

# Chapter 17

## Network Flow Algorithms

CS 473: Fundamental Algorithms, Spring 2011

March 29, 2011

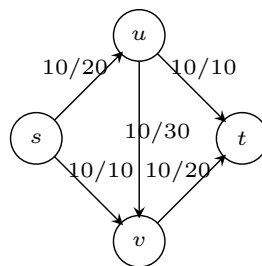
### 17.1 Algorithm(s) for Maximum Flow

#### 17.1.0.1 Greedy Approach

1. Begin with  $f(e) = 0$  for each edge
2. Find a  $s$ - $t$  path  $P$  with  $f(e) < c(e)$  for every edge  $e \in P$
3. *Augment* flow along this path
4. Repeat augmentation for as long as possible.

#### 17.1.0.2 Greedy Approach: Issues

1. Begin with  $f(e) = 0$  for each edge
2. Find a  $s$ - $t$  path  $P$  with  $f(e) < c(e)$  for every edge  $e \in P$
3. Augment flow along this path



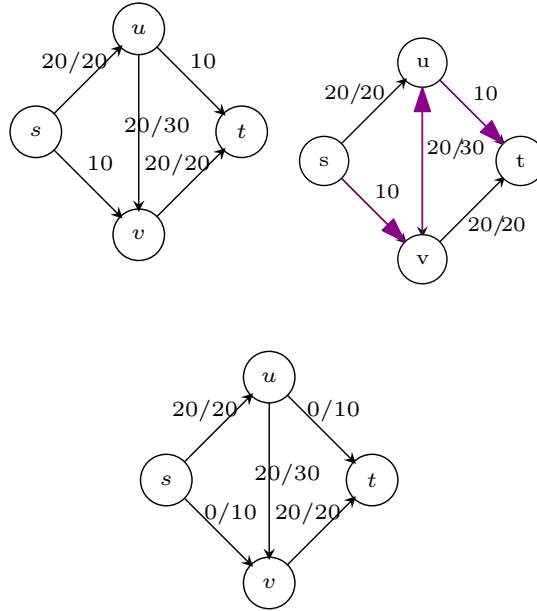


Figure 17.1: Flow in red edges

- Repeat augmentation for as long as possible.

Greedy can get stuck in sub-optimal flow!  
Need to “push-back” flow along edge  $(u, v)$

## 17.2 Ford-Fulkerson Algorithm

### 17.2.0.3 Residual Graph

**Definition 17.2.1** For a network  $G = (V, E)$  and flow  $f$ , the residual graph  $G_f = (V', E')$  of  $G$  with respect to  $f$  is

- $V' = V$
- *Forward Edges:* For each edge  $e \in E$  with  $f(e) < c(e)$ , we  $e \in E'$  with capacity  $c(e) - f(e)$
- *Backward Edges:* For each edge  $e = (u, v) \in E$  with  $f(e) > 0$ , we  $(v, u) \in E'$  with capacity  $f(e)$

### 17.2.0.4 Residual Graph Example

### 17.2.0.5 Residual Graph Property

**Observation:** Residual graph captures the “residual” problem exactly.

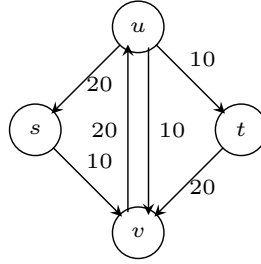


Figure 17.2: Residual Graph

**Lemma 17.2.2** *Let  $f$  be a flow in  $G$  and  $G_f$  be the residual graph. If  $f'$  is a flow in  $G_f$  then  $f + f'$  is a flow in  $G$  of value  $v(f) + v(f')$ .*

**Lemma 17.2.3** *Let  $f$  and  $f'$  be two flows in  $G$  with  $v(f') \geq v(f)$ . Then there is a flow  $f''$  of value  $v(f') - v(f)$  in  $G_f$ .*

Definition of  $+$  and  $-$  for flows is intuitive and the above lemmas are easy in some sense but a bit messy to formally prove.

