



# Teoria dos Grafos e Computabilidade

— Network Flow —

Silvio Jamil F. Guimarães

Graduate Program in Informatics – PPGINF

Laboratory of Image and Multimedia Data Science – IMScience

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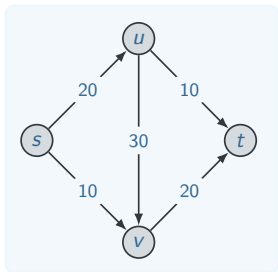
# Maximum Flow and Minimum Cut

- ▶ Two rich algorithmic problems.
- ▶ Fundamental problems in combinatorial optimization.
- ▶ Beautiful mathematical duality between flows and cuts.
- ▶ Numerous non-trivial applications:
  - ▶ Bipartite matching.
  - ▶ Data mining.
  - ▶ Project selection.
  - ▶ Airline scheduling.
  - ▶ Baseball elimination.
  - ▶ Image segmentation.
  - ▶ Network connectivity.
  - ▶ Open-pit mining.
  - ▶ Network reliability.
  - ▶ Distributed computing.
  - ▶ Egalitarian stable matching.
  - ▶ Security of statistical data.
  - ▶ Network intrusion detection.
  - ▶ Multi-camera scene reconstruction.
  - ▶ Gene function prediction.

- ▶ Use directed graphs to model transportation networks :
  - ▶ edges carry traffic and have capacities.
  - ▶ nodes act as switches.
  - ▶ *source* nodes generate traffic, *sink* nodes absorb traffic.

# Flow Networks

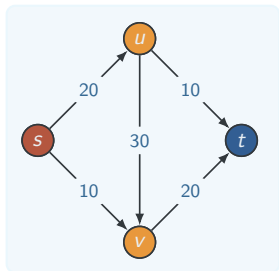
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  - ▶ Each edge  $e \in E$  has a capacity  $c(e) > 0$ .

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- ▶ Use directed graphs to model **transportation networks**:
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- ▶ A **flow network** is a directed graph  $G = (V, E)$ 
  - ▶ Each edge  $e \in E$  has a capacity  $c(e) > 0$ .
  - ▶ There is a single **source** node  $s \in V$ .
  - ▶ There is a single **sink** node  $t \in V$ .
  - ▶ Nodes other than  $s$  and  $t$  are **internal**.

# Defining Flow

- ▶ In a flow network  $G = (V, E)$ , an **s-t flow** is a function  $f : E \rightarrow \mathbb{R}^+$  such that

- (i) **Capacity conditions** For each  $e \in E$ ,  $0 \leq f(e) \leq c(e)$ .
- (ii) **Conservation conditions** For each internal node  $v$ ,

$$\sum_{e \text{ into } v} f(e) = \sum_{e \text{ out of } v} f(e)$$

- ▶ The **value** of a flow is  $\nu(f) = \sum_{e \text{ out of } s} f(e)$ .

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- ▶ The **value** of a flow is  $\nu(f) = \sum_{e \text{ out of } s} f(e)$ .
- ▶ Useful notation:

$$f^{\text{out}}(v) = \sum_{e \text{ out of } v} f(e)$$

For  $S \subseteq V$ ,

$$f^{\text{out}}(S) = \sum_{e \text{ out of } S} f(e)$$

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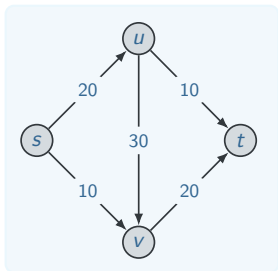


# Maximum-Flow Problem

## MAXIMUM FLOW

**INSTANCE** A flow network  $G$

**SOLUTION** The flow with largest value in  $G$



► **Assumptions:**

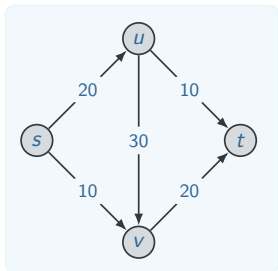
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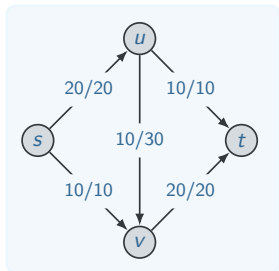
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3. All edge capacities are **integers**.

