then we'd have  $|f| = c(S^*, T^*)$ .

2. Show that  $|f|$  cannot exceed the capacity of **any** cut in  $G$ .
3. Conclude that  $f$  is a maximum flow.

net flow across any  $s$ - $t$  cut =  $|f|$

Recall that

$$\blacktriangleright f^{\text{out}}(S) = \sum_{e \text{ out of } S} f(e)$$

$$\blacktriangleright f^{\text{in}}(S) = \sum_{e \text{ into } S} f(e)$$

$$\blacktriangleright \text{net flow across } (S, T) \triangleq f^{\text{out}}(S) - f^{\text{in}}(S)$$

### Lemma 10.

*Let  $f$  be any  $s$ - $t$  flow, and  $(S, T)$  any  $s$ - $t$  cut. Then*

$$|f| = f^{\text{out}}(S) - f^{\text{in}}(S).$$



First, rewrite the flow out of  $s$  in terms of the flow on the vertices on  $S$ :

$$|f| = f^{\text{out}}(s) = \sum_{v \in S} \left( f^{\text{out}}(v) - f^{\text{in}}(v) \right) \quad (7)$$

since

- ▶  $f^{\text{in}}(s) = 0$ ;
- ▶ for every  $v \in S - \{s\}$ , the terms in the right-hand side of (7) cancel out because of flow conservation constraints.

Next, rewrite the right-hand side of equation 7 in terms of the *edges* that participate in these sums.

There are three types of edges.

## Proof of Lemma 10 (cont'd)

1. Edges with both endpoints in  $S$ : such edges appear once in the first sum in equation 7 and once in the second, hence their flows cancel out.
2. Edges with the tail in  $S$  and head in  $T$  (out of  $S$ ): such edges contribute to the first sum,  $\sum_{v \in S} f^{\text{out}}(v)$ , in equation 7 so they appear with a  $+$ .
3. Edges with the tail in  $T$  and head in  $S$  (into  $S$ ): such edges contribute to the second sum,  $\sum_{v \in S} f^{\text{in}}(v)$ , in equation 7 so they appear with a  $-$ .

In effect, the right-hand side of equation 7 becomes

$$\sum_{e \text{ out of } S} f(e) - \sum_{e \text{ into } S} f(e).$$

The lemma follows.

The value of a flow cannot exceed capacity of any cut

### Corollary 11.

*Let  $f$  be any  $s$ - $t$  flow and  $(S, T)$  any  $s$ - $t$  cut. Then*

$$|f| \leq c(S, T).$$

Proof.

$$|f| = f^{\text{out}}(S) - f^{\text{in}}(S) \leq f^{\text{out}}(S) \leq c(S, T).$$



# Putting everything together

- ▶ By Corollary 11, the value of a flow cannot exceed the capacity of any cut; in particular,

$$|f| \leq c(S^*, T^*).$$

- ▶ By Lemma 10,  $|f|$  equals the net flow across any  $s$ - $t$  cut; in particular,

$$|f| = f^{\text{out}}(S^*) - f^{\text{in}}(S^*).$$

- ▶ From (6), the net flow across  $(S^*, T^*)$  equals  $c(S^*, T^*)$ . Hence the above becomes

$$|f| = f^{\text{out}}(S^*) - f^{\text{in}}(S^*) = c(S^*, T^*).$$

- ⇒ Thus the flow computed by Ford-Fulkerson is a maximum flow because it cannot be increased anymore.

# The max-flow min-cut theorem

## Theorem 12.

*If  $f$  is an  $s$ - $t$  flow such that there is no  $s$ - $t$  path in  $G_f$ , then there is an  $s$ - $t$  cut  $(S^*, T^*)$  in  $G$  such that  $|f| = c(S^*, T^*)$ .  
Therefore,  $f$  is a max flow and  $(S^*, T^*)$  is a cut of min capacity.*

## Theorem 13 (Max-flow Min-cut).

*In every flow network, the maximum value of an  $s$ - $t$  flow equals the minimum capacity of an  $s$ - $t$  cut.*

# Integrality theorem

Recall the following claim.

## Claim 4.

During execution of the Ford-Fulkerson algorithm, the flow values  $\{f(e)\}$  and the residual capacities in  $G_f$  are **all integers**.

