

$u \rightarrow \geq 20$

CS473 Algorithms - Lecture 19 (2022-03-31)

logistics = - part 7 due FIT
- part 8 due FIT

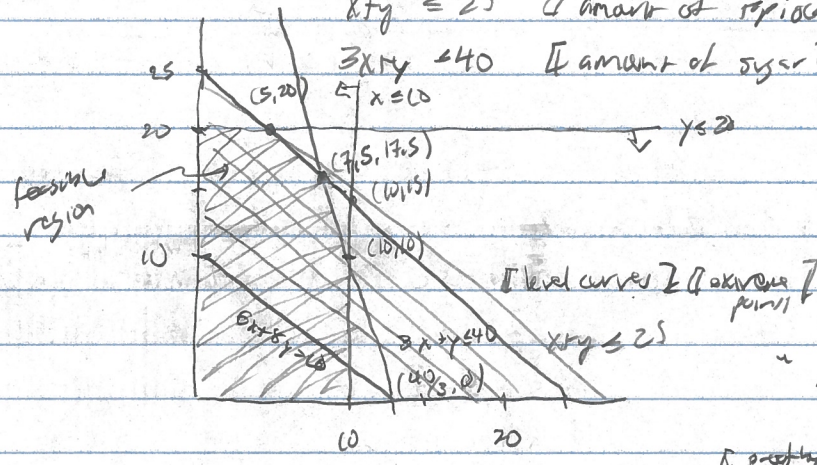
last lecture: linear programming

- example [2D]
- def
- reduction to canonical form [format]
- LP vs - max flow [idea] [bounded]
- min cut [discrete vs continuous]
- [randomized rounding]

today = linear programming
 [last lecture] [profit] [black] [example]
 [constraint] [shop] [max] $6x + 8y$

- $x, y \geq 0$ [nonneg tea per day]
- $x \leq 10$ [black tea bags]
- $y \leq 20$ [green tea bags]
- $x + y \leq 25$ [amount of replace pods]
- $3x + y \leq 40$ [amount of sugar]

kernel
crossed
reader
so $x, y \geq 0$
less



$6 \cdot 5 + 8 \cdot 20 = 190$
 $6 \cdot 7.5 + 8 \cdot 17.5 = 195$
 $6 \cdot 10 + 8 \cdot 10 = 140$
 \Rightarrow "opt is 190"

Q: how to prove

max $6x + 8y = 190$? [pretty picture] [formal/symbolic proof]

idea = derive new constraints from original constraints

eg = $x, y \geq 0$
 $x \leq 10$
 $y \leq 20$
 $x + y \leq 25$
 $3x + y \leq 40$

$x \leq 10 \Rightarrow 6x \leq 60$
 $+ y \leq 20 \Rightarrow 8y \leq 160$
 $x + y \leq 25 \Rightarrow 6x + 8y \leq 220$
 $4 \text{ vs } x + y \leq 25$

depend function
 \Rightarrow max $6x + 8y \leq 220$
 $st \dots$

$8(x+y) \leq 8 \cdot 25$
 $-2(x \geq 0)$ [crossing with inequality]
 $6x + 8y \leq 200$
 objective

$8(x+y) \leq 8 \cdot 25 = 200$
 $6x + 8y$
 or $6x + 8y$ [coordinate] $x, y \geq 0$

$2(3x+y) \leq 2 \cdot 40$
 $6x + 2y \leq 80$
 $6x + 8y \leq 200$
 objective

$6(x+y) \leq 6 \cdot 25$
 $2 \cdot y \leq 2 \cdot 20$
 $6x + 8y \leq 190$
 \Rightarrow max $6x + 8y \leq 190$
 $st \dots$

complementary slackness
 order of constraints

best $z = 170$ at $(x,y) = (5,20)$ $\rightarrow 6x + 8y = 170$

\Rightarrow 170 is optimum value

Q: mechanize this process? find best ub?

