

CSE 421

Introduction to Algorithms

Lecture 20: Linear Programming:

A really very extremely big hammer

Note: Wed Nov 22
will introduce a
new topic!

Midterm

Midterm grades will be released at the end of this class.

- Breathe!
- These grades don't count for that much.
 - In past 421 I've had a student with a midterm grade in the mid-60's end with a 4.0 in the course

Histogram

90s **10**

80s **24**

70s **23**

60s **13**

50s **10**

40s **11**

<40 **10**

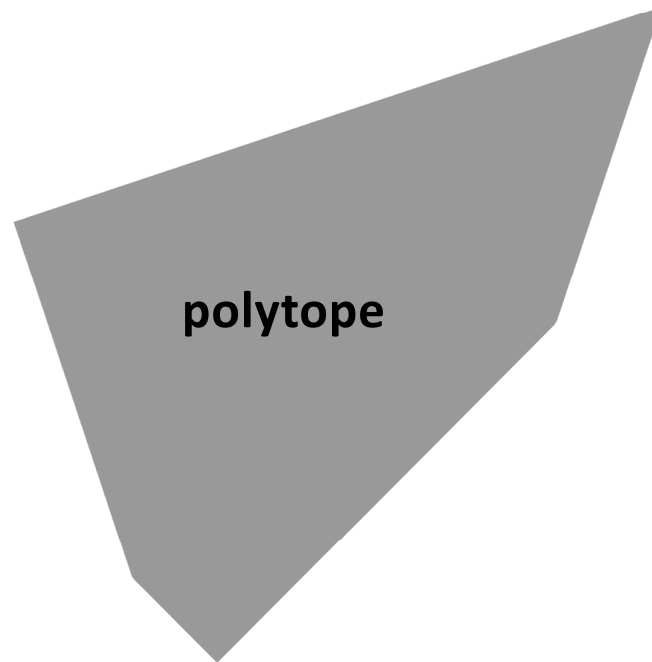
Median **74**

Average **67.71**

Standard Deviation **20.32**

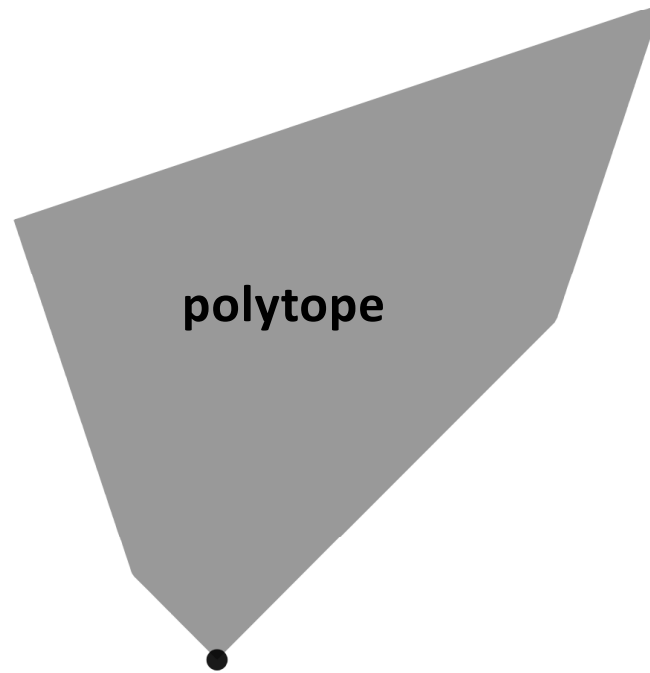
Given: a polytope

Find: the *lowest* point in the polytope



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Find: the *lowest* point in the polytope



**We have fast
algorithms for this!**

Maximize $z_1 + 2z_3$

subject to:

