

Network Flow

Flow Network

Defn: A flow network is a directed graph

$$G = (V, E)$$

with source $s \in V$ and sink $t \in V$, and all $e \in E$ has capacity $c_e \geq 0$

Feasible Flow

An **s-t flow** is a function $f : E \rightarrow \mathbb{R}^+$.

A feasible flow satisfies:

- (i) Capacity constraint: $\forall e \in E : 0 \leq f(e) \leq c_e$
- (ii) Conservation of flow: $\forall v \in V$ except s, t : $\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$

Residual Network

Defn: Let G be a flow network and f be a flow on G . We define G_f to be the residual network of f on G such that:

- G_f has the same nodes as G
- For each edge $e = (u, v)$ where $f(e) < c_e$, we include $e = (u, v)$ in G_f with a capacity of $c_e - f(e)$ (forward edges)
- For each edge $e = (u, v)$ where $f(e) > 0$, we include $e' = (v, u)$ in G_f with a capacity of $f(e)$ (backward edges)

The **Residual Capacity** is the capacity of an edge in G_f .

Ford-Fulkerson Algorithm

Let P be a simple (does not revisit any node) s-t path in G_f . Define $\text{bottleneck}(P, f)$ to be the minimum residual capacity of any edge in P due to f .

Augment is a function that yields a new flow f' in G :

Algorithm 1: $\text{augment}(f, P)$

```
1 Let  $b \leftarrow \text{bottleneck}(P, f)$ ;  
2 for each edge  $(u, v) \in P$  do  
3   if  $e = (u, v)$  is a forward edge then  
4     increase  $f(e)$  in  $G$  by  $b$ ;  
5   else  
6      $(u, v)$  is a backward edge, and let  $e = (v, u)$ ;  
7     decrease  $f(e)$  in  $G$  by  $b$ ;  
8   end  
9 end  
10 return  $f$ ;
```

Algorithm 2: $\text{Ford-Fulkerson}(G, s, t)$

```
1 Initially  $f(e) = 0$  for all  $e$  in  $G$ ;  
2 while there is an s-t path in the residual graph  $G_f$  do  
3   Let  $P$  be a simple s-t path in  $G_f$ ;  
4    $f' = \text{augment}(f, P)$ ;  
5   Update  $f$  to be  $f'$ ;  
6   Update the residual graph  $G_f$  to be  $G_{f'}$ ;  
7 end  
8 return  $f$ ;
```

Min-Cut

For any node $v \in V$ in G let us define:

$$f^{\text{out}}(v) = \sum_{e \text{ out of } v} f(e)$$

$$f^{\text{in}}(v) = \sum_{e \text{ into } v} f(e)$$

Defn: An s - t cut is a partition (S, T) such that:

- $s \in S$
- $t \in T$
- $S \cup T = V$

The **capacity** of (S, T) is the sum of capacities of edges that go from S into T :

$$c(S, T) = \sum_{u \in S, v \in T} c(u, v)$$

The **flow** of (S, T) with respect to a flow f is:

$$v(f) = f^{\text{out}}(S) - f^{\text{in}}(S)$$

A **Min-Cut** is a cut with minimum capacity among all s - t cuts.

Max-Flow Min-Cut Theorem

Let G be a flow network with source s and sink t .

A feasible flow f is a max-flow \iff there is no s - t path in the residual network G_f (e.g. no augmenting path in G_f)
 $\iff v(f) = c(S, T)$ for some min-cut s - t cut (S, T)

Complexity

Undecidable Decision Problems

There exists undecidable decision problems, such as the Halting problem:

Given a program P and input x , decide whether $P(x)$ terminates.

Class P and NP

P: The class of decision problems solvable in polynomial time.

NP: The class of decision problems whose solutions can be verified in polynomial time.

Defn: A decision problem $f \in NP$ if there exists a program $A \in P$ such that:

$$f(x) = 1 \iff \exists \text{ witness } w \in \{0, 1\}^* \text{ such that } A(x, w) = 1$$

Karp Reduction

Defn: Problem f polynomial Karp-reduces to g if there exists an algorithm $R \in P$ such that:

$$f(x) = 1 \iff g(R(x)) = 1$$

And we write $f \leq_m g$

NP-Complete

Defn: A problem g is NP-complete if $\forall f \in NP$:

$$f \leq_m g$$

Cook-Levin Theorem

The Boolean Satisfiability Problem (SAT) is NP-complete.