

Lecture 13: Linear Programming

- Reminder: def.
 - Some examples
 - shortest paths
 - L_1 regression
 - Duality of LPs.
- max flow* (with arrow pointing to 'shortest paths')

objective fn

maximize $2x_1 + 3x_2$

subject to

- $4x_1 + 8x_2 \leq 12$
- $2x_1 + x_2 \leq 3$
- $3x_1 + 2x_2 \leq 4$
- $x_1, x_2 \geq 0$

$l_i(x) \leq b_i$

$a_1x_1 + a_2x_2 + \dots + a_nx_n \leq b$

LP = linear program

$$\begin{cases}
 \text{maximize} & 2x_1 + 3x_2 \\
 \text{subject to} & 4x_1 + 8x_2 \leq 12 \\
 & 2x_1 + x_2 \leq 3 \\
 & 3x_1 + 2x_2 \leq 4 \\
 & x_1, x_2 \geq 0.
 \end{cases}$$

"general form"

$$\begin{aligned}
 & \max c^T x \\
 & Ax \leq b \\
 & x \geq 0
 \end{aligned}$$

↑
componentwise

$$\begin{aligned}
 \text{Minimize} & 12y_1 + 3y_2 + 4y_3 \\
 \text{subject to} & 4y_1 + 2y_2 + 3y_3 \geq 2 \\
 & 8y_1 + y_2 + 2y_3 \geq 3 \\
 & y_1, y_2, y_3 \geq 0.
 \end{aligned}$$

$$c^T = [2 \quad 3]$$

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$A = \begin{bmatrix} 4 & 8 \\ 2 & 1 \\ 3 & 2 \end{bmatrix}$$

$$b = \begin{bmatrix} 12 \\ 3 \\ 4 \end{bmatrix}$$

$$\begin{aligned}
 & \min c^T x \\
 & Ax \leq b \\
 & Cx \geq d \\
 & x \geq 0
 \end{aligned}$$

$$d^T = [12 \quad 3 \quad 4]$$

$$A^0 = \begin{pmatrix} 4 & 2 & 3 \\ 8 & 1 & 2 \end{pmatrix}$$

$$f = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$$\begin{aligned}
 & \min d^T y \\
 & \text{st } Ey \geq f \\
 & y \geq 0
 \end{aligned}$$

$$y = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix}$$

① max flow as LP

$$G = (V, E)$$

$$\max \sum_{u: su \in E} f(su)$$

← total flow leaving s

$$\sum_{u: uv \in E} f(uv) = \sum_{w: vw \in E} f(vw)$$

∀ vertex $v \neq s, t$

← flow conservation

$$0 \leq f(e) \leq u_e \quad \forall e$$

← capacities are respected

$$f(e) = 0$$

∀ edges entering s
leaving t

~~$f(e) \in \mathbb{Z} \quad \forall \quad u_e \in \mathbb{Z}_e$~~

Min Cost Max flow:

① solve LP to find max flow value F^*

②
$$\min \sum_e c_e f(e)$$

$$\sum_{u: s \rightarrow u} f(su) = F^*$$

$$\sum_{u: u \rightarrow v} f(uv) = \sum_{w: v \rightarrow w} f(vw) \quad \forall v \neq s, t$$

$$0 \leq f(e) \leq u_e \quad \forall e$$

~~no~~ no flow leaves t , enters s

① figure out vars

② figure out constraints

③ write them down as linear constraints.

L₁-regression

vars: m, c

$$\min \sum_{i=1}^n |b_i - (ma_i + c)|^2$$

↑ absolute value!!
not an LP.

