

**Cliff Stein**

**Department of Industrial Engineering and Operations  
Research**

**Department of Computer Science  
School of Engineering and Applied Science  
Columbia University**

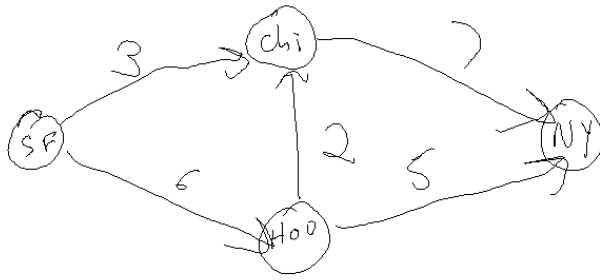
# Part 1: Network Flows

- Maximum Flows (in detail)
- Minimum Cost Flows
- Multicommodity Flows

## Internet Routing Example

Acme Routing Company wants to route traffic over the internet from San Francisco to New York. It owns some wires that go between San Francisco, Houston, Chicago and New York. The table below describes how many kilobytes can be routed on each wire in a second. Figure out a set of routes that maximizes the amount of traffic that goes from San Francisco to New York.

Cities	Maximum number of kbytes per second
S.F. - Chicago	3
S.F. - Houston	6
Houston - Chicago	2
Chicago - New York	7
Houston - New York	5



**One commodity, one source, one sink**

# Maximum Flows

- A **flow network**  $G = (V, E)$  is a directed graph in which each edge  $(u, v) \in E$  has a nonnegative **capacity** .
- If  $(u, v) \notin E$  , we assume that  $c(u, v) = 0$  .
- We distinguish two vertices in a flow network: a **source**  $s$  and a **sink**  $t$  .

A **flow** in  $G$  is a real-valued function  $f : V \times V \rightarrow \mathbb{R}$  that satisfies the following two properties:

**Capacity constraint:** For all  $u, v \in V$  , we require  $0 \leq f(u, v) \leq c(u, v)$  .

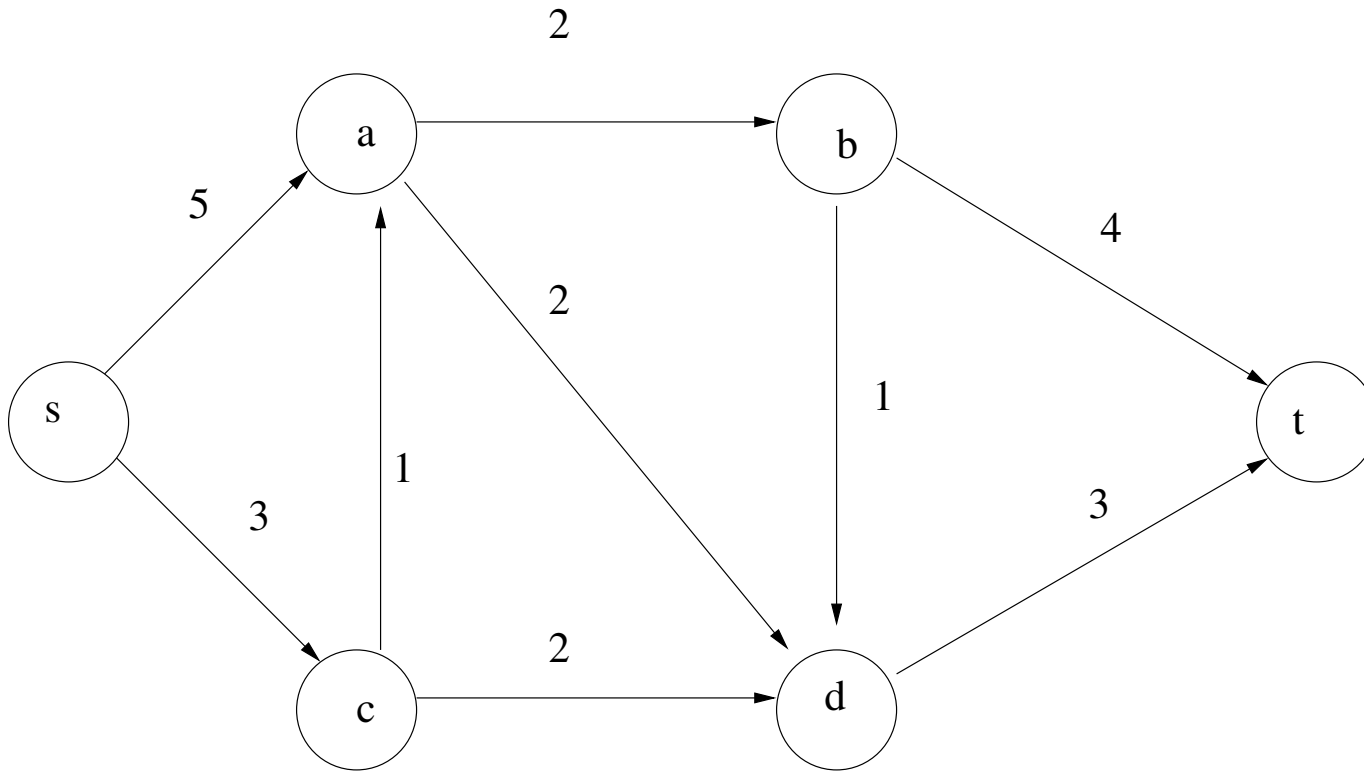
**Flow conservation:** For all  $u \in V - \{s, t\}$  , we require

$$\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v) .$$

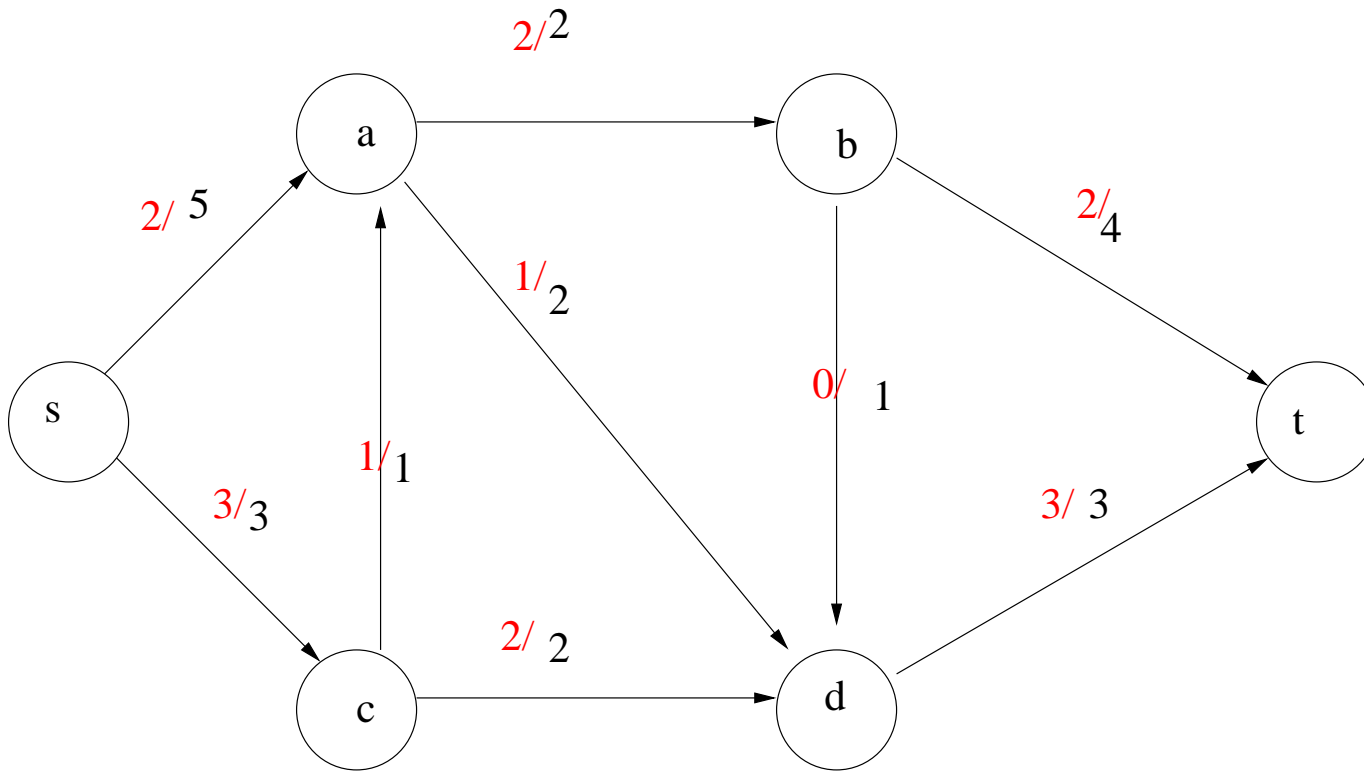
The **value** of a flow  $f$  is defined as

$$|f| = \sum_{v \in V} f(s, v) - \sum_{v \in V} f(v, s) , \tag{1}$$

# Example



# Solutions



# Internet Advertising



**CHRONIC BACK PAIN**

**DISCOVERY**

www.11.com/for/BackPain.com

**BREAKTHROUGH**

FDA CLEARED

NON-SURGICAL

PAIN RELIEF

[Click Here For More Info](#)

\*NO DOWN TIME

As by Google

Advertise on NYTimes.com

**Candidate's Platform: Jobs. Experience: N.B.A.**

By THOMAS KAPLAN



Chris Dudley campaigned in Beaverton, Ore., to become Oregon's next governor. Chris Dudley, who stands 6 feet 11 inches and played for the Nets and the Knicks during a 16-year career, stands out while campaigning to become Oregon's governor.

**No Sign of Favre at Camp, but the Vikings Don't Seem to Mind**

By PAT BORD

The players, coaches and fans in Minnetonka, Minn., for the Vikings' training camp seem certain that Brett Favre will be joining the team soon.

**Agency Role Could Limit Basketball Broker's Power**

By PETE THWILL

William Wesley's affiliation with Creative Artists Agency could curtail his basketball recruiting.

**Assuming Leading Role, Jets' Sanchez Acts the Part**

By GREG BISHOP

On the first day of his second season, Mark Sanchez moved with swagger in his step and spoke with urgency in his voice.

**In Guillen's Strong Words, Mets See a Nugget of Truth**

By DAVID WALDBSTEIN 18 minutes ago

Comments by White Sox Manager Ozzie Guillen about a lack of Spanish-language translators in baseball hit home for some Mets.

