

### **Lecture Notes - Week III**

**Flows** 

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# CHAPTER 1 Flows and Cuts

In **graph theory**, **flow network** is a **directed** graph G = (V, E) where each edge has a **capacity**  $u: E \to \mathbb{R}_+$  and each edge receives a **flow**  $f: E \to \mathbb{R}_+$ , where the amount of flow allowed in each edge cannot surpass its capacity  $(f(e) \le u(e), e \in E)$ . Hence, the *excess* of a flow f at  $v \in V$ :

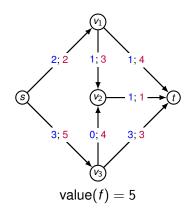
$$\operatorname{\mathsf{ex}}_f(v) := \sum_{e \in \delta^-(v)} f(e) - \sum_{e \in \delta^+(v)} f(e)$$

 $\delta^-(v) = \{e \in E : e = (u, v)\}$  incoming edges

 $\delta^+(v) = \{e \in E \colon e = (v, u)\}$  outgoing edges

The flow in this type of graph also have the satisfy **flow conservation** which state that:

**Definition 1** The total net flow entering a node v is zero for **all nodes** in the network except the source s and sink t.





A **cut** in graph theory corresponds to a **partition** of the nodes in a graph splitting them into **disjoint subsets**. For example, see Figure 1.1.

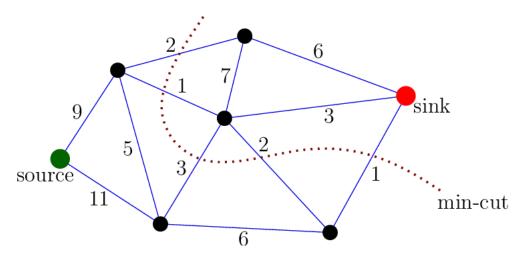


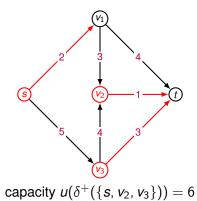
Figure 1.1: Example of a cut in a graph

A specific type of cut is a *s-t-cut*  $\delta^+(S)$  where  $S \subseteq V$  and  $s \in S$ ,  $t \notin S$ . Therefore:

$$\delta^+(S) = \{e = (u, v) \in E \colon u \in S, v \in V \setminus S\}$$

The capacity of such cut can be expressed as:

$$u(\delta^+(S)) = \sum_{e \in \delta^+(S)} u(e)$$





#### 1.1 WEEK DUALITY

Using the definitions of flows and cuts, we can establish the following conclusion:

**Lemma 1** For any  $S \subseteq V$  with  $s \in S$ ,  $t \notin S$  and any s-t-flow f:

1. 
$$value(f) = \sum_{e \in \delta^+(S)} f(e) - \sum_{e \in \delta^-(S)} f(e)$$

2. value(
$$f$$
)  $\leq u(\delta^+(S))$ 

**Proof 1** From the flow conservation for  $v \in S \setminus \{s\}$ :

$$\begin{aligned} \text{value}(f) &= -\mathsf{ex}_f(s) \\ &= \sum_{e \in \delta^+(s)} f(e) - \sum_{e \in \delta^-(s)} f(e) \\ &= \sum_{v \in S} \left( \sum_{e \in \delta^+(v)} f(e) - \sum_{e \in \delta^-(v)} f(e) \right) \\ &= \sum_{e \in \delta^+(S)} f(e) - \sum_{e \in \delta^-(S)} f(e) \end{aligned}$$

This can also expressed as:

$$0 \le f(e) \le u(e)$$



# CHAPTER 2

### Maximum Flows and Minimal Cuts

Once again, the task of find which flow and which cuts a graph can accept is **not challenging**. However, whenever **optimal values** (either minimal or maximal) are required, the configuration of such problems becomes challenging.

First, we state both problems:

**Problem 1** Maximum Flow Problem (MaxFlow) Given a flow network represent as a digraph G = (v, E) with capacities u and unique source and unique sink s and t respectively, such that  $s, t \in V$ . The goal is to find an s-t-flow of **maximum** value.