$$\min \sum_{i=1}^n z_i$$

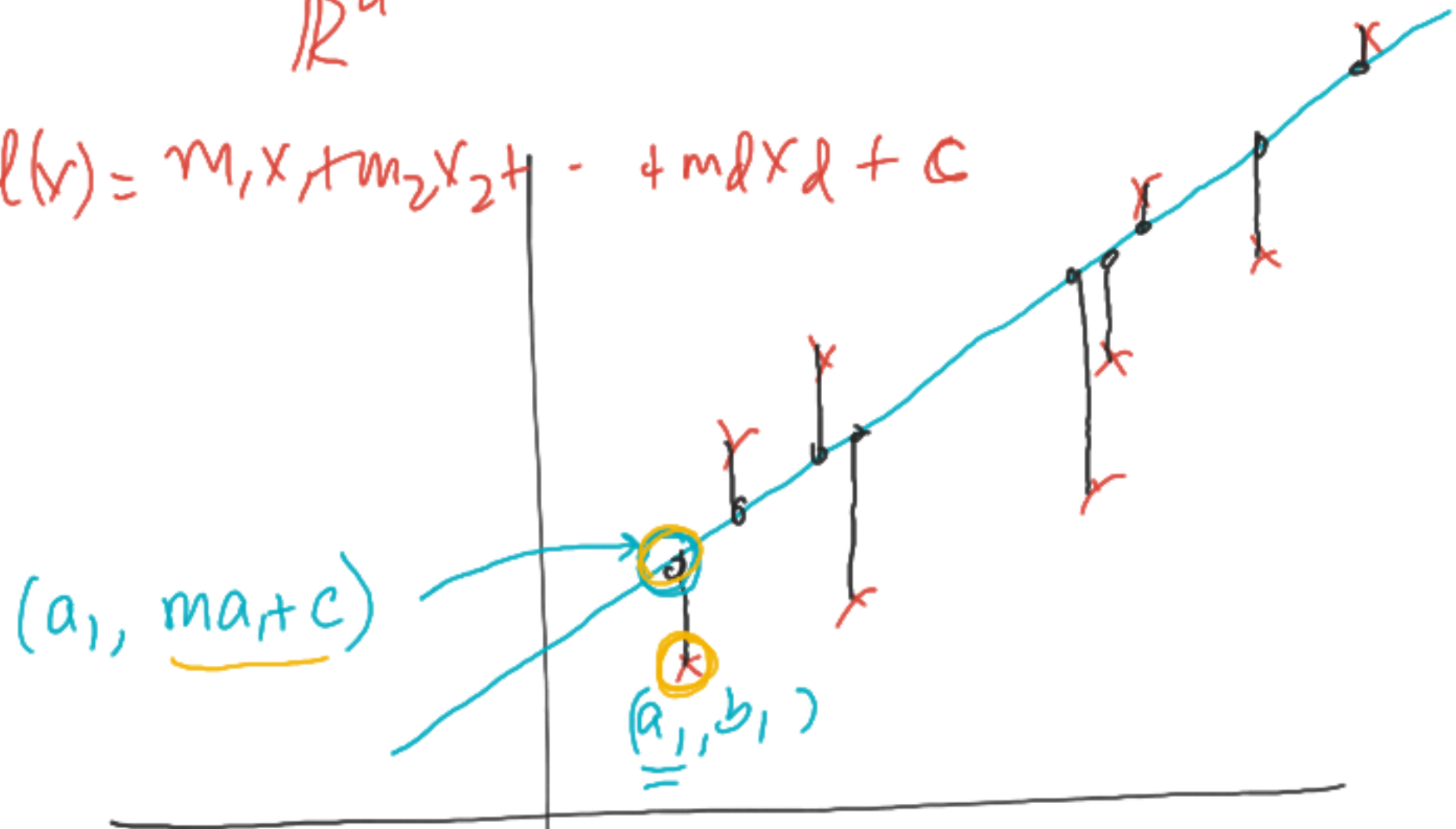
$$z_i \geq b_i - (ma_i + c)$$

$$z_i \geq -(b_i - (ma_i + c))$$

$$z_i \geq |b_i - (ma_i + c)|$$

data $(a_1, b_1) \dots (a_n, b_n)$
↓
 \mathbb{R}^d

$$l(x) = m_1x_1 + m_2x_2 + \dots + m_dx_d + c$$



slope

$$\underline{\text{line}} = mx + c$$

↑ offset

Shortest Paths:

directed G , source s , sink t \rightarrow destination
find shortest $s-t$ path
length $l(e)$ on edges.
 ≥ 0



$$\min \sum_e l(e) x_e$$

$$\sum_{e \text{ leaving } s} x(e) = 1$$

$$\sum_{e \text{ entering } t} x(e) = 1$$

$$\sum_{e \text{ entering } v} x(e) = \sum_{e \text{ leaving } v} x(e)$$

$$0 \leq x_e \leq 1$$

Solve this min cost flow problem
Every flow carrying path
is a shortest path.

