

# CS 341: ALGORITHMS

Lecture 17: max flow  
Readings: CLRS 26.2

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## QUICK REVIEW OF LAST TIME

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## RECALL: MAX-FLOW MIN-CUT THEOREM

- **Theorem 3:** every max  $s$ - $t$  flow has value equal to the capacity of a min  $s$ - $t$  cut
- We give an **algorithmic proof** of this theorem
  - (showing that one algorithm solves both max-flow and min-cut at the same time)

## FORD-FULKERSON METHOD

Algorithm development  
(mixed together with proof of max-flow min-cut theorem)

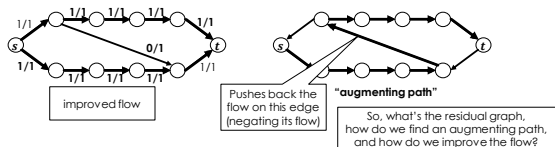
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## FORD-FULKERSON METHOD

Same Ford as in Bellman-Ford :)

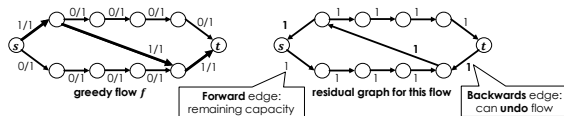
- Can **undo** previous decisions to improve the flow
  - Can effectively "push back" some flow using an **augmenting path** through a **residual graph**



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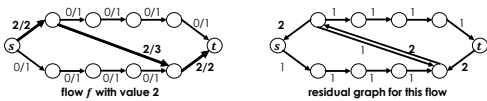
## RESIDUAL GRAPH

- A **residual graph**  $R_f$  is defined for a **given flow**  $f$  and **graph**  $G$
- $R_f$  has the same vertices as  $G$
- For each edge  $e = uv$  in  $G$ ,
  - If  $f(e) < c(e)$ , then  $R_f$  contains a **forward edge**  $(u, v)$  with the **remaining capacity**  $c(e) - f(e)$
  - If  $f(e) > 0$ , then  $R_f$  contains a **backwards edge**  $(v, u)$  with **capacity**  $f(e)$  representing flow that could be "pushed back"



**ANOTHER EXAMPLE RESIDUAL GRAPH**

- Recall: for each edge  $e = uv$  in  $G$ ,
  - If  $f(e) < c(e)$ , then  $R_f$  contains a **forward edge**  $(u, v)$  with the **remaining capacity**  $c(e) - f(e)$
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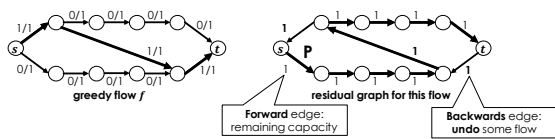
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CONTINUING WITH NEW MATERIAL

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**FORD-FULKERSON METHOD**

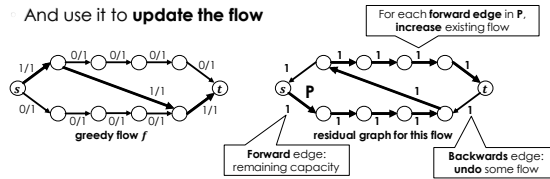
- Find a **shortest path P** from  $s$  to  $t$  in the **residual graph**
  - If it **improves** the flow, we call it an **augmenting path**
  - And use it to **update the flow**



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**FORD-FULKERSON METHOD**

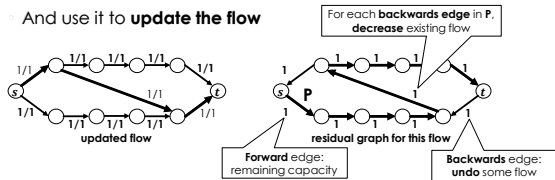
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**FORD-FULKERSON METHOD**

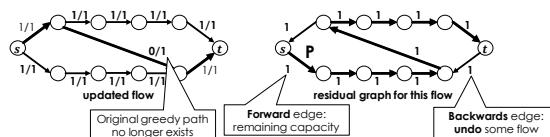
- Find a **shortest path P** from  $s$  to  $t$  in the **residual graph**
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**FORD-FULKERSON METHOD**