Combine with Theorem 12 to conclude:

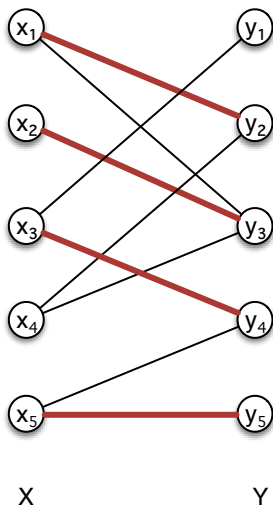
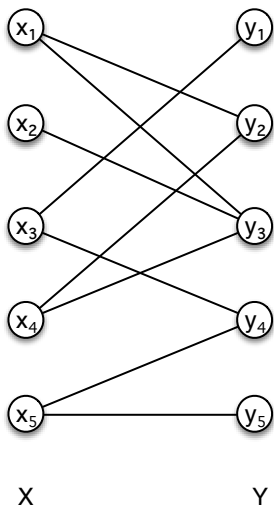
## Theorem 14 (Integrality theorem).

*If all capacities in a flow network are integers, then **there is** a maximum flow for which **every** flow value  $f(e)$  is an **integer**.*

# Today

- 1 Flow networks
  - Applications
- 2 The residual graph and augmenting paths
- 3 The Ford-Fulkerson algorithm for max flow
- 4 Correctness of the Ford-Fulkerson algorithm
- 5 Application: max bipartite matching**

# Bipartite Matching





## Definition 15.

A matching  $M$  is a **subset of edges** where every vertex in  $X \cup Y$  appears at most once.

Example:  $\{(x_1, y_2), (x_2, y_3), (x_3, y_4), (x_5, y_5)\}$  is a matching.

**Perfect** matching: every vertex in  $X \cup Y$  appears exactly **once**.

- ▶ Not always possible: e.g.,  $|X| \neq |Y|$ .

**Maximum** matching still desirable in applications.

- ▶ If we had an algorithm to find maximum matching then we could also find a perfect matching, if one exists (*why?*).

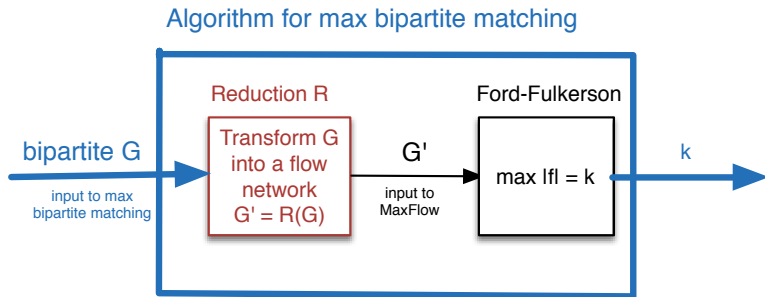
# Finding maximum matchings in bipartite graphs

**Idea:** Use the Ford-Fulkerson algorithm to find maximum (or perfect) matchings in bipartite graphs.

**Strategy:** reformulate the problem as a max flow problem which we know how to solve (reduction).

To this end, we need to transform our input bipartite graph into a flow network.

# A diagram of the algorithm for max bipartite matching



## Remark 2.

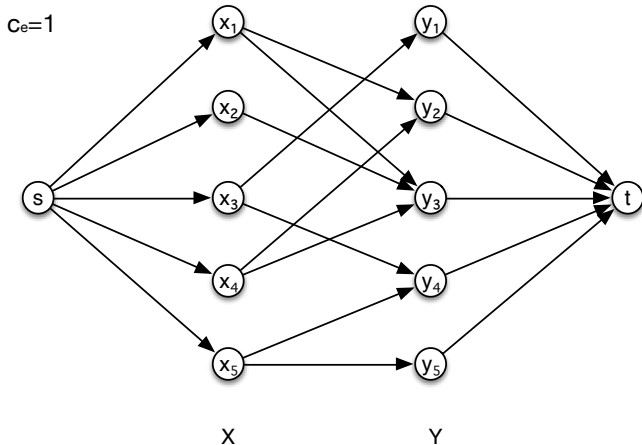
1. The reduction  $R$  must be efficient (polynomial in the size of  $G$ ).
2.  $G$  and  $G'$  should be equivalent, in the sense that  $G$  has a max matching of size  $k$  if and only if the max flow in  $G'$  has value  $k$ .

# Deriving a flow network given a bipartite graph

Given a bipartite graph  $G = (X \cup Y, E)$ , we construct a **flow network**  $G'$  as follows.

