

# Math 409: Discrete Optimization

Today: total unimodularity (cont')

Recall:

Def: A matrix  $A \in \mathbb{R}^{m \times n}$  is totally unimodular (TU) if every square submatrix has determinant  $0, \pm 1$ .

Prop: If  $A$  is TU,  $b \in \mathbb{Z}^m$  and  $l, u \in \mathbb{Z}^n$  then all vertices of each of the following are integer:

- 1)  $\{x \in \mathbb{R}^n : Ax \leq b, x \geq 0\}$
- 2)  $\{x \in \mathbb{R}^n : Ax \leq b, l \leq x \leq u\}$
- 3)  $\{x \in \mathbb{R}^n : Ax = b, x \geq 0\}$
- 4)  $\{x \in \mathbb{R}^n : Ax = b, l \leq x \leq u\}$

Prop: The node-edge incidence matrix of any directed graph is totally unimodular.

## Applications to Network Flow

$G = (V, E)$  directed graph,  $s, t \in V$ ,  $c: E \rightarrow \mathbb{Z}_{\geq 0}$

Equations defining s-t flows  $f: E \rightarrow \mathbb{R}_{\geq 0}$ ,  $f(e) = x_e$

$$\sum_{e \in \mathcal{I}^{\text{in}}(v)} x_e = \sum_{e \in \mathcal{I}^{\text{out}}(v)} x_e \quad \forall v \in V \setminus \{s, t\}$$

Encode into node edge incidence matrix  $A \in \{0, \pm 1\}^{V \times E}$

$$A_{ve} = \begin{cases} 1 & \text{if } e \in \delta^{\text{in}}(v) \\ -1 & \text{if } e \in \delta^{\text{out}}(v) \\ 0 & \text{o.w.} \end{cases}$$

$e = (w, v)$  for some  $w$   $v$   
 $e = (v, w)$  for some  $w$   $w$

For  $x \in \mathbb{R}^E$ ,  $Ax \in \mathbb{R}^V$  with  $(Ax)_v = \sum_{e \in \delta^{\text{in}}(v)} x_e - \sum_{e \in \delta^{\text{out}}(v)} x_e$  ← netflow into  $v$

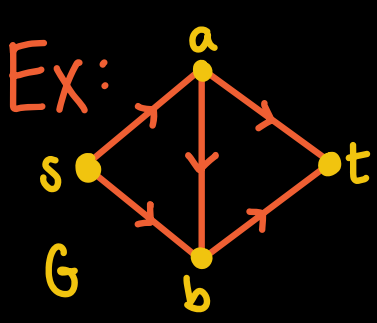
Cor: For integer edge capacities  $c: E \rightarrow \mathbb{Z}_{\geq 0}$ , the set of  $s$ - $t$  flows equals the polytope

$$P_{\text{flow}} = \{x \in \mathbb{R}^E : \tilde{A}x = 0, 0 \leq x \leq c\}$$

where  $\tilde{A}$  is the  $(V \setminus \{s, t\}) \times E$  submatrix of  $A$ .

Moreover, all vertices of  $P_{\text{flow}}$  are integer.

(Proof)  $A \text{ TU} \Rightarrow \tilde{A} \text{ TU} \Rightarrow$  integer vertices  
↑  
by Cor about  $\{x \in \mathbb{R}^n : Ax = b, l \leq x \leq u\}$



$$A = \begin{matrix} & \begin{matrix} sa & sb & ab & at & bt \end{matrix} \\ \begin{matrix} s \\ a \\ b \\ t \end{matrix} & \begin{pmatrix} -1 & -1 & 0 & 0 & 0 \\ 1 & 0 & -1 & -1 & 0 \\ 0 & 1 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 1 \end{pmatrix} \end{matrix}$$

$$Ax = \begin{pmatrix} -x_{sa} - x_{sb} \\ x_{sa} - x_{ab} - x_{at} \\ -x_{sb} + x_{ab} + x_{bt} \\ x_{at} + x_{bt} \end{pmatrix}$$

$\tilde{A}$   $\tilde{A}x$

Cor: For  $c: E \rightarrow \mathbb{R}_{\geq 0}$ , the max value of an  $s$ - $t$  flow is achieved by an integer flow  $f: E \rightarrow \mathbb{Z}_{\geq 0}$