### 17.2.0.6 Residual Graph Property: Implication

*Recursive* algorithm for finding a maximum flow:

```

MaxFlow( $G, s, t$ ):
  If the flow from  $s$  to  $t$  is 0
    return 0
  Find any flow  $f$  with  $v(f) > 0$  in  $G_f$ 
  Recursively compute a maximum flow in  $G_{f+f'}$ 
  Output the flow  $f + f'$ 

```

*Iterative* algorithm for finding a maximum flow:

```

MaxFlow( $G, s, t$ ):
  Start with flow  $f$  that is 0 on all edges
  While there is a flow  $f'$  in  $G_f$ 
     $f = f + f'$ 
    Update  $G_f$ 
  endwhile
  Output  $f$ 

```

### 17.2.0.7 Ford-Fulkerson Algorithm

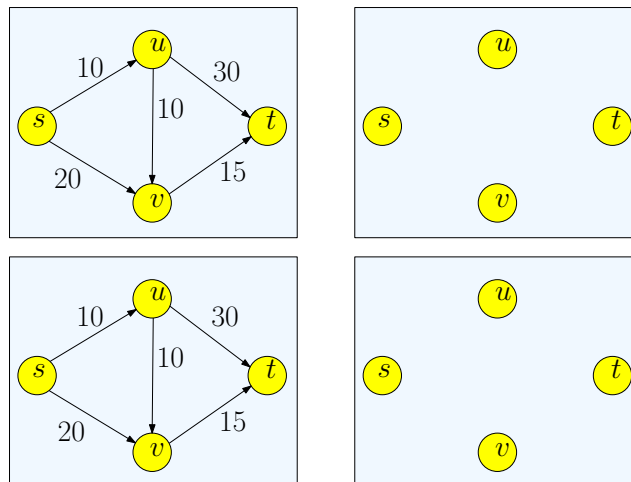
```

algFordFulkerson
for every edge  $e$ ,  $f(e) = 0$ 
 $G_f$  is residual graph of  $G$  with respect to  $f$ 
while  $G_f$  has a simple  $s$ - $t$  path do
  let  $P$  be simple  $s$ - $t$  path in  $G_f$ 
   $f = \text{augment}(f, P)$ 
  Construct new residual graph  $G_f$ 
  