**SPORTS OF THE TIMES**

**Jets' Ryan Is Leaner and Says He's Learning**

By WILLIAM C. RHODEN

It was one thing for Jets Coach Rex Ryan to change his lifestyle; now he wants the rest of the team to follow suit.

**ON BASEBALL**

**Showalter Looks Up Again, From Far Below**

By TYLER KEEPER 16 minutes ago

Starting Tuesday, Buck Showalter begins the job of resurrecting the Baltimore Orioles, whose last winning season was 1997.

**SPORTS OF THE TIMES**

**The Cameras Are Rolling, and the Jets Expect to Be, Too**

By WILLIAM C. RHODEN

With the presence of HBO's "Hard Knocks" crew at training camp, the Jets' mission this season hasn't changed.

**Bats**

**Rodriguez's Waiting Game**

August 2, 2010 6:00 PM ET



**Showalter Leads the Charge**

August 2, 2010 5:59 PM ET

Want Instant Replay in Baseball? Try the Little League World Series

August 2, 2010 4:42 PM ET

Go to the Bats Blog »

**The Fifth Down**

**Jets' McKnight Quiets Fitness Concerns**



MLB | P.G.A. | L.P.G.A. | M.L.S. | A.T.P. | W.T.A.

**MLB Scoreboard**

AMERICAN LEAGUE

Toronto	8	Bot
NY Yankees	5	8th
Cleveland	6	Bot
Boston	2	7th
Minnesota	2	Bot
Tampa Bay	4	7th
Kansas City		10:05 ET
Oakland		
NATIONAL LEAGUE		
Cincinnati	3	Bot

**NEWS FROM AP & REUTERS**

**Roddick Finding Form With U.S. Open on The Horizon**

27 minutes ago

Sabres Agree to 2-Year Deal With D. Morrison

40 minutes ago

Osborne: Move to Big Ten About Stability

40 minutes ago

Cowboys Hurt Rookie Bryant Skill Catcher Balls

41 minutes ago

NBA Surpasses 2 Million Followers on Twitter

50 minutes ago

Vandy Drops Interim From Caldwell Title

8:03 p.m.

**Time to rebuild your credit?**

**Orchard Bank® Visa® Classic.**

**It's your choice. It's your credit.**

**Pre-qualify now.**

# Availability Queries

**Given:**

- Forecasts on ad impressions for each publisher's slot
- A reservation request

**Compute:** The maximum number of reservations that can be satisfied from the reservation request.

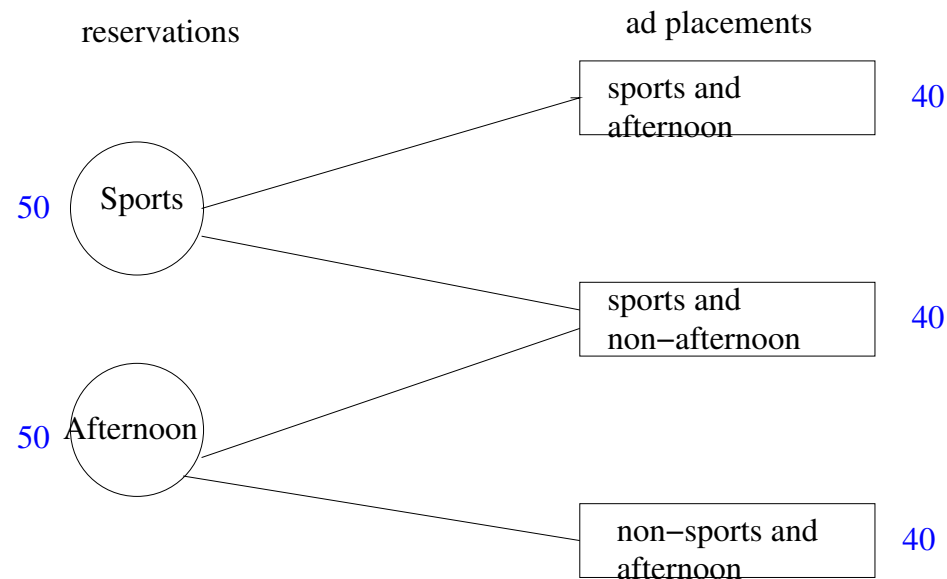
# Example

An advertiser wants:

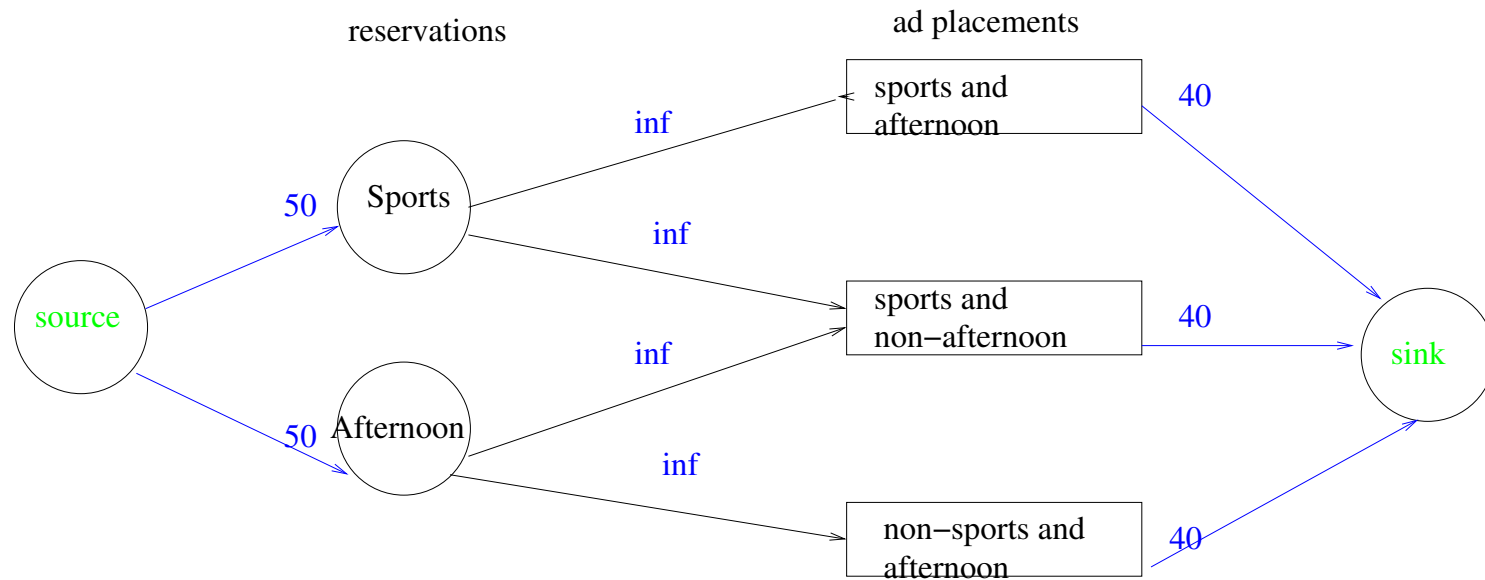
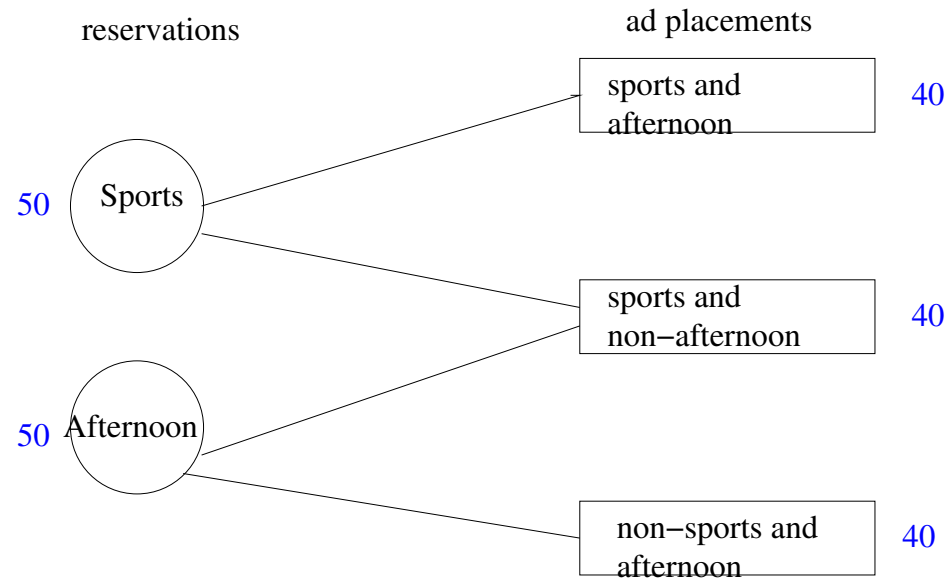
- 50 placements on sports slots
- 50 placements on afternoon slots

Publisher is offering

- 40 placements on slots that are sports and non-afternoon
- 40 placements on slots that are sports and afternoon
- 40 placements on slots that are non-sports and afternoon



# Example is a max flow problem



# Algorithm: Ford Fulkerson

Greedy send flow from source to sink.

Ford-Fulkerson-Method  $(G, s, t)$

- 1 initialize flow  $f$  to 0
- 2 while there exists an augmenting path  $p$
- 3     augment flow  $f$  along  $p$
- 4 return  $f$

For this to work, we need a notion of a **residual graph**

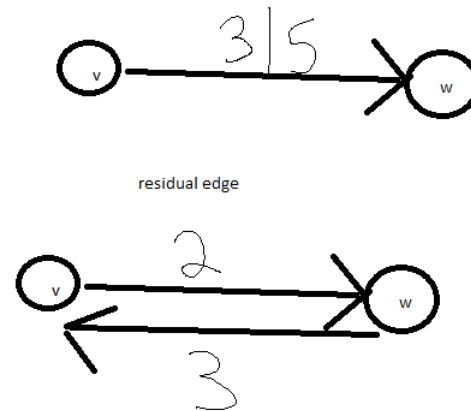
# Residual Graph

The residual graph is the graph of edges on which it is possible to push flow from source to sink.