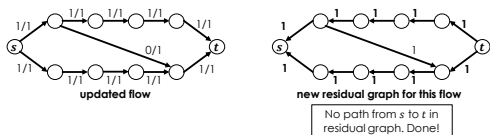
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### FORD-FULKERSON METHOD

- Find a **shortest path P** from  $s$  to  $t$  in the **residual graph**
  - If it **improves** the flow, we call it an **augmenting path**
  - And use it to **update the flow**



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### IMPROVING A FLOW $f$ GIVEN AN AUGMENTING PATH $P$

- An augmenting path w.r.t a flow  $f$  is a **simple  $s$ - $t$  path** in  $R_f$
- Let  $P$  be an augmenting path w.r.t  $f$
- Let  $\text{bottleneck}(f, P)$  be the minimum capacity of an edge in  $P$
- We show this subroutine  $\text{augment}(f, P)$  always improves the value of flow  $f$

no cycles!

```

1 augment(f, P)
2   let b = bottleneck(f, P)
3   for each edge e = (u,v) in P
4     if e is a forward edge
5       f(e) = f(e) + b
6     else if e is a backwards edge
7       let e' = (v,u)
8       f(e') = f(e') - b
    
```

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### LEMMA 4: AUGMENT() IMPROVES FLOW $f$

- Let  $f$  be a flow in  $G$  with  $f^{in}(s) = 0$ , and  $P$  be an augmenting path w.r.t  $f$
- Let  $f'$  be the resulting flow after running  $\text{augment}(f, P)$
- Then  $f'$  is a flow with  $\text{value}(f') = \text{value}(f) + \text{bottleneck}(f, P)$
- That is,  $\text{augment}(f, P)$  increases the flow by  $\text{bottleneck}(f, P)$

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### PROOF

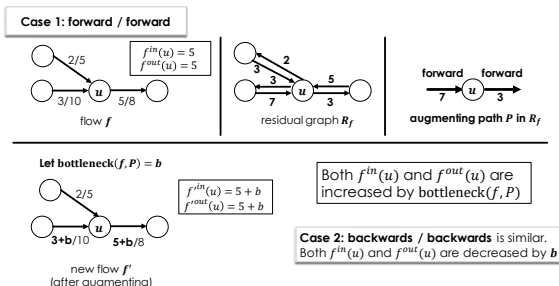
- Claim:  $\text{augment}(f, P)$  increases the flow by  $\text{bottleneck}(f, P)$
- First check  $f'$  is a flow
  - Prove capacity and conservation constraints, and  $f'^{in}(s) = 0$
- Are capacity constraints satisfied?**
  - We add/subtract  $\text{bottleneck}(f, P)$  to/from each edge
  - And  $\text{bottleneck}(f, P)$  is the minimum of the smallest remaining capacity, and the current flow
  - So capacity constraints are satisfied

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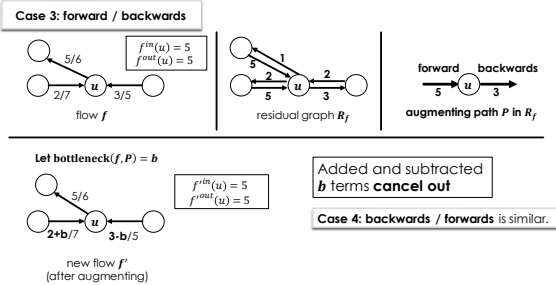
### PROOF

- Claim:  $\text{augment}(f, P)$  increases the flow by  $\text{bottleneck}(f, P)$
- How about conservation of flow?**
  - Consider how the flow into and out of each vertex  $u \notin \{s, t\}$  changes as a result of running  $\text{augment}(f, P)$
  - We show the change in  $f^{in}(u)$  is the same as the change in  $f^{out}(u)$
  - There are 4 cases, depending on whether the edges entering/leaving  $u$  are **forward** or **backward** edges

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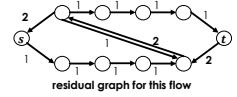
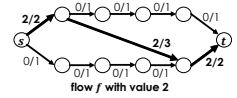
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SHOWING  $f'^{in}(s) = 0$

- Last step in showing  $f'$  is a flow
  - Prove:  $s$  still has no flow into it
- Since  $f$  is a flow,  $f^{in}(s) = 0$
- To get  $f'^{in}(s) > 0$ , an augmenting path must include an edge **into**  $s$
- But then an augmenting path starts at  $s$ , then returns to  $s$ , forming a cycle -- contradiction!



