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CSE 421

Linear Programming

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Linear Programming

System of Linear Equations

Find a solution to

$$x_3 - x_1 = 4$$

$$x_3 - 2x_2 = 3$$

$$x_1 + 2x_2 + x_3 = 7$$

Can be solved by Gaussian elimination method in $O(n^3)$
when we have n variables/ n constraints

Linear Algebra Premier

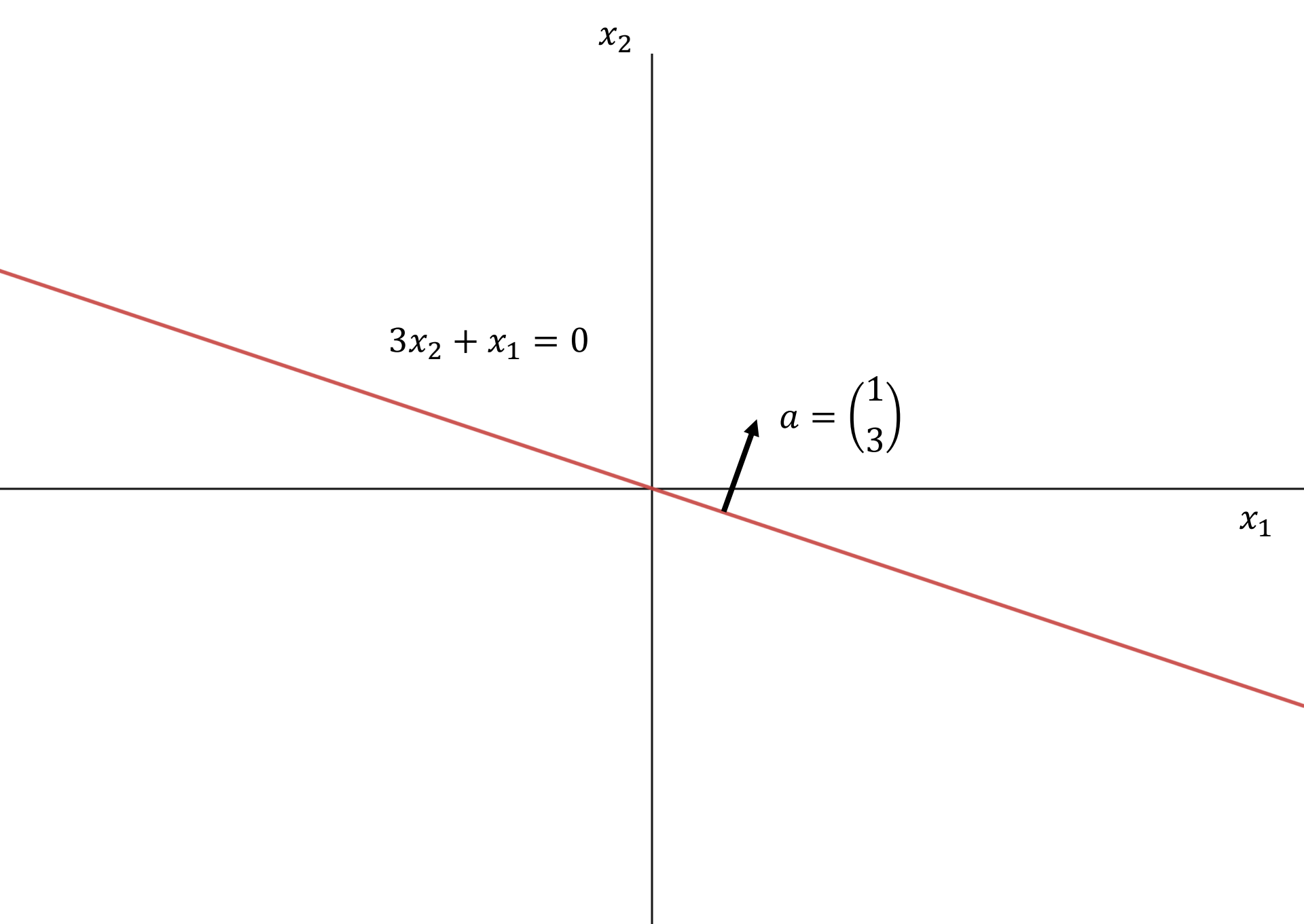
Let a be a column vector in \mathbb{R}^d and x a column vector of d variables.

