4.5 Polynomial-time variations and extensions of Ford and Fulkerson algorithm

Assume throughout this section that n = O(m). |V| = |V|

This is not restrictive, because if m < n-1, then the graph is disconnected and we can determine the connected component containing the source and focus on that one.

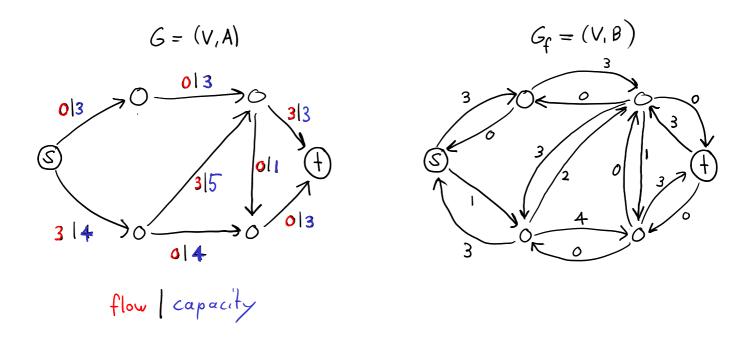
Moreover, let U := u(A) (sum of all capacities)

We will discuss 2 efficient maximum flow algorithms:

- (a) The capacity scaling algorithm
- (b) Edmonds Karp algorithm

#### **Definition 4.30:** $G_{f,\Delta}$

Let f be an s-t flow in the directed graph G=(V,A) with capacities  $u\colon A\to \mathbb{Z}_{\geq 0}$  and let  $\Delta\in\mathbb{R}_{\geq 0}$ . We denote by  $G_{f,\Delta}$  the subgraph of the residual graph  $G_f=(V,B)$  containing only the arcs with residual capacity of at least  $\Delta$ .



 $G_{f,2}$ 

 $\frac{3}{3}$ 

#### **Algorithm 6:** Capacity scaling algorithm for maximum s-t flows

**Input**: Directed graph G=(V,A) with arc capacities  $u\colon A \to \mathbb{Z}_{\geq 0}$  and  $s,t\in V,s\neq t$ .

**Output:** A maximum s-t flow f.

$$f(a) = 0 \ \forall a \in A.$$

// We start with the zero flow.

 $\Delta = 2^{\lfloor \log_2(U) \rfloor}.$ 

while  $\Delta \geq 1$  do

// These iterations are called phases.

**while**  $\exists f$ -augmenting path P in  $G_{f,\Delta}$  **do** 

$$\Delta = \frac{\Delta}{2}$$
.

return f

#### Theorem 4.31

Algorithm 6 returns a maximum s-t flow.

### Proof

Notice that throughout algorithm, we have  $\Delta \in \mathbb{Z}$ .

 $\rightarrow$  In last iteration, we have  $\Delta = 1$ .

However,  $G_{f,1} = G_f$ , because f is integral throughout algo.

This iteration finishes when there is no augmenting path in  $G_{f,l} = G_f$ .

Theorem 4.13 ) Returned flow f is maximum s-t flow.

This is poly-time because the input size is  $\Theta\left(m + \sum_{a \in A} \log \left(u(a) + 1\right)\right) = \Theta\left(m + \log \left(\prod \left(u(a) + 1\right)\right)\right) = \Omega\left(m + \log U\right)$ 

#### Theorem 4.32

Algorithm 6 runs in  $O(m^2 \log U)$  time.

Proof

# phases = O(log U)

We show that each phase takes O(m²) time.

```
Algorithm 6: Capacity scaling algorithm for maximum s-t flows

Input: Directed graph G=(V,A) with arc capacities u\colon A\to\mathbb{Z}_{\geq 0} and s,t\in V,s\neq t.
Output: A maximum s-t flow f.
f(a)=0 \ \forall a\in A. \qquad // \ \text{We start with the zero flow.}
\Delta=2^{\lfloor\log_2(U)\rfloor}.
while \Delta\geq 1 do // \ \text{These iterations are called } phases.
\begin{array}{c} \text{while } \exists f\text{-}augmenting \ path \ P \ in \ G_{f,\Delta} \ do \\ & \bot \ \text{Augment } f \ \text{along } P \ \text{and set } f \ \text{to the augmented flow.} \\ & \bot \ \Delta=\frac{\Delta}{2}. \\ \text{return } f \end{array}
```

Consider current flow f at start of some phase (which is defined by value of s).