# Teoria dos Grafos e Computabilidade

— Ford-Fulkerson Algorithm —

Silvio Jamil F. Guimarães

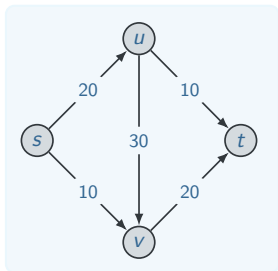
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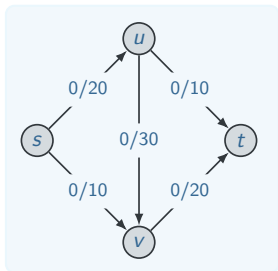
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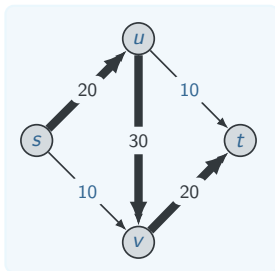
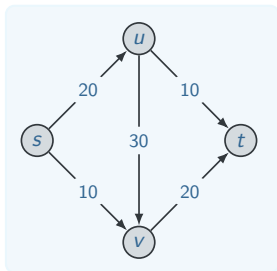
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- ▶ A **flow network** is a directed graph  $G = (V, E)$
- ▶ Let us take a greedy approach.
  1. Start with **zero flow** along all edges.



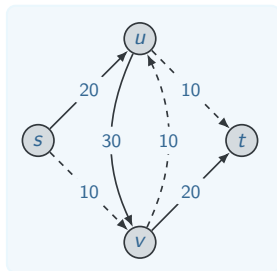
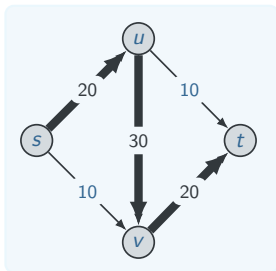
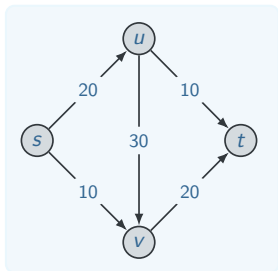
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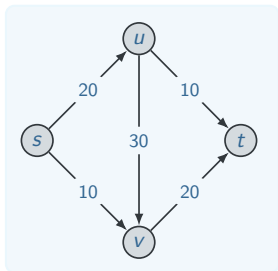
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  1. Start with **zero flow** along all edges.
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  3. **Key idea**: Push flow along edges with **leftover capacity** and **undo flow** on edges already carrying flow.





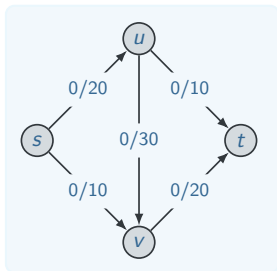
# Residual Graph

- ▶ Given a flow network  $G = (V, E)$  and a flow  $f$  on  $G$ , the residual graph  $G_f$  of  $G$  with respect to  $f$  is a directed graph such that
  - (i) Nodes –  $G_f$  has the same nodes as  $G$ .



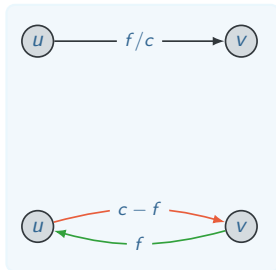
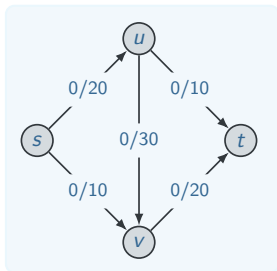
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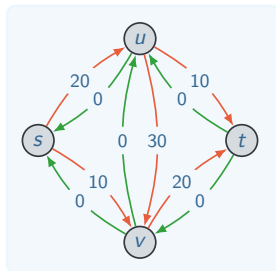
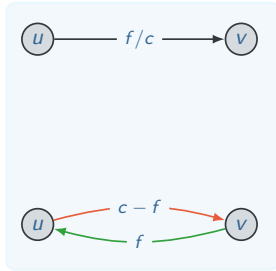
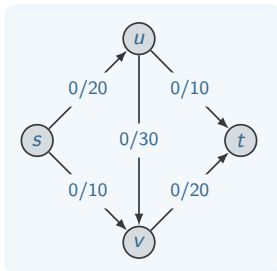
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  - (ii) **Forward** edges – For each edge  $e = (u, v) \in E$  such that  $f(e) < c(e)$ ,  $G_f$  contains the edge  $(u, v)$  with a **residual capacity**  $c(e) - f(e)$ .
  - (iii) **Backward** edges – For each edge  $e \in E$  such that  $f(e) > 0$ ,  $G_f$  contains the edge  $e' = (v, u)$  with a residual capacity  $f(e)$ .



