

# Chapter 5

## Flows

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## Course outcomes

Course learning outcomes	
CLO.1	Understanding of the basic concepts of graphs Special types of graph, computer based graph representation, isomorphism, planar graph, connectivity in graph, graph traversal.
CLO.2	Describe definition of path and circuit Identify the existence of Euler path & circuit Identify the existence of Hamilton path & circuit
CLO.3	Compute minimum spanning tree in a (weighted) graph Use algorithms: Prim, Kruskal
CLO.4	Determine shortest path in a weighted graph Use algorithms: Dijkstra, Bellman-Ford, Floyd-Warshall
CLO.5	Solve maximum flow problem Use Ford-Fulkerson's algorithm

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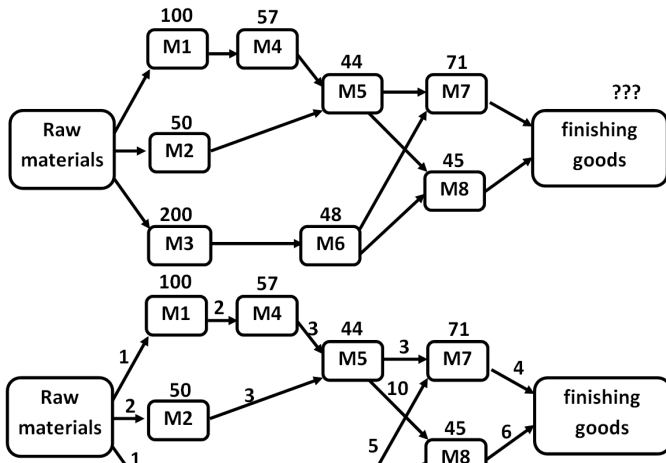
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- Distributed manufacturing system :  

$$(((M1 \wedge M4) \vee M2) \wedge M5) \vee (M3 \wedge M6)) \wedge (M7 \vee M8)$$
- Production capacity of each branch is defined in graph  $G$
- How to determine the production capacity (e.g. pieces/min)?
- How to determine the production paths with the minimum transportation cost ?



# Maximum flow problem

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## Given data

- A directed graph  $G = (V, E)$  with source node  $s$  and sink node  $t$
- capacity function  $c : E \rightarrow \mathcal{R}$ , i.e.  $c(u, v) \geq 0$  for any edge  $(u, v) \in E$

## Objective

Send as much flow as possible with flow  $f : E \rightarrow \mathcal{R}^+$  such that

- $f(u, v) \leq c(u, v)$ , for all  $(u, v) \in E$
- $\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v)$ , for  $u \neq s, t$

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## Given data

- A directed graph  $G = (V, E)$  with source node  $s$  and sink node  $t$
- capacity function  $c : E \rightarrow \mathcal{R}$ , i.e.  $c(u, v) \geq 0$  for any edge  $(u, v) \in E$
- cost function  $a : E \rightarrow \mathcal{R}$ , i.e.  $a(u, v) \geq 0$  for any edge  $(u, v) \in E$

## Objective

Send as much flow as possible with minimum cost such that

- $f(u, v) \leq c(u, v)$ , for all  $(u, v) \in E$
- $\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v)$ , for  $u \neq s, t$
- $\sum_{(u, v) \in E} a(u, v) f(u, v)$  should be minimized

## Flow Algorithms

- Linear programming
- Ford-Fulkerson algorithm  $O(E \max |f|)$
- Edmond-Karp algorithm  $O(VE^2)$
- Dinitz blocking flow algorithm  $O(V^2E)$
- General push-relabel maximum flow algorithm  $O(V^2E)$
- Push-relabel algorithm with FIFO vertex selection rule  $O(V^3)$
- Dinitz blocking flow algorithm with dynamic trees  $O(VE \log(V))$
- Push-relabel algorithm with dynamic trees  $O(VE \log(V^2/E))$
- Binary blocking flow algorithm  $O(E \min(V^{2/3}, \sqrt{E}) \log(V^2/E) \log(U))$   
with  $U = \max c(u, v)$

