CSS.201.1 Algorithms

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TIFR 2024, Aug-Dec

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Contents

CHAPTER 1	Maximum Flow	Page 3
1.1	Flow	3
1.2	Ford-Fulkerson Algorithm	4
	1.2.1 Max Flow Min Cut1.2.2 Edmonds-Karp Algorithm	6 8
13	Preflow-Puch Puch Relabel Algorithm	Q

Maximum Flow

1.1 **Flow**

Suppose we are given a directed graph G = (V, E) with a source vertex s and a target vertex t. And additionally for every edge $e \in E$ we are given a number $c_e \in \mathbb{Z}_0$ which is called the capacity of the edge.

Definition 1.1.1: Flow

An s-t flow is a function $f: E \to \mathbb{R}_0$ which satisfies the following:

$$\bigcirc$$
1) $\forall e \in E, f(e) \leq c_e$

①
$$\forall e \in E, f(e) \le c_e$$
② $\forall v \in V \setminus \{s, t\}, \sum_{e \in in(v)} f(e) = \sum_{e \in out(v)} f(e)$

Also the value of a flow f is denoted by $|f| := \sum_{e \in out(s)} f(e)$.

Before proceeding into the setup and the problem first we will assume some things

Assumption. • $in(s) = \emptyset$ i.e. there is no edge into s.

- $out(t) = \emptyset$ i.e. there is no edge out of t.
- There are no parallel edges

Lemma 1.1.1

For any flow f, $|f| = \sum_{e \in in(t)} f(e)$

Proof: We have for every edge $e \in E$, $\exists v \in V$ such that $e \in in(v)$ and $\exists u \in V$ such that $e \in out(u)$. Hence we get

$$\sum_{e \in E} f(e) = \sum_{v \in V} \sum_{e \in in(v)} f(e) = \sum_{v \in V} \sum) e \in out(v) f(e) \implies \sum_{v \in V} \left[\sum_{e \in in(v)} f(e) - \sum_{e \in out(v)} f(e) \right] = 0$$

Now we know $\forall v \in V \setminus \{s, t\}$. $\sum_{e \in in(v)} f(e) = \sum_{e \in out(v)} f(e)$. Therefore we get

$$\sum_{v \in V} \left[\sum_{e \in in(v)} f(e) - \sum_{e \in out(v)} f(e) \right] = 0 \implies \sum_{v \in \{s,t\}} \left[\sum_{e \in in(v)} f(e) - \sum_{e \in out(v)} f(e) \right] = 0 \implies \sum_{e \in out(s)} f(e) - \sum_{e \in in(t)} f(e)$$

Hence we have $|f| = \sum_{e \in in(t)} f(e)$.

Max Flow

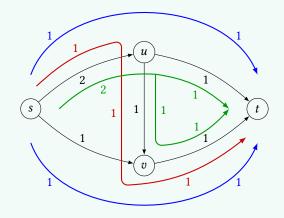
Input: A directed graph G = (V, E) with source vertex s and target vertex t and for all edge $e \in E$ capacity

of the edge $c_e \in \mathbb{Z}_+$

Question: Given such a graph and its capacities find an s - t flow which has the maximum value

Example 1.1.1

Consider the following directed graph with capacities: $V = \{s, t, u, v\}$, $c_{s,u} = 2$, $c_{s,v} = c_{u,t} = c_{v,t} = c_{u,v} = 1$. Firstly the following function: f': f'(s, u) = 2 = f(u, t). It is not a flow since $f(u, t) = 2 > 1 = c_{u,t}$. Now we define three different flow functions:



- f: f(s, u) = f(u, v) = f(v, t) = 1 and otherwise 0. Therefore |f| = 1
- g: g(s, u) = g(u, t) = 1, g(s, v) = g(v, t) = 1 and otherwise 0. Therefore |g| = 2
- h: h(s, u) = 2, h(u, t) = h(u, v) = h(v, t) = 1 and otherwise 0. Therefore |h| = 2

Notice here q and h has the maximum flow value.

1.2 Ford-Fulkerson Algorithm

Definition 1.2.1: Residual Graph

Given a directed graph G = (V, E) and capacities C_e for all $e \in E$ and an s - t flow f the residual graph $G_f = (V, E_f)$ has the edges with the following properties:

- ① If $(u,v) \in E$ and f(u,v) > 0 then $(v,u) \in E_f$ and $c_{v,u}^f = f(u,v)$. Such an edge is called a *backward* edge.
- ② If $(u, v) \in E$ and $f(u, v) < c_{u,v}$ then $(u, v) \in E_f$ and $c_{u,v}^f = c_{u,v} f(u,v)$. It is called *forward* edge.