$$2z_1 - z_2 + 3z_3 \leq 1$$

$$-z_1 + z_2 - z_3 \leq 5$$

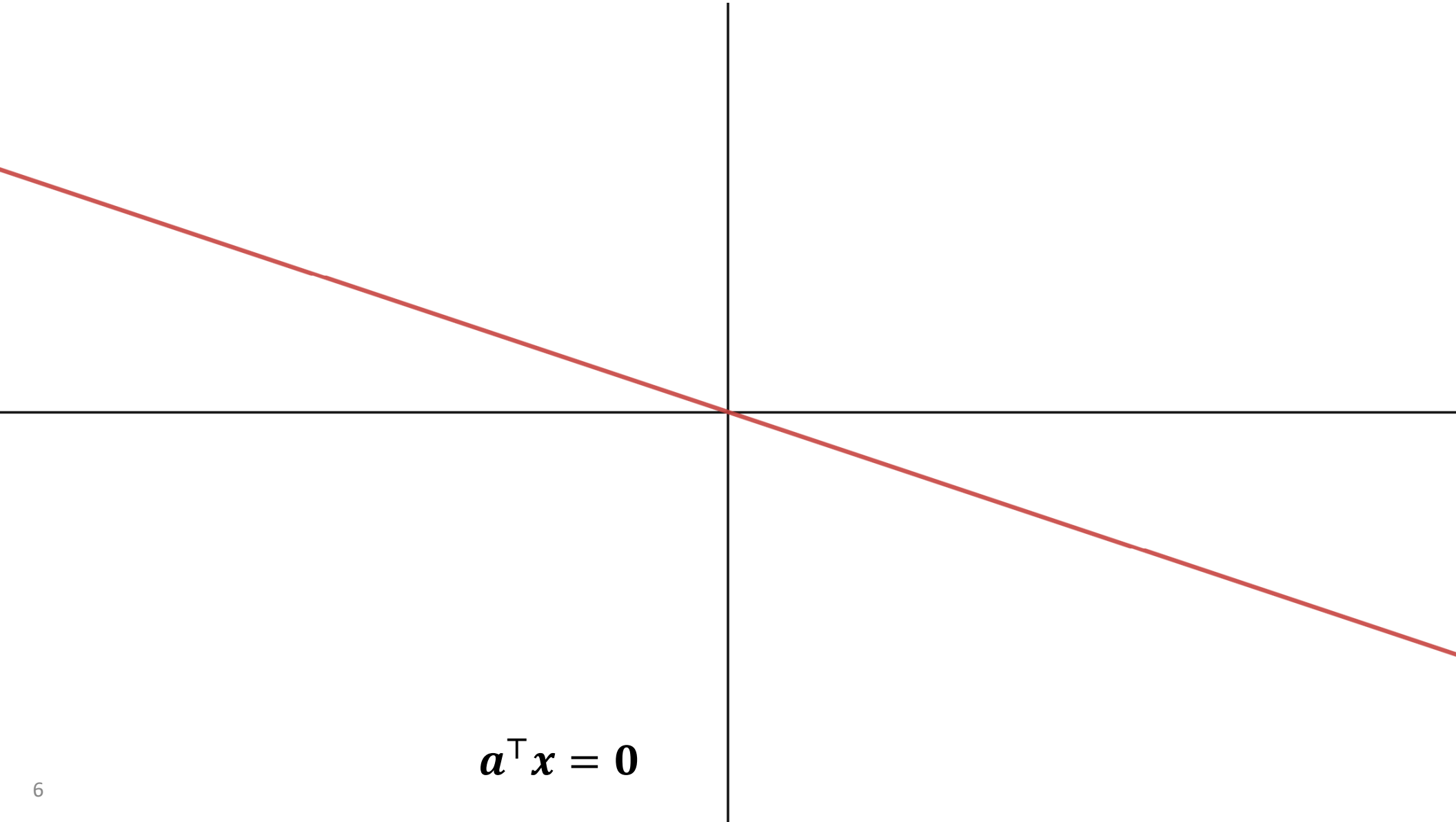
Linear Algebra primer

For $\mathbf{a}, \mathbf{x} \in \mathbb{R}^n$ we think of \mathbf{a} and \mathbf{x} as column vectors