**Problem 2** Minimum Cut Problem (MinCut) Given a flow network represent as a digraph G = (v, E) with capacities u and unique source and unique sink s and t respectively, such that  $s, t \in V$ . The goal is to find an s-t-cut of **minimum capacity**.

Although those two problems might seem **unrelated** or even **contradictory**, they can be directly connected via the following lemmas:

**Lemma 2** Let G = (V, E) be a digraph with capacities u and s,  $t \in V$ . Then

$$\max\{\text{value}(f): f \text{ } s\text{-}t\text{-flow}\} \leq \min\{u(\delta^+(S)): \delta^+(S) \text{ } s\text{-}t\text{-cut}\}.$$

**Lemma 3** Let G = (V, E) be a digraph with capacities u and  $s, t \in V$ . Let f be an s-t-flow and  $\delta^+(S)$  be an s-t-cut. If

$$value(f) = u(\delta^+(S))$$

then f is a maximal flow and  $\delta^+(S)$  is a minimal cut.

Hence, a single algorithm is enough to solve **both** problems.

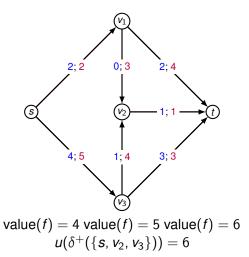
**Remark:** in combinatorics, many problems can be expressed as another. This is a key point for future lectures.



#### 2.1 IDEA FOR FINDING MAXIMAL FLOWS

If there exists non-saturated s-t-path (f(e) < u(e) for all edges), then the flow f can be increased along this path. This means that if the path is not satured, more flow can be put into that path.

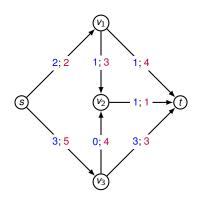
However, non-existence of such a path does not guarantee optimality.



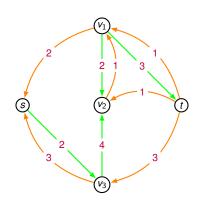
In this context, we introduced another concept: **residual graphs**. Considering that G = (V, E) is a digraph with capacities u, f be an s-t-flow, a residual graph is the graph  $G_f = (V, E_f)$  with  $E_f = E_+ \cup E_-$  and capacity  $u_f$ :

- forward edges  $+e \in E_+$ : for  $e = (u, v) \in E$  with f(e) < u(r), add +e = (u, v) with residual capacity  $u_f(+e) = u(e) - f(e)$
- backward edges  $-e \in E_-$ : for  $e = (u, v) \in E$  with f(e) > 0, add -e = (v, u) with residual capacity  $u_f(-e) = f(e)$

**Remark:**  $G_f$  can have parallel edges even if G is simple.







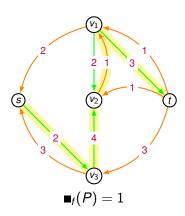
In addition, we can also define *f*-augmenting paths:

**Definition 2** An s-t-path P in  $G_f$  is called augmenting path. The value:

$$\blacksquare_f(P) = \min_{a \in E(P)} u_f(a)$$

is called residual capacity of P.

**Remark:**  $\blacksquare_f(P) > 0$  as  $u_f(a) > 0$  for all  $a \in E_f$ .



With this definition in mind, the following theorem is established.

**Theorem 1** An s-t-flow is optimal if and only if there exists no f-augmenting path.

Proof idea:

⇒ *P f*-augmenting path. Construct *s*-*t*-flow

$$\bar{f}(e) = \begin{cases} f(e) + \blacksquare_f(P) & \text{if } + e \in E(P) \\ f(e) - \blacksquare_f(P) & \text{if } - e \in E(P) \\ f(e) & \text{otherwise} \end{cases}$$

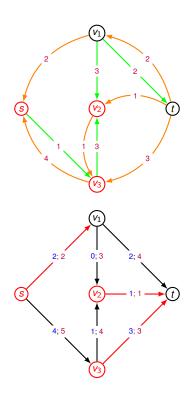
with higher value.