```

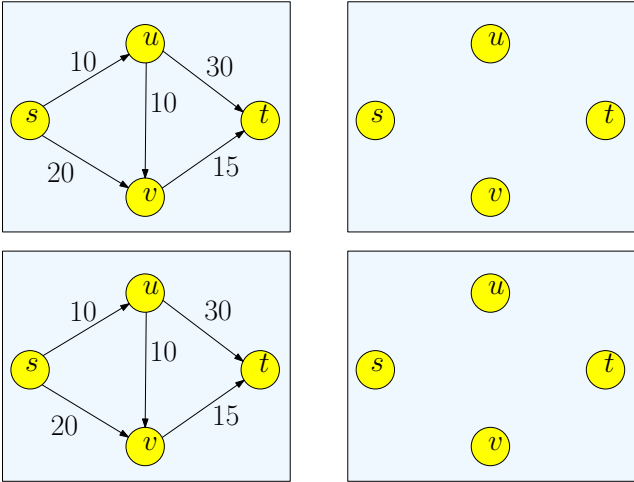
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augment( $f, P$ )
let  $b$  be bottleneck capacity,
  i.e., min capacity of edges in  $P$  (in  $G_f$ )
for each edge  $(u, v)$  in  $P$  do
  if  $e = (u, v)$  is a forward edge then
     $f(e) = f(e) + b$ 
  else (*  $(u, v)$  is a backward edge *)
    let  $e = (v, u)$  (*  $(v, u)$  is in  $G$  *)
     $f(e) = f(e) - b$ 
return  $f$ 
  
```

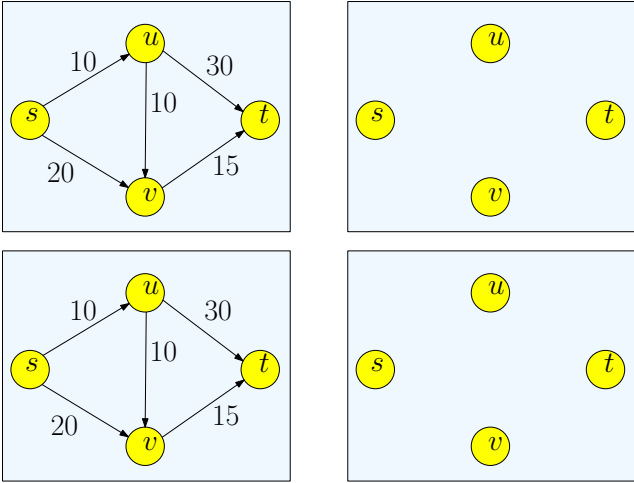
### 17.2.0.8 Example



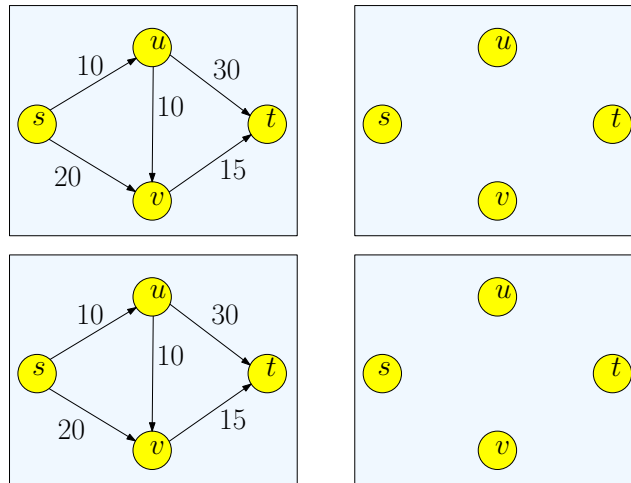
17.2.0.9 Example continued



17.2.0.10 Example continued



### 17.2.0.11 Example continued



## 17.3 Correctness and Analysis

### 17.3.1 Termination

#### 17.3.1.1 Properties about Augmentation: Flow

**Lemma 17.3.1** *If  $f$  is a flow and  $P$  is a simple  $s$ - $t$  path in  $G_f$ , then  $f' = \text{augment}(f, P)$  is also a flow.*

*Proof:* Verify that  $f'$  is a flow. Let  $b$  be augmentation amount.

- *Capacity constraint:* If  $(u, v) \in P$  is a forward edge then  $f'(e) = f(e) + b$  and  $b \leq c(e) - f(e)$ . If  $(u, v) \in P$  is a backward edge, then letting  $e = (v, u)$ ,  $f'(e) = f(e) - b$  and  $b \leq f(e)$ . Both cases  $0 \leq f'(e) \leq c(e)$ .
- *Conservation constraint:* Let  $v$  be an internal node. Let  $e_1, e_2$  be edges of  $P$  incident to  $v$ . Four cases based on whether  $e_1, e_2$  are forward or backward edges. Check cases (see fig next slide). ■

#### 17.3.1.2 Properties about Augmentation: Conservation Constraint

#### 17.3.1.3 Properties about Augmentation: Integer Flow

**Lemma 17.3.2** *At every stage of the Ford-Fulkerson algorithm, the flow values  $f(e)$  and the residual capacities in  $G_f$  are integers*

*Proof:* Initial flow and residual capacities are integers. Suppose lemma holds for  $j$  iterations. Then in  $(j + 1)$ st iteration, minimum capacity edge  $b$  is an integer, and so flow after augmentation is an integer. ■

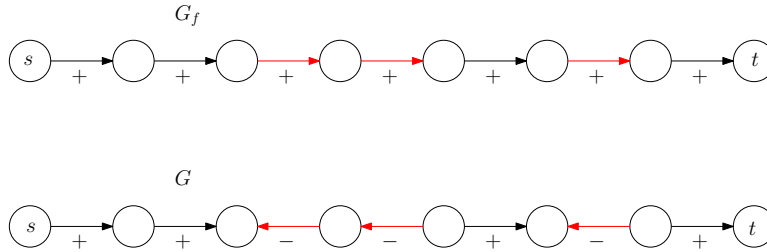


Figure 17.3: Augmenting path  $P$  in  $G_f$  and corresponding change of flow in  $G$ . Red edges are backward edges.

### 17.3.1.4 Progress in Ford-Fulkerson

**Proposition 17.3.3** *Let  $f$  be a flow and  $f'$  be flow after one augmentation. Then  $v(f) < v(f')$ .*

*Proof:* Let  $P$  be an augmenting path, i.e.,  $P$  is a simple  $s$ - $t$  path in residual graph

- First edge  $e$  in  $P$  must leave  $s$
- Original network  $G$  has no incoming edges to  $s$ ; hence  $e$  is a forward edge
- $P$  is simple and so never returns to  $s$
- Thus, value of flow increases by the flow on edge  $e$

■

### 17.3.1.5 Termination Proof

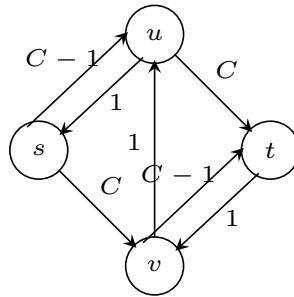