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# Ford-Fulkerson's algorithm for solving Max Flow Problem

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**Input:** graph  $G$  with flow capacity  $c$ , a source node  $s$ , and a sink node  $t$

**Output:** a maximum flow  $f$  from  $s$  to  $t$

$k = 0$ ;  $G^{(0)} = G$ ;

$c^{(0)}(u, v) = c(u, v)$ ,  $c^{(0)}(v, u) = 0$ ,  $\forall (u, v) \in G^{(0)}$ ;

**While**  $\exists$  a path  $\Pi^{(k)}(s, t)$  in  $G^{(k)}$  such that  $c^{(k)}(u, v) > 0$ ,  $\forall (u, v) \in \Pi^{(k)}$  **do**

Find  $f(\Pi^{(k)}) = \min\{c^{(k)}(u, v) | (u, v) \in \Pi^{(k)}\}$  ;

**For** each edge  $(u, v) \in \Pi^{(k)}$  **do**

If  $(u, v) \in G$  then

$$c^{(k+1)}(u, v) = c^{(k)}(u, v) - f(\Pi^{(k)});$$

$$c^{(k+1)}(v, u) = c^{(k)}(v, u) + f(\Pi^{(k)});$$

**Else**

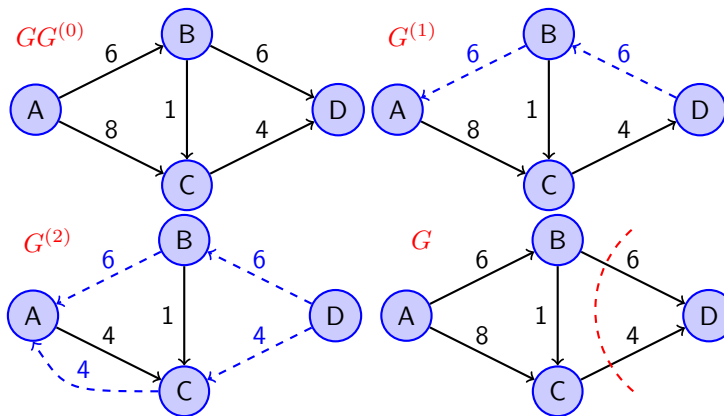
$$c^{(k+1)}(u, v) = c^{(k)}(u, v) + f(\Pi^{(k)});$$

$$c^{(k+1)}(v, u) = c^{(k)}(v, u) - f(\Pi^{(k)});$$

$k++$ ;



## Example 1



$k$	$\Pi^{(k)}$	(A,B)	(A,C)	(B,C)	(B,D)	(C,D)	$f(\Pi^{(k)})$
0	$\{(A,B), (B,D)\}$	6	-	-	6	-	6
1	$\{(A,C), (C,D)\}$	-	4	-	-	4	4
1	$\{(A,C), (C,D)\}$	-	4	-	-	4	4
Stop		6	4	-	6	4	10

