

# Algorithms and datastructures II

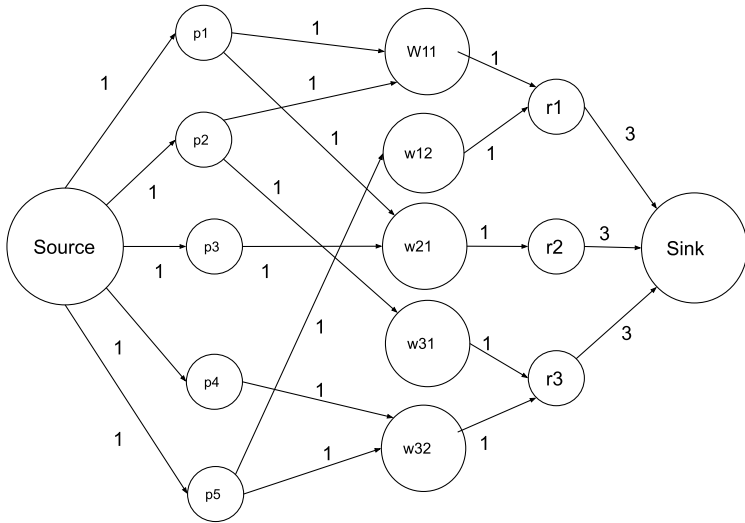
## Lecture 3: network flows

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# Network flow



# Network flow

## Definition (Network)

Network is an 4-tuple  $N = (V, E, s, t, c)$  where

- ①  $(V, E)$  is a directed graph,
- ②  $s \in V$  is a **source** vertex,
- ③  $t \in V$  is a **sink** vertex,
- ④  $c : E \rightarrow \mathbb{R}_0^+$  is a function assigning every edge a **capacity**.

- $f^+(v) = \sum_{u, (u,v) \in E} f(u, v)$  (**flow into a vertex**)
- $f^-(v) = \sum_{u, (v,u) \in E} f(v, u)$  (**flow out of a vertex**)
- $f^\Delta(v) = f^+(v) - f^-(v)$  (**surplus or excess**)

Here  $f : E \rightarrow \mathbb{R}_0^+$

## Definition (Flow)

Function  $f : E \rightarrow \mathbb{R}_0^+$  is **flow** if it satisfies

- ① **Capacity constraint**:  $(\forall_{e \in E}) : f(e) \leq c(e)$
- ② **Conservation of flows (Kirchoff's law)**:  $(\forall_{v \in V \setminus \{s, t\}}) : f^\Delta(v) = 0$

Value of the flow:  $|f| = f^\Delta(t)$ .

# Network flow

## Network flow problem

Given network  $N = (V, E, s, t, c)$  find flow  $f$  maximizing  $|f|$  (a **maximum flow**).

Naive approach: Start with 0 flow and keep improving as long as there is path from source to sink that can be improved.

Today: we show that naive approach works and give an effective algorithm to solve this problem

## Ford–Fulkerson algorithm, 1956



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Definition (Residual capacity)

$$r(u, v) = c(u, v) - f(u, v) + f(v, u)$$

Definition (Augumenting path)

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## FordFulkerson( $V, E, s, t, c$ )

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If capacities are integers, FordFulkerson terminates

Homework: Time complexity when all edges have capacity 1.



Yefim Dinitz

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Flow  $f$  is **blocking flow** if for every (oriented) path from  $s$  to  $t$  contains edge  $e$  with  $f(e) = c(e)$ .

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- 1 Using BFS( $s$ ) determine layers in  $(V, E)$ .
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## LayeredNetwork( $R = (V, E, s, t, r)$ )

- ① Using BFS( $s$ ) determine layers in  $(V, E)$ .
- ② Remove all layers after  $t$ .
- ③ Remove edges inside layers and edges going backwards.
- ④ Remove dead ends:
- ⑤  $Q = \{v \neq z : \text{deg}^{\text{out}}(v) = 0\}$
- ⑥ While there exists  $v \leftarrow \text{Dequeue}(Q)$ :
  - ⑦ Remove  $v$  and all edges containing  $v$ .
  - ⑧ If removal of some edge decreased  $\text{deg}^{\text{out}}$  of some vertex to 0, add it to  $Q$ .
- ⑨ Return  $R$ .

# Constructing blocking flow of layered network

BlockingFlow (Layered  $R = (V, E, s, t, r)$ )

- ①  $g \leftarrow$  zero flow.
- ② While there exists oriented path  $P$  from  $s$  to  $t$ :
- ③  $\epsilon \leftarrow \min_{e \in P} (r(e) - g(e))$ .
- ④ For every  $e \in P : g(e) \leftarrow g(e) + \epsilon$ .
- ⑤ Remove from  $E$  all  $e \in P$  such that  $g(e) = r(e)$ .
- ⑥ Remove dead ends like in previous algorithm.
- ⑦ Return  $g$ .

# Constructing blocking flow of layered network

BlockingFlow (Layered  $R = (V, E, s, t, r)$ )

- ①  $g \leftarrow$  zero flow.
- ② While there exists oriented path  $P$  from  $s$  to  $t$ :
- ③  $\epsilon \leftarrow \min_{e \in P} (r(e) - g(e))$ .
- ④ For every  $e \in P : g(e) \leftarrow g(e) + \epsilon$ .
- ⑤ Remove from  $E$  all  $e \in P$  such that  $g(e) = r(e)$ .
- ⑥ Remove dead ends like in previous algorithm.
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Time complexity:  $O(|V||E|)$

# Dinic algorithm

Dinic ( $V, E, s, t, c$ )

- ①  $f \leftarrow$  zero flow.
- ② Repeat:
- ③ Build residual network  $R$  and remove all edges  $e$  with  $r(e) = 0$ .
- ④  $\ell \leftarrow$  length of the shortest oriented path from  $s$  to  $t$  in  $R$ .
- ⑤ If there is no such path return  $f$ .
- ⑥  $L \leftarrow$  LayeredNetwork ( $R$ ).
- ⑦  $g \leftarrow$  BlockingFlow ( $L$ ).
- ⑧ Improve flow  $f$  using  $g$ .



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Lemma (On improving flows)

For every flow  $f$  in network  $N = (V, E, s, t, c)$  and flow  $g$  in  $R(S, f) = (V, E, s, t, r)$  one can construct in time  $O(m)$  flow  $h$  in  $S$  such that  $|h| = |f| + |g|$ .

Proof.

For every  $(u, v)$  such that  $\{u, v\} \in E$  compute  $h^*(u, v) = f(u, v) - f(v, u) + g(u, v) - g(v, u)$ .  
If  $h^*(u, v) \geq 0$  put  $h(u, v) = h^*(u, v)$  and  $h(v, u) = 0$ .



# Dinic algorithm

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- ⑧ Improve flow  $f$  using  $g$ .

## Lemma

*Every iteration of the loop increases  $\ell$  by at least 1.*

Loop iterates at most  $|V|$  times. It follows that the runtime of algorithm is  $O(|V|^2|E|)$ .

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## Proof.

Denote by  $R_i$  the residual network  $R$  at iteration  $i$ .  
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## Theorem

*Dinic algorithm will terminate in time  $O(|V|^2|E|)$  and will return maximum flow.*

Maximality of flow returned follows from the same analysis as in Ford–Fulkerson's algorithm.