# EECS 336: Introduction to Algorithms Network Flow

Lecture 8

Ford-fulkerson, duality, minimum cut

**Reading:** 7.0-7.5

Announcements: midterm tuesday

- closed book, closed notes.
- one handwritten cheat sheet.
- dynamic programming.
- focus:
  - writing Parts I-II.
  - writing Parts III-IV (given Parts I-II).

#### Last time:

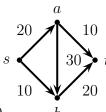
- reductions
- Network flow defn
- Bipartite matching
- reduction: matching  $\Rightarrow$  flow.

#### Today:

- Network flow
- duality: max flow = min cut

### **Network Flow**

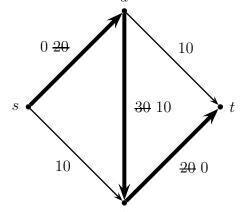
Example:



Max flow = 30.

**Idea:** repeatedly pus flow on s-t paths until can't push anymore.

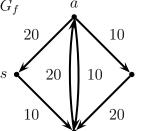
**Example:** Push 20 on P = (s, a, b, t)



**Note:** when pushing flow, we can undo flow already pushed.

**Def:** the <u>residual graph</u>  $G_f$  for flow f on G is the graph that represents capacity constraints for flows after pushing f

Example:  $G_f$ 



Construction:  $G_f \stackrel{b}{=} (V, E_f), c_f(\cdot)$ : For each  $e = (u, v) \in E$ ,

(if 
$$f(e) = c(e)$$
 discard  $e$ )

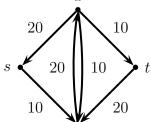
- if f(e) < c(e),
  - add e to  $E_f$
  - $c_f(e) = c(e) f(e)$
- if f(e) > 0
  - let e' = (v, u)
  - add e' to  $E_f$
  - $\bullet \ c_f(e') = c(e') + f(e)$

**Def:** the <u>residual capacity</u> of e in  $E_f$  is  $c_f(e)$ .

**Def:** the <u>bottleneck</u> capacity of s-t path P in  $G_f$  is minimm residual capacity of any edge in P.

**Def:** an augmenting path P in a residual graph  $G_f$  is a path with positive bottleneck capacity.

**Example:**  $G_f$  after pushing 20 on P = (s, a, b, t)



Augmenting path P = b(s, b, a, t) with bottleneck capacity 10.

Augment f with flow of 10 on P:

- $f(s,b) \leftarrow f(s,b) + 10$
- $f(a,b) \leftarrow f(a,b) 10$
- $f(a,t) \leftarrow f(a,t) + 10$

Note: can find augmenting paths with BFS.

**Algorithm:** Augment f with P

• b = bottleneck $(P, G_f)$ .

- for e in P:
  - $\bullet$  if e a foward edge:

$$f(e) \leftarrow f(e) + b$$

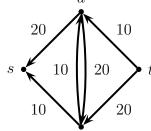
 $\bullet$  if e a back edge:

let 
$$e' = \text{back edge}$$

$$f(e') \leftarrow f(e) - b$$
.

**Example:**  $G_f$  after augmenting with  $P = G_f$ 

(s,b,a,t)



No more augmenting paths!

Algorithm: Ford-Fulkerson

- $f \leftarrow \text{null flow}$ .
- $G_f \leftarrow G$ .
- while exists s-t path P in  $G_f$  (by BFS)
  - augment f with P.
  - $G_f \leftarrow \text{residual graph for } G \text{ and } f.$
- return f.

## Runtime

Each iteration:

- construct  $G_f$ : O(m).
- find P: O(m).
- augmentation: O(n).
- (Total: O(m))

Fact: the value of flow increases by bottleneck capacity in each iteration.

**Theorem:** if C is upper bound on max flow and all capacities are integral then algorithm terminates in O(C) iterations with runtime O(mC)

**Proof:** (by "measure of progress")

- 1. bottleneck capacities integral:
  - current residual capacities integral
    - $\Rightarrow$  integral bottleneck capacity
    - $\Rightarrow$  next residual capacities integral
  - induction!
- 2. bottleneck capacities  $\geq 1$
- 3. flow increases by 1 each iteration
- 4. terminates in  $\leq C$  iterations.

**QED** 

Note:  $C \leq \sum_{e \text{ out of } s} c(e)$ .

**Note:** Clever choice of augmenting paths gives runtime  $O(m^2 \log C)$ .

#### Correctness

- 1. f is feasible.
- 2. f is optimal.

**Lemma:** f is feasible.

**Proof:** induction!

## Max flow = min cut

"duality: for maximization problem there is corresponding minimization problem"

**Recall:** an s-t cut (A, B) is partition of V into A and B with  $s \in A$  and  $t \in B$ .

**Def:** the capacity of cut 
$$(A, B)$$
 is  $c(A, B) = \sum_{e \text{ from } A \text{ to } B} c(e)$ 

Goal: flow algorithm is optimal

**Proof Approach:** primal = dual.

Claim 1: any flow f and any cut (A, B)then  $|f| \le c(A, B)$ .

Claim 2: for flow  $f^*$  with no augmenting path in  $G_{f^*}$  then exists cut  $(A^*, B^*)$  with  $|f^*| = c(A^*, B^*)$ 

#### Picture:

**Proof:** (of theorem)

• all flows 
$$|f| \underset{\text{by Claim 1}}{\leq} c(A^*, B^*) \underset{\text{by Claim 2}}{=} |f^*|.$$

**Corollary:** value of max flow = capacity of min cut

**Lemma:** for any flow f, cut (A, B) then,  $|f| = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e)$ 

**Proof:** (by picture, see text for formal proof)

**Proof:** (of Claim 1)

From Lemma:

Hermite.
$$|f| = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ in to } A} f(e)$$

$$\leq \sum_{e \text{ out of } A} f(e)$$

$$\leq \sum_{e \text{ out of } A} c(e)$$

**Proof:**  $(\overline{of} C(AB))$  no s-t path in  $G_f$ :

- let  $A^*$  be vertices connected to s.  $(B^* = V \setminus A^*)$
- $(A^*, B^*)$  is cut:
  - $s \in S^*$
  - $t \in B^*$
- for all e = (u, v) out of  $A^*$  in G:
  - $e \notin G_f$  $\Rightarrow f^*(e) = c(e)$
- for all e = (u, v) in to  $A^*$  in G:
  - $e' = (v, u) \notin G_f$  $\Rightarrow f^*(e) = 0$
- Lemma

$$\Rightarrow |f| = \sum_{e \text{ out of } A^*} f(e) - \sum_{e \text{ in to } A^*} f(e)$$
$$= \sum_{e \text{ out of } A^*} c(e) - 0$$
$$= c(A^*, B^*)$$

### Summary

- algorithm: augmenting paths in residual graph.
- correctness: max-flow min-cut theorem.
- many problems can be reduced to network flows.
- entire courses on network flows.