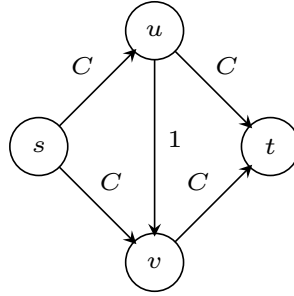
**Theorem 17.3.4** *Let  $C$  be the minimum cut value; in particular  $C \leq \sum_{e \text{ out of } s} c(e)$ . Ford-Fulkerson algorithm terminates after finding at most  $C$  augmenting paths.*

*Proof:* The value of the flow increases by at least 1 after each augmentation. Maximum value of flow is at most  $C$ .

■

### Running time

- Number of iterations  $\leq C$
- Number of edges in  $G_f \leq 2m$
- Time to find augmenting path is  $O(n + m)$
- Running time is  $O(C(n + m))$  (or  $O(mC)$ ).



### 17.3.1.6 Efficiency of Ford-Fulkerson

Running time =  $O(mC)$  is not polynomial. Can the running time be as  $\Omega(mC)$  or is our analysis weak? Ford-Fulkerson can take  $\Omega(C)$  iterations.

## 17.3.2 Correctness

### 17.3.2.1 Correctness of Ford-Fulkerson Augmenting Path Algorithm

*Question:* When the algorithm terminates, is the flow computed the maximum  $s$ - $t$  flow?

Proof idea: show a cut of value equal to the flow. Also shows that maximum flow is equal to minimum cut!

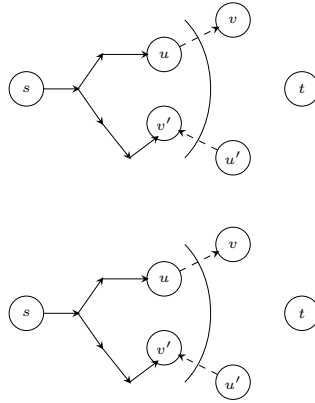
### 17.3.2.2 Recalling Cuts

**Definition 17.3.5** Given a flow network an  $s$ - $t$  cut is a set of edges  $E' \subset E$  such that removing  $E'$  disconnects  $s$  from  $t$ : in other words there is no directed  $s \rightarrow t$  path in  $E - E'$ . Capacity of cut  $E'$  is  $\sum_{e \in E'} c(e)$ .

Let  $A \subset V$  such that

- $s \in A, t \notin A$
- $B = V - A$  and hence  $t \in B$





Define  $(A, B) = \{(u, v) \in E \mid u \in A, v \in B\}$

**Claim 17.3.6**  $(A, B)$  is an  $s$ - $t$  cut.

Recall: Every *minimal*  $s$ - $t$  cut  $E'$  is a cut of the form  $(A, B)$ .

### 17.3.2.3 Ford-Fulkerson Correctness

**Lemma 17.3.7** *If there is no  $s$ - $t$  path in  $G_f$  then there is some cut  $(A, B)$  such that  $v(f) = c(A, B)$*

*Proof:* Let  $A$  be all vertices reachable from  $s$  in  $G_f$ ;  $B = V \setminus A$

- $s \in A$  and  $t \in B$ . So  $(A, B)$  is an  $s$ - $t$  cut in  $G$
- If  $e = (u, v) \in G$  with  $u \in A$  and  $v \in B$ , then  $f(e) = c(e)$  (*saturated* edge) because otherwise  $v$  is reachable from  $s$  in  $G_f$

■

### 17.3.2.4 Lemma Proof Continued

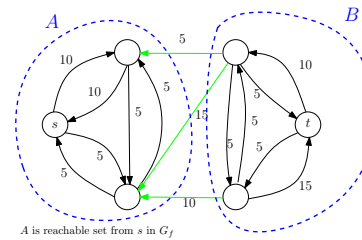
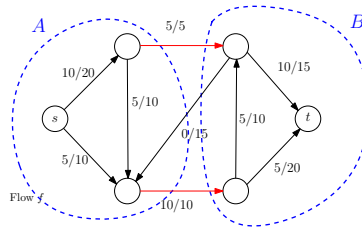
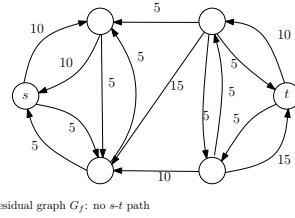
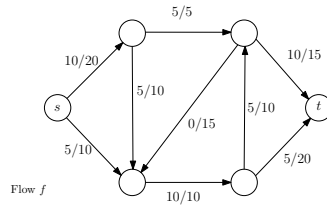
*Proof:*

- If  $e = (u', v') \in G$  with  $u' \in B$  and  $v' \in A$ , then  $f(e) = 0$  because otherwise  $u'$  is reachable from  $s$  in  $G_f$
- Thus,

$$\begin{aligned}
 v(f) &= f^{\text{out}}(A) - f^{\text{in}}(A) \\
 &= f^{\text{out}}(A) - 0 \\
 &= c(A, B) - 0 \\
 &= c(A, B)
 \end{aligned}$$

■