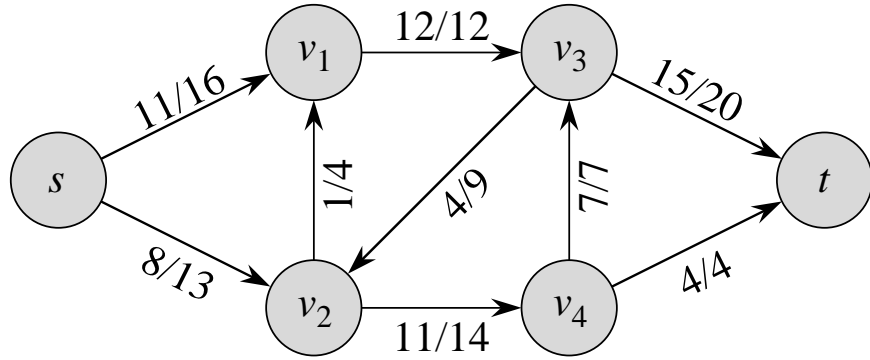
- The **residual capacity** of  $(u, v)$ , is

$$c_f(u, v) = \begin{cases} c(u, v) - f(u, v) & \text{if } (u, v) \in E, \\ f(v, u) & \text{if } (v, u) \in E, \\ 0 & \text{otherwise.} \end{cases} \quad (2)$$

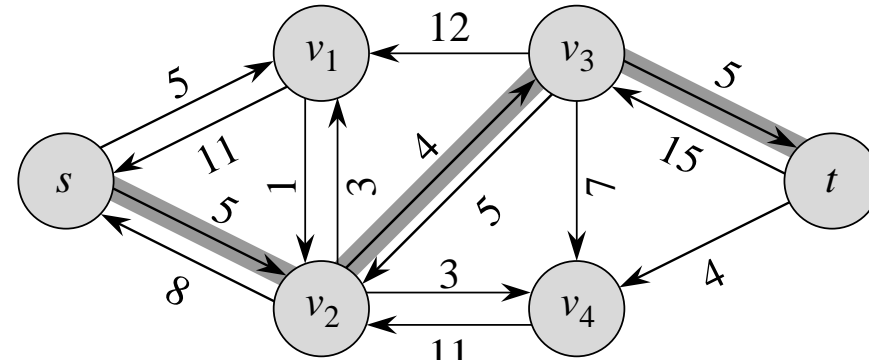
- The residual graph  $G_f$  is the graph consisting of edges with positive residual capacity



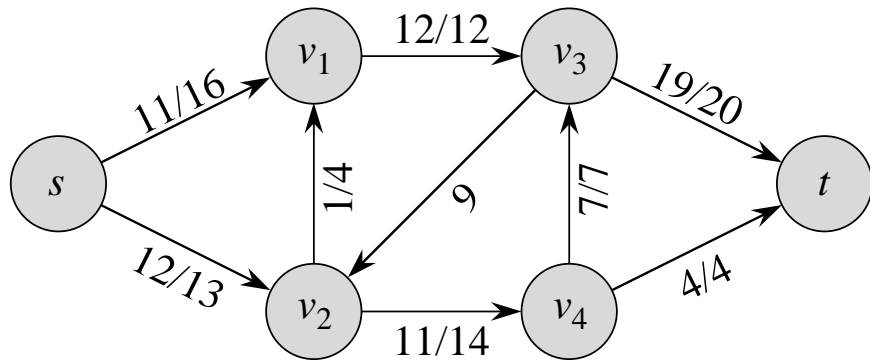
# Residual Network



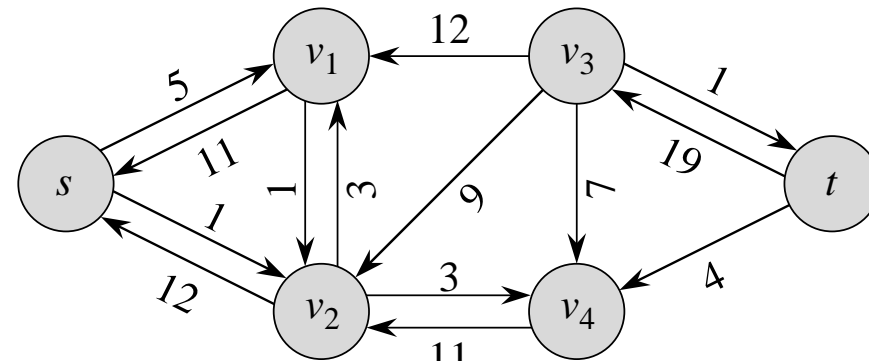
(a)



(b)



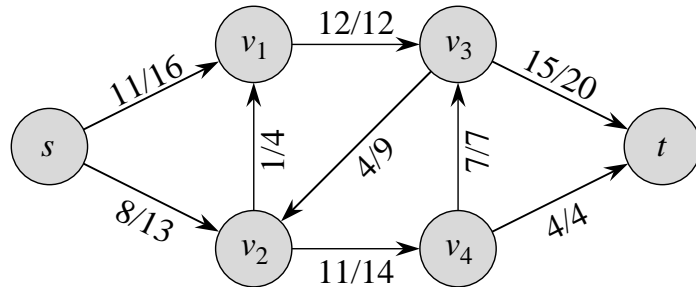
(c)



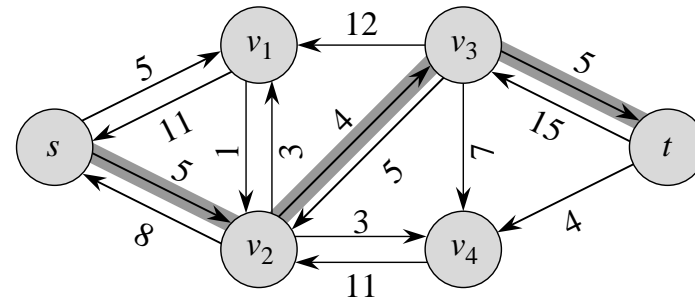
(d)

# Updating a Flow

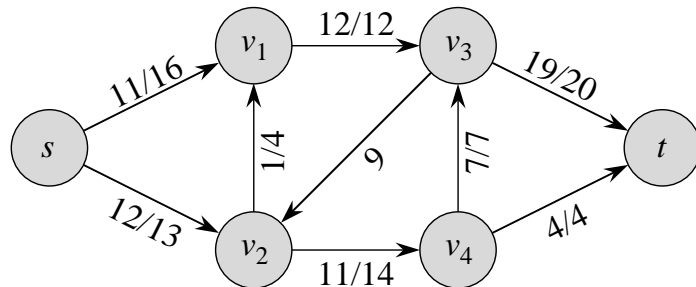
- Send flow along the path defined by the residual graph.
- Amount: minimum of capacity of all residual edges in the augmenting path.
- If a residual edge is a graph edge, then **add** the flow.
- If a residual edge is a reverse edge, then **subtract** the flow.



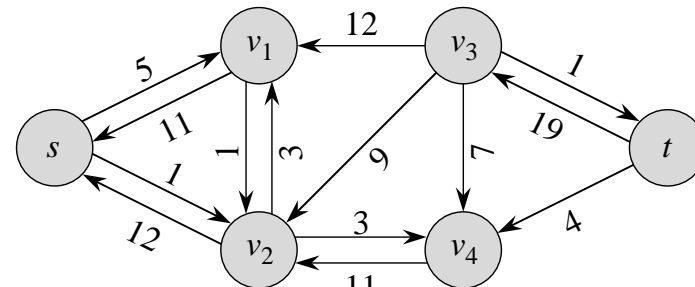
(a)



(b)



(c)



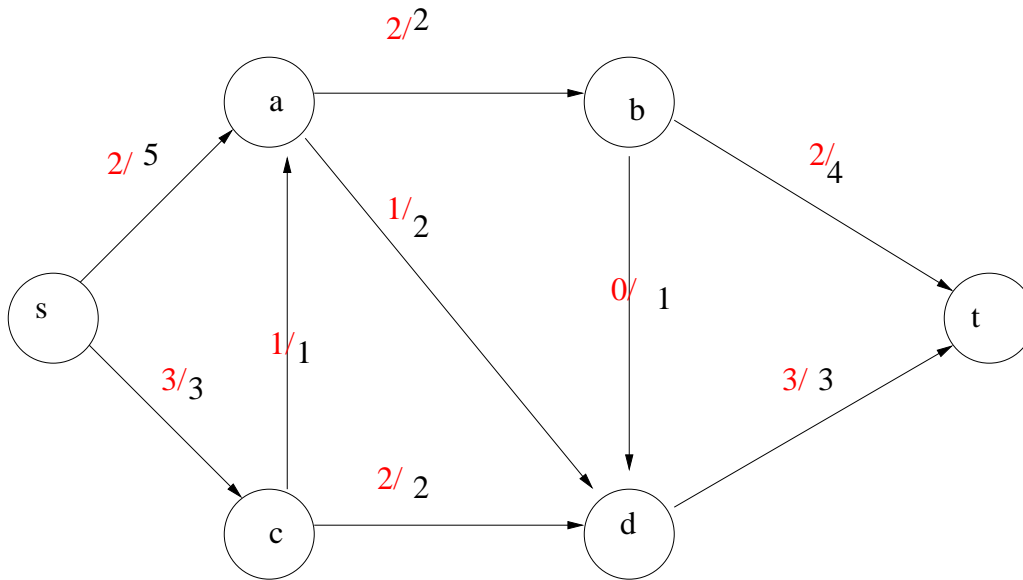
(d)