- ▶ Add a source  $s$ .
- ▶ Add a sink  $t$ .
- ▶ Add  $(s, x)$  edges for all  $x \in X$ .
- ▶ Add  $(y, t)$  edges for all  $y \in Y$ .
- ▶ Direct all  $e \in E$  from  $X$  to  $Y$ .
- ▶ Assign to every edge capacity of 1.

# The flow network for the example bipartite graph



# Computing matchings in $G$ from flows in $G'$

- ▶  $G = (X \cup Y, E)$  is the bipartite graph
- ▶  $G'$  is the derived flow network

## Claim 5.

*The size of the maximum matching in  $G$  equals the value of the maximum flow in  $G'$ . The edges of the matching are the edges that carry flow from  $X$  to  $Y$  in  $G'$ .*

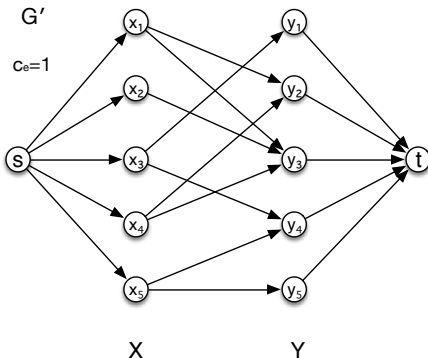
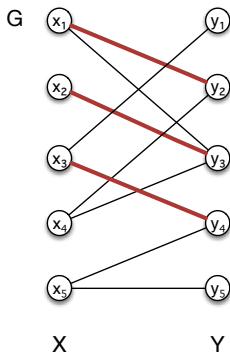
## Proof of Claim 5

The claim follows if we show the following two statements (*why?*).

1. ( $\Rightarrow$  **Forward direction**) For any matching  $M$  in  $G$ , we can construct a flow  $f$  in  $G'$  with value equal to the size of  $M$ , that is,  $|f| = |M|$ .
2. ( $\Leftarrow$  **Reverse direction**) Given a max flow  $f'$  in  $G'$ , we can construct a matching  $M'$  in  $G$ , with size equal to the value of the max flow, that is,  $|M'| = |f'|$ .

(1.  $\Rightarrow$ ) From a matching  $M$  to a flow  $f$  with  $|f| = |M|$

Let  $|M| = k$ .

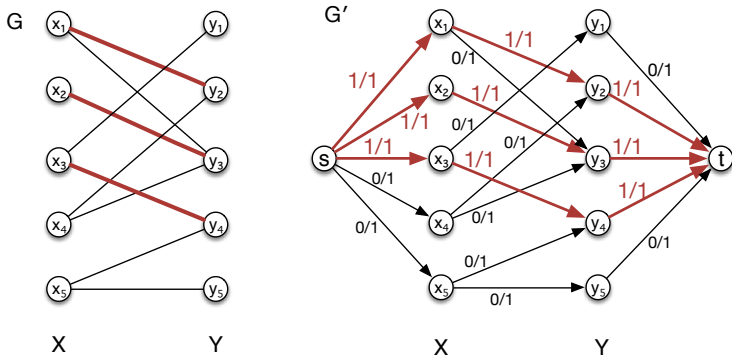


Given matching  $M$  (the red edges in  $G$ ), construct an integral flow  $f$  in  $G'$ , such that the value of  $f$  equals the number of edges in  $M$ .



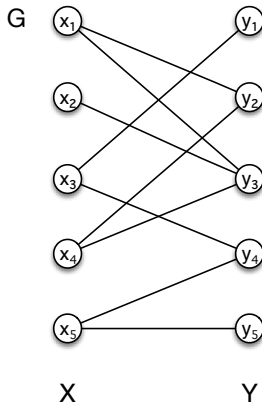
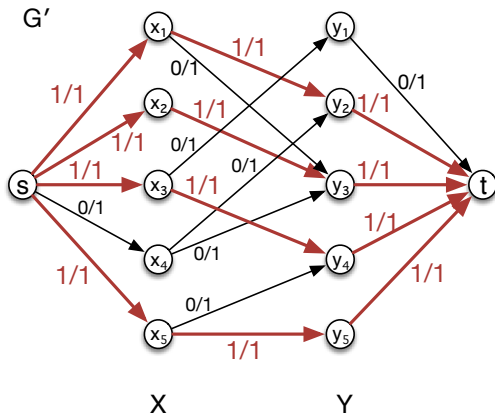
(1.  $\Rightarrow$ ) From a matching  $M$  to a flow  $f$  with  $|f| = |M|$

Let  $|M| = k$ . Send one unit of flow along each of the  $k$  edge-disjoint  $s$ - $t$  paths that use the edges in  $M$ ; then  $|f| = k$ .