### 17.3.2.5 Example



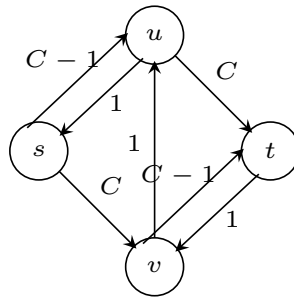
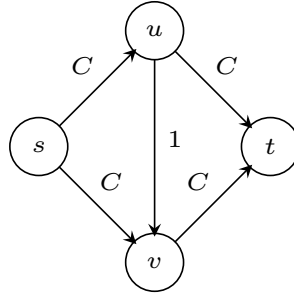
### 17.3.2.6 Ford-Fulkerson Correctness

**Theorem 17.3.8** *The flow returned by the algorithm is the maximum flow.*

*Proof:*

- For any flow  $f$  and  $s-t$  cut  $(A, B)$ ,  $v(f) \leq c(A, B)$
- For flow  $f^*$  returned by algorithm,  $v(f^*) = c(A^*, B^*)$  for some  $s-t$  cut  $(A^*, B^*)$
- Hence,  $f^*$  is maximum

■



### 17.3.2.7 Max-Flow Min-Cut Theorem and Integrality of Flows

**Theorem 17.3.9** For any network  $G$ , the value of a maximum  $s$ - $t$  flow is equal to the capacity of the minimum  $s$ - $t$  cut.

*Proof:* Ford-Fulkerson algorithm terminates with a maximum flow of value equal to the capacity of a (minimum) cut. ■

### 17.3.2.8 Max-Flow Min-Cut Theorem and Integrality of Flows

**Theorem 17.3.10** For any network  $G$  with integer capacities, there is a maximum  $s$ - $t$  flow that is integer valued.

*Proof:* Ford-Fulkerson algorithm produces an integer valued flow when capacities are integers. ■

## 17.4 Polynomial Time Algorithms

### 17.4.0.9 Efficiency of Ford-Fulkerson

Running time =  $O(mC)$  is not polynomial. Can the upper bound be achieved?

### 17.4.0.10 Polynomial Time Algorithms

*Question:* Is there a polynomial time algorithm for maxflow?

*Question:* Is there a variant of Ford-Fulkerson that leads to a polynomial time algorithm?  
Can we choose an augmenting path in some clever way? Yes! Two variants.

- Choose the augmenting path with largest bottleneck capacity.
- Choose the shortest augmenting path.

## 17.4.1 Capacity Scaling Algorithm

### 17.4.1.1 Augmenting Paths with Large Bottleneck Capacity

- Pick augmenting paths with largest bottleneck capacity in each iteration of Ford-Fulkerson
- How do we find path with largest bottleneck capacity?
  - Assume we know  $\Delta$  the bottleneck capacity
  - Remove all edges with residual capacity  $\leq \Delta$
  - Check if there is a path from  $s$  to  $t$
  - Do binary search to find largest  $\Delta$
  - Running time:  $O(m \log C)$
- Can we bound the number of augmentations? Can show that in  $O(m \log C)$  augmentations the algorithm reaches a max flow. This leads to an  $O(m^2 \log^2 C)$  time algorithm.

### 17.4.1.2 Augmenting Paths with Large Bottleneck Capacity

How do we find path with largest bottleneck capacity?

- Max bottleneck capacity is one of the edge capacities. Why?
- Can do binary search on the edge capacities. First, sort the edges by their capacities and then do binary search on that array as before.
- Algorithm's running time is  $O(m \log m)$ .
- Different algorithm that also leads to  $O(m \log m)$  time algorithm by adapting Prim's algorithm.