## $s - t$ Cuts

An  $s - t$  cut satisfies

- $s \in S$  ,  $t \in T$
- $S \cup T = V$  ,  $S \cap T = \emptyset$



**Capacity of a cut** (only forward edges)

$$c(S, T) = \sum_{u \in S, v \in T} c(u, v)$$

**Flow crossing a cut** (net flow)

$$(S, T) = \sum_{u \in S, v \in T} f(u, v) - \sum_{u \in T, v \in S} f(u, v)$$

# Properties of cuts and flows

**Capacity of a cut** (only forward edges)

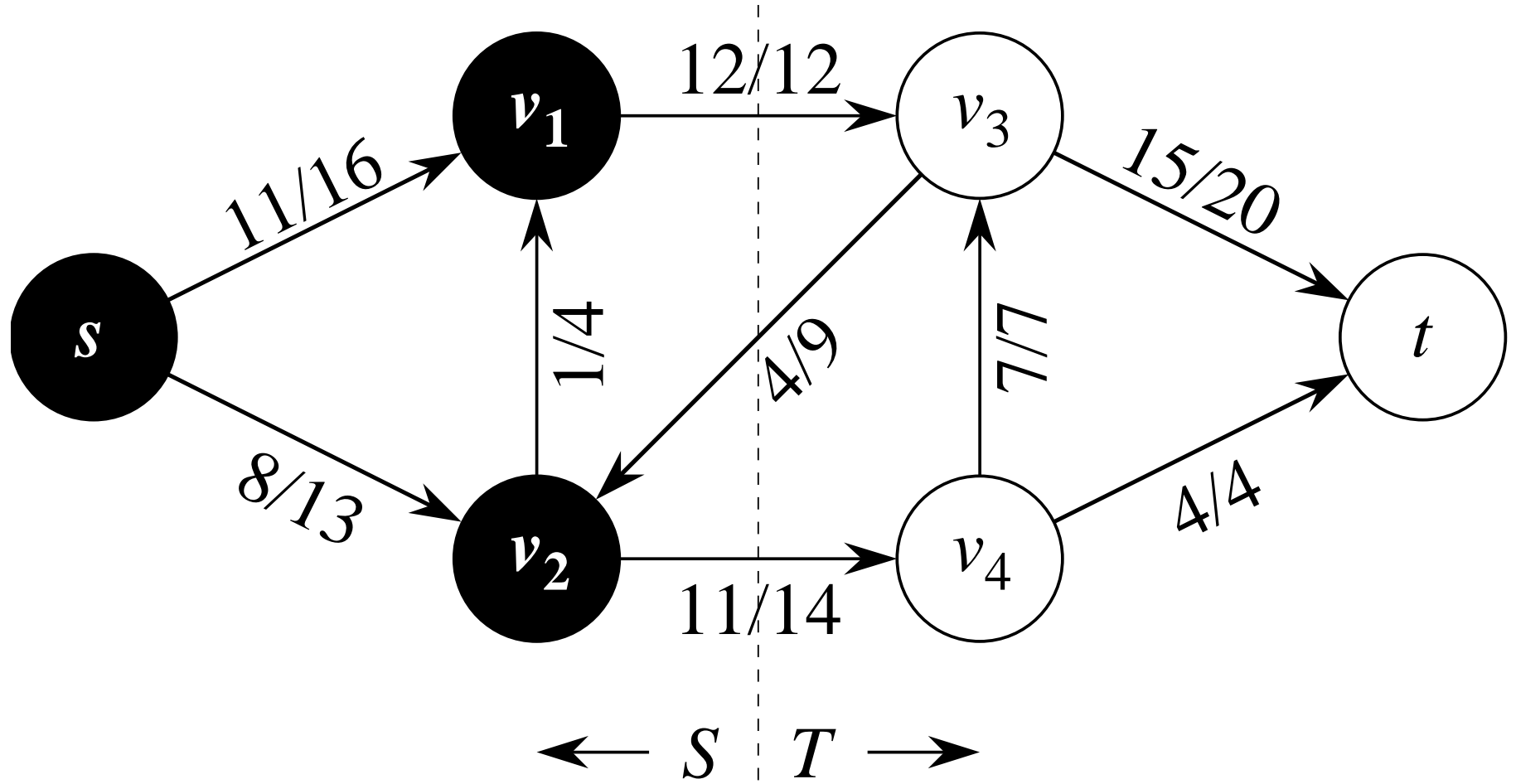
$$c(S, T) = \sum_{u \in S, v \in T} c(u, v)$$

**Flow crossing a cut** (net flow)

$$f(S, T) = \sum_{u \in S, v \in T} f(u, v) - \sum_{u \in T, v \in S} f(u, v)$$

- For all cuts  $(S, T)$  and all feasible flows  $f$ ,  $f(S, T) \leq c(S, T)$  (weak duality).
- For all pairs of cuts  $(S_1, T_1)$  and  $(S_2, T_2)$ , and all feasible flows  $f$ ,  $f(S_1, T_1) = f(S_2, T_2)$ .

## Examples of cuts



## Max-flow min-cut theorem

If  $f$  is a flow in a flow network  $G = (V, E)$  with source  $s$  and sink  $t$ , then the following conditions are equivalent:

1.  $f$  is a maximum flow in  $G$ .
2. The residual network  $G_f$  contains no augmenting paths.
3.  $|f| = c(S, T)$  for some cut  $(S, T)$  of  $G$ .

# Proof

# Ford Fulkerson expanded

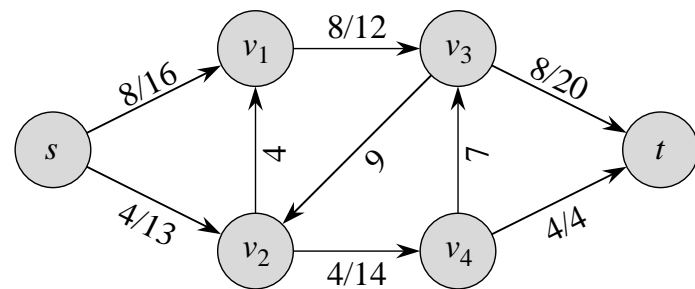
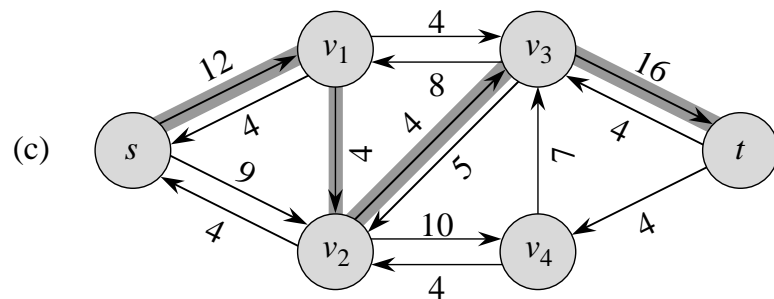
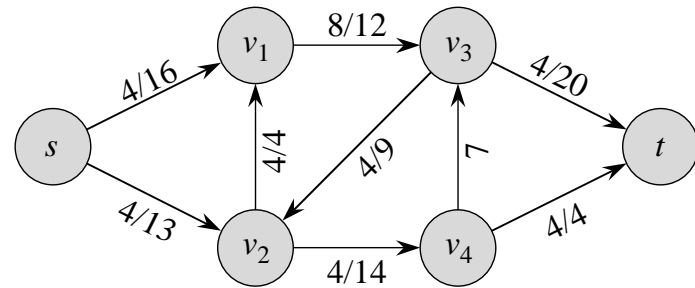
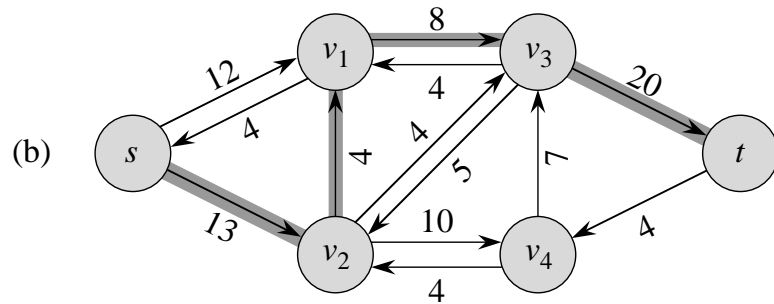
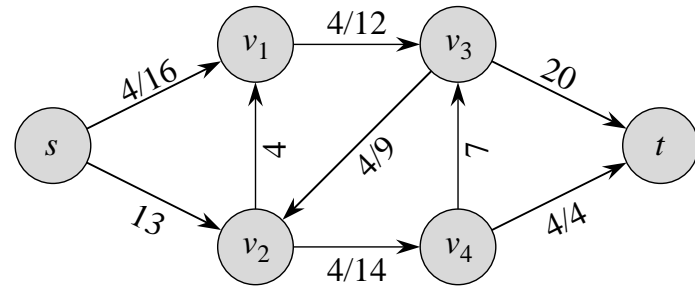
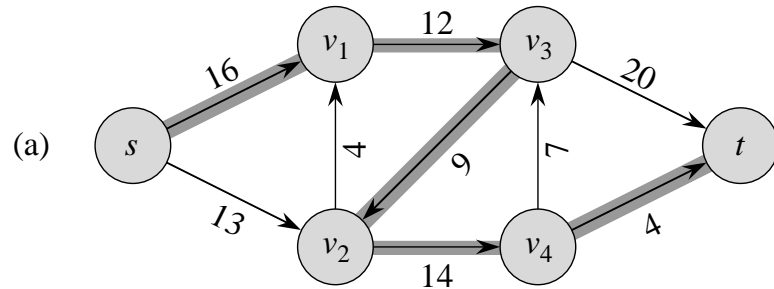
*Ford – Fulkerson*( $G, s, t$ )

```
1  for each edge  $(u, v) \in G.E$ 
2       $(u, v).f = 0$ 
3  while there exists a path  $p$  from  $s$  to  $t$  in the residual network  $G_f$ 
4       $c_f(p) = \min\{c_f(u, v) : (u, v) \text{ is in } p\}$ 
5      for each edge  $(u, v)$  in  $p$ 
6          if  $(u, v) \in E$ 
7               $(u, v).f = (u, v).f + c_f(p)$ 
8          else  $(v, u).f = (v, u).f - c_f(p)$ 
```

# Algorithm

Residual graph (Capacities)