#### Proof idea:

 $\Leftarrow$  There exists no f-augmenting path. Consider s-t-cut  $\delta^+(S)$  defined by connected component S of s in  $G_f$ . Show that

$$value(f) = u(\delta^+(S)).$$



With this previous theorem in mind, we can conclude that:

#### Theorem 2 (Ford and Fulkerson, 1956; Dantzig and Fulkerson, 1956)

In a digraph G with capacities u, the maximum value of an s-t-flow equals the minimum capacity of an s-t-cut.



# CHAPTER 3

## Finding Maximal Flows

The most common algorithm for maximum flow was first published by L. R. Ford Jr. and D. R. Fulkerson in in 1956. It is commonly known as **Ford-Fulkerson algorithm**. The algorith is as follows:

```
Algorithm: FORD-FULKERSON ALGORITHM
```

**Input:** digraph G = (V, E), capacities  $u: E \to \mathbb{Z}_+$ ,  $s, t \in V$ 

Output: maximal s-t-flow f 1 set f(e) = 0 for all  $e \in E$ 

2 while there exists f-augmenting path in G<sub>f</sub> do

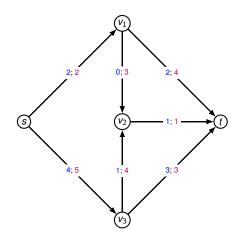
choose *f*-augmenting path *P* 

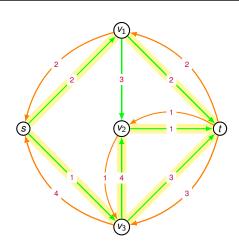
4 set  $\blacksquare_f(P) = \min_{a \in E(P)} u_f(a)$ 

augment f along P by  $\blacksquare_f(P)$ 

6 update  $G_f$ 

#### 7 return f





$$\blacksquare_f(P) = 3$$

$$■_f(P) = 2$$

$$\blacksquare_f(P)=1$$

Analysing the previous algorithms allow us to infer a few details. Lines 1, 4, 5 and 6 can be calculated in **linear time** in terms the number of edges m in a graph. An efficient algorithm to apply in Line 3 is actually **DFS** (Depth-First Search) which is also **linear** in the number of edges m. The *WHILE* loop requires up to  $n \cdot U$ , where n is the number of nodes and U is  $max_{e \in E}u(e)$ . The entire algorithm has a runtime proportional to  $O(n \cdot m \cdot U)$  (**polynomial**).



**Remark**: flow *f* is integer.

An improved version of this algorithm allows for real values in the capacities. In this case, for non-integer capacities,  $\blacksquare_f$  can be arbitrarily small when P is not chosen carefully, resulting in a runtime  $O(n \cdot m^2)$ .

The resulting algorithm represent such adaption:

```
Algorithm: EDMONDS-KARP ALGORITHM
```

**Input:** digraph G = (V, E), capacities  $u: E \to \mathbb{R}_+$ ,  $s, t \in V$ 

Output: maximal s-t-flow f 1 set f(e) = 0 for all  $e \in E$ 

**2 while** there exists f-augmenting path in  $G_f$  do

choose f-augmenting path P with minimal number of edges 3

set  $\blacksquare_f(P) = \min_{a \in E(P)} u_f(a)$ 

augment f along P by  $\blacksquare_f(P)$ 5

update  $G_f$ 6

7 return f

Last but not least, there is also linear programming formulation for this problem. See full model below:

$$\sum_{e \in \delta^+(s)} f_e \tag{3.1a}$$

max 
$$\sum_{e \in \delta^+(s)} f_e \tag{3.1a}$$
 s.t. 
$$\sum_{e \in \delta^-(v)} f_e - \sum_{e \in \delta^+(v)} f_e = 0 \qquad v \in V \setminus \{s,t\} \tag{3.1b}$$

$$f_e \le u(e)$$
  $e \in E$  (3.1c)

$$f_e \ge 0$$
  $e \in E$  (3.1d)

The flow conservation flow conversation constraints (3.1b) are part of many LPs and IPs, e.g. for shortest path. The coefficient matrix of flow conversation constraints is node-arc-incidence matrix and it is totally unimodular, i.e., all extreme points are integer.