with  $f_{max} = 10$

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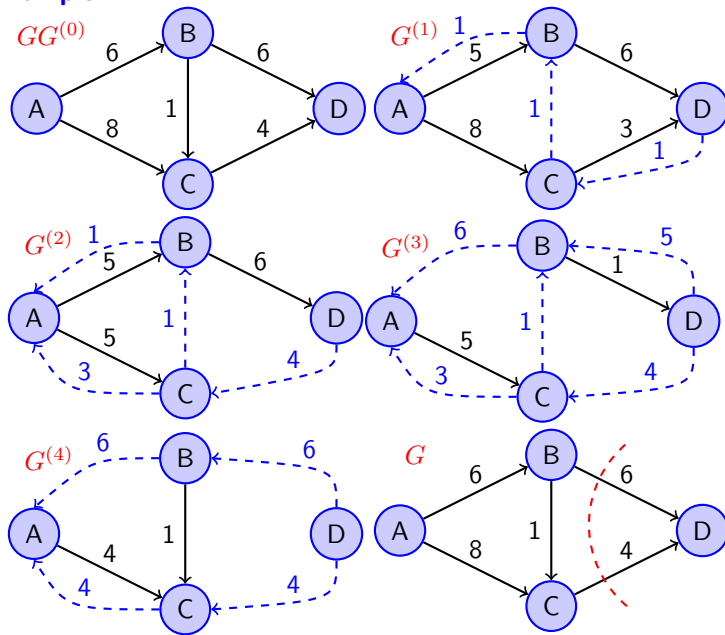
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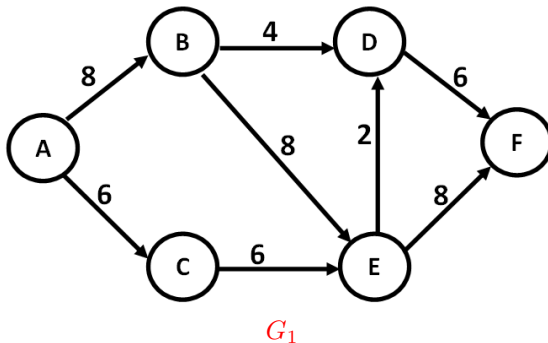
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Find the maximum flow and the min-cut in the following network.



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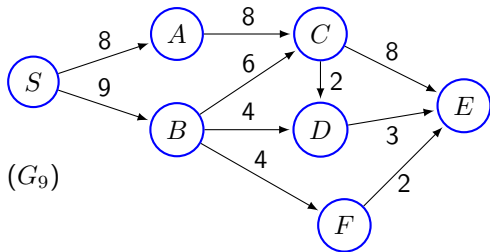
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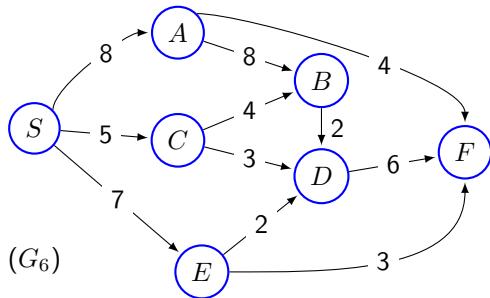
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**Input:** graph  $G$  with flow capacity  $c$ , a source node  $s$ , and a sink node  $t$

**Output:** a maximum flow  $f$  from  $s$  to  $t$

$k = 0$ ;  $G^{(0)} = G$ ;  $c^{(0)}(u, v) = c(u, v)$ ,  $c^{(0)}(v, u) = 0$ ,  $\forall (u, v) \in G^{(0)}$ ;

**While**  $\exists$  a **shortest** path  $\Pi^{(k)}(s, t)$  in  $G^{(k)}$  such that  $c^{(k)}(u, v) > 0$ ,  
 $\forall (u, v) \in \Pi^{(k)}$  **do**

Find  $f(\Pi^{(k)}) = \min\{c^{(k)}(u, v) | (u, v) \in \Pi^{(k)}\}$ ;

**For** each edge  $(u, v) \in \Pi^{(k)}$  **do**

**If**  $(u, v) \in G$  **then**

$$c^{(k+1)}(u, v) = c^{(k)}(u, v) - f(\Pi^{(k)});$$

$$c^{(k+1)}(v, u) = c^{(k)}(v, u) + f(\Pi^{(k)});$$

**Else**

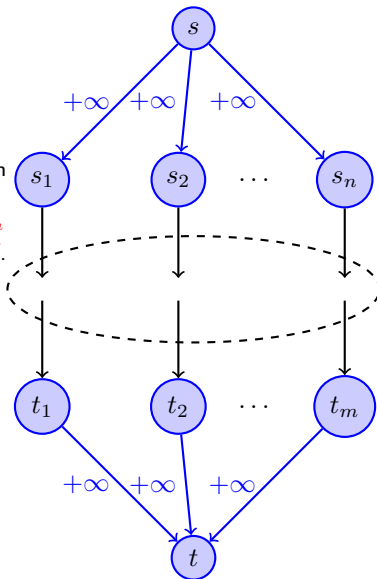
$$c^{(k+1)}(u, v) = c^{(k)}(u, v) + f(\Pi^{(k)});$$

$$c^{(k+1)}(v, u) = c^{(k)}(v, u) - f(\Pi^{(k)});$$

$k++$ ;