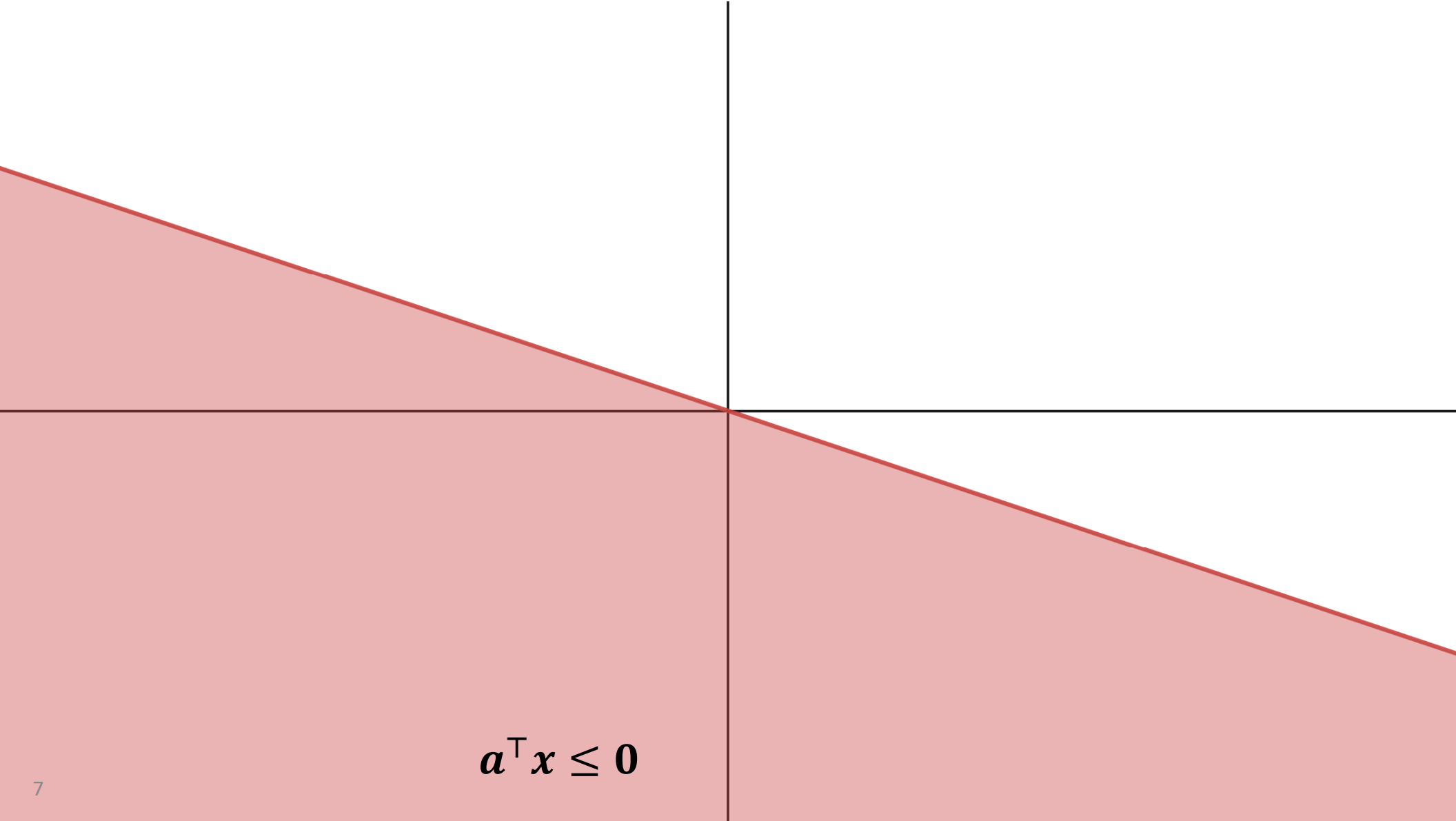
$$\mathbf{a}^\top \mathbf{x} = a_1 x_1 + \cdots + a_n x_n$$

The set of \mathbf{x} satisfying $\mathbf{a}^\top \mathbf{x} = 0$ is *hyperplane*