Algorithm 1: Ford-Fulkerson

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Input: Directed graph G=(V,E), source s, target t and edge capacities C_e for all e\in E
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Output: Flow f with maximum value

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1 begin
2 | for e \in E do
3 | \int f(e) = 0
4 | while \exists s \leadsto t \ path \ P \ in \ G_f do
5 | \delta \longleftarrow \min_{e \in P} \{c_e^f\} for e = (u, v) \in P do
6 | if e is Forward Edge then
7 | \int f(u, v) \longleftarrow f(u, v) + \delta
8 | else
9 | \int f(u, v) \longleftarrow f(v, u) - \delta
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Page 5 Chapter 1 Maximum Flow

Lemma 1.2.1

At any iteration the f' obtained after the flow augmentation of the flow f is a valid flow

Proof: At any iteration let P be the path from $s \rightsquigarrow t$ and $\delta = \min_{e \in P} c_f(e)$. Let f' be the new function such that for each $(u,v) \in P$ if (u,v) is forward edge in G_f then $f'(u,v) = f(u,v) + \delta$ and if (u,v) is backward edge in G_f then $f'(v,u) = f(v,u) - \delta$ and for other edges $e \in E \setminus P$, f'(e) = f(e).

Now since $\delta = \min_{e \in P} c_f(e)$, $c_f(e) \ge \delta$ for all $e \in P$. Hence if (u, v) is backward edge then $(v, u) \in E$ and $c_f(u, v) = f(u, v)$. Hence $f'(v, u) = f(v, u) - \delta \ge 0$. Therefore for all $e \in E$, $f'(e) \ge 0$.

Now first we will show $f'(e) \le c_e$ for all $e \in E$. If $(u,v) \in P$ is a forward edge then $(u,v) \in E$ and $c_f(u,v) = c_{u,v}f(u,v)$. Therefore $f'(u,v) = f(u,v) + \delta \le f(u,v) + c_{u,v} - f(u,v) = c_{u,v}$. Now if $(u,v) \in P$ is a backward edge then $(v,u) \in E$ and $c_f(u,v) = f(u,v)$. Therefore $f'(u,v) = f(u,v) - \delta \le f(u,v) \le c_{u,v}$. For other edges $e \in E \setminus P$, $f'(e) = f(u) \le c_e$. Therefore $f'(e) \le c_e$ for all $e \in E$

Now we will prove for all $v \in V \setminus \{s, t\}$, $\sum_{e \in in(v)} f'(e) = \sum_{e \in out(v)} f'(e)$. If v is not in the path P in G_f then, f'(e) = f(e)

for all edges $e \in in(v) \cup out(v)$. Hence the condition is satisfied for such vertices. Suppose v is in the path P. Then there are two edges e_1 and e_2 in P which are incident on e. If both are forward edges or both are backward edges then one of them is in in(v) and other one is in out(v). WLOG suppose $e_1 \in in(v)$ and $e_2 \in out(v)$ we have

$$\sum_{e \in in(v)} f'(e) = \sum_{e \in in(v) \setminus \{e_1\}} f(e) + f(e_1) \pm \delta = \sum_{e \in out(v) \setminus \{e_2\}} f(e) + f(e_2) \pm \delta = \sum_{e \in out(v)} f'(e)$$

If one of e_1 , e_2 forward edge and other one is backward edge then either e_1 , $e_2 \in in(v)$ (when e_1 is forward and e_2 is backward) or e_1 , $e_2 \in out(v)$ (when e_1 is backward and e_2 is forward). Now if e_1 , $e_2 \in in(v)$, $f'(e_1) + f'(e_2) = f(e_1) + \delta + f(e_2) - \delta = f(e_1) + f(e_2)$ and if e_1 , $e_2 \in out(v)$ then $f'(e_1) + f'(e_2) = f(e_1) - \delta + f'(e_2) + \delta = f(e_1) + f(e_2)$. Hence

$$\sum_{e \in in(v)} f'(e) = \sum_{e \in in(v)} f(e) = \sum_{e \in out(v)} f(e) = \sum_{e \in out(v)} f'(e)$$

Hence f' is a valid flow.

Lemma 1.2.2

At any iteration Given G_f if the flow, f' obtained after flow augmentation of f by δ then

$$|f'| = |f| + \delta$$

Proof: Since we augment flow along an $s \rightsquigarrow t$ path, the first edge of the path is always in out(s). Let the first edge is e = (s, u). Now e has to be a forward edge because otherwise $(u, s) \in E$ and then there is an incoming edge in G which is not possible. Hence

$$|f'| = \sum_{e \in out(s)} f'(e) = \sum_{e \in out(s) \setminus \{e\}} f(e) + f'(e) = \sum_{e \in out(s) \setminus \{e\}} f(e) + f(e) + \delta = \sum_{e \in out(s)} f(e) + \delta = |f| + \delta$$

Hence we have the lemma.

