# Computer Vision I

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**Excursus:** Maximum st-Flow and Minimum st-Cut

- $\blacktriangleright$  Maximum st-Flow Problem
- ▶ Residual networks and augmenting paths
- $\blacktriangleright$  Minimum st-Cut Problem
- $\blacktriangleright$  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  $UW := \{uv \in E \mid u \in U \wedge w \in W\}$ .



 $\mathsf{Definition} \ 1.$  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

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

<span id="page-3-2"></span>
$$
\varphi_U^- = \sum_{vu \in U^c U} f_{vu} \tag{2}
$$

<span id="page-3-1"></span><span id="page-3-0"></span>
$$
\varphi_U = \varphi_U^+ - \varphi_U^- \tag{3}
$$

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 outflux, influx and flux of  $u$  in  $(V, E)$  wrt.  $f$ .



<span id="page-5-0"></span>**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.

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

□

**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

<span id="page-7-1"></span>
$$
\forall e \in E \quad 0 \le f_e \le c_e \tag{5}
$$

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

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

<span id="page-7-0"></span>
$$
\forall v \in V - \{s, t\} \quad \varphi_v = 0 \tag{7}
$$

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

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

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

<span id="page-8-1"></span><span id="page-8-0"></span>
$$
\forall v \in V - \{s, t\} \quad \sum_{vw \in E} f_{vw} = \sum_{uv \in E} f_{uv} \quad . \tag{10}
$$

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  ${\sf residual}$  network of  $N$  wrt.  $f$  is the network  $N'=(V,E',s,t,c')$  such that

$$
E' = E^+ \cup E^-
$$
  
\n
$$
E^+ = \{vw \in E \mid c_{vw} - f_{vw} > 0\}
$$
  
\n
$$
E^- = \{vw \in V^2 \mid wv \in E \land f_{wv} > 0\}
$$

and

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

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  $f$ .

<span id="page-10-0"></span>**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

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

is an  $st$ -preflow in  $N$  wrt. which

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

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  $Y = \{vw \in E | v \in X \wedge w \notin X\}.$



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

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

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

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

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

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

<span id="page-13-0"></span>**Lemma 3.** For every network  $N = (V, E, s, t, c)$ , every st-flow f in N, and every st-cutset  $X \subseteq V$ ,

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

#### Proof.



Lemma [3](#page-13-0) 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

- <span id="page-14-0"></span>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  $(8)-(10)$  $(8)-(10)$  $(8)-(10)$ .
- 3. No augmenting path of  $f$  exists.

### Proof.

- (1) implies (2) by virtue of Lemma [3.](#page-13-0)
- (2) implies (3) by virtue of Lemma [2.](#page-10-0)

We prove that (3) implies (1):

- $\blacktriangleright$  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).
- $\blacktriangleright$  Moreover, ...

#### ▶ Moreover,

- $\varphi_s = \sum$ v∈S by [\(7\)](#page-7-0) and  $t \notin S$  $=\varphi_S$  by Lemma [1](#page-5-0)  $= \sum f_{vw} - \sum$  $vw ∈ S S<sup>c</sup>$   $v w ∈ S<sup>c</sup> S$ by definition of  $\varphi_S$  $=$   $\Sigma$  $vw \in SSc$ by the arguments below.
- ▶ For any  $vw \in SS^c$ , we have  $f_{vw} = c_{vw}$  (otherwise, the contradiction  $w \in S$  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)

<span id="page-17-0"></span>**Input:** Network  $N = (V, E, s, t, c)$ **Output:**  $f : E \to \mathbb{N}_0$ for all  $vw \in E$  $f_{vw} := 0$ while  $∃n ∈ ℕ ∃augmenting path p = (v₁w₁, …, v_nw_n)$  of f  $\delta := \min_{vw \in p([n])} c'_{vw}$ for all  $vw \in E$  $f_{vw} :=$  $\sqrt{ }$  $\int$  $\mathbf{I}$  $f_{vw} + \delta$  if  $vw \in P \wedge vw \in E$  $f_{vw} - \delta$  if  $vw \in P \wedge wv \in E$  $f_{vw}$  otherwise

**Theorem 2.** Algorithm [1](#page-17-0) terminates. The output  $f$  is a maximum st-flow in  $N$ .

Proof. Termination.

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

▶ Moreover,

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

- ▶ Therefore, only finitely many augmenting paths are processed.
- $\blacktriangleright$  Thus, the algorithm terminates.

Optimality:

- $\blacktriangleright$  Throughout the execution, f is an st-flow in N.
- $\blacktriangleright$  When the algorithm terminates, no augmenting path exists.
- $\blacktriangleright$  Thus, f is a maximum st-flow in N (by Theorem [1\)](#page-14-0).