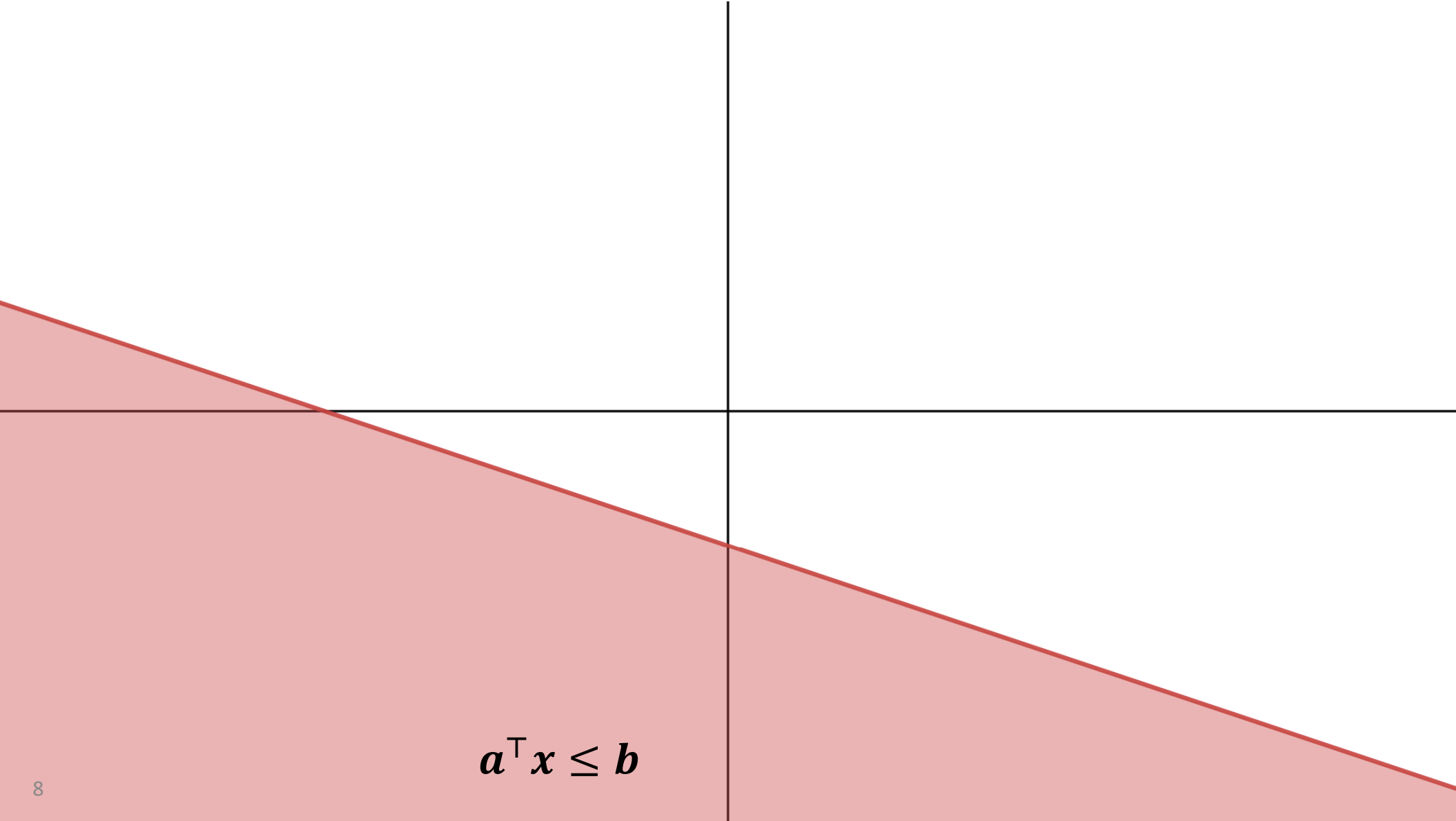




$$\mathbf{a}^T \mathbf{x} = 0$$



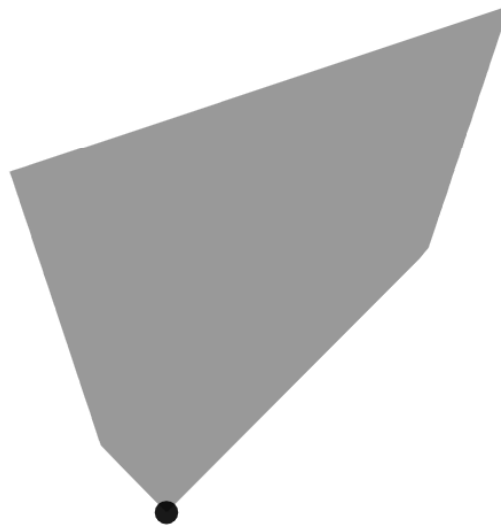
$$\mathbf{a}^T \mathbf{x} \leq 0$$



$$a^T x \leq b$$

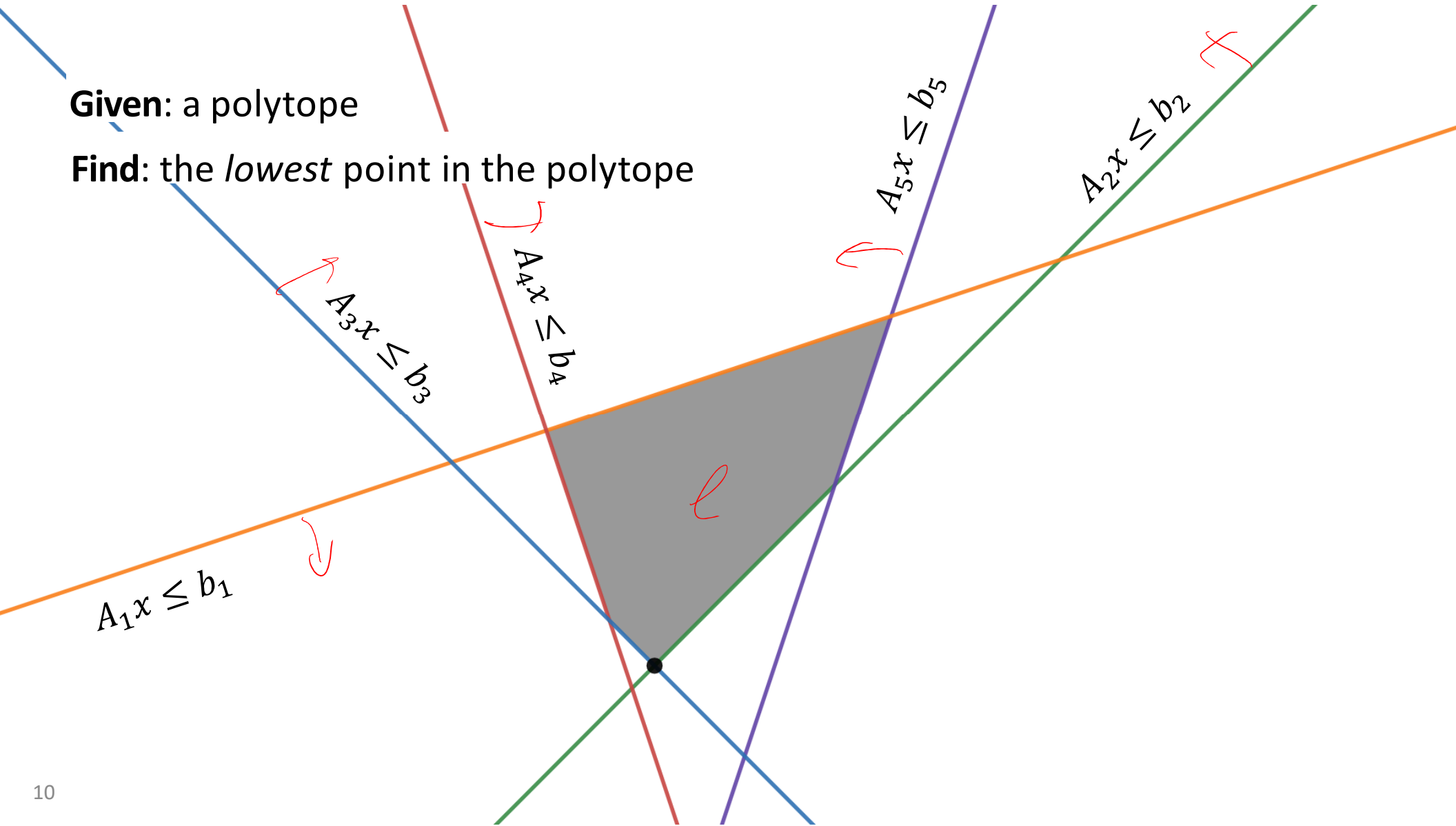
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Linear Algebra primer