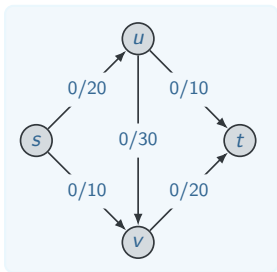
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# Augmenting Paths in a Residual Graph

- ▶ Let  $P$  be a **simple  $s$ - $t$  path** in  $G_f$ .
- ▶  **$\text{bottleneck}(P, f)$**  is the minimum residual capacity of any edge in  $P$ .
- ▶ The following operation  $\text{augment}(f, P)$  yields a new flow  $f'$  in  $G$ :



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## Algorithm: Augmented path

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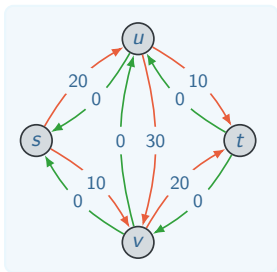
**output**: The distances of the vertices from  $s$

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1 Let  $b = \text{bottleneck}(P, f)$  ;
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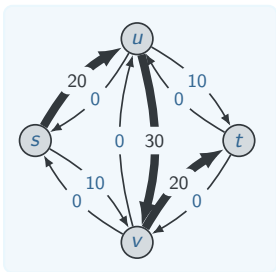
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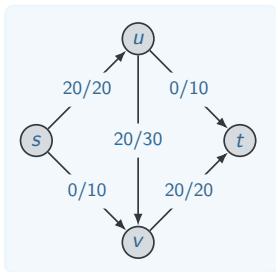
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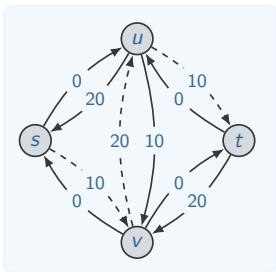
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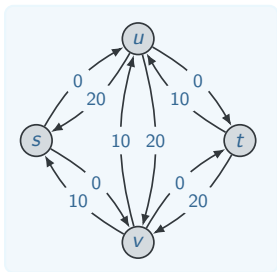
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- ▶ A simple  $s$ - $t$  path in the residual graph is an augmenting path.
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  - ▶ Conservation condition on internal node  $v \in P$ . Four cases to work out.

# Ford-Fulkerson Algorithm

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**Algorithm:** Ford-Fulkerson Algorithm

