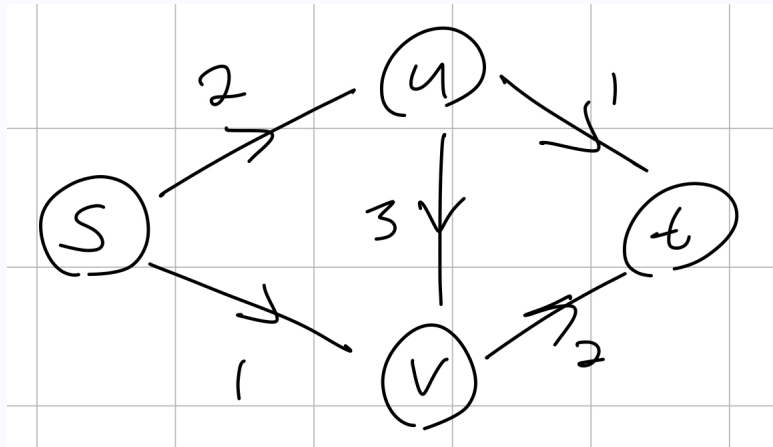


19 Network Flow

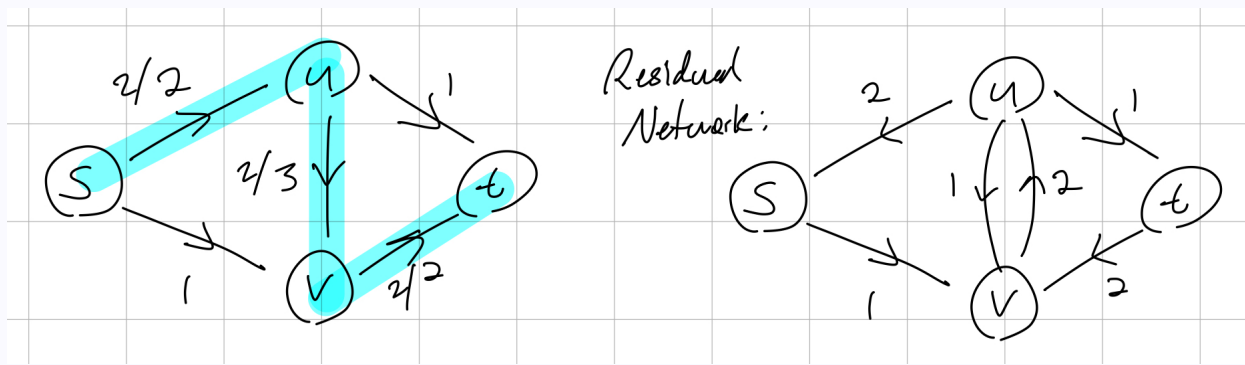
19.1 Flow Networks

Example 19.1

Take the following flow network from last lecture:



Given the example flow from last class, the residual flow network will be the following:



Definition 19.2

An **augmenting path** is a simple path from s to t in the residual network with positive capacity.

19.1.1 Ford-Fulkerson Algorithm

How do we find an augmenting path? We can just use BFS.

How do we augment? For a path P and flow f , let $\text{bottleneck}(P, f)$ be the smallest capacity of any edge on P in the residual network corresponding to f .

Augment(f, P):

```

let b = bottleneck(P, f)
for each edge e = (u, v) along P
  if e is a forward edge
    increase f(e) by b
  else (e is a backward edge)
    decrease f((v, u)) by b
  endif
    
```

Why does Augment(f, P) produce a valid flow?

- If e is a forward edge, we increase the flow by b . Every edge along P has residual capacity at least b , so we satisfy the capacity constraint.

- If e is a backward edge, then we decrease the flow by b (on the original network), which is at most $f(e)$, so the resulting flow is non-negative.
- Whenever we add flow into an internal (not source or sink) vertex, we add the same flow going out.

MaxFlow(G, c, s, t):

Let $f(e) = 0$ for all edges e of G

While there is a simple path P from s to t
in the residual network of flow f

 Update f to Augment(f, P)

 update residual network to use new flow f

Endwhile

return f

19.1.2 Termination and Running Time

Lemma 19.3

The value of the flow strictly increases at every step of the algorithm.

Proof. The first edge of the augmenting path P starts from s . The flow along this edge is increased by $b = \text{bottleneck}(P, f) > 0$. Since P is simple, this is the only edge of P that involves s , so the value of the flow is increased by b . \square

Lemma 19.4

At each step of the algorithm, the flow values and residual capacities are all integers.

Proof. This is true initially. Every step just involves adding/subtracting flows and capacities, so this remains true for the whole algorithm. \square

Thus, the value of the flow increases by at least 1 at every step.

The largest possible flow value is at most

$$C = \sum_{e \in E \text{ leaving } s} c_e$$

So, the algorithm goes through the loop at most C times.

Since each pass through the loop can be implemented in time $O(m+n)$, the overall running time is $O(C \cdot (m+n))$.

19.1.3 Maximum Flows and Minimum Cuts

A **cut** in $G = (V, E)$ is a bipartition of V ($V = A \cup B$, $A \cap B = \emptyset$).

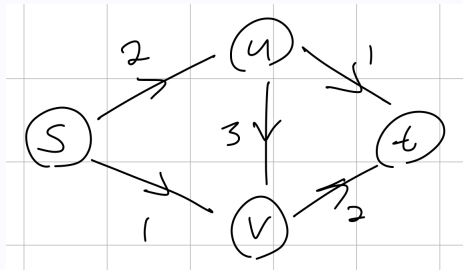
This is an $s - t$ cut if $s \in A$ and $t \in B$.

In a flow network, the **capacity** of an $s - t$ cut is the total capacity of all edges leaving A .

$$c(A, B) = \sum_{e \text{ out of } A} c_e$$

Example 19.5

Take the flow network from before:



The cut $A = \{s\}$ has capacity 3.

The cut $A = \{s, u, v\}$ has capacity 3.

The cut $A = \{s, u\}$ has capacity 5.

The cut $A = \{s, v\}$ has capacity 4.