Given matching  $M$  (the red edges in  $G$ ), construct the integral flow  $f$  in  $G'$ . Then the value of  $f$  equals the number of edges in  $M$ .

(2.  $\Leftarrow$ ) From max flow  $f'$  to  $M'$  with  $|M'| = |f'|$



Given integral max flow  $f'$  in  $G'$ , construct a matching  $M'$  in  $G$  so that the number of edges in  $M'$  equals the value of  $f'$ .

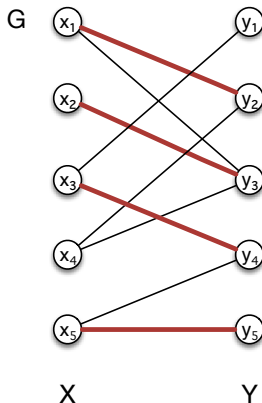
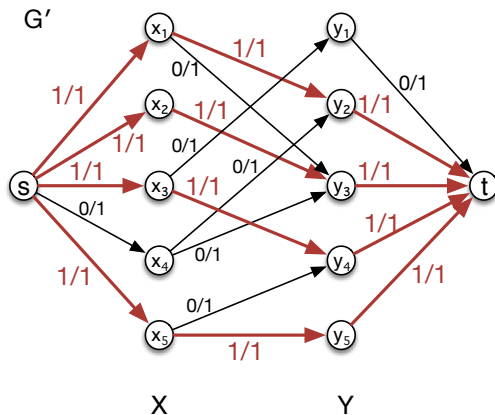
## From max flow $f'$ to matching $M'$ with $|M'| = |f'|$

Given a max flow  $f'$  in  $G'$  with  $|f'| = k$ , we want to select a set of edges  $M'$  in  $G$  so that  $M'$  is a matching of size  $k$ .

- ▶ By the integrality theorem, there is an **integer-valued** flow  $f$  of value  $k$ .
- ▶ Then for every edge  $e$ ,  $f(e) = 0$  or  $f(e) = 1$  (*why?*).
- ▶ Define the following matching  $M'$ :

$$M' = \left\{ e = (x, y) : x \in X, y \in Y \text{ and } f(e) = 1 \right\}.$$

# Obtaining a matching $M'$ from an integral flow $f$



Given integral flow  $f$  in  $G'$ , construct matching  $M'$  (the red edges in  $G$ ), so that the number of edges in  $M'$  equals the value of  $f$ .

$M'$  is a matching of size  $k$

We need to show the following two facts.

1. **Fact 1:**  $M'$  is a matching.
2. **Fact 2:**  $M'$  has size  $k$ .

### Proof of Fact 1.

Must show that every node in  $G'$  appears at most once in  $M'$ .

- ▶ Each node in  $X$  is the tail of at most one edge in  $M'$   
(*flow conservation constraints*).
- ▶ Each node in  $Y$  is the head of at most one edge in  $M'$   
(*flow conservation constraints*).



$M'$  has size  $k$

## Proof of Fact 2.

- ▶ Consider the cut  $(S, T)$  where  $S = \{s\} \cup X$ ,  $T = Y \cup \{t\}$ .
- ▶ We will compute its **net flow**.

1. By definition, the **net flow** of  $(S, T)$  is

$$f^{\text{out}}(S) - f^{\text{in}}(S) = |M'|$$

since

- ▶ the only edges that carry flow out of  $S$  are the edges in  $M'$ ;
  - ▶ the flow into  $S$  is 0 (no edges enter  $S$ ).
2. By Lemma 10, the **net flow** across  $(S, T)$  equals  $|f|$ ; hence **net flow** across  $(S, T) = k$ .

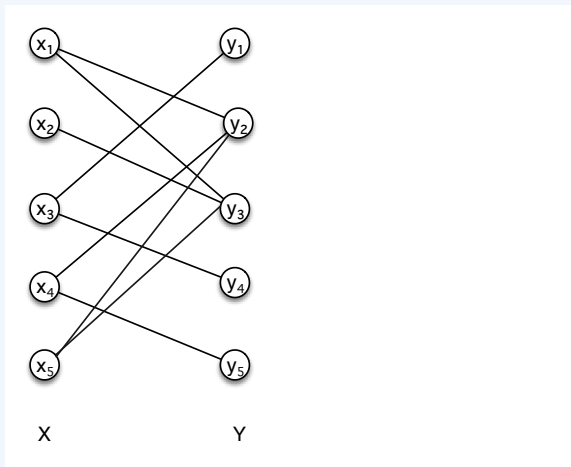
$\Rightarrow$  Thus  $|M'| = k$ .