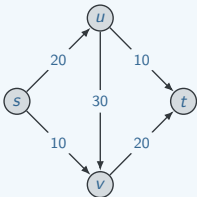
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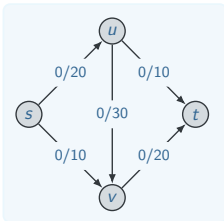
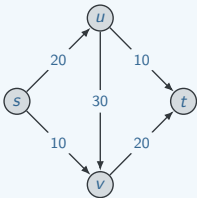
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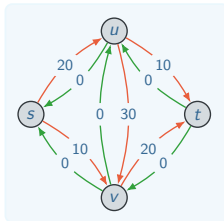
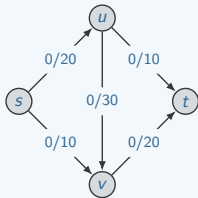
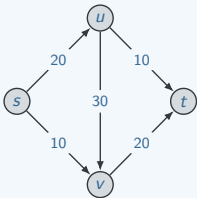
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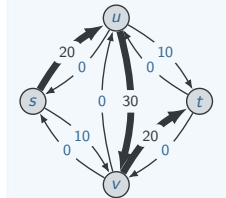
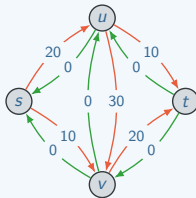
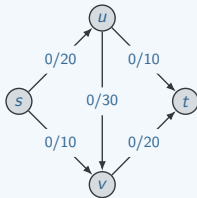
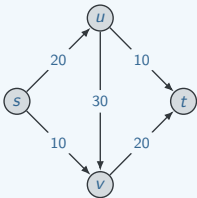
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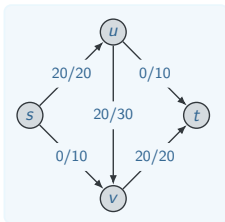
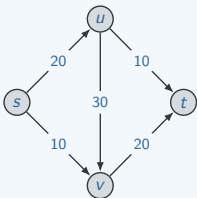
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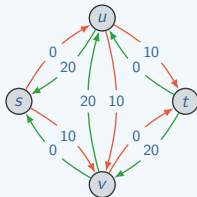
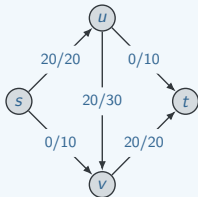
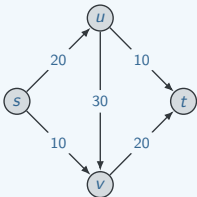
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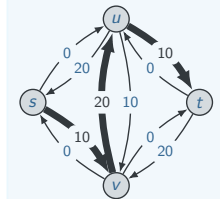
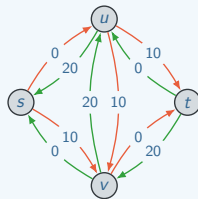
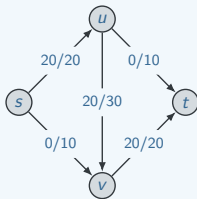
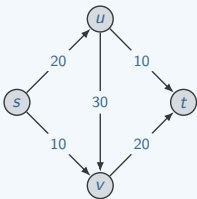
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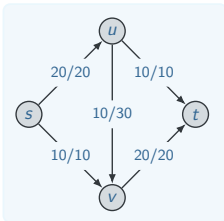
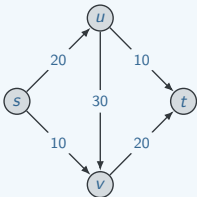
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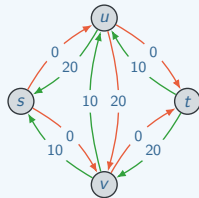
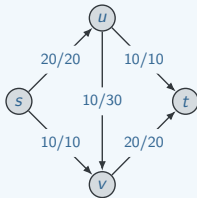
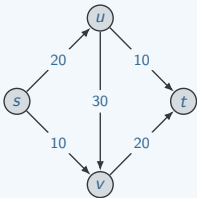
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- ▶ Idea: An  **$s$ - $t$  cut** is a partition of  $V$  into sets  $A$  and  $B$  such that  $s \in A$  and  $t \in B$ .
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- ▶ Question: Is the reverse true? Is the smallest capacity of a cut at most the maximum flow?
- ▶ Answer: Yes, and the Ford-Fulkerson algorithm computes this **cut**!

- ▶ Let  $\bar{f}$  denote the flow computed by the Ford-Fulkerson algorithm.
- ▶ Enough to show  $\exists s-t$  cut  $(A^*, B^*)$  such that  $\nu(\bar{f}) = c(A^*, B^*)$ .
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- ▶ Claim: If  $f$  is an  $s-t$  flow such that  $G_f$  has no  $s-t$  path, then there is an  $s-t$  cut  $(A^*, B^*)$  such that  $\nu(f) = c(A^*, B^*)$ .
  - ▶ Claim applies to *any* flow  $f$  such that  $G_f$  has no  $s-t$  path, and not just to the flow  $\bar{f}$  computed by the Ford-Fulkerson algorithm.

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- ▶ Corollary: If all capacities in a flow network are integers, then there is a maximum flow  $f$  where every flow value  $f(e)$  is an integer.

# Teoria dos Grafos e Computabilidade

## — Scaling Max-Flow Algorithm —

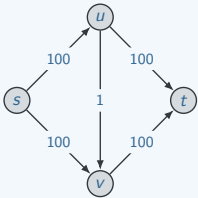
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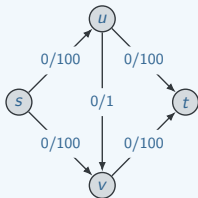
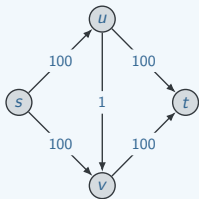
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# Bad Augmenting Paths