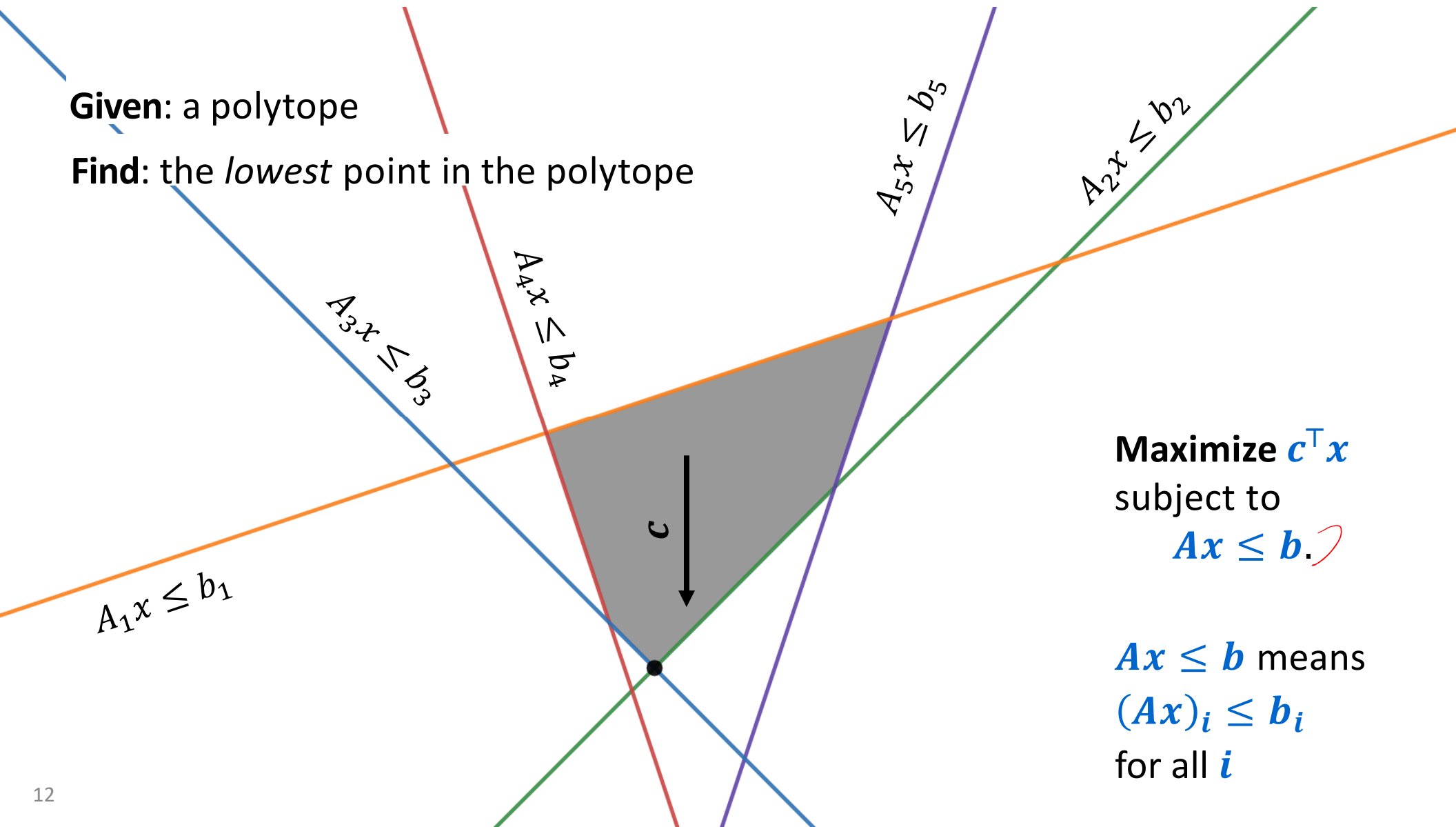
For $\mathbf{a}, \mathbf{x} \in \mathbb{R}^n$ we think of \mathbf{a} and \mathbf{x} as column vectors