$$\langle a, x \rangle = a^T x = a_1 x_1 + a_2 x_2 + \cdots + a_d x_d$$

Hyperplane: A hyperplane is the set of points x such that $\langle a, x \rangle = b$ for some $b \in \mathbb{R}$

Halfspace: A halfspace is the set of points on one side of a hyperplane.

$$\{x: \langle a, x \rangle \leq b\} \quad \text{or} \quad \{x: \langle a, x \rangle \geq b\}$$

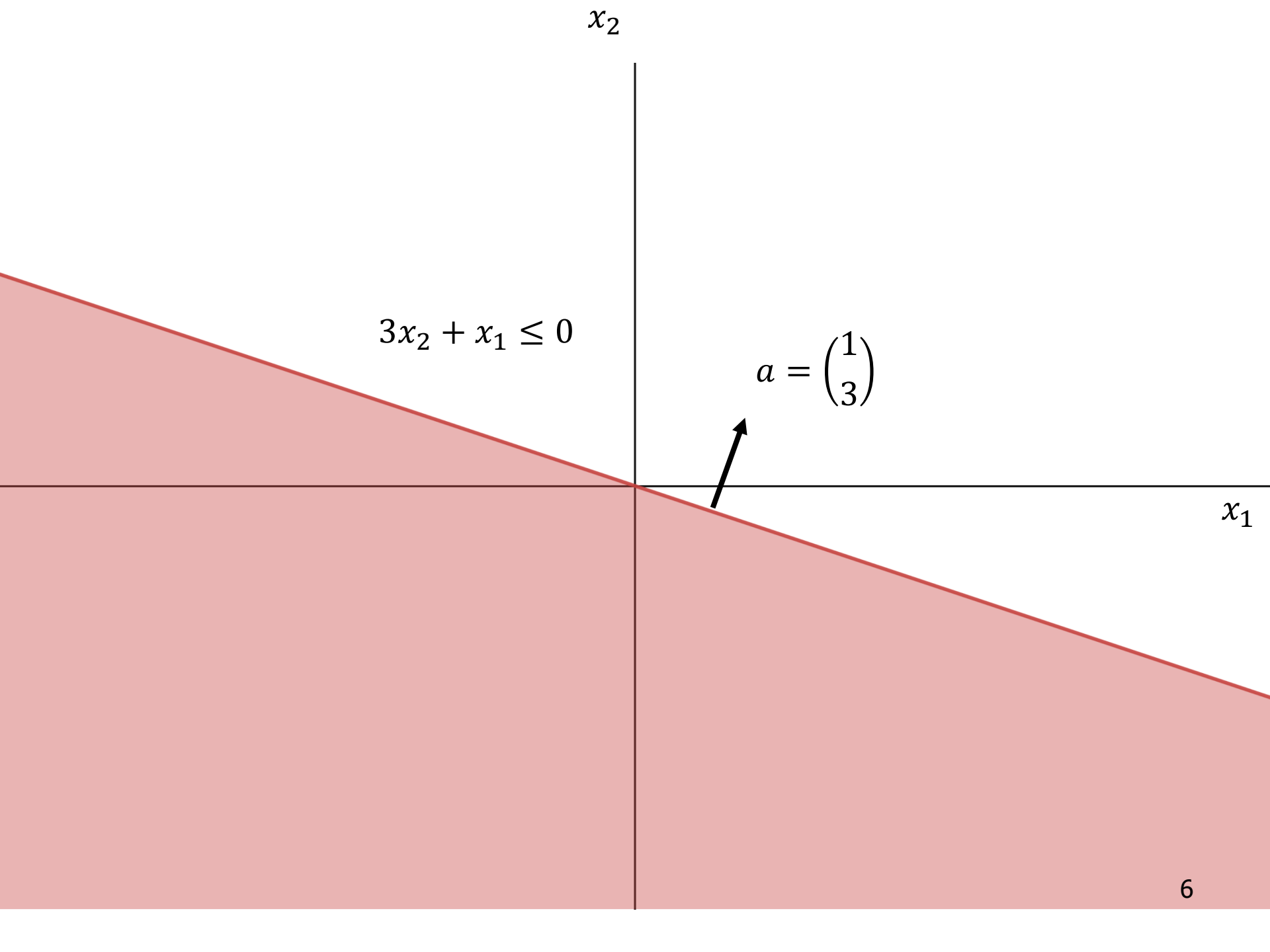


x_2

$$3x_2 + x_1 \leq 0$$

$$a = \begin{pmatrix} 1 \\ 3 \end{pmatrix}$$

x_1



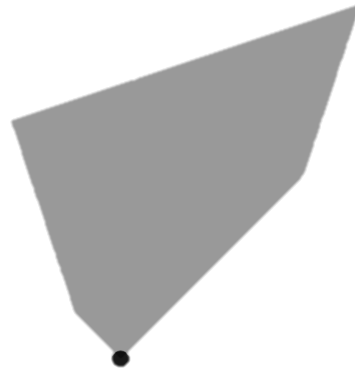
x_2

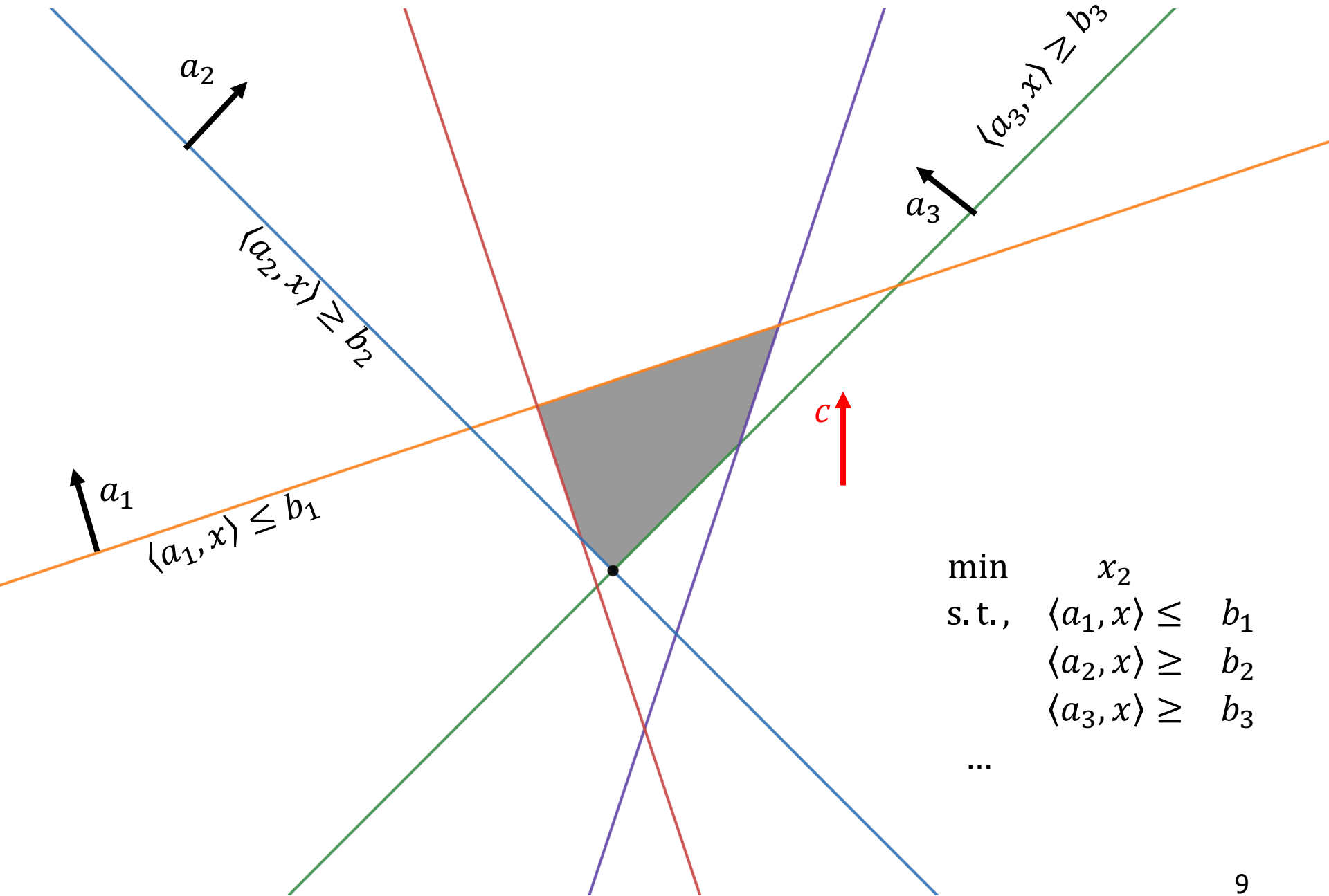
$$3x_2 + x_1 \leq -3$$

x_1

$a = \begin{pmatrix} 1 \\ 3 \end{pmatrix}$

Find the smallest point in a polytope





Linear Programming

Optimize a linear function subject to linear inequalities

obj

$$\begin{array}{ll} \min & 3x_1 - 4x_3 \\ \text{s.t.} & x_1 + x_2 \leq 5 \\ & x_3 - x_1 = 4 \\ & x_3 - x_2 \geq -5 \\ & x_1, x_2, x_3 \geq 0 \end{array}$$

is linear
 $\langle C, x \rangle$
 \hookrightarrow cost vector

\rightarrow equalities

\rightarrow ineq

$\underbrace{\hspace{10em}}$
non-neg const

- We can have equalities and inequalities,
- We can have a linear objective functions