# Multi-source Multi-sink Maximum Flow Problem

- Given a network  $\mathcal{N} = (V, E)$  with a set of sources  $S = s_1, \dots, s_n$  and a set of sinks  $T = t_1, \dots, t_m$
- find the maximum flow across  $\mathcal{N}$ .
- $\implies$  transform into a maximum flow problem by adding a super source connecting to each vertex in  $S$  and a super sink connected by each vertex in  $T$  with infinite capacity on each edge



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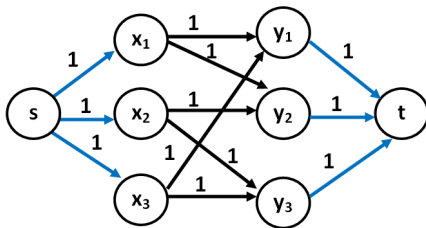
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- Given a bipartite graph  $G = (X \cup Y, E)$
- find a maximum cardinality matching in  $G$ , the maximum number of edges we can choose such that no two edges share a common vertex.
- $\Rightarrow$  transform into a maximum flow problem by constructing a network

$\mathcal{N} = (X \cup Y \cup \{s, t\}, E')$ :

- $E'$  contains the edges in  $G$  directed from  $X$  to  $Y$ .
- $(s, x) \in E'$  for each  $x \in X$  and  $(y, t) \in E'$  for each  $y \in Y$ .
- $c(e) = 1$  for each  $e \in E'$ .





# Minimum Path Cover in Directed Acyclic Graph

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- Given a directed acyclic graph  $G = (V, E)$ , we are to find the minimum number of paths to cover each vertex in  $V$ . We can construct a bipartite graph  $G' = (V_{out} \cup V_{in}, E')$  from  $G$ , where
  - $V_{out} = \{v \in V: v \text{ has positive out-degree}\}$ .
  - $V_{in} = \{v \in V: v \text{ has positive in-degree}\}$ .
  - $E' = \{(u, v) \in (V_{out}, V_{in}): (u, v) \in E\}$ .
- Then it can be shown that  $G'$  has a matching of size  $m$  iff there exists  $n - m$  paths that cover each vertex in  $G$ , where  $n$  is the number of vertices in  $G$ .
- Therefore, the problem can be solved by finding the maximum cardinality matching in  $G'$  instead.

# Maximum Flow Problem with Vertex Capacities

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- Given a network  $\mathcal{N} = (V, E)$ , in which there is capacity at each node in addition to edge capacity, that is, a mapping  $c : V \rightarrow \mathbb{R}^+$ , denoted by  $c(v)$ , such that the flow  $f$  has to satisfy not only the capacity constraint and the conservation of flows, but also the vertex capacity constraint  $\sum_{i \in V} f_{i,v} \leq c(v), \forall v \in V \setminus s, t$
- $\implies$  the amount of flow passing through a vertex cannot exceed its capacity.
- To find the maximum flow across  $\mathcal{N}$ , we can transform the problem into the maximum flow problem in the original sense by expanding  $\mathcal{N}$ .
  - each  $v \in V$  is replaced by  $v_{in}$  and  $v_{out}$ 
    - $v_{in}$  is connected by edges going into  $v$
    - $v_{out}$  is connected to edges coming out from  $v$ ,
  - assign capacity  $c(v)$  to the edge connecting  $v_{in}$  and  $v_{out}$
- In this expanded network, the vertex capacity constraint is removed and therefore the problem can be treated as the original maximum flow problem.