$$\mathbf{a}^\top \mathbf{x} = a_1 x_1 + \cdots + a_n x_n$$

Write $m \times n$ matrix A , for $A\mathbf{x} = \begin{bmatrix} A_1 \mathbf{x} \\ A_2 \mathbf{x} \\ A_3 \mathbf{x} \\ \dots \\ A_m \mathbf{x} \end{bmatrix}$ where A_1, \dots, A_m are rows of A .

Given: a polytope

Find: the *lowest* point in the polytope



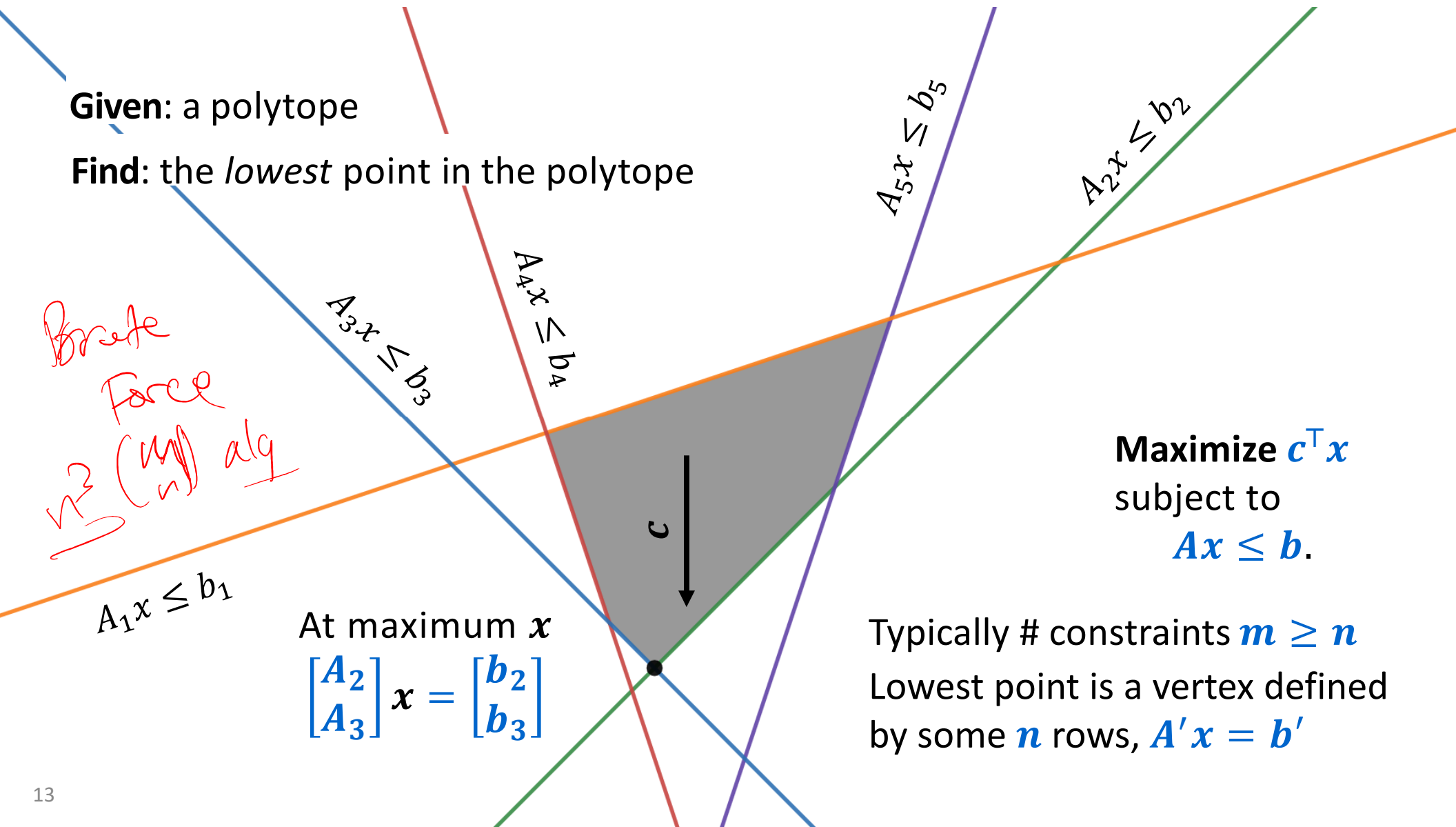
Maximize $c^T x$
subject to
 $Ax \leq b.$