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FINISHING LEMMA 4: AUGMENT() IMPROVES FLOW

- Finally we argue  $value(f') = value(f) + bottleneck(f, P)$
- $f$  and  $f'$  are flows, so  $value(f') = f'^{out}(s)$  and  $value(f) = f^{out}(s)$
- We thus show  $f'^{out}(s) = f^{out}(s) + bottleneck(f, P)$
- The augmenting path  $P$  is a **simple** path (leaving  $s$  exactly once)
- And there is no flow into  $s$ , so the edge  $e \in P$  leaving  $s$  is a **forward edge**
- This means  $augment(f, P)$  **adds**  $bottleneck(f, P)$  to  $f(e)$
- So  $f'^{out}(s) = f^{out}(s) + bottleneck(f, P)$

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FORD-FULKERSON METHOD

- By Lemma 4, starting from any flow  $f$ , if we can **find an augmenting path**  $P$  w.r.t  $f$  in  $R_f$ , then we can use  $augment(f, P)$  to **improve our flow**
- Ford-Fulkerson does this repeatedly **starting from an empty flow**

```

1 FordFulkerson(G=(V,E))
2   for e in E
3     f(e) = 0
4
5   while there is a simple s-t path P in Rf do
6     augment(f, P)
7     and update the residual graph Rf
    
```

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What we have proved so far: **augmenting improves flow.**

We don't know yet if

- we can actually obtain the max flow, or
- whether max-flow = min-cut.

MAX-FLOW MIN-CUT THEOREM PROOF

PROOF STRATEGY

- Claim: when there is **no augmenting path**, there is a **cut with capacity equal to the value of the current flow.**
- Proving this will simultaneously
  - prove the max-flow min-cut theorem,
  - prove correctness of the Ford-Fulkerson method,
  - solve the max flow problem, and
  - solve the min cut problem

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### PROVING MAX FLOW = MIN CUT

Two directions:  
**max flow ≤ min cut and max flow ≥ min cut**

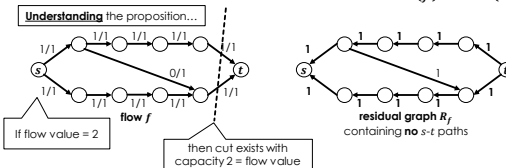
We actually proved the ≤ direction already (Lemma 2 last time) when discussing upper bounds for max flow

It remains to prove the ≥ direction.

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### PROVING MAX FLOW ≥ MIN CUT

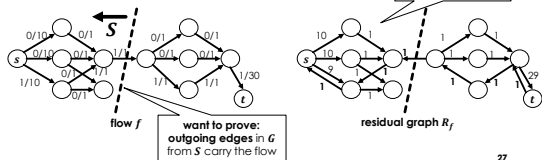
Proposition: if  $f$  is an  $s-t$  flow such that there is no  $s-t$  path in the residual graph  $R_f$ , then there is an  $s-t$  cut  $S$  s.t.  $\text{value}(f) = c^{\text{out}}(S)$



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### PROVING THE PROPOSITION

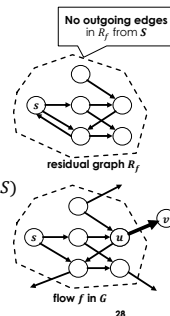
- Since there is no  $s-t$  path in  $R_f$ , there is a subset  $S$  of vertices with  $s \in S, t \notin S$  such that  $S$  has **no outgoing edges** in  $R_f$
- What does this statement look like?