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- Given a directed graph  $G = (V, E)$  and two vertices  $s$  and  $t$ ,
- Find the maximum number of independent paths from  $s$  to  $t$ .
- Two paths are said to be independent if they do not have a vertex in common apart from  $s$  and  $t$ .
- We can construct a network  $\mathcal{N} = (V, E)$  from  $G$  with vertex capacities, where
  - ①  $s$  and  $t$  are the source and the sink of  $\mathcal{N}$  respectively.
  - ②  $c(v) = 1$  for each  $v \in V$ .
  - ③  $c(e) = 1$  for each  $e \in E$ .
- Then the value of the maximum flow is equal to the maximum number of independent paths from  $s$  to  $t$ .

# Maximum Edge-disjoint Path

- given a directed graph  $G = (V, E)$  and two vertices  $s$  and  $t$
- find the maximum number of edge-disjoint paths from  $s$  to  $t$ .
- This problem can be transformed to a maximum flow problem by constructing a network  $\mathcal{N} = (V, E)$  from  $G$  with  $s$  and  $t$  being the source and the sink of  $\mathcal{N}$  respectively and assign each edge with unit capacity.

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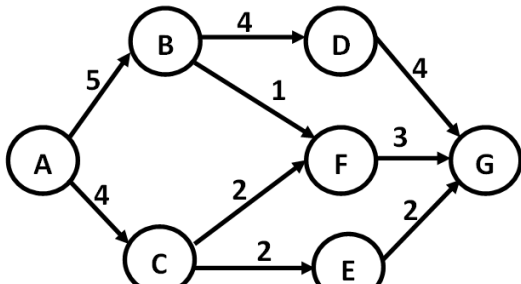
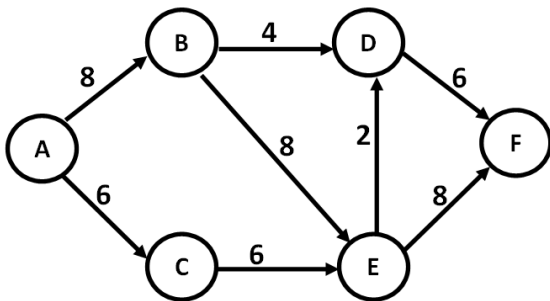
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Find the maximum flow in the following networks



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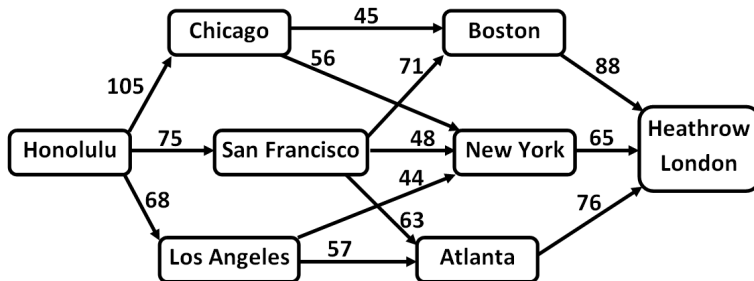
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- Whole pineapples are served in a restaurant in London.
- To ensure freshness, the pineapples are purchased in Hawaii and air freighted from Honolulu to Heathrow in London.
- The following network diagram outlines the different routes that the pineapples could take.



