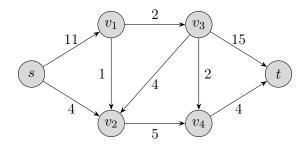
COL751 - Lecture 8

1 Flows and Cuts

A flow network is a (directed/undirected) graph G = (V, E, c) satisfying $c(e) \ge 0$, for each edge e.



For any flow $f: E \to \mathbb{R}^+ \cup \{0\}$, the following constraints must be satisfied:

- 1. Capacity constraint: Flow f(e) through edge e is bounded by its capacity c(e).
- 2. Flow conservation: Flow entering a node x is identical to flow exiting a node x, for every $x \neq s, t$.

We define

$$f_{out}(x) = \sum_{(x, y) \in E} f(x, y).$$

Similarly,

$$f_{in}(x) = \sum_{(y, x) \in E} f(y, x).$$

Definition 1 The value of a flow f, denoted by value(f), is defined as $f_{out}(s)$.

Definition 2 An (s,t)-cut is any partition (A,B) of vertices satisfying $s \in A$, $t \in B$. The capacity of cut (A,B) is defined as

$$\sum_{\substack{(x,y)\in\\(A\times B)\cap E}}c(x,y).$$

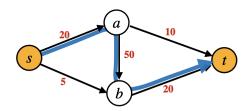
Lemma 3 For any (s,t)-cut (A,B) and any flow f,

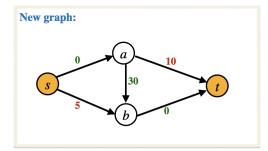
$$value(f) = f_{out}(A) - f_{in}(A).$$

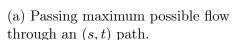
Corollary 4 For any flow f, we have $f_{out}(s) = f_{in}(t)$.

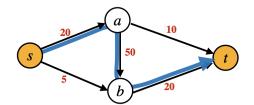
2 Computing (s, t) max-flow

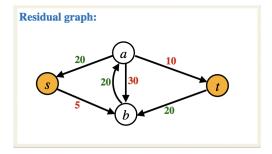
Consider the problem of computing a valid flow f for a G with source s and destination t that maximizes $f_{out}(s)$. A naive greedy approach would be to pass maximum possible flow along a path, and cancel the saturated edges. However, this strategy doesn't work. See Figure (a).











(b) Introducing reverse edges to cancel flow through existing edges.

Given a flow f, we construct a residual graph G_f with respect to f as below:

If $c(x, y) - f(x, y) > 0$	Include (x, y) in G_f with $c_r(x, y) = c(x, y) - f(x, y)$	Forward Edge
If $f(x,y) > 0$	Include (y, x) in G_f with $c_r(y, x) = f(x, y)$	Backward Edge

We will argue that algorithm below computes a max-flow. Note that

- 1. Capacity constraint is satisfied for each edge.
- 2. Flow at each node other than source/destination is conserved.
- 3. Flow increases in each round.
- 4. If all the edges have integer capacity then number of rounds is at most $\sum_{e \in E} c(e)$, or alternatively $O(m \cdot \text{maxflow}(s, t, G))$.

Lemma 5 An (s,t)-flow f is a max-flow if and only if there is no (s,t) path in residual graph G_f .

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1 Initialise f = 0;

2 while (\exists s \to t \ path \ in \ G_f) do

3 | Compute residual graph G_f, and find an (s,t) path P in G_f;

4 | Let c_{min} = \min\{c(e) \mid e \in P\};

5 | foreach (x,y) \in P do

6 | if (x,y) is a forward edge then f(x,y) = f(x,y) + c_{min};

7 | if (x,y) is a backward edge then f(x,y) = f(x,y) - c_{min};

8 | end

9 end
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Algorithm 1: Ford-Fulkerson-algorithm(G, s, t)

Proof: Let f be the max-flow computed from Ford-Fulkerson algorithm. Let A be vertices reachable from s in G_f , and let $B = V \setminus A$. Then, it can be argued that

- 1. For each edge $(x,y) \in A \times B$, f(x,y) = c(x,y).
- 2. For each edge $(x, y) \in B \times A$, f(x, y) = 0.

This proves the claim.

Theorem 6 For any directed/undirected graph $G = (V, E, c : E \to \mathbb{Z}^+)$ with a source s and destination t, an (s, t)-max-flow can be computed in $O(m \cdot \max flow(s, t, G) + n)$ time.

In proof of Lemma 5, we see that there exists an (s, t)-cut (A, B) satisfying c(A, B) = value of (s, t)-max-flow. Since all min-cuts have same size, and their size is lower bounded by (s, t)-flow value, we get the following theorem.

Theorem 7 (Max-Flow Min-Cut Theorem) For any directed/undirected graph G = (V, E, c), the maximum amount of flow passing from a source s to a sink t is equal to the total weight of the edges in a minimum (s, t) cut, i.e., the smallest total weight of the edges which if removed would disconnect the source from the sink.

3 An application of Max-Flow Min-Cut Theorem

Definition 8 An undirected graph G = (V, E) with at least three vertices is said to be

- k-edge connected if for all distinct pairs $(x,y) \in V \times V$, there are k-edge disjoint paths between x and y in G.
- k-vertex connected if for all distinct pairs $(x,y) \in V \times V$, there are k-internally-vertex disjoint paths between x and y in G.

Homework Let G be a k-edge-connected graph. Show how to compute in O(mk) time a sparse subgraph H of G with O(nk) edges such that H is also k-edge-connected.