Linear Algebra Premier

Let a be a column vector in \mathbb{R}^d and x a column vector of d variables.

$$\langle a, x \rangle = a^T x = a_1 x_1 + a_2 x_2 + \cdots + a_d x_d$$

$$A = \begin{bmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_m^T \end{bmatrix} \quad \longrightarrow \quad Ax = \begin{pmatrix} \langle a_1, x \rangle \\ \langle a_2, x \rangle \\ \vdots \\ \langle a_m, x \rangle \end{pmatrix}$$

$$Ax \leq b \quad \longrightarrow \quad \begin{array}{l} \langle a_1, x \rangle \leq b_1 \\ \langle a_2, x \rangle \leq b_2 \\ \vdots \\ \langle a_m, x \rangle \leq b_m \end{array}$$

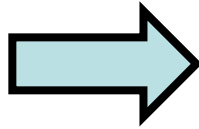
Linear Programming Standard Form

$$\begin{array}{ll} \max & \langle c, x \rangle \\ \text{s.t.}, & Ax \leq b \\ & \underbrace{x \geq 0} \end{array} \quad \text{no equalities}$$

you can turn any LP into standard form.

Transforming to Standard Form

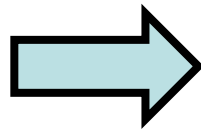
$$\begin{array}{ll} \min & y_1 - 2y_2 \\ \text{s.t.}, & y_1 + 2y_2 = 3 \\ & y_1 - y_2 \geq 1 \\ & y_1, y_2 \geq 0 \end{array}$$



$$\begin{array}{ll} \max & -y_1 + 2y_2 \\ \text{s.t.}, & y_1 + 2y_2 \leq 3 \\ & -(y_1 + 2y_2) \leq -3 \\ & -(y_1 - y_2) \leq -1 \\ & y_1, y_2 \geq 0 \end{array}$$

$$\begin{array}{ll} \max & y_1 \\ \text{s.t.}, & y_1 + y_2 \leq 3 \\ & y_2 \geq 0 \end{array}$$

Replace y_1
with $z_1 - z'_1$



$$\begin{array}{ll} \max & z_1 - z'_1 \\ \text{s.t.}, & (z_1 - z'_1) + y_2 \leq 3 \\ & z_1, z'_1, y_2 \geq 0 \end{array}$$

Applications of Linear Programming

Generalizes: $Ax=b$, 2-person zero-sum games, shortest path, max-flow, matching, multicommodity flow, MST, min weighted arborescence, ...

Why significant?

- We can solve linear programming in polynomial time.
- Useful for approximation algorithms
- We can model many practical problems with a linear model and solve it with linear programming

Linear Programming in Practice:

- There are very fast implementations: IBM CPLEX, Gorubi in Python, CVX in Matlab,
- CPLEX can solve LPs with millions of variables/constraints in minutes

Example 1: Diet Problem

Suppose you want to schedule a diet for yourself. There are four category of food: veggies, meat, fruits, and dairy. Each category has its own (p)rice, (c)alory and (h)appiness per pound:

	veggies	meat	fruits	dairy
price	p_v	p_m	p_f	p_d
calorie	c_v	c_m	c_f	c_d
happiness	h_v	h_m	h_f	h_d

Linear Modeling: Consider a linear model: If we eat 0.5lb of meat, 0.2lb of fruits we will be $0.5 h_m + 0.2 h_f$ happy

- You should eat 1500 calories to be healthy
- You can spend 20 dollars a day on food.

Goal: Maximize happiness?

Diet Problem by LP

- You should eat 1500 calories to be healthy
- You can spend 20 dollars a day on food.