$Ax \leq b$ means
 $(Ax)_i \leq b_i$
for all i

Given: a polytope

Find: the *lowest* point in the polytope

Brute
Force
 n^2 (m) alg




Maximize $c^T x$
subject to
 $Ax \leq b$.

At maximum x

$$\begin{bmatrix} A_2 \\ A_3 \end{bmatrix} x = \begin{bmatrix} b_2 \\ b_3 \end{bmatrix}$$

Typically # constraints $m \geq n$
Lowest point is a vertex defined
by some n rows, $A'x = b'$

Standard Form


$$\begin{aligned} &\text{Maximize } c^T x \\ &\text{subject to} \\ &Ax \leq b \\ &x \geq 0 \end{aligned}$$

$$\begin{aligned} &\text{Maximize } z_1 + 2z_3 \\ &\text{subject to} \end{aligned}$$

$$\begin{aligned} 2z_1 - z_2 + 3z_3 &\leq 1 \\ -z_1 + z_2 - z_3 &\leq 5 \end{aligned}$$

replace each z_i by
 $x_{i,a} - x_{i,b}$
for $x_{i,a}, x_{i,b} \geq 0$

$$\begin{aligned} &\text{Maximize } (x_{1,a} - x_{1,b}) + 2(x_{3,a} - x_{3,b}) \\ &\text{subject to} \end{aligned}$$

$$\begin{aligned} 2(x_{1,a} - x_{1,b}) - (x_{2,a} - x_{2,b}) + 3(x_{3,a} - x_{3,b}) &\leq 1 \\ -(x_{1,a} - x_{1,b}) + (x_{2,a} - x_{2,b}) - (x_{3,a} - x_{3,b}) &\leq 5 \\ x &\geq 0 \end{aligned}$$

$$c = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$
$$A = \begin{bmatrix} 2 & -1 & 3 \\ -1 & 1 & -1 \end{bmatrix}$$
$$b = \begin{bmatrix} 1 \\ 5 \end{bmatrix}$$

Max Flow

Given: A Flow Network $G = (V, E)$ with source s , sink t , and $c: E \rightarrow \mathbb{R}^{\geq 0}$

Maximize flow out of s

subject to

- respecting capacities
- flow conservation at internal nodes

LP Variables:

x_e for each $e \in E$ representing flow on edge e

Maximize $\sum_{e \text{ out of } s} x_e$

subject to

$0 \leq x_e \leq c(e)$ for every $e \in E$

$$\sum_{e \text{ out of } v} x_e = \sum_{e \text{ into } v} x_e$$

for every node $v \in V - \{s, t\}$

Max Flow

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subject to

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for every node $v \in V - \{s, t\}$

Replace equality constraints by a pair of inequalities



LP
Maximize $c^T x$
subject to

$$Ax \leq b$$
$$x \geq 0$$

This is for the c above.
Nothing to do with capacities!

1. $c_e = \begin{cases} 1 & \text{if } e \text{ out of } s \\ 0 & \text{otherwise} \end{cases}$
2. $x_e \leq c(e)$
3. $\sum_{e \text{ out of } v} x_e - \sum_{e \text{ into } v} x_e \leq 0$
4. $\sum_{e \text{ into } v} x_e - \sum_{e \text{ out of } v} x_e \leq 0$
5. $x \geq 0$

Minimization or Maximization

Minimize $c^T x$

subject to

$$Ax \geq b$$

$$x \geq 0$$



Maximize $(-c)^T x$

subject to

$$(-A)x \leq (-b)$$

$$x \geq 0$$

*different
objective
function
value
(Sign flip)*

Shortest Paths

Given: Directed graph $G = (V, E)$
vertices s, t in V

Find: (length of) shortest path from s to t

Claim: Length ℓ of the shortest path is the solution to this program.

Proof sketch: A shortest path yields a solution of cost ℓ . Optimal solution must be a combination of flows on shortest paths also cost ℓ ; otherwise there is a part of the 1 unit of flow that gets counted on more than ℓ edges.