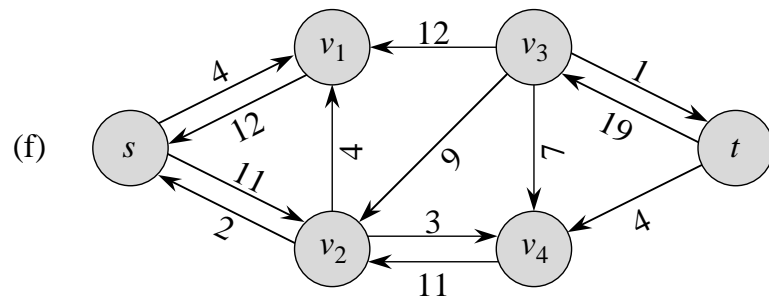
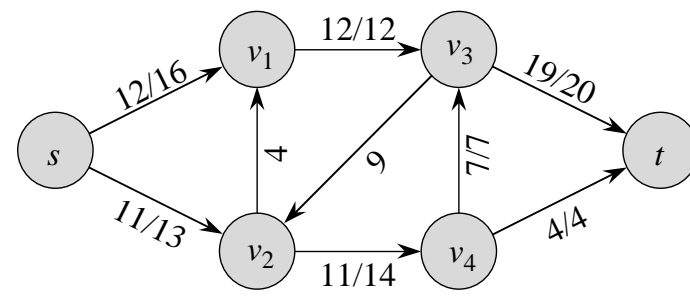
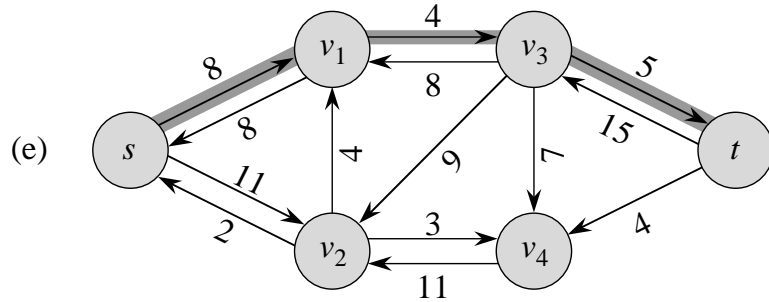
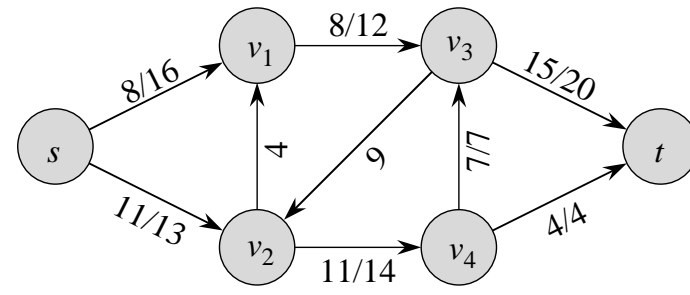
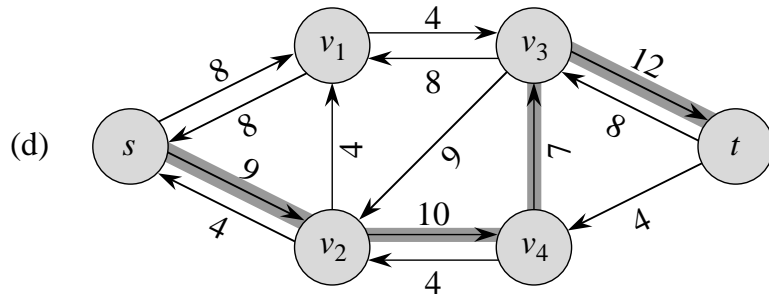
Graph (Flow/Capacity)



# Algorithm continued

Residual graph (Capacities)

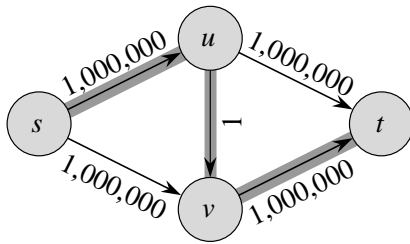
Graph (Flow/Capacity)



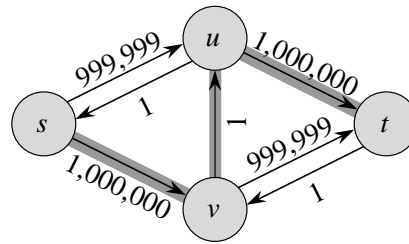


# Analysis

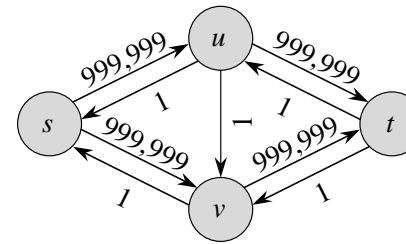
- 1 iteration of FF takes  $O(E + V)$  time (breadth-first search plus book-keeping).
- Each iteration sends at least one unit of flow.
- Total time  $O(f^*E)$ .
- This algorithm is only pseudo-polynomial.



(a)



(b)



(c)

# A polynomial algorithm– Shortest Augmenting Path

Algorithm due to Edmonds and Karp, Dinic

## Algorithm

- Run Ford-Fulkerson, but always choose the shortest augmenting path in the residual graph.
- Breadth-first search implements this algorithm.

# Idea for Analysis

## Intuition

- Augment along shortest path in residual graph.
- Short paths get “saturated” and disappear.
- Future augmenting paths are along longer paths
- Eventually algorithm terminates because no more paths exist (remember that no augmenting path can have length greater than  $V$ ).

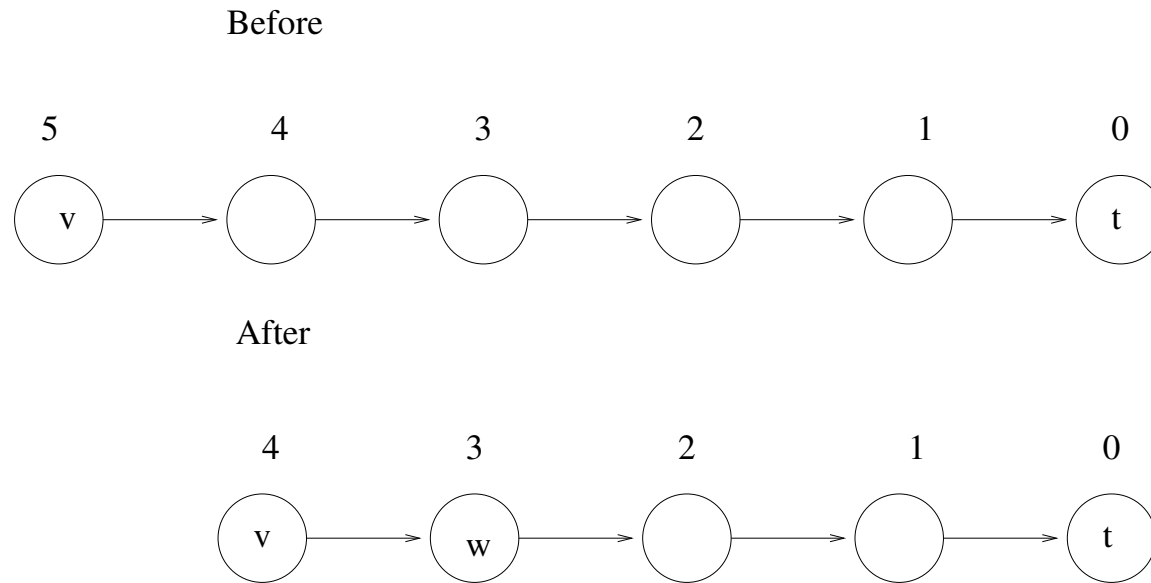
This doesn't quite work, but is close

# Analysis

- Let  $\delta(v)$  be the shortest path distance from  $v$  to  $t$  in  $G_f$ .
- Lemma:  $\delta(v)$  is monotonically increasing over time.

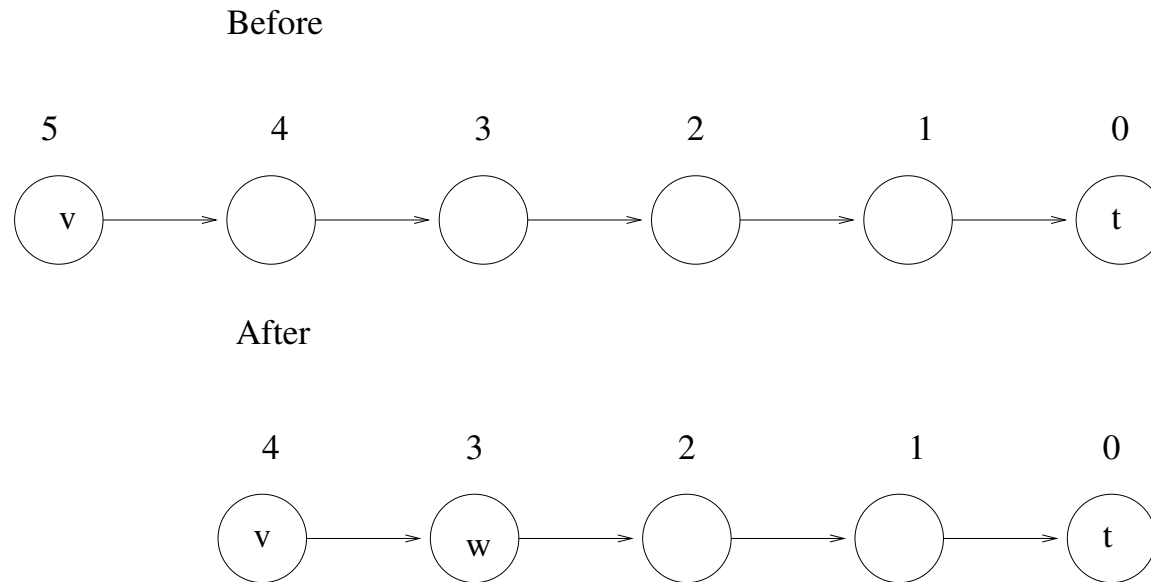
## Proof:

- Let  $\delta$  be shortest path distances before an augmentation.
- Let  $\delta'$  be shortest path distances after an augmentation.
- Suppose, f.p.o.c. that for some  $v$ ,  $\delta'(v) > \delta(v)$ , and  $v$  is the minimum vertex with this property (w.r.t  $\delta'$ )



How could this happen?

# How could this happen?



- $(v, w) \in G'_f$  but not in  $G_f$ . (why?)
- We must have sent flow on  $(w, v)$  in the augmenting path
- $\delta(w) = 6$
- $\delta'(w) < \delta(w)$ , which contradicts  $v$  being the minimum vertex whose label decreased.

**Conclusion:**  $\delta(v)$  never decreases.

## Analysis Continued

- Each  $\delta(v)$  increases at most  $V$  times.
- Total number of times any  $\delta(v)$  increases is at most  $V^2$ .
- We need to tie this to an augmenting path.
- **Problem:** A particular augmenting path may not increase any  $\delta(v)$ .  
(How can this happen?)

**Solution** We need to find some other event that happens on every augmenting path and then relate this event to increasing distances.

**Conclusion:**  $\delta(v)$  never decreases.

# Analysis Continued

- $0 \leq \delta(v) \leq V$
- $\delta(v)$  never decreases.
- Each  $\delta(v)$  increases at most  $V$  times.
- Total number of times some  $\delta(v)$  increases is at most  $V^2$ .
- We need to tie this to an augmenting path.
- **Problem:** A particular augmenting path may not increase any  $\delta(v)$ .  
(How can this happen?)

**Solution** We need to find some other event that happens on every augmenting path and then relate this event to increasing distances.

## Edge Saturation

- If an augmenting path send  $c_f(v, w)$  units of flow on edge  $(v, w)$  then we call the push **saturating**.
- Every augmenting path saturates at least one edge. (Why?)

# Analysis Continued

## Edge Saturation

- If an augmenting path send  $c_f(v, w)$  units of flow on edge  $(v, w)$  then we call the push **saturating**.
- Every augmenting path saturates at least one edge.