a	$x \leq 10$	\rightarrow	$ax \leq 10a$	\boxed{I}	$a \geq 0$	<small>standard inequality</small>
b	$y \leq 20$	\rightarrow	$by \leq 20b$	\boxed{II}	$b \geq 0$	
c	$x+y \leq 25$	\rightarrow	$cx + cy \leq 25c$	\boxed{III}	$c \geq 0$	
d	$3x+y \leq 40$	\rightarrow	$3dx + dy \leq 40d$	\boxed{IV}	$d \geq 0$	
e	$x \geq 0$	\rightarrow	$ex \leq 0$	\boxed{V}	$e \leq 0$	<small>inverse</small>
f	$y \geq 0$	\rightarrow	$fy \leq 0$	\boxed{VI}	$f \leq 0$	

$$(a+e)x + (b+f)y \leq 10a + 20b + 25c + 40d + 0 + 0$$

$= 6$

$= 8$

\Rightarrow

$\forall 6x + 8y$ for feasible (x,y)

Q: best ub? $\left[\begin{array}{l} \text{min} \\ \text{max} \end{array} \right]$

LP min $10a + 20b + 25c + 40d$

$\left[\begin{array}{l} \text{min ub} \\ \text{min vs max} \end{array} \right]$

st $a + c + 3d + e = 6$

$b + c + d + f = 8$

$a, b, c, d \geq 0$

$e, f \leq 0$

$\left[\begin{array}{l} \text{slack var} \\ \text{I} \end{array} \right]$

$\left[\begin{array}{l} \text{canon form} \\ \text{min} \end{array} \right]$

min $10a + 20b + 25c + 40d$

st $-(a + c + 3d) \leq -6$

$-(b + c + d) \leq -8$

$a, b, c, d \geq 0$

$\left[\begin{array}{l} \text{remove slack} \\ \text{I} \end{array} \right]$

ans = any feasible (a,b,c,d) yields ub on $6x + 8y$ $\forall (x,y)$ feasible

\Rightarrow max $6x + 8y$

min $10a + 20b + 25c + 40d$

st $x \leq 10$

st $-(a + c + 3d) \leq -6$

$y \leq 20$

$-(b + c + d) \leq -8$

$x + y \leq 25$

$3x + y \leq 40$

$a, b, c, d \geq 0$

$x, y \geq 0$

and

is equality

$6 \cdot 5 + 8 \cdot 20 = 170 = 10 \cdot 0 + 20 \cdot 2 + 25 \cdot 6 + 40 \cdot 0$

and

var = 2

4

non axis constraints = 4

2

Q: generalize?

def: primal linear program

$$P = \begin{cases} \text{max } \langle c, x \rangle \\ \text{st } Ax \leq b \\ x \geq 0 \end{cases}$$

$$\begin{aligned} c \in \mathbb{R}^n \\ A \in \mathbb{R}^{m \times n} \\ b \in \mathbb{R}^m \end{aligned}$$

if cost
if in non axis
constraints
if in axis
constraints

has dual $|| = \min \langle b, y \rangle$ m variables

$$\text{st } \begin{cases} A^T y \geq c \\ y \geq 0 \end{cases} \leftarrow n \text{ (non axis) constraints}$$

matrix transpose $(A^T)_{ij} = (A)_{ji}$

convention: all vectors as column vectors

note: dual $||$ is canonical form, but can be made to be

$$\begin{cases} \text{max } \langle c, x \rangle \\ Ax \leq b \\ x \geq 0 \\ A \text{ } m \times n \end{cases}$$

or

$$\begin{cases} \text{min } \langle b, y \rangle \\ A^T y \geq c \\ y \geq 0 \\ A^T \text{ } n \times m \end{cases}$$

canonical
minimization

$$\begin{cases} \text{max } \langle -b, y \rangle \\ (-A^T) y \leq -c \\ y \geq 0 \\ A^T \text{ } n \times m \end{cases}$$

- many choices for "the" dual

$$\text{eg: } \begin{cases} \text{max } x \\ \text{st } x \leq 1 \\ x \geq 0 \end{cases} \rightarrow$$

$$\begin{cases} \text{min } y \\ y \geq 1 \\ y \geq 0 \end{cases}$$

isomorphism

$$\begin{cases} \text{max } x \\ \text{st } x \leq 1 \\ x \leq 2 \\ x \geq 0 \end{cases} \rightarrow$$

$$\begin{cases} \text{min } y + z \\ \text{st } y + z \geq 1 \\ y, z \geq 0 \end{cases}$$

