# Computer Vision I

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https://mlcv.cs.tu-dresden.de/courses/24-winter/cv1/

Winter Term 2024/2025

Excursus: Maximum st-Flow and Minimum st-Cut

- Maximum st-Flow Problem
- Residual networks and augmenting paths
- Minimum st-Cut Problem
- ► Maximum *st*-Flow/Minimum *st*-Cut Theorem
- ► Ford-Fulkerson-Algorithm

For any directed graph (V, E), any  $U \subseteq V$  and any  $W \subseteq V$  let  $\frac{UW}{W} := \{uv \in E \mid u \in U \land w \in W\} \ .$ 



**Definition 1.** For any directed graph (V, E) and any  $f \in \mathbb{N}_0^E$ , the maps  $\varphi^+, \varphi^-, \varphi: 2^V \to \mathbb{Z}$  such that

$$\forall U \in 2^V \quad \varphi_U^+ = \sum_{uv \in UU^c} f_{uv} \tag{1}$$

$$\varphi_U^- = \sum_{vu \in U^c U} f_{vu} \tag{2}$$

$$\varphi_U = \varphi_U^+ - \varphi_U^- \tag{3}$$

are called the **outflux**, influx and flux in (V, E) wrt. f.



For any  $u \in V$ ,

$$\varphi_u^+ := \varphi_{\{u\}}^+$$
$$\varphi_u^- := \varphi_{\{u\}}^-$$
$$\varphi_u^- := \varphi_{\{u\}}$$

are called the  ${\rm outflux},$  influx and flux of u in (V,E) wrt. f.



Lemma 1. For any directed graph (V, E), any  $f \in \mathbb{N}_0^E$  and any  $U \subseteq V$ 

$$\varphi_U = \sum_{u \in U} \varphi_u \quad . \tag{4}$$



# Proof.

$$\begin{split} \varphi_U &= \sum_{uv \in UU^c} f_{uv} - \sum_{vu \in U^c U} f_{vu} \\ &= \left( \sum_{uv \in UV} f_{uv} - \sum_{uu' \in UU} f_{uu'} \right) - \left( \sum_{vu \in VU} f_{vu} - \sum_{u'u \in UU} f_{uu'} \right) \\ &= \sum_{uv \in UV} f_{uv} - \sum_{vu \in VU} f_{vu} \\ &= \sum_{u \in U} \left( \sum_{vw \in \{u\} \{u\}^c} f_{vw} - \sum_{vw \in \{u\}^c \{u\}} f_{vw} \right) \\ &= \sum_{u \in U} \varphi_u . \end{split}$$

**Definition 2.** A 5-tuple N = (V, E, s, t, c) is called a **network** iff (V, E) is a directed graph and  $s \in V$  and  $t \in V$  and  $s \neq t$  and  $c \in \mathbb{N}^{E}$ .

The nodes s and t are called the **source** and the **sink** of N, respectively.

For any edge  $e \in E$ ,  $c_e$  is called the **capacity** of e in N.

**Definition 3.** A map  $f \in \mathbb{N}_0^E$  is called an *st*-preflow in a network N = (V, E, s, t, c) iff

$$\forall e \in E \quad 0 \le f_e \le c_e \tag{5}$$

$$\forall v \in V - \{s\} \quad \varphi_v \le 0 \quad . \tag{6}$$

An st-preflow f in N is called an st-flow in N iff, in addition,

$$\forall v \in V - \{s, t\} \quad \varphi_v = 0 \quad . \tag{7}$$

Definition 4. The instance of the Maximum  $st\mbox{-Flow Problem}$  wrt. a network N=(V,E,s,t,c) is to

$$\max_{f \in \mathbb{N}_0^E} \quad \sum_{sv \in E} f_{sv} - \sum_{vs \in E} f_{vs} \tag{8}$$

subject to  $\forall e \in E \quad 0 \le f_e \le c_e$  (9)

$$\forall v \in V - \{s, t\} \quad \sum_{vw \in E} f_{vw} = \sum_{uv \in E} f_{uv} \quad . \tag{10}$$

Note:

$$\sum_{sv \in E} f_{sv} - \sum_{vs \in E} f_{vs} = \varphi_s$$

**Definition 5.** For any network N = (V, E, s, t, c) and any *st*-preflow *f* in *N*, the **residual network** of *N* wrt. *f* is the network N' = (V, E', s, t, c') such that

$$E' = E^+ \cup E^-$$
  

$$E^+ = \{vw \in E \mid c_{vw} - f_{vw} > 0\}$$
  

$$E^- = \{vw \in V^2 \mid wv \in E \land f_{wv} > 0\}$$

and

$$\forall vw \in E' \quad c'_{vw} = \begin{cases} c_{vw} - f_{vw} & \text{if } vw \in E^+ \\ f_{wv} & \text{if } vw \in E^- \end{cases}$$
(11)

For any  $e \in E'$ ,  $c'_e$  is called the **residual capacity** of e wrt. f.