# Time for finding max matching in bipartite graphs

1. Ford-Fulkerson:  $O(mnU) = O(mn)$
2. Improved:  $O(m\sqrt{n})$  [HopcroftKarp, Karzanov 1973]
3. Improved further for **sparse** ( $m = O(n)$ ) graphs:  
 $\tilde{O}(m^{10/7})$  [Madry2013]

# Perfect matchings in bipartite graphs with $|X| = |Y| = n$ (Hall's theorem)

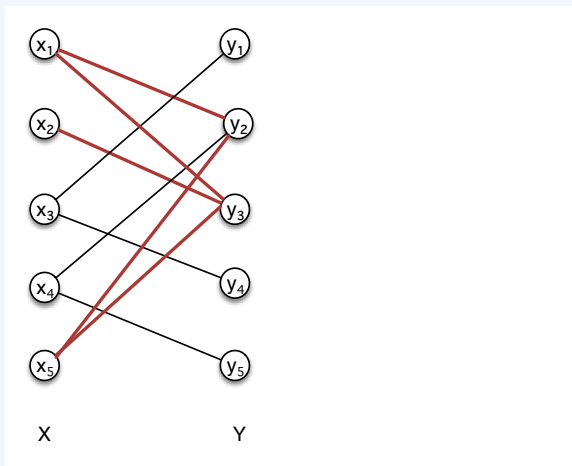


Is there a matching of size  $n$ , that is, a *perfect* matching in  $G$ ?



# A necessary condition for a perfect matching to exist

For every subset  $A$  of nodes in  $X$ , there are at least as many neighbors of  $A$  in  $Y$ . In symbols,  $\forall A \subseteq X, |N(A)| \geq |A|$ .

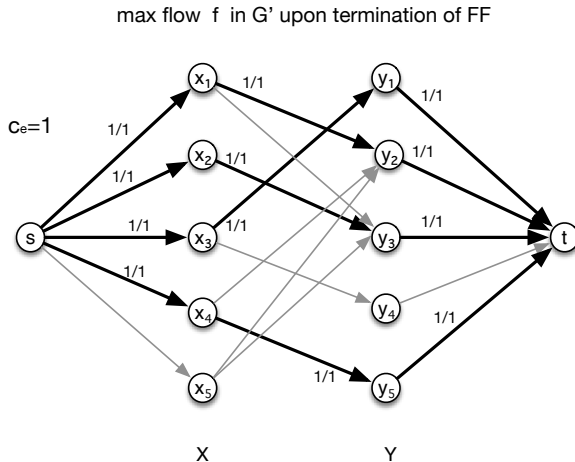


## *Is this a sufficient condition as well?*

*That is, if  $G$  does not have a perfect matching, is there always a subset  $A \subseteq X$  such that  $|N(A)| < |A|$ ?*

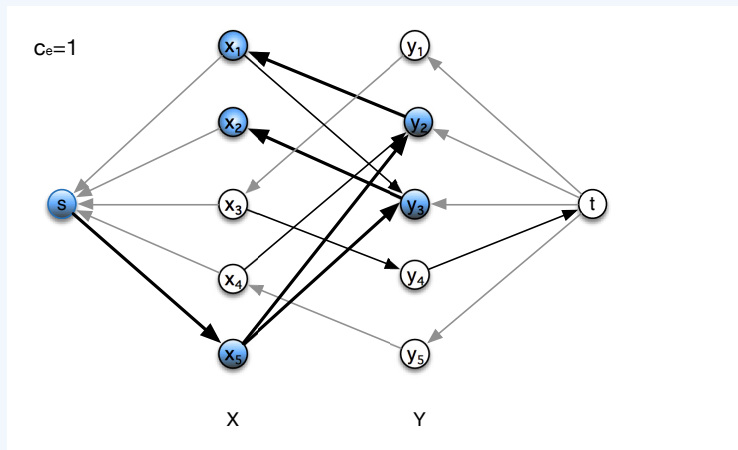
- ▶ Given bipartite  $G$ , transform it into a flow network  $G'$ .
- ▶ Run Ford-Fulkerson on  $G'$ .
- ▶ Assume  $\max |f| < n$ . We want to exhibit a set  $A$  as above.
- ▶ Since  $\max |f| < n$ , we know that  $\min_{(S,T)} c(S,T) = \max |f| < n$ .
- ▶ Consider the residual graph  $G_f$  upon termination of FF.
- ▶ Define the cut  $(S^*, T^*)$  as before, that is,  $S^*$  consists of all vertices reachable from  $s$  and  $T^*$  of everything else.
- ▶ We claim that the set  $A = S^* \cap X$  satisfies  $|A| > |N(A)|$ .