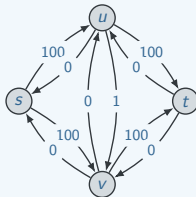
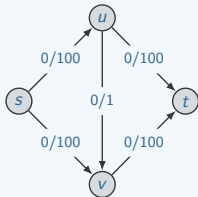
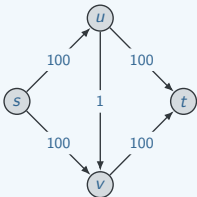


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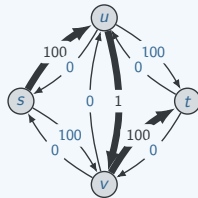
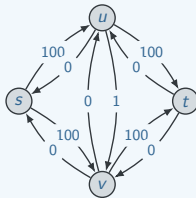
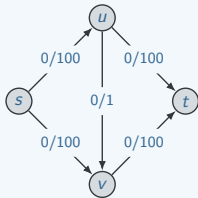
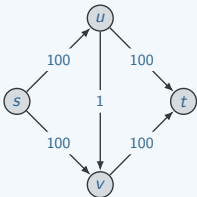




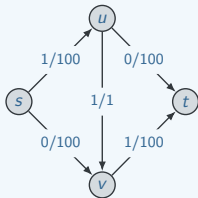
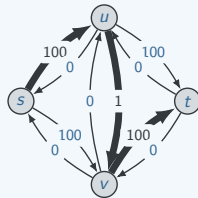
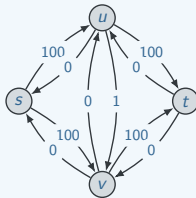
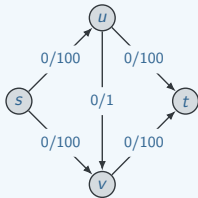
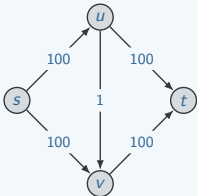
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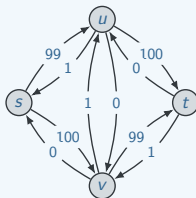
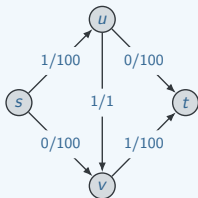
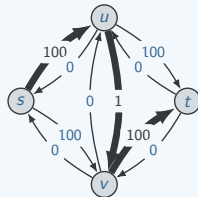
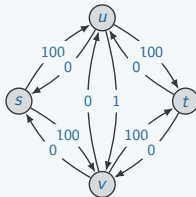
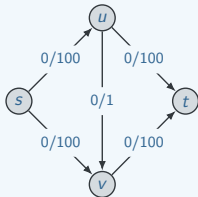
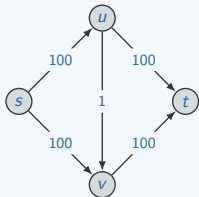
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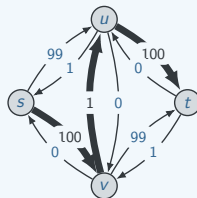
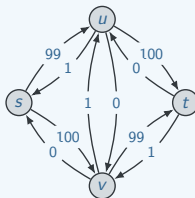
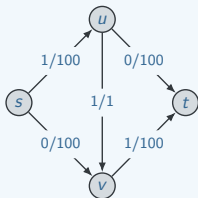
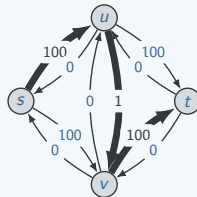
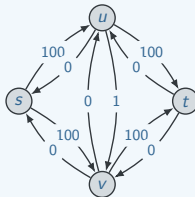
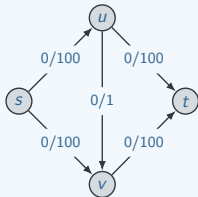
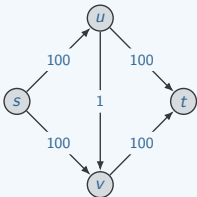
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# Improving Ford-Fulkerson Algorithm

- ▶ Bad case for Ford-Fulkerson algorithm is when the bottleneck edge is the augmenting path has a low capacity.
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# Other Maximum Flow Algorithms

- Desire a **strongly polynomial** algorithm: running time depends only on the *size* of the graph and is *independent* of the numerical values of the capacities (as long as numerical operations).