Any path in (V, E') from s to t (if such a path exists) is called an **augmenting** path of f.

**Lemma 2.** Let N = (V, E, s, t, c) be a network and f an st-preflow in N. Assume that an  $n \in \mathbb{N}$  and an augmenting path  $p = (v_1w_1, \ldots, v_nw_n)$  of f exist.

Let

$$\delta := \min_{vw \in p([n])} c'_{vw} \quad . \tag{12}$$

Then,  $f' \in \mathbb{N}_0^E$  such that

$$\forall vw \in E': \quad f'_{vw} = \begin{cases} f_{vw} + \delta & \text{if } vw \in p([n]) \land vw \in E \\ f_{vw} - \delta & \text{if } vw \in p([n]) \land wv \in E \\ f_{vw} & \text{otherwise} \end{cases}$$
(13)

is an st-preflow in N wrt. which

$$\varphi'_s = \varphi_s + \delta \quad . \tag{14}$$

Moreover, if f is an *st*-flow in N, so is f'.

**Definition 6.** Let (V, E) be a directed graph. Let  $s \in V$  and  $t \in V$  and  $s \neq t$ .

- $X \subseteq V$  is called an *st*-cutset of (V, E) iff  $s \in X$  and  $t \notin X$ .
- ▶  $Y \subseteq E$  is called an *st*-cut of (V, E) iff there exists an *st*-cutset X such that  $Y = \{vw \in E | v \in X \land w \notin X\}.$



Definition 7. The instance of the Minimum  $st\mathchar`-Cut$  Problem wrt. a network N=(V,E,s,t,c) is to

$$\min_{x \in \{0,1\}^V} \quad \sum_{vw \in E} x_v (1 - x_w) c_{vw} \tag{15}$$

subject to 
$$x_s = 1$$
 (16)

$$x_t = 0 \tag{17}$$

Note: With  $X := \{v \in V | x_v = 1\}$ , we have

$$\sum_{vw\in E} x_v (1-x_w) c_{vw} = \sum_{vw\in XX^c} c_{vw}$$

**Lemma 3.** For every network N = (V, E, s, t, c), every st-flow f in N, and every st-cutset  $X \subseteq V$ ,

$$\varphi_s \le \sum_{vw \in XX^c} c_{vw} \quad . \tag{18}$$

#### Proof.



Lemma 3 does **not** hold analogously for every *st*-preflow, because, wrt. an *st*-preflow,  $\varphi_S$  need not be an upper bound on  $\varphi_s$ .

**Theorem 1.** For any network N = (V, E, s, t, c), any  $s, t \in V$  such that  $s \neq t$ , and any *st*-flow f in N, the following three conditions are equivalent

- 1. There exists an st-cut whose capacity is equal to  $\varphi_s$ .
- 2. The st-flow f is optimal, i.e., a solution of (8)-(10).
- 3. No augmenting path of f exists.

## Proof.

- (1) implies (2) by virtue of Lemma 3.
- (2) implies (3) by virtue of Lemma 2.

We prove that (3) implies (1):

- Let f be an st-flow such that no augmenting path exists.
- Let S be the set of all nodes  $v \in V$  such that there exists a path in the residual network wrt. f from s to v. Let S also include s itself.
- ▶ Then,  $t \notin S$  (otherwise, the path from s to t in the residual network would be an augmenting path).
- ► Moreover, ...

#### Moreover,

- $$\begin{split} \varphi_s &= \sum_{v \in S} \varphi_v & \text{by (7) and } t \notin S \\ &= \varphi_S & \text{by Lemma 1} \\ &= \sum_{vw \in SS^c} f_{vw} \sum_{vw \in S^c S} f_{vw} & \text{by definition of } \varphi_S \\ &= \sum_{vw \in SS^c} c_{vw} & \text{by the arguments below.} \end{split}$$
- For any vw ∈ SS<sup>c</sup>, we have f<sub>vw</sub> = c<sub>vw</sub> (otherwise, the contradiction w ∈ S follows by construction of S and by definition of the residual network).
- For any  $vw \in S^cS$ , we have  $f_{vw} = 0$  (otherwise, the contradiction  $v \in S$  follows by construction of S and by definition of the residual network).

Algorithm 1. (Ford and Fulkerson, 1956)

**Theorem 2.** Algorithm 1 terminates. The output f is a maximum st-flow in N.

Proof. Termination.

• For every augmenting path processed,  $\varphi_s$  increases by at least 1.

► Moreover,

$$arphi_s \leq \sum_{vw \in \{s\}\{s\}^c} c_{vw}$$
 (by Lemma 3)

- Therefore, only finitely many augmenting paths are processed.
- ► Thus, the algorithm terminates.

Optimality:

- Throughout the execution, f is an st-flow in N.
- When the algorithm terminates, no augmenting path exists.
- Thus, f is a maximum *st*-flow in N (by Theorem 1).