variables: distance of vertex v from s . d_v

if want true distances to all nodes from s ,

$$\max \sum_{v \in V} d_v$$

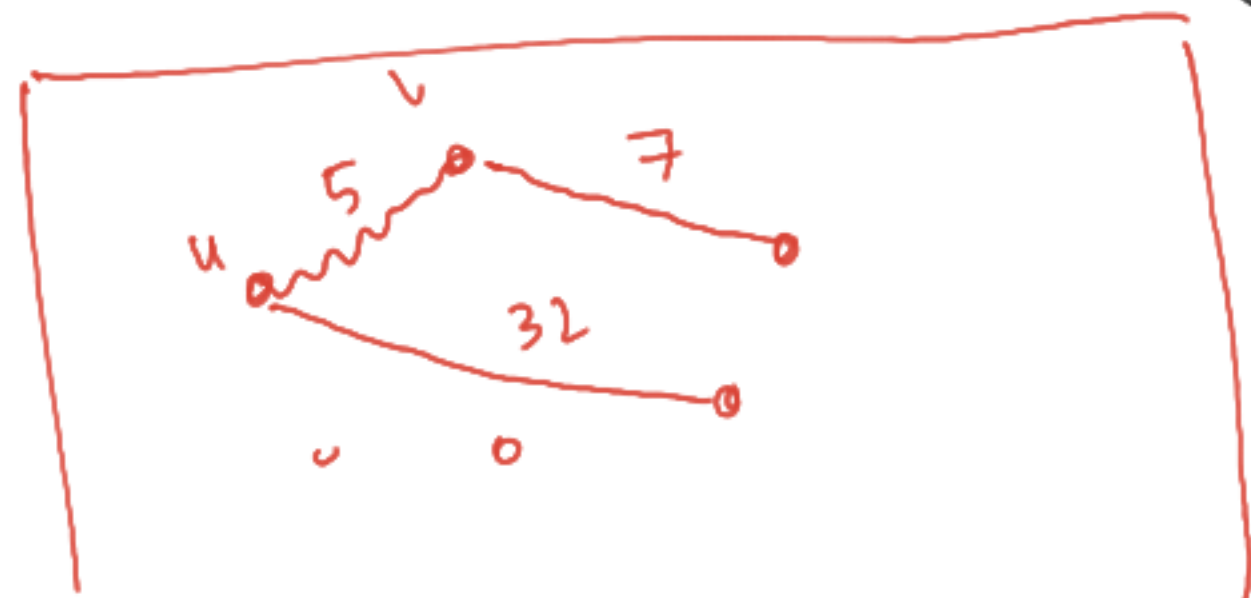
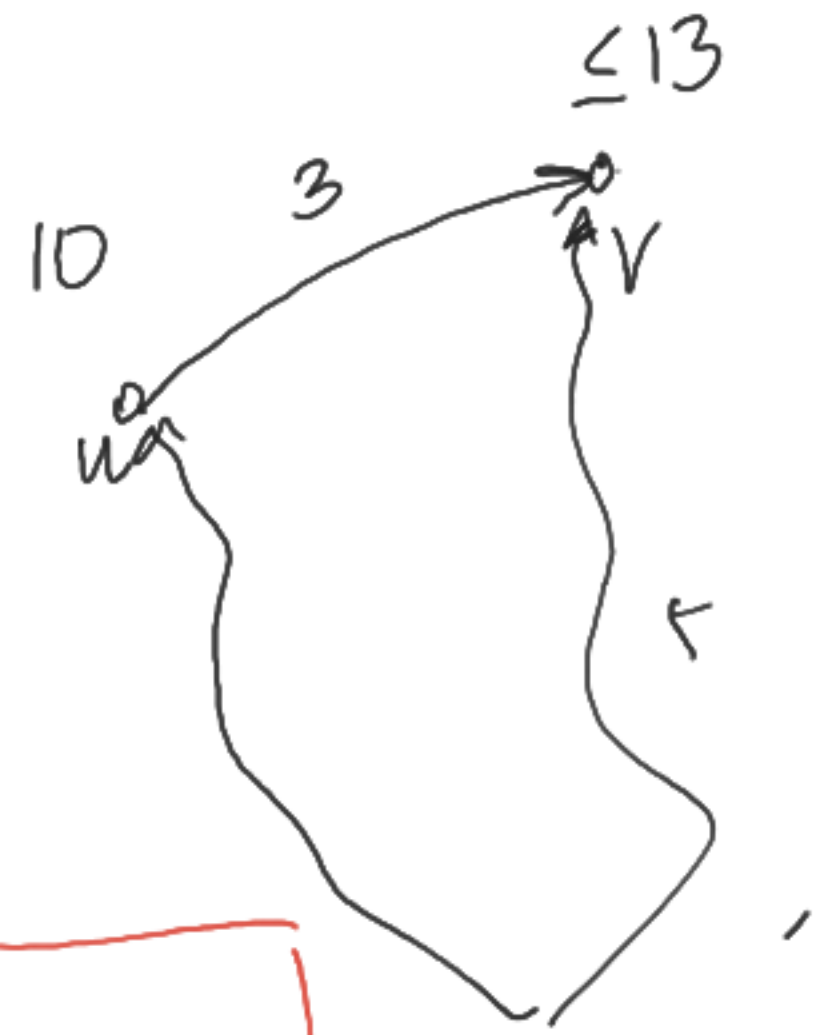
max d_E