# Other Maximum Flow Algorithms

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- ▶ **Edmonds-Karp, Dinitz**: choose augmenting path to be the shortest path in  $G_f$  (use breadth-first search).

# Teoria dos Grafos e Computabilidade

— Exercises —

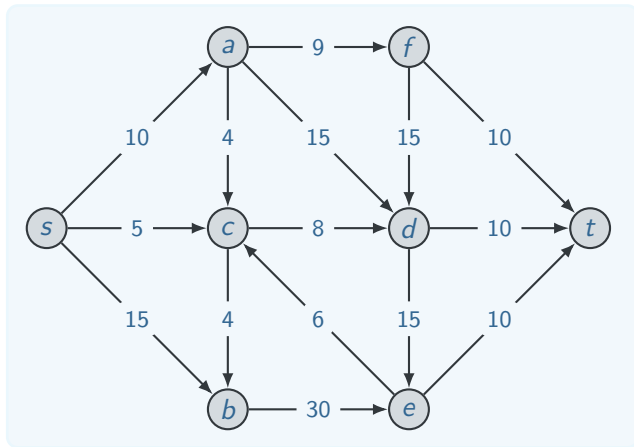
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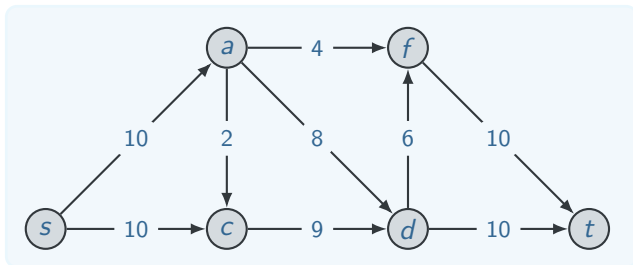
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# Compute the maximum flow



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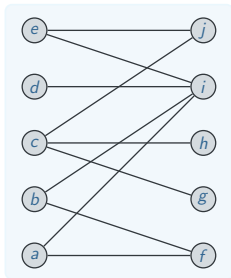


# Bipartite graph matching

## BIPARTITE GRAPH MATCHING

**INSTANCE** Let  $G = (L \cup R, E)$  be an undirected graph.  $M \subseteq E$  is a **matching** if each node appear in, at most, one edge in  $M$ .

**SOLUTION** Find a **max cardinality** matching.

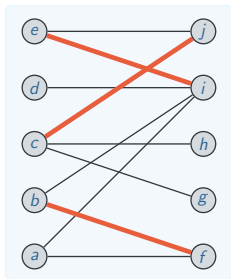


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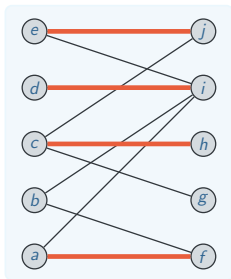


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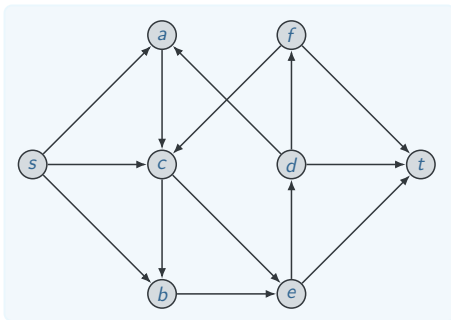


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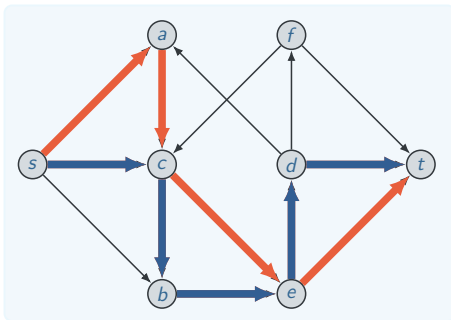


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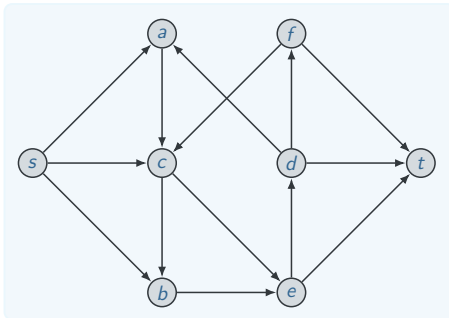
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