Lemma 1.2.3

At every iteration of the Ford-Fulkerson Algorithm the flow values and the residual capacities of the residual graph are non-negative integers.

Proof: Initial flow and the residual capacities are non-negative integers. Let till i^{th} iteration the flow values and the residual capacities were non-negative integers. Let the flow after i^{th} iteration was f. Hence $\forall e \in E, f(e) \in \mathbb{Z}_0$. Therefore in the G_f for all $e \in E_f$, $c_f(e) \in \mathbb{Z}_0$. Hence $\delta \in \mathbb{Z}_0$. Therefore $\forall e \in E, f'(e) \in \mathbb{Z}_0$. And therefore for all $e \in E_{f'}$ where $G_{f'}$ is the residual graph of the flow f', $c_{f'}(e) \in \mathbb{Z}_0$. Hence by mathematical induction the lemma follows.

At any iteration let P be the path from $s \rightsquigarrow t$. Then for all $e \in P$, $c_f(e) > 0$. Therefore $\delta = \min_{e \in P} c_f(e) \ge 1$. Therefore the algorithm must stop in at most $\sum_{e \in out(s)} c_e$ since we can have the value of a flow to be at max the value of the sum of capacities of edges in out(s) and therefore we can increase the flow at max that many times.

Lemma 1.2.4

If f is a max flow then there is no $s \rightsquigarrow t$ path in G_f .

Proof: Suppose there is an $s \rightsquigarrow t$ path P in G_f . We will show that then f is not a max flow following the algorithm. Then $\forall e \in P$, $c_f(e) > 0$. Hence $\delta = \min_{e \in P} c_f(e) \ge 1$. Now after the flow augmentation process of f by δ we get a new valid flow f' by Lemma 1.2.1 and by Lemma 1.2.2 we have $|f'| = |f| + \delta > ||f|$. Hence f is not a maximum flow. Hence contradiction. Therefore there is no $s \rightsquigarrow t$ path in G_f .

1.2.1 Max Flow Min Cut

Definition 1.2.2: Cut Set

For a graph G = (V, E) and a subset $A \subseteq V$, the cut $(A, V \setminus A)$ is a bipartition of V where the edges E_A of the graph $G_A = (A, V \setminus A, E_A)$ is the set $E_A = E \cap (A \times (V \setminus A))$.

Now if s, t are two vertices of G then an s - t Cut $(A, V \setminus A)$ is a cut such that $s \in A$ and $t \in V \setminus A$.

Now we define for a cut $(A, V \setminus A)$ the Capacity of the Cut $(A, V \setminus A) = \sum_{e \in E_A} c_e$. For an s - t cut $(A, V \setminus A)$ we denote the capacity of the cut by cap(A) A Min s - t Cut is a s - t cut of minimum capacity. Then we have the following relation between cut and flow.

Lemma 1.2.5

Given a graph G = (V, E), $s, t, c_e \in \mathbb{Z}_0$ for all $e \in E$ for any flow f and a s - t cut $(A, V \setminus A)$

$$|f| \le cap(A)$$

Proof: Given f and the s - t cut $(A, V \setminus A)$ we have

$$|f| = \sum_{e \in out(s)} f(e)$$

$$= \sum_{v \in A} \left[\sum_{e \in out(v)} f(e) - \sum_{e \in in(v)} f(e) \right]$$

$$= \sum_{e = (u,v), \ u \notin A, v \notin A} f(e) - \sum_{e = (u,v), \ u \notin A, v \in A} f(e)$$

$$= \sum_{e \in out(A)} f(e) - \sum_{e \in in(A)} f(e)$$

$$\leq \sum_{e \in out(A)} f(e) \leq \sum_{e \in out(A)} c_e = cap(A)$$
[Edges for both endpoints in A are canceled out]

Hence we have the lemma.

Having this lemma we have for any flow f and s - t cut $(A, V \setminus A)$ we have

$$|f| \le cap(A) \implies \max_{f} |f| \le \min_{s-t \ cut \ (A, V \setminus A)} cap(A)$$

So we have the following theorem that the value of maximum flow is equal to the capacity of minimum cut.