(This is termination eviterion of previous

This is termination eviterion of previous phase.

(And is clearly true for first phase.)

Let  $C = \{v \in V : 7 \text{ s-v path in } G_{1,2\Delta} \}$   $\Rightarrow$  by previous point, C is an s-t cut.  $\Rightarrow u_{f}(a| \leq 2\Delta) \quad \forall u \in S_{G_{f}}^{+}(C)$ , by definition of C.  $\Rightarrow u_{f}(S_{G_{f}}^{+}(C)) \leq |S_{G_{f}}^{+}(C)| \cdot 2\Delta \leq 2\Delta m$   $\Rightarrow u_{f}(S_{G_{f}}^{+}(C)) \leq |S_{G_{f}}^{+}(C)| \cdot 2\Delta \leq 2\Delta m$ 

Moveover,  $u_f(\delta_{G_f}^t(C))$  is upper bound on how much f can be increased in terms of value.

$$2\Delta m \ge u_f(\delta_G^+(C)) = u(\delta^+(C)) - f(\delta^+(C)) + f(\delta^-(C)) = u(\delta^+(C)) - v(f)$$

$$vesidual capacities = -v(f)$$

$$of arcs \delta^+(C)$$

$$Lemma 4.3$$

$$(veak max-flow min-cut)$$

$$heaven$$

=) Augmentations in phase & can augment flow by no more than 2 sm.

Each augmentation in phase & hæs augmentation volume ≥ A.

 $\Rightarrow$  # augmentations in phase  $\Delta = O(m)$ 

Each augmentation takes O(m) time via BFS. (recall n = O(m))

=) time per phase: O(m2).

#

## 4.5.2 Edmonds-Karp algorithm

# Idea: Augment always on shortest paths.

#### Algorithm 7: Edmonds-Karp algorithm

**Input**: Directed graph G = (V, A) with arc capacities  $u: A \to \mathbb{Z}_{\geq 0}$  and  $s, t \in V, s \neq t$ .

**Output:** A maximum s-t flow f.

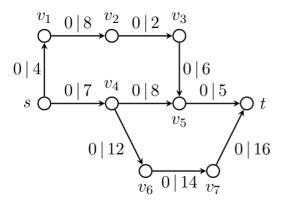
 $f(a) = 0 \ \forall a \in A.$ 

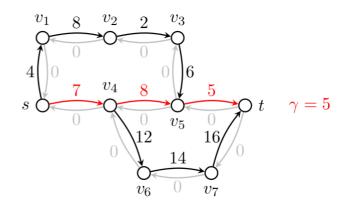
**while**  $\exists f$ -augmenting path in  $G_f$  **do** 

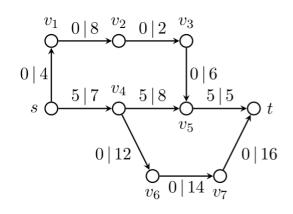
Find an f-augmenting path P in  $G_f$  minimizing |P|.

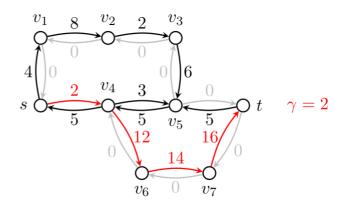
Augment f along P and set f to augmented flow.

