

Algorithm Design and Analysis

**CSE
565**

LECTURES 24

Linear programming

- Definitions
- Duality

(Based on Erickson,
Chapter 26)

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Converting a LP into Canonical Form

- Replacing minimization with maximization

Replace $\min_{\vec{x}} \vec{c} \cdot \vec{x}$ with $\max_{\vec{x}} -\vec{c} \cdot \vec{x}$

- Getting rid of unconstraint variables

Replace each variable x_j with two new variables x_j^+ , x_j^-

- add the inequalities $x_j^+ \geq 0$ and $x_j^- \geq 0$
- substitute $x_j^+ - x_j^-$ instead of x_j

- Getting rid of equalities

Replace every equality $\sum_j a_{ij}x_j = b_i$ with

- $\sum_j a_{ij}x_j \geq b_i$ and $\sum_j a_{ij}x_j \leq b_i$

- Getting rid of \geq signs

Replace every lower bound $\sum_j a_{ij}x_j \geq b_i$ with

- $\sum_j -a_{ij}x_j \leq -b_i$

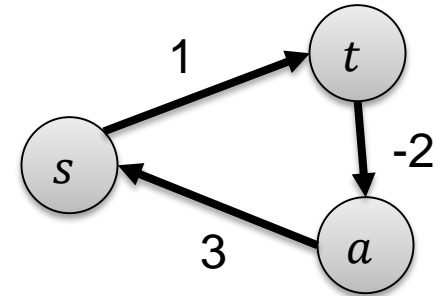
Example: Max flow as LP

- Given a graph, formulate an LP to determine the maximum s - t flow:
 - **Variables:** f_e for each edge $e \in E$
 - **Constraints:**
 - $f_e \geq 0$
 - $f_e \leq c_e$
 - $\sum_w f_{wv} - \sum_u f_{vu} = 0$ for all vertices $v \in V \setminus \{s, t\}$
 - **Objective:**
 - Maximize $\sum_u f_{su} - \sum_w f_{ws}$

Shortest Paths as an LP

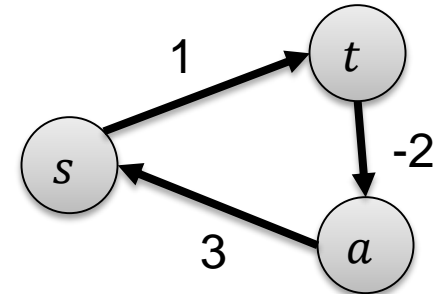
- Given: Graph $G = (V, E)$, lengths $\ell_{u,v}$, vertices $s, t \in V$
- **Variables:** d_v for each $v \in V$
- **LP:** maximize d_t
s.t. $d_v - d_u \leq \ell_{u,v} \quad \forall (u, v) \in E$
 $d_s = 0$
 - What is the value of this LP?
- **Claim:** LP value is the distance from s to t
 - Lower bound: If G has no negative cost cycles, then vector of shortest path distances from s is feasible.
 - Upper bound:
Add the inequalities for edges along shortest s - t path:
 $d_t \leq (\text{length of shortest } s\text{-}t \text{ path})$

Example: Shortest paths



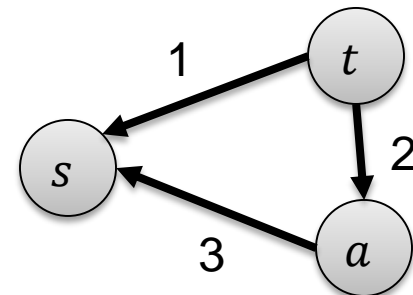
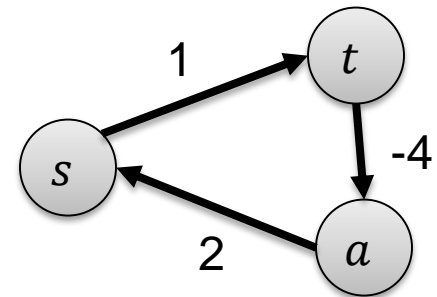
Shortest paths as an LP

Good case: shortest paths well-defined



What happens to this LP if...

- G has a negative cost cycle?
 - LP is not feasible:
Adding inequalities along cycle leads to contradiction
- G has no negative cost cycles, but t is not reachable from s in G ?
 - LP is feasible but unbounded



Physical interpretation

- When G is undirected and edge lengths are positive, we can imagine
 - $|V|$ magnets on a whiteboard
 - For every edge $(u, v) \in E$, connect magnets u and v by a string of length $\ell_{u,v}$
 - Magnet s always at the bottom of the white board
- Interpreting the LP
 - $d_v =$ height of magnet v on board (above s)
 - Feasible points: possible heights of magnets in some arrangement (with strings unbroken)
 - Value of LP: largest possible height for magnet t

Min-Cost Flow

- **Input:** G, s, t, c , cost function $\$,$ and demand value d
- **Output:** flow f of value d that minimizes $\sum_{e \in E} \$(e) \cdot f(e)$
- Give an LP for this problem
 - **Variables:** f_e for each edge $e \in E$
 - **Constraints:**
 - $f_e \geq 0$ for all edges $e \in E$
 - $f_e \leq c_e$ for all edges $e \in E$
 - $\sum_w f_{wv} - \sum_u f_{vu} = 0$ for all vertices $v \in V \setminus \{s, t\}$
 - $\sum_u f_{su} - \sum_w f_{ws} = d$
 - **Objective:**
 - Maximize $\sum_{e \in E} \$(e) \cdot f_e$