Page 7 Chapter 1 Maximum Flow

Theorem 1.2.6 Max Flow Min Cut

Given a graph G = (V, E), $s, t, c_e \in \mathbb{Z}_0$ for all $e \in E$. Then the following are equivalent:

- (1) f is a maximum flow.
- (2) There is no $s \rightsquigarrow t$ path in G_f
- (3) There exists an s t cut of capacity |f|

Proof:

- $(1) \Longrightarrow (2)$: This is by Lemma 1.2.4.
- (2) \Longrightarrow (3): We are given a flow f such that there is no $s \rightsquigarrow t$ path in G_f . We will construct a s-t cut which has the capacity |f|. Now take A to be all the vertices reachable from s in G_f . This is a valid s-t cut since $s \in A$ and as there is no $s \rightsquigarrow t$ path in G_f , $t \notin A$. Now

$$|f| = \sum_{e \in out(A)} f(e) - \sum_{e \in in(A)} f(e)$$

Now $\forall e = (u, v) \in E$ where $u \in A$ and $v \notin A$ we have $c_{u,v} = f(u,v) \implies c_{u,v} - f(u,v) = 0$ since otherwise $c_{u,v} - f(u,v) \neq 0 \implies c_{u,v} > f(u,v) \implies (u,v) \in E_f$ and therefore v is reachable from $v \notin A$ contradiction. Therefore $v \in A$ and $v \in A$ we have $v \in A$ and $v \in A$ we have $v \in A$ and $v \in A$ we have $v \in A$ and $v \in A$ and $v \in A$ and $v \in A$ contradiction. Hence we have

$$|f| = \sum_{e \in out(A)} f(e) - \sum_{e \in in(A)} f(e) = \sum_{e \in out(A)} c_e = cap(A)$$

(3) \Longrightarrow (1): Now by Lemma 1.2.5 we have for any flow f and s - t cut

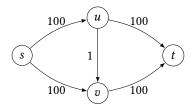
$$|f| \le cap(A) \implies \max_{f} |f| \le \min_{s-t \ cut \ (A, V \setminus A)} cap(A)$$

Now given f there exists an s - t cut of capacity |f|. Hence f is a max flow.

Hence at the end of the Ford-Fulkerson Algorithm let the flow returned by the algorithm is f. The algorithm terminates when there is no $s \rightsquigarrow t$ path in G_f . Hence by Max Flow Min Cut Theorem we have f is a maximum flow. This completes the analysis of the Ford-Fulkerson Algorithm.

Since the capacities of the edges can be very large we want an algorithm return the maximum flow with running time $poly(n, m, \log c_e)$ where n is the number of vertices and m is number of edges and $\log c_e$ basically means number of bits at most needed to represent the capacities.

But Ford-Fulkerson algorithm takes does not run in $poly(n, m, \log c_e)$ instead $poly(n, m, c_e)$ as the while loop in the algorithm takes $poly(c_e)$ many iterations. For example in the following graph: it takes around 100 steps



and in general Ford-Fulkerson takes $O(|f_{\text{max}}|)$ time. For this reason we will now discuss a modification of the Ford-Fulkerson Algorithm which takes $poly(n, m, \log c_e)$ time, Edmonds-Karp Algorithm.

1.2.2 Edmonds-Karp Algorithm

To get a $poly(n, m, \log c_e)$ time algorithm we will always pick the shortest $s \rightsquigarrow t$ path in the residual graph. This algorithm is known as the Edmonds-Karp Algorithm

Suppose f_i be the total flow after i^{th} iteration. And G_{f_i} be the residual graph with respect f_i . Then $f_0(e) = 0$ for all $e \in E$ and $G_{f_0} = G$. Also suppose $dist_i(v) = \text{Shortest } s \rightsquigarrow v$ path distance in the residual graph G_{f_i} . Hence $dist_i(s) = 0$ for all i and $dist_i(t) = \infty$ at the end of the algorithm.

Note:-

In i^{th} iteration of the Ford-Fulkerson Algorithm or Edmonds-Karp Algorithm if P is the $s \leadsto t$ in the residual graph G_{f_i} where $e = (u, v) \in P$ and $c_{f_i}(u, v) = \delta = \min_{e \in P} c_{f_i}(e)$ then the edge (u, v) is not present in the next residual graph $G_{f_{i+1}}$. Thus at least one edge disappears in each iteration of Ford-Fulkerson or Edmonds-Karp Algorithm.

Now we will prove following two lemmas which will help us to prove that the Edmond-Karp algorithm takes O(mn) iterations.

Lemma 1.2.7

At any iteration $i, \forall v \in V, dist_i(v) \leq dist_{i+1}(v)$

Lemma 1.2.8

For any edge $e = (u, v) \in E$ the number of iterations where either (u, v) appears or (v, u) appears is at most O(n) i.e.

$$\left|\left\{i\colon (u,v)\notin G_{f_{i}},(u,v)\in G_{f_{i+1}}\right\}\right|+\left|\left\{i\colon (v,u)\notin G_{f_{i}},(v,u)\in G_{f_{i+1}}\right\}\right|=O(n)$$

1.3 Preflow-Push/Push Relabel Algorithm