## Focus on a particular pair of vertices $(v, w)$

- Suppose that  $(v, w) \in G_f$
- What has to happen between two consecutive saturations of  $(v, w)$



# Events Between 2 edge saturations



saturate ( $v,w$ )



saturate ( $w,v$ )



saturate ( $v,w$ )

- Both  $v$  and  $w$  have to be relabeled.
- So each edge can be saturated at most  $V/2$  times.

## Putting it together

- Each edge (or its reverse) can be saturated at most  $V/2$  times.
- There are most  $EV$  edge saturations.
- Each augmenting causes at least one edge saturation
- Therefore, there are at most  $EV$  augmenting paths.
- Recall  $O(E)$  time per augmenting path.

**A max flow can be found in  $O(E^2V)$  time .**

Faster augmenting path based algorithms exist, most based on finding the augmenting paths more efficiently.

- $O(VE \log V)$  (Sleator and Tarjan)
- $O(\min E^{3/2}, V^{2/3}E \log(V^2/E) \log C)$  (Goldberg and Rao)

# Push relabel algorithms

## Algorithm due to Goldberg and Tarjan

### Ideas

- Maintain variables  $d(v)$  which are “distances”, estimates (lower bounds) on the distance to the sink (or source) in the residual graph
- Push flow over one edge at a time, pushing from a higher distance vertex to a lower distance vertex.
- Allow **excess** flow to accumulate at a vertex.
- Maintain a **preflow** rather than a flow.

Capacity constraint: For all  $u, v \in V$ , we require  $0 \leq f(u, v) \leq c(u, v)$ .

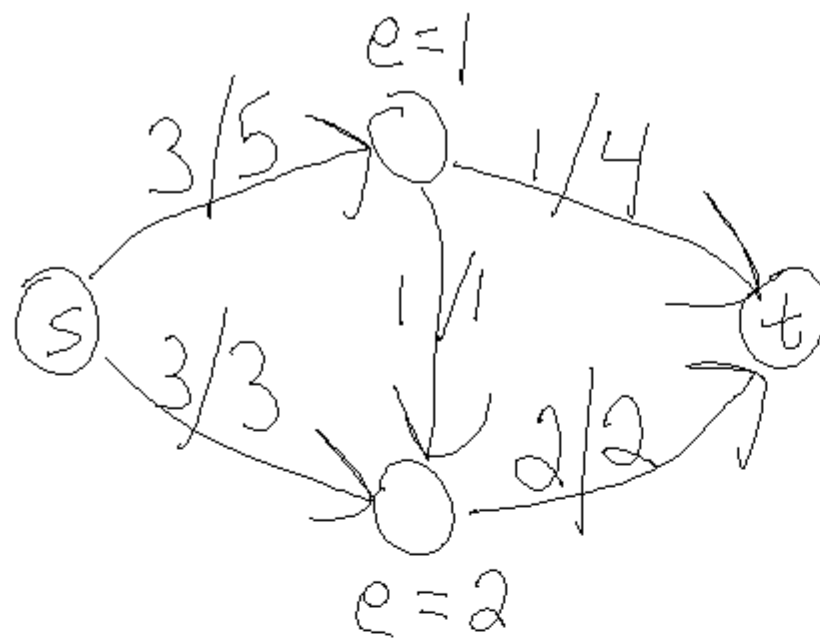
Relaxed Flow conservation: For all  $u \in V - \{s, t\}$ , we require

$$\sum_{v \in V} f(v, u) \geq \sum_{v \in V} f(u, v) .$$

Define excess

$$e(u) = \sum_{v \in V} f(v, u) - \sum_{v \in V} f(u, v) \tag{3}$$

# Example



## Comparison with augmenting paths

- Augmenting paths
  - Always maintains a flow (feasibility)
  - Works towards optimality (maximum flow)
- Push/relabel
  - Always maintains a preflow and works towards a flow
  - Maintains a superoptimal preflow and moves towards feasibility.

# Basic Operations

- If  $e(v) > 0$  then  $v$  is **active**.
- If  $c_f(v, w) > 0$  and  $d(v) = d(w) + 1$  then  $(v, w)$  is admissible.

**Push** from active vertices over admissible edges.

*Push*( $u, v$ )

- 1 // Applies when:  $u$  is active,  $c_f(u, v) > 0$ , and  $u.d = v.d + 1$ .
- 2 // Action: Push  $f(u, v) = \min(u.e, c_f(u, v))$  units of flow from  $u$  to  $v$ .
- 3  $f(u, v) = \min(u.e, c_f(u, v))$
- 4 if  $(u, v) \in E$
- 5      $(u, v).f = (u, v).f + f(u, v)$
- 6 else  $(v, u).f = (v, u).f - f(u, v)$
- 7  $u.e = u.e - f(u, v)$
- 8  $v.e = v.e + f(u, v)$

**Relabel** When a vertex has all its outgoing residual edges pointing uphill, we relabel it.

*Relabel*( $u$ )

- 1 // Applies when:  $u$  is active and for all  $v \in V$  such that  $(u, v) \in E_f$ ,  
we have  $u.d \leq v.d$ .
- 2 // Action: Increase the label of  $u$ .
- 3  $u.d = 1 + \min\{v.d : (u, v) \in E_f\}$

# Generic Push Relabel Algorithm

*Initialize – Preflow*( $G, s$ )

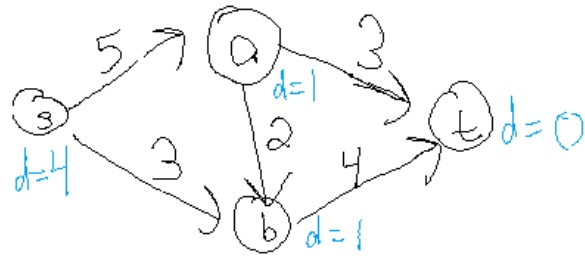
- 1 **for each vertex**  $v \in G.V$
- 2      $v.d = 0$
- 3      $v.e = 0$
- 4 **for each edge**  $(u, v) \in G.E$
- 5      $(u, v).f = 0$
- 6  $s.d = |G.V|$
- 7 **for each vertex**  $v \in s.Adj$
- 8      $(s, v).f = c(s, v)$
- 9      $v.e = c(s, v)$
- 10      $s.e = s.e - c(s, v)$

*Generic – Push – Relabel*( $G$ )

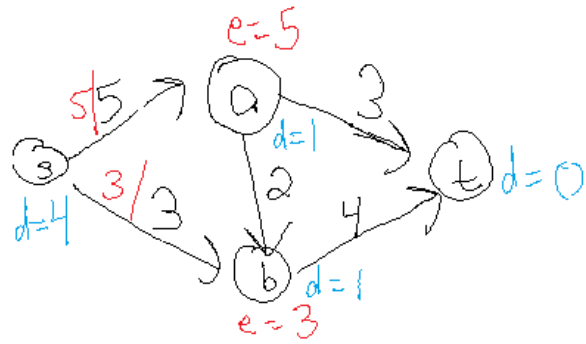
- 1 INITIALIZE-PREFLOW( $G, s$ )
- 2 **while** there exists an applicable push or relabel operation
- 3     select an applicable push or relabel operation and perform it

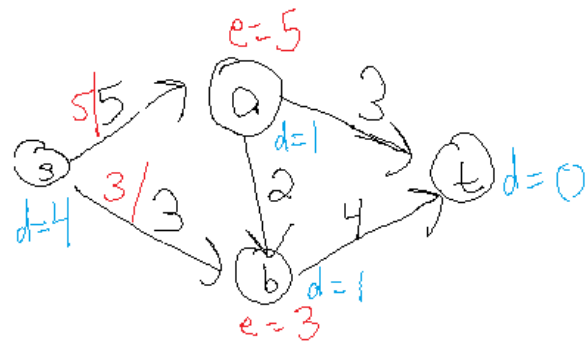


# Run Through Example

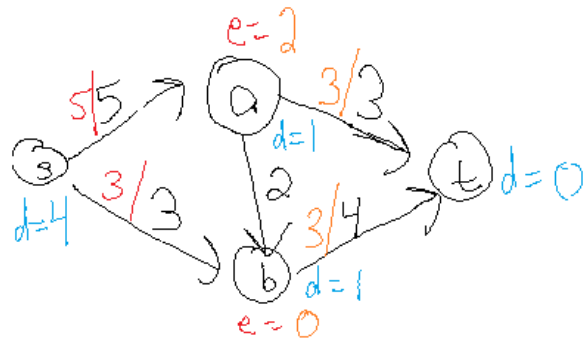


init

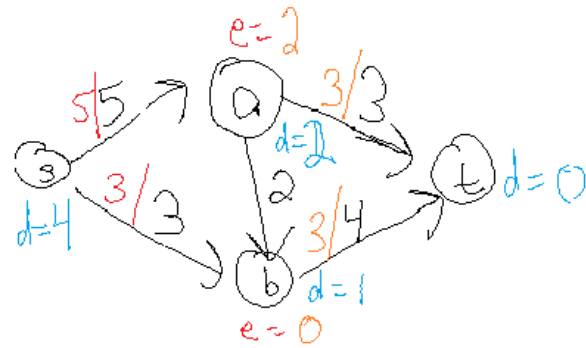
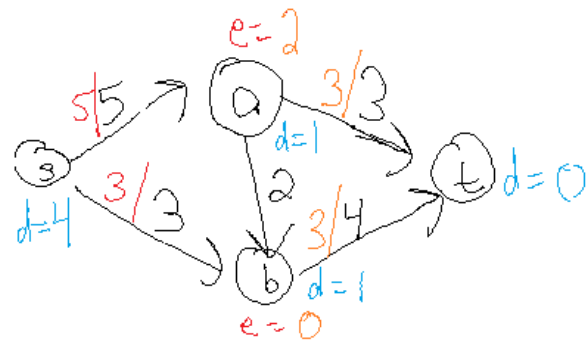


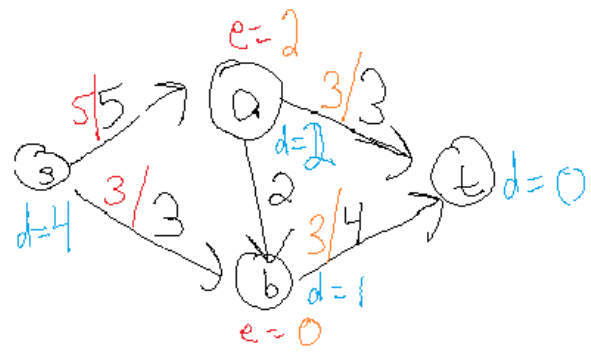


push(a, t)  
push(b, t)