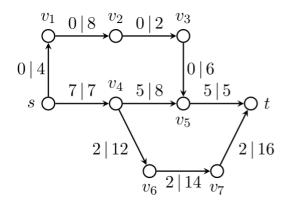
return f

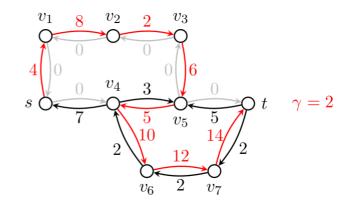


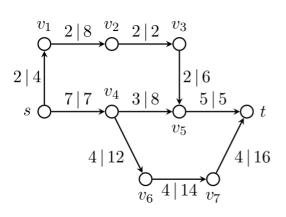


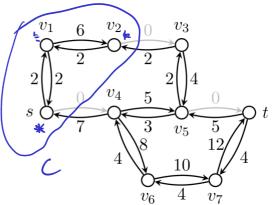












Key property:

more
formally

Distances from s and distances to t become larger in residual graphs, when only considering arcs with strictly positive residual capacity.

#### **Lemma 4.33**

Let G=(V,A) be a directed graph with arc capacities  $u\colon A\to\mathbb{Z}_{\geq 0}$ , and let  $s,t\in V$  with  $s\neq t$ . Moreover, let  $f_1$  be an s-t flow in G, and let  $f_2$  be an s-t flow obtained by augmenting  $f_1$  along a shortest augmenting path P in  $G_{f_1}$ . Then,

$$\begin{array}{ll} d_{f_1}(s,v) \leq d_{f_2}(s,v) & \forall v \in V \ , \ \text{and} \\ d_{f_1}(v,t) \leq d_{f_2}(v,t) & \forall v \in V \ , \end{array}$$

where  $d_f(v, w)$  denotes, for  $v, w \in V$  and an s-t flow f, the length (in terms of number of arcs) of a shortest v-w path in  $G_f$  that only uses arcs with strictly positive f-residual capacity.

Proof

It suffices to show first statement: df (s,v) \le df\_2 (s,v) \text{ V eV}

Indeed, second one can be reduced to first one by

reversing arc directions and flows on arcs, and

exchange rolex of s and t.

Notice that Gh and Gh are same graphs (V, B) with

different arc capacities up and up, respectively.

Let  $B_i = ab \in B$ :  $u_{f_i}(b) > ob$   $\forall i \in d_{1,2}$ .  $d_{f_i}(v,w) \leftarrow v - w$  distance in  $(V_3B_i)$ .

Assume by sake of contradiction  $\forall v \in V$  s.t.  $d_{f_i}(s,v) > d_{f_2}(s,v)$ .

re Among all such v, we choose one where  $d_{f_2}(s,v)$  is smallest. Let  $P_2$  be a shortest s-v path in  $(V,P_2)$ .  $\rightarrow d_{f_2}(s,v) = |P_2|$ .

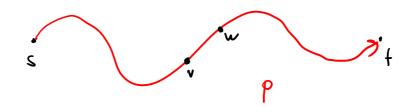


..... P2 \ B,

I These ares must have been used in apposite direction by P.

Claim:  $(w,v) \notin B_1$ Because  $df_2(s,v)$  is smallest among all v fulfilling (v), we have  $df_1(s,w) \leq df_2(s,w) = df_2(s,v) - 1$ If we had  $(w,v) \in B_1 \implies df_1(s,v) \leq df_1(s,v) + 1$   $(v) \in B_1(s,v) \leq df_2(s,v)$  contradicting  $(v) \in B_1(s,v) \leq df_2(s,v)$  contradicting  $(v) \in B_1(s,v) \leq df_2(s,v)$ 

Hence,  $(u, v) \in B_2 \setminus B_1$ . =)  $(v, w) \in P$ . P is shorlest set path in  $(V, B_1)$  containing (v, w). =)  $df_1(s, w) = df_1(s, v) + 1$ 



 $\Rightarrow d_{f_1}(s,v) < d_{f_1}(s,w)$  \( \rightarrow \delta\_{f\_2}(s,v) > d\_{f\_2}(s,w) \) \( \rightarrow \text{ Contradiction with choice of } v \) \( \rightarrow \text{ one could have chosen } w \text{ instead.} \)

-> see script for more details.