(objective function =  $\sum_{e \in \delta^{\text{out}}(s)} x_e - \sum_{e \in \delta^{\text{in}}(s)} x_e$ , linear in  $x_e$ )

# Application: Min Cost Max Flow

Given prices  $p: E \rightarrow \mathbb{R}_{\geq 0}$  with

$p(e)$  = cost of sending one unit of flow on  $e$ ,

cost of flow  $f: E \rightarrow \mathbb{R}_{\geq 0}$  is  $\sum_{e \in E} p(e)f(e)$

## MIN COST MAX FLOW

Input: digraph  $G=(V,E)$ ,  $s,t \in V$ ,  $c: E \rightarrow \mathbb{Z}_{\geq 0}$ ,  $p: E \rightarrow \mathbb{R}$

Goal: Find max value s-t flow of minimum cost

capacities  $\swarrow$   
prices  $\swarrow$

We can solve this in polynomial time with LP:

1) Compute max value  $m$  of an s-t flow

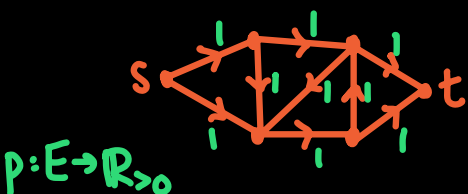
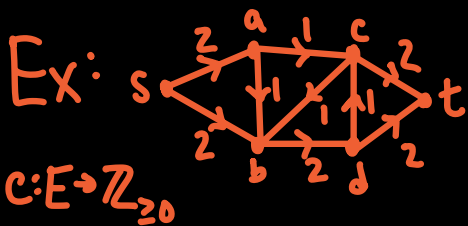
2) Find  $\min \sum_{e \in E} p(e)x_e$  s.t.  $0 \leq x \leq c$ ,  $Ax = m(\mathbb{1}_t - \mathbb{1}_s)$

$t^{\text{th}}$  entry =  $m$   
 $s^{\text{th}}$  entry =  $-m$   
rest =  $0$

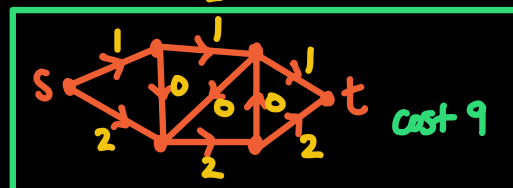
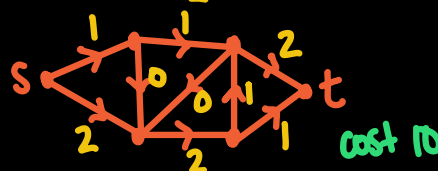
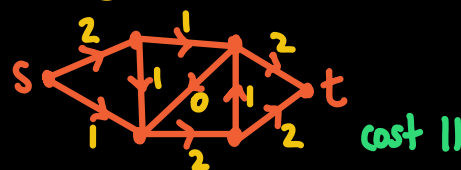
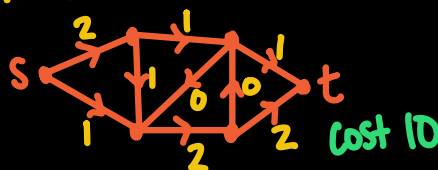
Moreover, since  $A$  is TU, the

optimal value is achieved by an integer-valued

flow  $f: E \rightarrow \mathbb{Z}_{\geq 0}$ .



Max flows of value = 3



min cost max flow  $\uparrow$

Encoded as an LP:

$$\min x_{sa} + x_{sb} + \dots + x_{dt} \text{ s.t.}$$

$$0 \leq x_{sa} \leq 2, 0 \leq x_{sb} \leq 2, \dots, 0 \leq x_{dt} \leq 2 \text{ and}$$

$$\begin{array}{c} s \\ a \\ b \\ c \\ d \\ t \end{array} \begin{pmatrix} & sa & sb & \dots & dt \\ -1 & -1 & & & 0 \\ 1 & 0 & -1 & & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & & -1 \\ 0 & 0 & & & 1 \end{pmatrix} \begin{pmatrix} x_{sa} \\ x_{sb} \\ \vdots \\ x_{dt} \end{pmatrix} = \begin{pmatrix} -3 \\ 0 \\ 0 \\ 0 \\ 0 \\ 3 \end{pmatrix}$$

full node-edge adjacency matrix  $A$