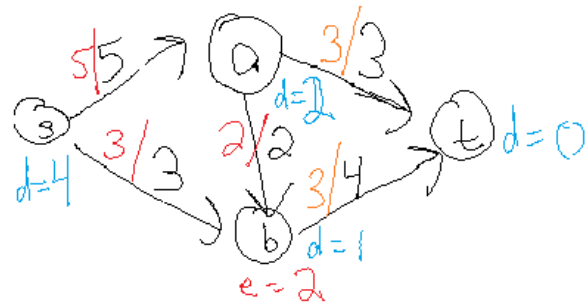


relabel a

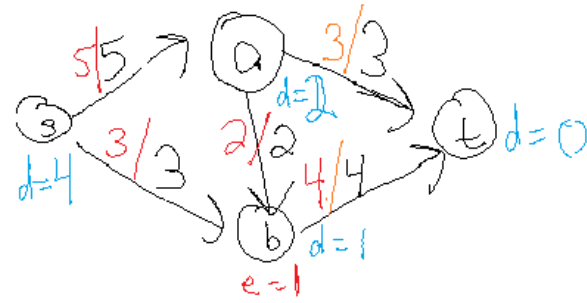
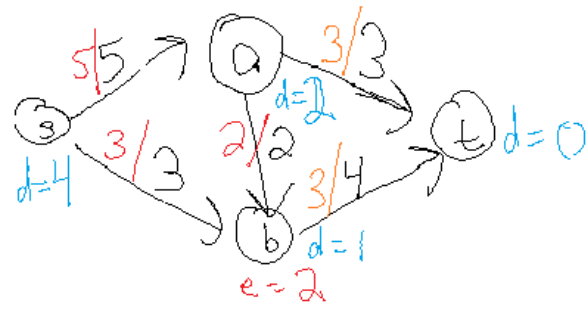


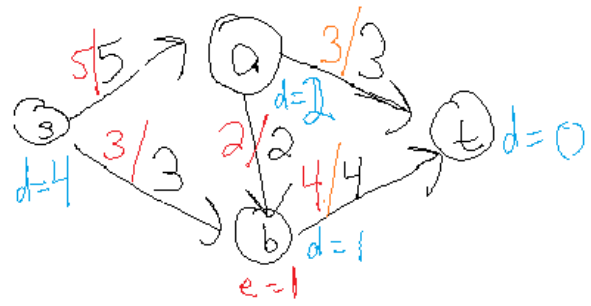


push(a,b)

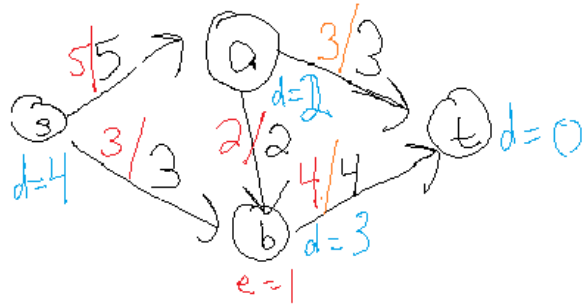


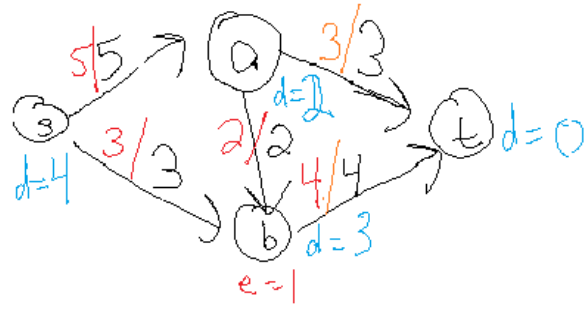
push(b, t)



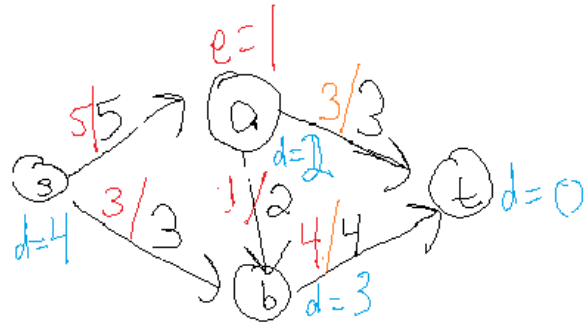


relabel b

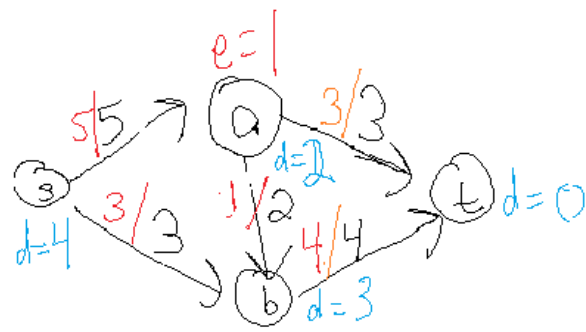




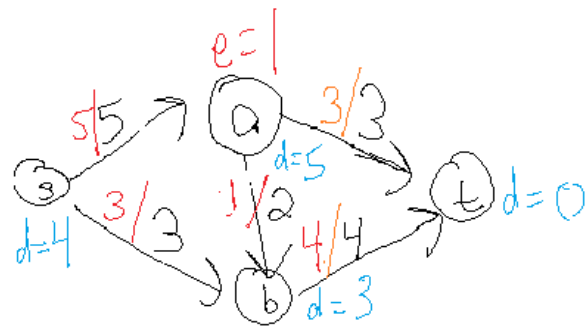
push(b,a)



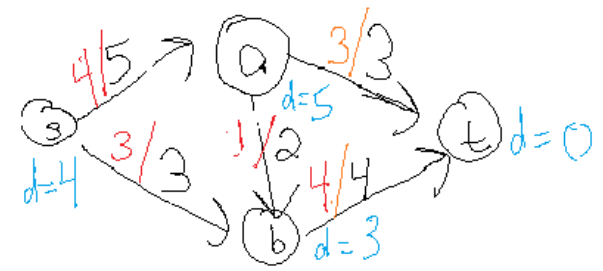
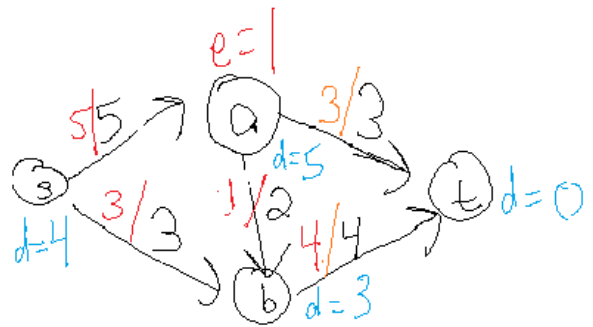




relabeled a



push(a,s)



Flow!

## Properties of Algorithm (without proofs)

- At any point, any active vertex either has a path to the source or the sink in the residual graph.
- There is never an  $s-t$  path in  $G_f$ .
- Let  $\delta(v)$  be the minimum of distance to sink and  $V$  plus distance to source in  $G_f$ .
- If  $v$  is active then  $d(v) \leq \delta(v)$ . (inductive proof, similar to shortest augmenting path, new short paths are not created).
- $d(v) \leq d(w) + 1$  for any  $(v, w) \in G_f$ .
- If  $e(v) > 0$  then either
  - $v$  has an outgoing admissible edge
  - $v$  can be relabeled (and the label will increase)
- If  $e(v) = 0$  for all  $v \in V - \{s, t\}$ , then  $f$  is a maximum flow.

### Proof of last statement

- $f$  is a flow by the definition of preflow.
- $f$  is maximum because there is no  $s-t$  path in  $G_f$ .

# Running Time

We will need to bound

1. Number of relabels
2. Time per relabel
3. Number of pushes
4. Time per push
5. Bookkeeping time

## Easy ones:

*Relabel(u)*

- 1 // Applies when:  $u$  is active and for all  $v \in V$  such that  $(u, v) \in E_f$ , we have  $u.d \leq v.d$ .
- 2 // Action: Increase the label of  $u$ .
- 3  $u.d = 1 + \min\{v.d : (u, v) \in E_f\}$

### Time per relabel

- Easy bound:  $O(V)$
- Time to relabel each vertex once :  $O(V^2)$

## Easy ones:

*Relabel(u)*

- 1 // Applies when:  $u$  is active and for all  $v \in V$  such that  $(u, v) \in E_f$ , we have  $u.d \leq v.d$ .
- 2 // Action: Increase the label of  $u$ .
- 3  $u.d = 1 + \min\{v.d : (u, v) \in E_f\}$

### Time per relabel

- Easy bound:  $O(V)$
- Time to relabel each vertex once :  $O(V^2)$
- Better bound:  $O(\text{degree}(v))$
- Time to relabel each vertex once:  $\sum_v \text{degree}(v) = O(E)$

(This is an example of amortized analysis)

### Number of relabels

- Vertex labels are at most  $2V$  , so each vertex is relabelled at most  $2V$  times.
- Total time spent relabelling =  $V \sum_v \text{degree}(v) = O(EV)$

# Pushes

*Push(u, v)*

- 1 // Applies when:  $u$  is active,  $c_f(u, v) > 0$ , and  $u.d = v.d + 1$ .
- 2 // Action: Push  $f(u, v) = \min(u.e, c_f(u, v))$  units of flow from  $u$  to  $v$ .
- 3  $f(u, v) = \min(u.e, c_f(u, v))$
- 4 if  $(u, v) \in E$
- 5      $(u, v).f = (u, v).f + f(u, v)$
- 6 else  $(v, u).f = (v, u).f - f(u, v)$
- 7  $u.e = u.e - f(u, v)$
- 8  $v.e = v.e + f(u, v)$

## Easy part

- Time per push =  $O(1)$  .
- Bookkeeping (can be amortized against other operations with reasonable data structures and rules for choosing push/relabel operations).

## Two types of Pushes

- **Saturating push:** Sends  $c_f(v, w)$  flow on  $(v, w)$
- **Non-saturating push:** Sends less flow. ( $e(v)$ ).

# Bounding Saturating Pushes

**Consider** a saturating push on  $(v, w)$ . What has to happen before the next saturating push on  $(v, w)$ ?



# Bounding Saturating Pushes

**Consider** a saturating push on  $(v, w)$ . What has to happen before the next saturating push on  $(v, w)$ ?

- $w$  has to be relabelled
- A push on  $(w, v)$  must occur
- $v$  has to be relabelled.

