## Augmenting Paths Math 482, Lecture 25

Misha Lavrov

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### Lecture plan

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- See it get stuck.

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We know how to find a max flow using an LP. But this is inefficient; there are many many algorithms that are faster.

Plan for today:

- Describe a simple greedy algorithm that tries to find a max flow.
- See it get stuck.
- Make the algorithm more powerful.

The	greedy	algorithm	
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Augmenting paths

## Directed *s*, *t*-paths

#### Definition

In a network, a *directed path* from s to t is a sequence

 $s, v_1, v_2, \ldots, v_k, t$ 

where  $v_1, v_2, ..., v_k \in N$  and  $(s, v_1), (v_1, v_2), ..., (v_k, t) \in A$ .



The	greedy	algorithm
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Example:



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Example: directed path  $s \rightarrow a \rightarrow b \rightarrow t$ 



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Whenever we have a directed path from s to t and all arcs along the path are below capacity, we can use it to increase the flow.



# The greedy algorithm oco Augmenting paths oco The residual graph oc Using a directed path Image: Comparison of the comparison

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At this point, there are no more directed paths where all arcs are below capacity. But is this the maximum flow?

 The greedy algorithm
 Augmenting paths
 The residual graph

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### A further improvement

If we redirect some  $a \to b \to t$  flow to go  $a \to d \to t$ , we can send more flow along the path  $s \to c \to d \to t$ ...



 The greedy algorithm
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Increase flow along red edges, decrease flow along blue edge.

 Augmenting paths
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Increase flow along red edges, decrease flow along blue edge. Note:  $s \rightarrow c \rightarrow b \leftarrow a \rightarrow d \rightarrow t$  is *almost* a directed path.

The greedy algorithm	Augmenting paths	The residual graph
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### Augmenting paths

#### Definition

Given a network (N, A) and a feasible flow **x**, an augmenting path for **x** is a sequence of nodes

$$s = v_0, v_1, v_2, \ldots, v_k, v_{k+1} = t$$

such that for each pair  $v_i$ ,  $v_{i+1}$ :

 either e = (v<sub>i</sub>, v<sub>i+1</sub>) is an arc below capacity (e ∈ A and x<sub>e</sub> < c<sub>e</sub>)

The greedy	algorithm

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The greedy	algorithm

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The sequence

$$s \rightarrow c \rightarrow b \leftarrow a \rightarrow d \rightarrow t$$

we found on the previous slide was an augmenting path.

The greedy algorithm 000	Augmenting paths 00●0	The residual graph
Using an augmenting pa	th to improve <b>x</b>	

To augment a feasible flow  $\mathbf{x}$  along an augmenting path:

- $\textbf{0} \quad \text{Let } \delta > \textbf{0} \text{ be the largest value such that}$ 
  - $x_e \leq c_e \delta$  for all forward arcs on the path;
  - $x_e \ge \delta$  for all backward arcs on the path.

The greedy algorithm	Augmenting paths	The residual graph
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**2** For each forward arc *e*, increase  $x_e$  by  $\delta$ .

The greedy algorithm 000	Augmenting paths 00●0	The residual graph
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When we do this, flow is still conserved at internal nodes of the augmenting path. There are four possible cases:

$$\cdots \xrightarrow{+\delta} p \xrightarrow{+\delta} \cdots \qquad \cdots \xrightarrow{+\delta} q \xleftarrow{-\delta} \cdots \\ \cdots \xleftarrow{-\delta} r \xrightarrow{+\delta} \cdots \qquad \cdots \xleftarrow{-\delta} s \xleftarrow{-\delta} \cdots$$

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**(**) Find the augmenting path  $s \rightarrow c \rightarrow b \leftarrow a \rightarrow d \rightarrow t$ .

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- **(**) Find the augmenting path  $s \rightarrow c \rightarrow b \leftarrow a \rightarrow d \rightarrow t$ .
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- **②** Find the value  $\delta$  we can augment by: here,  $\delta = 2$ .

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- **③** Increase or decrease the flow along each edge by  $\delta$ .

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idual graph

## The residual graph

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Idea: augmenting paths for  ${\bf x}$  are directed paths in the residual graph.



We construct the residual graph for our next-to-last feasible flow:



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We can find the augmenting path  $s \rightarrow b \leftarrow a \rightarrow d \rightarrow t$  by following only *forward* arrows (of either color) in the residual graph;  $\delta = 2$  is the smallest number along those arrows.