Minimize $\sum_e x_e$ Sum of flow on all edges

subject to

$x \geq 0$

$\sum_{e \text{ out of } s} x_e = 1$ Flow out of s is 1

$\sum_{e \text{ into } t} x_e = 1$ Flow into t is 1

$\sum_{e \text{ out of } v} x_e = \sum_{e \text{ into } v} x_e$
for every node $v \in V - \{s, t\}$
Flow conservation

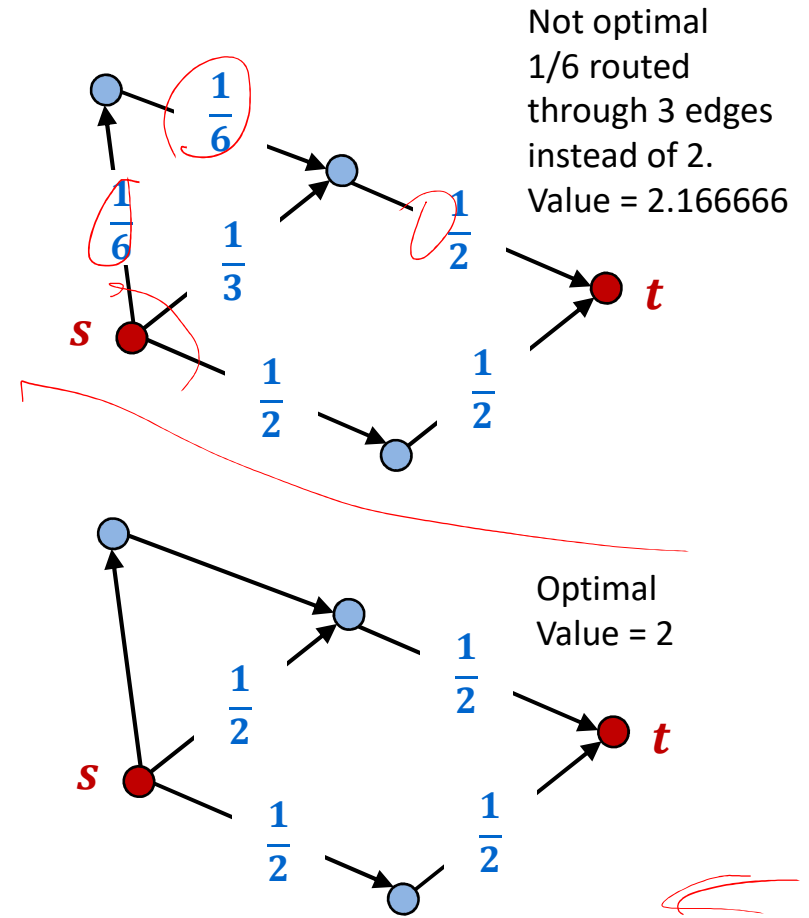
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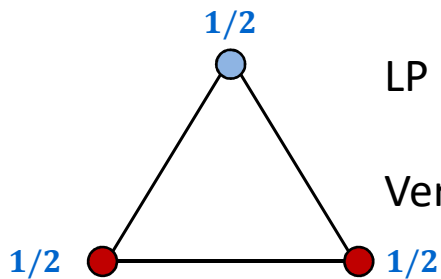


Vertex Cover

Given: Undirected graph $G = (V, E)$

Find: smallest set of vertices touching all edges of G .

Doesn't work: To define a set we need $x_v = 0$ or $x_v = 1$



LP minimum = $3/2$

Vertex Cover minimum = 2

Natural Variables for LP:

x_v for each $v \in V$

Minimize $\sum_v x_v$

subject to

X $0 \leq x_v \leq 1$ for each node $v \in V$

$x_u + x_v \geq 1$ for each edge $\{u, v\} \in E$

This LP optimizes for a different problem:
“fractional vertex cover”.

x_v indicates the fraction of vertex v that is chosen in the cover.

What makes Max Flow different?

For Vertex Cover we only got a fractional optimum but for Max Flow can get integers.

- Why?
 - Ford-Fulkerson analysis tells us this for Max Flow.
 - Is there a reason we can tell just from the LP view?

Recall: Optimum is at some vertex x satisfying $A'x = b'$ for some subset of exactly n constraints.

This means that $x = (A')^{-1}b'$.

Entries of the matrix inverse are quotients of determinants of sub-matrices of A' so, for integer inputs, optimum is always rational.

Fact: Every full rank submatrix of MaxFlow matrix A has determinant ± 1

\Rightarrow all denominators are $\pm 1 \Rightarrow$ integers. A is “totally unimodular”

Next: How **MaxFlow=MinCut** is an example of a general “duality” property of LPs