## **Conclusion**

- Between any two saturating pushes on  $(v, w)$ ,  $v$  must be relabelled.
- $v$  can be relabelled at most  $2V$  times.
- There are at most  $2V$  saturating pushes on  $(v, w)$ .
- The total number of saturating pushes is  $O(VE)$ .

# Non-saturating pushes

**Issue** After a non-saturating push, the graph does not change, nothing need be relabelled and another non-saturating push can occur on the edge before anything significant happens.

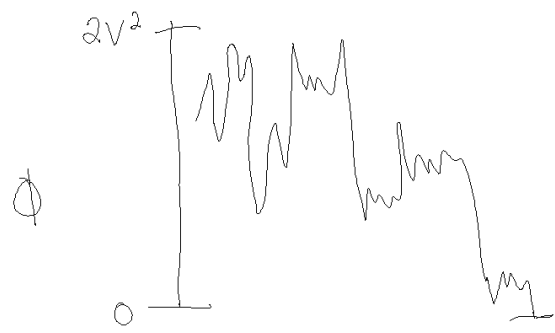
**Solution** As with shortest-augmenting-path, we won't count directly, but will bound other operations and related to non-saturating pushes. We will do so via means of a **potential function**.

$$\Phi = \sum_{v:e(v)>0} d(v)$$

# Analysis

$$\Phi = \sum_{v:e(v)>0} d(v)$$

- After initialization  $\Phi \leq 2V^2$ .
- At termination  $\Phi = 0$ .
- Let  $R$  be total increase in  $\Phi$  due to relabellings.
- Let  $S$  be total increase in  $\Phi$  due to saturating pushes.
- Each non-saturating push decreases  $\Phi$  by at least 1. (Why?).



## Putting these facts together

- Total decrease in  $\Phi$  associated with non-saturating pushes is at most  $2V^2 - 0 + R + S$ .
- Each non-saturating push decreases  $\Phi$  by at least 1. (Why?).
- Number of non-saturating pushes is at most  $2V^2 - 0 + R + S$ .

# Bounding R and S

$$\Phi = \sum_{v:e(v)>0} d(v)$$

## Relabellings

- Each relabelling must increase  $\Phi$  by at least 1.
- Total increase in  $\Phi$  associated with relabelling  $v$  is at most  $2V$ .
- Total increase associated with all relabellings is at most  $2V^2$ .

## Saturating Pushes

- A saturating push leaves excess at  $v$ . It adds excess to  $w$ .
  - If  $w$  already had excess, then  $\Phi$  is unchanged.
  - If  $w$  did not have excess, then  $\Phi$  increases by  $d(w)$ , which is at most  $2V$ .
- There are at most  $O(EV)$  saturating pushes, therefore total increase due to saturating pushes is  $O(EV^2)$ .

# Putting it Together

Number of non-saturating pushes is at most

$$2V^2 - 0 + R + S = O(V^2 + EV + EV^2) = O(EV^2)$$

.

**Conclusion:** Running time is  $O(EV^2)$  .

**Note:** Can be improved by

- Choosing operations more carefully.
- Better data structures to represent flow
- Best running times are close to  $VE$ .
- Winner in practice, assuming that one uses two additional heuristic ideas
  - gap heuristic
  - global relabellings

# Minimum Cost Flow

- We can add costs to a flow problem.
- There are several natural ways to formulate a minimum cost flow problem. Here is one.
- We have a flow network  $G$  with capacities  $u$  and edge costs  $a$
- We want to send  $d$  units from  $s$  to  $t$
- We want to find a flow that sends  $d$  units and minimizes cost.

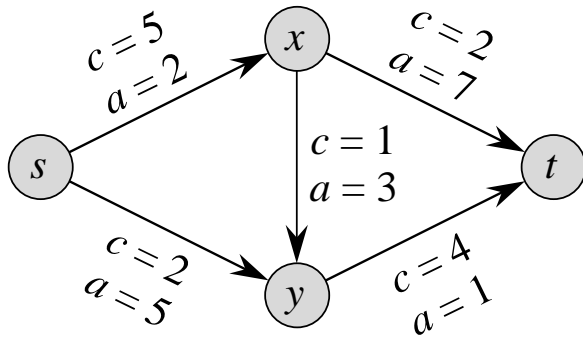
$$\text{minimize } \sum_{(u,v) \in E} a(u,v) f_{uv}$$

subject to

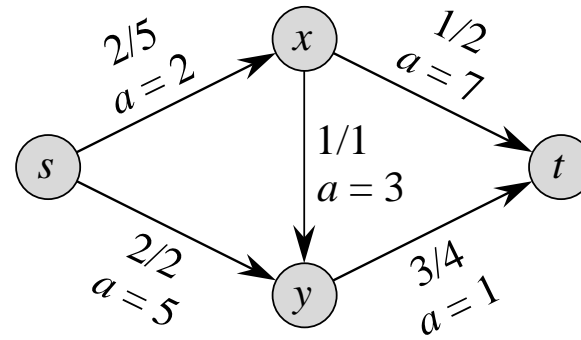
$$\begin{aligned} f_{uv} &\leq c(u,v) \quad \text{for each } u,v \in V, \\ \sum_{v \in V} f_{vu} - \sum_{v \in V} f_{uv} &= 0 \quad \text{for each } u \in V - \{s,t\}, \\ \sum_{v \in V} f_{sv} - \sum_{v \in V} f_{vs} &= d, \\ f_{uv} &\geq 0 \quad \text{for each } u,v \in V. \end{aligned}$$



# Example



(a)



(b)

# Minimum Cost Flow

- Generalizes maximum flow
- Generalizes shortest paths
- Generalizes transportation/transshipment and assignment problems

## Example Problem

A power company, located in Philadelphia, wants to supply power to New York, Boston and Washington D.C. The power company can generate 10 kw/hour and the demands for the three cities are 2, 3, and 5 kw/h respectively. The power grid contains wires between some of the pairs of cities. Each wire has a cost, representing the cost of transmitting one kw/h over that wire, and a capacity, which limits the number of kw that travel over the wire per hour. The data appears below. Formulate a plan for the power company to supply power at the minimum possible cost.

Cities	Cost	Capacity
Philadelphia - New York	\$1	10
Philadelphia - Boston	\$6	10
New York - Boston	\$2	2
New York - Washington	\$7	3
Boston - Washington	\$4	5

# Algorithms

- Efficient polynomial time algorithms exist.
- Augment along shortest path in residual graph
- A push relabel type algorithm
- Capacity scaling algorithms
- ...

# Multicommodity Flow

A new phone company has a network that can route 2000 calls between each of the following pairs of cities: (SF, LA), (SF, Bos), (SF, Chi.) (SF, Dallas), (LA, Chi.) , (LA, Dallas), (Chi. Dallas) , (Chi, Bos), (Chi., NY), (Dal., NY), (NY, Bos).

The company has customers who have the following demands for calls per hour:

Cities	Demand
SF - NY	1000
SF - Boston	3000
LA - Chicago	4000
Dallas - Boston	2000

Give a set of routes that allow all the calls to occur simultaneously.

# Multicommodity Flow

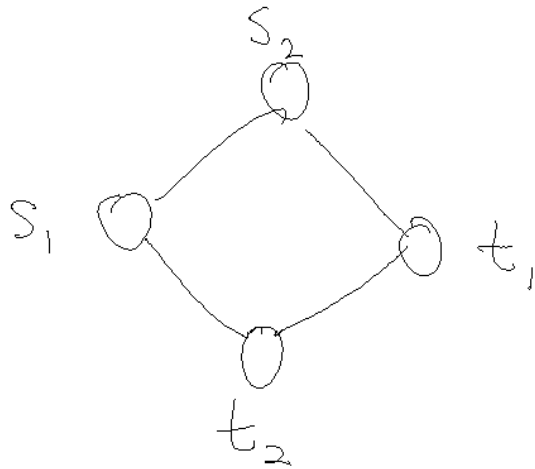
- You have a set of  $k$  commodities, each with its own source and sink that need to be routed simultaneously.
- Each commodity is its own flow.
- The capacity constraints are joint constraints.

$$\begin{aligned} \sum_{i=1}^k f_{iuv} &\leq c(u, v) \quad \text{for each } u, v \in V, \\ \sum_{v \in V} f_{iuv} - \sum_{v \in V} f_{ivu} &= 0 \quad \text{for each } i = 1, 2, \dots, k \text{ and} \\ &\quad \text{for each } u \in V - \{s_i, t_i\}, \\ \sum_{v \in V} f_{i, s_i, v} - \sum_{v \in V} f_{i, v, s_i} &= d_i \quad \text{for each } i = 1, 2, \dots, k, \\ f_{iuv} &\geq 0 \quad \text{for each } u, v \in V \text{ and} \\ &\quad \text{for each } i = 1, 2, \dots, k. \end{aligned}$$

# Multicommodity Flow Algorithms

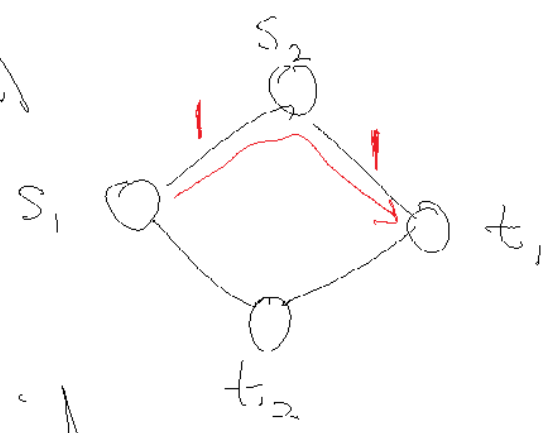
- Multicommodity flow, like minimum cost flow and maximum flow is a **linear program** (min/max a linear objective subject to linear constraints.)
- Polynomial-sized linear programs can be solved in polynomial time.
- Linear programs, in general, do not necessarily return integer solutions.
- Max flow and min-cost flow, given integral data, do have optimal integral solutions. Standard algorithms find optimal integral solutions.
- Multicommodity flow does not, in general have optimal integral solutions.
- Finding an optimal (real-valued) solution to a multicommodity flow problem is much harder (but still polynomial) than a similar sized maximum or minimum-cost flow problem. In practice, it's pretty slow.
- Finding an optimal integral solution to a multicommodity flow problem is NP-hard.

# Example





Integral



Fractional