Goal: Maximize happiness?

	veggies	meat	fruits	dairy
price	p_v	p_m	p_f	p_d
calorie	c_v	c_m	c_f	c_d
happiness	h_v	h_m	h_f	h_d

$$\begin{aligned} \max \quad & x_v h_v + x_m h_m + x_f h_f + x_d h_d \\ \text{s. t.} \quad & x_v p_v + x_m p_m + x_f p_f + x_d p_d \leq 20 \\ & x_v c_v + x_m c_m + x_f c_f + x_d c_d \leq 1500 \\ & x_v, x_m, x_f, x_d \geq 0 \end{aligned}$$

#pounds of veggies, meat, fruits, dairy to eat per day

Components of a Linear Program

- Set of variables
- Bounding constraints on variables,
 - Are they nonnegative?
- Objective function
- Is it a minimization or a maximization problem
- LP Constraints, make sure they are linear
 - Is it an equality or an inequality?

Example 2: Max Flow

Define the set of variables

- For every edge e let x_e be the flow on the edge e

Put bounding constraints on your variables

- $x_e \geq 0$ for all edge e (The flow is nonnegative)

Write down the constraints

- $x_e \leq c(e)$ for every edge e , (Capacity constraints)
- $\sum_{e \text{ out of } v} x_e = \sum_{e \text{ in to } v} x_e \quad \forall v \neq s, t$ (Conservation constraints)

Write down the objective function

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

Decide if it is a minimize/maximization problem

- **max**

Example 2: Max Flow

$$\begin{aligned} \max \quad & \sum_{e \text{ out of } s} x_e \\ \text{s.t.} \quad & \sum_{e \text{ out of } v} x_e = \sum_{e \text{ in to } v} x_e \quad \forall v \neq s, t \\ & x_e \leq c(e) \quad \forall e \\ & x_e \geq 0 \quad \forall e \end{aligned}$$

Q: Do we get exactly the same properties as Ford Fulkerson?

A: Not necessarily, the max-flow **may not be integral**

Example 3: Min Cost Max Flow

Suppose we can route 100 gallons of water from s to t .
But for every pipe edge e we have to pay $p(e)$
for each gallon of water that we send through e .

Goal: Send 100 gallons of water from s to t with minimum possible cost

$$\begin{array}{ll} \min & \sum_{e \in E} p(e) \cdot x_e \\ \text{s. t.} & \sum_{e \text{ out of } v} x_e = \sum_{e \text{ in to } v} x_e \quad \forall v \neq s, t \\ & \sum_{e \text{ out of } s} x_e = 100 \\ & x_e \leq c(e) \quad \forall e \\ & x_e \geq 0 \quad \forall e \end{array}$$

Linear Programming and Approximation Algorithms

Integer Program for Vertex Cover

Given a graph $G=(V,E)$ with costs c_v on the vertices. Find a vertex cover of G with minimum cost, i.e., $\min \sum_{v \in S} c_v$

Write LP with Integrality Constraint:

- Variables: One variable x_v for each vertex v
- Bound: $x_v \in \{0,1\}$
- Edge cover Constraints: $x_u + x_v \geq 1$ for every edge $(u, v) \in E$
- Obj: $\min \sum_v c_v x_v$

IP for Vertex Cover

$$\begin{aligned} \min \quad & \sum_v c_v x_v \\ \text{s.t.}, \quad & x_v + x_u \geq 1 \quad \forall (u, v) \in E \\ & x_v \in \{0, 1\} \quad \forall v \in V \end{aligned}$$

IP is NP-complete general!
But there are fast algorithms in
practice that often work

Fact:

min vertex cover.

Pf:

- First, any vertex cover S , $x_v = \begin{cases} 1 & \text{if } v \in S \\ 0 & \text{o.w.} \end{cases}$ is feasible
- For any feasible solution x , the $S = \{v: x_v = 1\}$ is a vertex cover

LP Relaxation Vertex Cover

$$\begin{aligned} \min \quad & \sum_v c_v x_v \\ \text{s.t.}, \quad & x_v + x_u \geq 1 \quad \forall (u, v) \in E \\ & 0 \leq x_v \leq 1 \quad \forall v \in V \end{aligned}$$

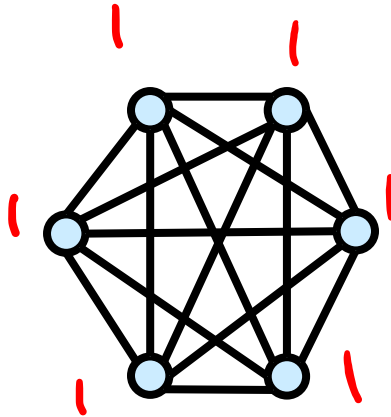
Fact: OPT-LP \leq Min Vertex Cover

Pf: Min vertex cover is a feasible solution of the LP

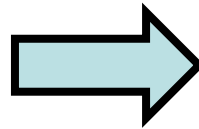
Q: Can we hope to get an integer solution?