Multicommodity Flow

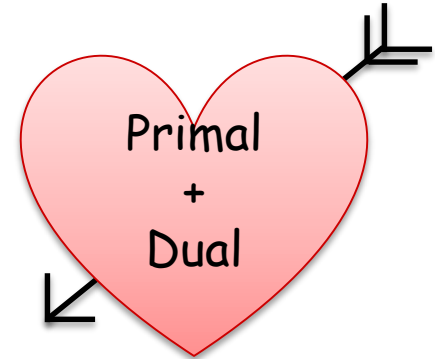
- **Input:** G, c , for each of k commodities: (s_i, t_i, d_i) , where s_i is the source, t_i is the sink, d_i is the demand for commodity i
- **Output:** flow f_i for each commodity i
- Give an LP for this problem
 - **Variables:** f_e^i for each edge $e \in E$ and commodity i
 - **Constraints:**
 - $f_e^i \geq 0$ for each i, e
 - $\sum_i f_e^i \leq c_e$ for each edge e
 - $\sum_w f_{wv}^i - \sum_u f_{vu}^i = 0$ for all vertices $v \in V \setminus \{s, t\}$ and commodities i
 - $\sum_u f_{su}^i - \sum_w f_{ws}^i = d_i$ for all commodities i
 - **Objective:**
 - Maximize 0

Min cut as LP

- Given a graph, formulate an LP to determine the minimum s - t cut:
 - **Variables:**
 - s_v for each vertex $v \in V$
 - Intuition: $s_v=1$ if $v \in S$ and 0 if $v \in T$
 - x_{uv} for each edge (u, v)
 - Intuition: $x_{uv}=1$ if $u \in S$ and $v \in T$
 - **Constraints:**
 - $x_{uv} + s_v - s_u \geq 0$ for all edges $(u, v) \in E$
 - $x_{uv} \geq 0$ for all edges $(u, v) \in E$
 - $s_s = 1$ and $s_t = 0$
 - **Objective:**
 - Minimize $\sum_{uv} c_{uv} \cdot x_{uv}$
- Requires some work to prove correctness

Duality of linear programs

- For every linear program Π , there is another linear program Π^* that is perfect for it
 - Π^* is “the” dual program of Π
 - The dual is not unique
 - The dual of Π^* is Π
- If we put Π in canonical form as a max. problem



$$\begin{array}{l} \max \vec{c} \cdot \vec{x} \\ \text{s.t. } Ax \leq \vec{b} \text{ and } x \geq 0 \end{array}$$

n variables
 m constraints

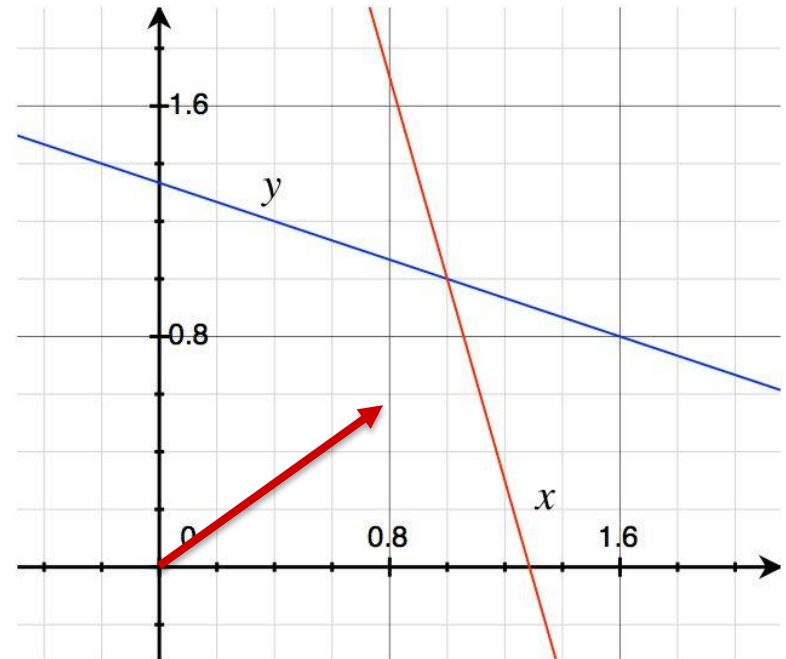
then we can write Π^* as a minimization problem

$$\begin{array}{l} \min \vec{b} \cdot \vec{y} \\ \text{s.t. } A^T y \geq \vec{c} \text{ and } y \geq 0 \end{array}$$

m variables
 n constraints

Examples

1. $\max 4x + 3y$
s.t. $\begin{pmatrix} 7 & 2 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \leq \begin{pmatrix} 9 \\ 4 \end{pmatrix}$
and $x, y \geq 0$



2. (Shortest paths variant)

LP: maximize $\sum_v d_v$

s.t. $d_v - d_u \leq \ell_{u,v} \forall (u, v) \in E$

$d_s = 0$