# A max flow in $G'$



# The residual graph upon termination of FF

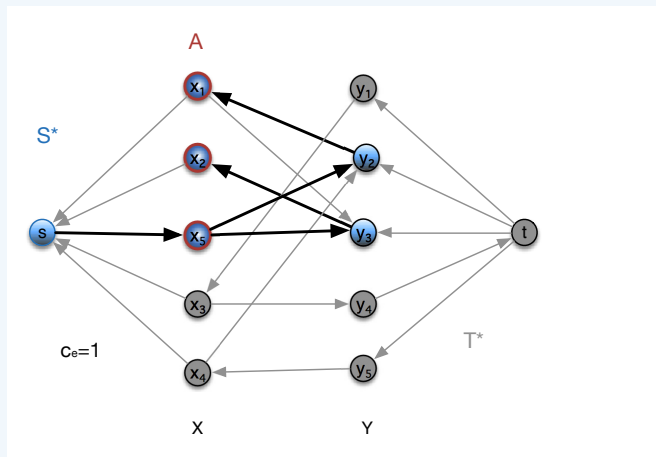
Define the cut  $(S^*, T^*)$  where  $S^*$  consists of all vertices reachable from  $s$  in  $G_f$  and  $T^*$  of everything else.



Here  $S^*$  consists of the blue vertices.

The set  $A$  that satisfies  $|A| > |N(A)|$

Clearly  $S^*$  consists of  $s$ , some vertices in  $X$  (*why?*) and possibly some vertices in  $Y$ . We set  $A = S^* \cap X$ .



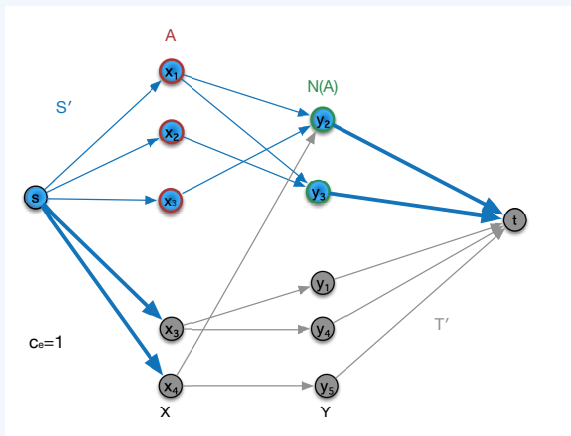
The residual graph slightly rearranged to group the vertices in  $A$  together.

# Augmenting $S^*$ to include all neighbors of $A$

Note that  $S^*$  might contain some neighbors of  $A$ .

Augment  $S^*$  by moving *all* neighbors of  $A$  into  $S^*$ .

Let  $S'$  be the resulting set,  $T' = V - S'$ .



Blue nodes belong to  $S'$ : among them, nodes with red contour are in  $A$  and nodes with green contour in  $N(A)$ . Silver nodes belong to  $T'$ .

## How does $c(S', T')$ compare to $c(S^*, T^*)$ ?

For every node  $u$  we move from  $T^*$  to  $S'$

- ▶ we add 1 to  $c(S^*, T^*)$  because of the edge  $(u, t)$  that now crosses from  $S'$  to  $T'$ ;
- ▶ we subtract at least 1 from  $c(S^*, T^*)$  because  $u$  is a neighbor of at least one node  $v \in A$ , hence the edge  $(v, u)$  no longer crosses from  $S^*$  to  $T^*$ .

Hence  $c(S', T') \leq c(S^*, T^*)$ .

Since  $(S^*, T^*)$  is a min cut, and  $\max |f| < n$ , we have

$$c(S', T') < n.$$

# The capacity of $(S', T')$

*What exactly is the capacity of  $(S', T')$ ?*

- ▶  $n - |A|$  edges cross from  $s \in S'$  to  $T'$
- ▶  $|N(A)|$  edges cross from  $S'$  to  $t \in T'$

Hence  $c(S', T') = n - |A| + |N(A)|$ .

Since  $c(S', T') < n$ , we have

$$n - |A| + |N(A)| < n.$$

Hence

$$|A| > |N(A)|.$$