Bad Optimum solutions

$$\begin{aligned} \min \quad & \sum_v c_v x_v \\ \text{s.t.}, \quad & x_v + x_u \geq 1 \quad \forall (u, v) \in E \\ & 0 \leq x_v \leq 1 \quad \forall v \in V \end{aligned}$$



K_n complete graph



A feasible solution:
Set $x_v = 0.5$ for all v
in the complete graph

If $c_v = 1$ for all v , then
Min vertex cover = $n - 1$
But OPT LP = $n/2$.

Approximation Alg for Vertex Cover

Given a graph $G=(V,E)$ with costs c_v on the edges. Find a vertex cover of G with minimum cost, i.e., $\min \sum_{v \in S} c_v$

Thm: There is a 2-approximation Alg for **weighted** vertex cover.

ALG: Solve LP. Let $S = \{v: x_v \geq 0.5\}$. Output S .

Pf: First, for every edge (u, v) , $x_u + x_v \geq 1$ So at least one is in S . So, S is a vertex cover.

Second,

$$\sum_{v \in S} c_v \leq \sum_{v \in S} c_v (2x_v) \leq 2 \text{OPTLP} \leq 2 \text{ Min Vertex Cov}$$

Duality

Intro to Duality

$$\begin{aligned} \max \quad & x_1 + 2x_2 \\ \text{s. t.}, \quad & x_1 + 3x_2 \leq 2 \\ & 2x_1 + 2x_2 \leq 3 \\ & x_1, x_2 \geq 0 \end{aligned}$$

Optimum solution: $x_1 = 5/4$ and $x_2 = 1/4$ with value $x_1 + 2x_2 = 7/4$
How can you prove an upper-bound on the optimum?

First attempt: Since $x_1, x_2 \geq 0$

$$x_1 + 2x_2 \leq x_1 + 3x_2 \leq 2$$

Second attempt:

$$x_1 + 2x_2 \leq \frac{2}{3}(x_1 + 3x_2) + \frac{1}{3}(2x_1 + 2x_2) \leq \frac{2}{3}(2) + \frac{1}{3}(3) = \frac{7}{3}$$

Third attempt:

$$x_1 + 2x_2 \leq \frac{1}{2}(x_1 + 3x_2) + \frac{1}{4}(2x_1 + 2x_2) \leq \frac{1}{2}(2) + \frac{1}{4}(3) = \frac{7}{4}$$

Dual Certificate

$$\begin{array}{ll} \max & x_1 + 2x_2 \\ \text{s. t.}, & x_1 + 3x_2 \leq 2 \\ & 2x_1 + 2x_2 \leq 3 \\ & x_1, x_2 \geq 0 \end{array} \quad \begin{array}{l} y_1 \\ y_2 \end{array}$$

Goal: Minimize $2y_1 + 3y_2$

But, we must make sure the sum of the LHS is at most objective, i.e.,

$$x_1 + 2x_2 \leq y_1(x_1 + 3x_2) + y_2(2x_1 + 2x_2)$$

In other words,

$$\begin{array}{l} 1 \leq 1 \cdot y_1 + 2 \cdot y_2 \\ 2 \leq 3 \cdot y_1 + 2 \cdot y_2 \end{array}$$

Finally, $y_1, y_2 \geq 0$ (else the direction of inequalities change)

Dual Program

$$\begin{array}{ll}\max & x_1 + 2x_2 \\ \text{s. t.}, & x_1 + 3x_2 \leq 2 \\ & 2x_1 + 2x_2 \leq 3 \\ & x_1, x_2 \geq 0\end{array}$$

$$\begin{array}{l}\text{OPT: } x_1 = 5/4 \text{ and } x_2 = 1/4 \\ \text{Value } 7/4\end{array}$$

$$\begin{array}{ll}\min & 2y_1 + 3y_2 \\ \text{s. t.}, & y_1 + 2y_2 \geq 1 \\ & 3y_1 + 2y_2 \geq 2 \\ & y_1, y_2 \geq 0\end{array}$$

$$\begin{array}{l}\text{OPT: } y_1 = 1/2 \text{ and } y_2 = 1/4 \\ \text{Value } 7/4\end{array}$$