ISO

draw
structure

then (weak duality) =

$$P = \begin{cases} \text{max } \langle c, x \rangle \\ Ax \leq b \\ x \geq 0 \end{cases}$$

$$D = \begin{cases} \text{min } \langle b, y \rangle \\ A^T y \geq c \\ y \geq 0 \end{cases}$$

both feasible

pt: x feasible for $P \Rightarrow Ax \leq b$

$$y \text{ feasible for } D \Rightarrow A^T y \geq c \Rightarrow y^T A \geq c^T$$

$$\begin{aligned} y^T A x &\leq y^T b \quad (\text{since } Ax \leq b) \\ &\geq c^T x \quad (\text{since } y^T A \geq c^T) \\ &= \langle c, x \rangle \end{aligned}$$

$\Rightarrow \langle c, x \rangle \leq \langle b, y \rangle$ for any x, y feasible

$$\Rightarrow |P| = \max_{x \text{ feasible}} \langle c, x \rangle \leq \min_{y \text{ feasible}} \langle b, y \rangle = |D|$$

max vs
min

cor: - dual unbounded $\Rightarrow |D| = -\infty \Rightarrow$ primal infeasible

- primal unbounded $\Rightarrow |P| = \infty \Rightarrow$ dual infeasible

- if x, y both feasible w/ $\langle c, x \rangle = \langle b, y \rangle \Rightarrow$ both optimal

if weak vs strong? then strong duality: if P feasible and bounded, then so is D and $|P| = |D|$

do example LP in non canonical form

rmk: - dual (dual) = primal

- \Leftrightarrow If feasible, bounded \Rightarrow Π is also

- strong duality \Rightarrow LPs have same opt of optimality

then weak duality: $\Pi = \begin{cases} \max \langle c, x \rangle \\ Ax = b \\ A'x \leq b' \\ x \geq 0 \end{cases} \leq \begin{cases} \min \langle b, y \rangle + \langle b', z \rangle \\ s.t. A^T y + (A')^T z \geq c \\ y, z \geq 0 \end{cases} = \Pi$

pk: $\langle c, x \rangle = x^T c \leq x^T (A^T y + (A')^T z) = (Ax)^T y + (A'x)^T z \leq b^T y + (b')^T z = \langle b, y \rangle + \langle b', z \rangle$

schubty to max flow
 prop: $G = (V, E)$ capacitated graph, $s, t \in V$ sources $(c_e)_{e \in E} \geq 0$

max flow = $\max \sum_{e: s \rightarrow \cdot} f_e - \sum_{e: \cdot \rightarrow s} f_e$
 $\sum_{e: v \rightarrow \cdot} f_e - \sum_{e: \cdot \rightarrow v} f_e = 0, \forall v \neq s, t$
 $f_e \leq c_e$
 $f_e \geq 0$

has dual

M.O. $\sum_e c_e x_e$
 s.t. $d_v \leq d_u + x_e \quad e: u \rightarrow v \in E$
 $d_s = 0$
 $d_t = 1$
 $x_e \geq 0$

prop: LP capacity min cut

pl: $\max -F_t$
 s.t. $\sum_{e: s \rightarrow \cdot} f_e - \sum_{e: \cdot \rightarrow s} f_e - F_s = 0$
 $\sum_{e: t \rightarrow \cdot} f_e - \sum_{e: \cdot \rightarrow t} f_e - F_t = 0$
 $\sum_{e: v \rightarrow \cdot} f_e - \sum_{e: \cdot \rightarrow v} f_e = 0 \quad \forall v \neq s, t$
 $x_e \geq 0 \rightarrow f_e \leq c_e$
 $f_e \geq 0$

schubty

min $d_s \cdot 0 + d_t \cdot 0 + \sum_{v \neq s, t} d_v \cdot 0 + \sum_e c_e x_e$
 s.t. $d_s = 0 \leftarrow F_s$ unconstrained $\forall v \neq s, t$
 $d_t = 1 \leftarrow F_t$ constrained $\forall v \neq s, t$
 $d_u - d_v + x_e \geq 0 \leftarrow f_e \geq 0 \quad \forall v \neq s, t$
 $= d_v \leq d_u + x_e$

cor: strong duality \Rightarrow max flow = min cut

rmk: - max flow = min cut crucial to flow also \Rightarrow strong duality crucial to LP duality

- FF required (integral) captures to prove, strong duality provides proof

today: linear programming - duality

next lecture = LP

- example - by hand
- def - mechanized
- weak duality = canonical
- dual of max flow = min cut
- feasible
- bounded
- arbitrary captures

- prop \Rightarrow dual
 - \Rightarrow dual
 - \Rightarrow dual