

CS 341: ALGORITHMS

Lecture 17: max flow
Readings: CLRS 26.2

Trevor Brown
<https://student.cs.uwaterloo.ca/~cs341>
trevor.brown@uwaterloo.ca

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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)

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FORD-FULKERSON METHOD

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

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FORD-FULKERSON METHOD

Some 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**

So, what's the residual graph, how do we find an augmenting path, and how do we improve the flow?

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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"

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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** $c(e) - f(e)$
 - If $f(e) > 0$, then R_f contains a **backwards edge** (v, u) with **capacity** $f(e)$ representing flow that could be "pushed back"



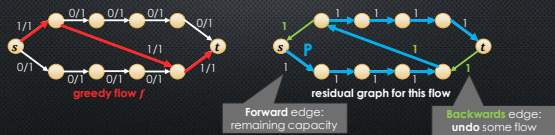
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CONTINUING WITH NEW MATERIAL

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FORD-FULKERSON METHOD

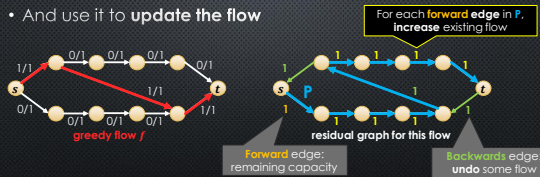
- Find a **shortest path P** from s to t in the **residual graph**
 - If it **improves** the flow, we call it an **augmenting path**
 - And use it to **update the flow**



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FORD-FULKERSON METHOD

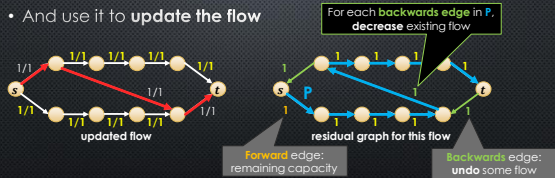
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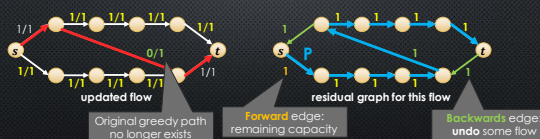
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FORD-FULKERSON METHOD

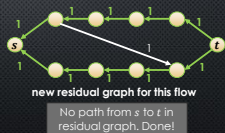
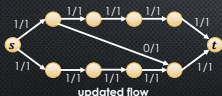
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FORD-FULKERSON METHOD

- 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

no cycles!

- 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

```

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, **augment(f, P) increases the flow by bottleneck(f, P)**

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PROOF

- Claim: **augment(f, P) increases the flow by 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: **augment(f, P) increases the flow by 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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Case 1: forward / forward

flow f

residual graph R_f

augmenting path P in R_f

Let $\text{bottleneck}(f, P) = b$

new flow f' (after augmenting)

Both $f^{in}(u)$ and $f^{out}(u)$ are increased by $\text{bottleneck}(f, P)$

Case 2: backwards / backwards is similar. Both $f^{in}(u)$ and $f^{out}(u)$ are decreased by b

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Case 3: forward / backwards

Let bottleneck(f, P) = b

new flow f' (after augmenting)

Added and subtracted b terms **cancel out**

Case 4: backwards / forwards is similar.

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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

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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**

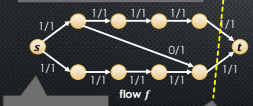
We actually proved the ≤ direction already (**Lemma 2 last time**)
 when discussing upper bounds for max flow

It remains to prove the ≥ direction.

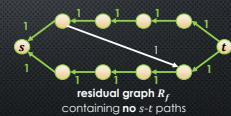
PROVING MAX FLOW ≥ MIN CUT

- 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. $\text{value}(f) = c^{\text{out}}(S)$

Understanding the proposition...



then cut exists with capacity 2 = flow value



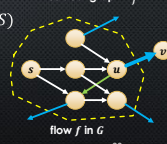
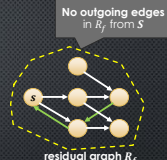
PROVING THE PROPOSITION

- 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?



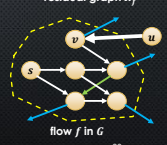
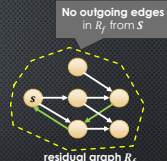
PROVING THE PROPOSITION

- 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



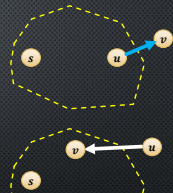
PROVING THE PROPOSITION

- 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



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**



Note this was the last thing remaining to prove the min-cut max-flow theorem, and the correctness of Ford-Fulkerson

TIME COMPLEXITY

of the Ford-Fulkerson method

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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?

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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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