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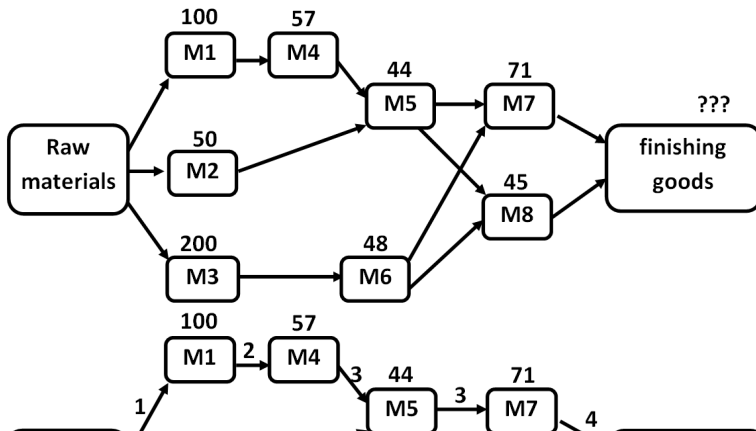
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# Production quantity measuring

- Distributed manufacturing system :  
$$(((M1 \wedge M4) \vee M2) \wedge M5) \vee (M3 \wedge M6)) \wedge (M7 \vee M8)$$
- Production capacity of each branch is defined in graph  $G$
- How to determine the production capacity (e.g. pieces/min)?
- How to determine the production paths with the minimum transportation cost ?



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- The table below gives the expenses for persons  $W$ ,  $X$ ,  $Y$  and  $Z$  to travel to places  $A$ ,  $B$ ,  $C$  and  $D$ .
- The objective is to send each person to one of the four places such that all places will be visited, whilst the total costs are as small as possible.
- Translate this problem into a maximum flow problem and solve it with the maximum flow algorithm.

	A	B	C	D
W	16	12	11	12
X	13	11	8	14
Y	10	6	7	9
Z	11	15	10	8





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- Consider the problem of assigning student to writing seminars.
- In class, we modeled a version of the problem where the total number of students exactly equals the number of available spots.
- In real applications, there are fewer students than available spots so some writing seminars are assigned fewer than 15 students.
- Model this problem as a minimum cost flow problem.
- Explain (in words and/or pictures) what are the vertices, supplies and demands, edges, and edge weights.

# Blood donation problem

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- Enthusiastic celebration of a sunny day at a prominent northeastern university has resulted in the arrival at the university's medical clinic of 169 students in need of emergency treatment.
- Each of the 169 students requires a transfusion of one unit of whole blood. The clinic has supplies of 170 units of whole blood.
- The number of units of blood available in each of the four major blood groups and the distribution of patients among the groups is summarized below.
  - type A patients can only receive type A or O;
  - type B patients can receive only type B or O;
  - type O patients can receive only type O;
  - type AB patients can receive any of the four types.

Blood type	A	B	O	AB
Supply	46	34	45	45
Demand	39	38	42	50

Give a max flow formulation that determines a distribution that satisfies the demands of a maximum number of patients.

# Energy supplying problem

Dining Services wonders how little money they can spend on food while still supplying sufficient energy (2000 kcal), protein (55g), and calcium (800mg) to meet the minimum Federal guidelines and avert a potential lawsuit. A limited selection of potential menu items along with their nutrient content and maximum tolerable quantities per day is given in the table below.

	Energy	Protein	Calcium	Cost per serving
	(kcal)	(g)	(mg)	(cents)
Oatmeal	110	4	2	3
Chicken	205	32	12	24
Eggs	160	13	54	13
Whole milk	160	8	285	9
Cherry pie	420	4	22	20
Pork with beans	260	14	80	19

Formulate a linear program to find the most economical menu.

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## Given data

- A directed graph  $G = (V, E)$  with source node  $s$  and sink node  $t$
- lower bound  $l(u, v)$  and upper bound  $u(u, v) \geq 0$  for any edge  $(u, v) \in E$
- cost function  $a : E \rightarrow \mathcal{R}$ , i.e.  $a(u, v) \geq 0$  for any edge  $(u, v) \in E$

## Objective

Send as much flow as possible with minimum cost such that

- $l(u, v) \leq f(u, v) \leq u(u, v)$ , for all  $(u, v) \in E$
- $\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v)$ , for  $u \neq s, t$
- $\sum_{(u, v) \in E} a(u, v) f(u, v)$  should be minimized

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