$$d_v \leq d_u + l(uv)$$

$$d_s = 0$$

strings are unshetshello

$\forall \text{ edge } (u,v) \in E$



Duality of LPs

$$\begin{array}{ll}
 \text{maximize} & 2x_1 + 3x_2 \\
 \text{subject to} & 4x_1 + 8x_2 \leq 12 \\
 & 2x_1 + x_2 \leq 3 \\
 & 3x_1 + 2x_2 \leq 4 \\
 & x_1, x_2 \geq 0.
 \end{array}$$

y_1
 y_2
 y_3

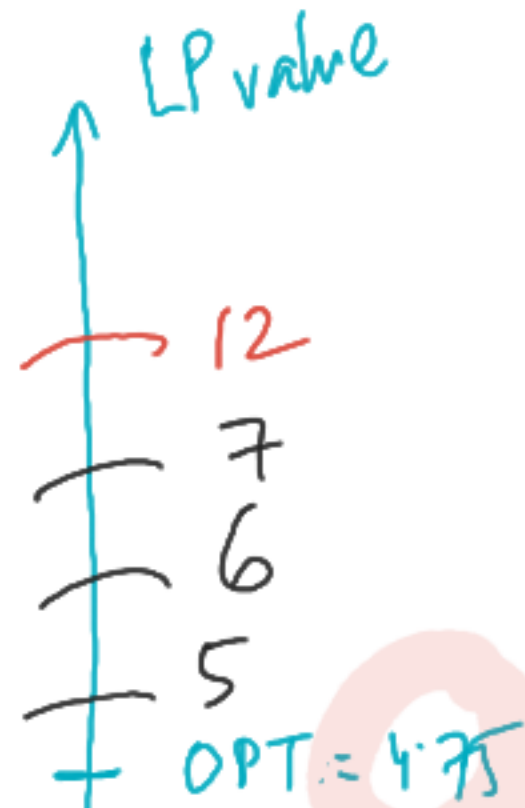
$$\begin{array}{ll}
 4y_1 + 2y_2 + 3y_3 \geq 2 & y_1, y_2, y_3 \geq 0 \\
 8y_1 + y_2 + 2y_3 \geq 3 \\
 \min 12y_1 + 3y_2 + 4y_3
 \end{array}$$

Give me upper bounds on OPT for this LP?

by nonneg

$$2x_1 + 3x_2 \leq \frac{1}{2}(4x_1 + 8x_2) \leq \frac{1}{2}12 = 6$$

$$\leq \frac{1}{3}(c_1 + c_2) \leq \frac{1}{3}(12 + 3) = 5$$



$$\begin{bmatrix} 2 & 3 \\ 4 & 8 \\ 2 & 1 \\ 3 & 2 \end{bmatrix} \quad \begin{bmatrix} 12 \\ 3 \\ 4 \end{bmatrix}$$

$$\begin{bmatrix} 12 & 3 & 4 \\ 4 & 2 & 3 \\ 8 & 1 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

Strong Duality Theorem

Thm: Consider any LP

$$\begin{aligned} \max \quad & c^T x \\ \text{s.t.} \quad & Ax \leq b \\ & x \geq 0 \end{aligned}$$

then \exists LP (dual)

and they are equal!

$$\min b^T y$$

$$A^T y \geq c$$

$$y \geq 0$$

s.t.

primal/dual	finite	unbounded	infeasible
finite	✓	x	x
unbdd	x	x	✓
infeasible	x	✓	⊗

max x	min $-5y_1 + 3y_2$
$-x \leq -5$	s.t. $-y_1 + y_2 \geq 1$
$x \leq 3$	$y_1, y_2 \geq 0$

$$\min -5y_1 + 3y_2$$

$$\text{s.t. } -y_1 + y_2 \geq 1$$

$$y_1, y_2 \geq 0$$

$$\begin{aligned} \max \quad & c^T x \\ & Ax \leq b \\ & x \geq 0 \end{aligned}$$

m constraints \leftarrow except non-neg
 n variables

$$A \in \mathbb{R}^{m \times n}$$

$$\begin{aligned} \min \quad & b^T y \\ & A^T y \geq c \\ & y \geq 0. \end{aligned}$$

n constraints \leftarrow except nonneg
 m variable



$$\max c^T x$$

$$Ax \geq b$$

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$x_1, x_2 \geq 0$$

def LPs.

modeled problem as LPs

LPs can be solved in poly time!

- simplex

{ - ellipsoid

- interior point