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### PROVING THE PROPOSITION

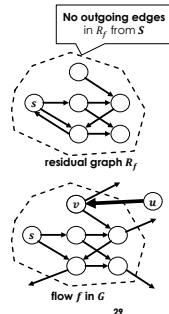
- Since there is no  $s-t$  path in  $R_f$ , there is a subset  $S$  of vertices with  $s \in S, t \notin S$  such that  $S$  has **no outgoing edges** in  $R_f$
- Claim:  $c^{\text{out}}(S) = \text{value}(f)$
- Consider two types of edges. Type 1:
  - $uv$  exiting  $S$  in  $G$  ( $uv \in \delta^{\text{out}}(S)$  in  $G, u \in S, v \notin S$ )
- Since  $S$  has no outgoing edge in  $R_f$ , we know  $uv \notin R_f$
- This implies  $f(uv) = c(uv)$ , as otherwise  $uv$  would be a forward edge in  $R_f$



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### PROVING THE PROPOSITION

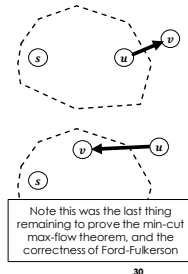
- Claim:  $c^{\text{out}}(S) = \text{value}(f)$
- Consider two types of edges. Type 2:
  - $uv$  entering  $S$  in  $G$  ( $uv \in \delta^{\text{in}}(S)$  in  $G, u \notin S, v \in S$ )
  - Since  $S$  has no outgoing edge in  $R_f$ , we know there is no edge  $vu \notin R_f$  (note  $vu$  would be directed out of  $S$ )
  - This implies  $f(uv) = 0$ , as otherwise  $vu$  would be a backwards edge in  $R_f$



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### PROVING THE PROPOSITION

- We just showed
  - For edge  $uv$  directed out of  $S$ ,  $f(uv) = c(uv)$
  - For edge  $uv$  directed into  $S$ ,  $f(uv) = 0$
- So  $f^{\text{out}}(S) - f^{\text{in}}(S) = c^{\text{out}}(S) - 0 = c^{\text{out}}(S)$
- This proves the proposition. I.e., given flow  $f$ , if there are no  $s-t$  paths in  $R_f$ , then **there is a cut matching the flow**



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### RUNTIME OF FORD-FULKERSON

- Depends on the implementation

```

1 FordFulkerson(G=(V,E))
2   for e in E
3     f(e) = 0
4
5   while there is a simple s-t path P in Rf do
6     augment(f, P)
7     and update the residual graph Rf
    
```

- How do we find an augmenting path?
- How many times do we need to augment before we terminate?

### TIME COMPLEXITY

of the Ford-Fulkerson method

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### RUNTIME OF FORD-FULKERSON

- Assume we find any arbitrary augmenting path  $P$ , using any technique, in  $O(n + m)$  time
- Then every time  $\text{augment}(f, P)$  is run, we know only that the flow **increases**
- If capacities are **integers**, the increase is at least 1
- In this case, **if max flow is  $k$  then runtime is  $O(k(n + m))$** 
  - For max flow we assume a connected graph, so this is  **$O(km)$**
  - Very bad if  $k$  is large**

If capacities are reals (and in particular some are irrational), this may **never** terminate!

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### WORST CASE FOR THIS APPROACH

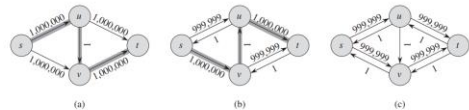


Figure 26.7 (a) A flow network for which FORD-FULKERSON can take  $\Theta(E/f^*)$  time, where  $f^*$  is a maximum flow, shown here with  $f^* = 2,000,000$ . The shaded path is an augmenting path with residual capacity 1. (b) The resulting residual network, with another augmenting path whose residual capacity is 1. (c) The resulting residual network.

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### EDMONDS-KARP APPROACH

- Use **BFS** to find a shortest path (in terms of number of edges) and use that as an augmenting path
- It turns out this always **terminates after  $O(nm)$  augmenting paths**
  - (even with real capacities)
- BFS takes  $O(n + m)$  time;  **$O(m)$**  since the graph is connected
- So total runtime is  $O(nm^2)$**

There are more sophisticated algorithms with  $O(V^2E)$  and even  $O(V^3)$  runtimes (optional: CLRS 26.4, 26.5)

In 2022, researchers found an *almost linear-time* algorithm, which leverages techniques from convex optimization and sophisticated data structures

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