

Max-flow via LP:

①

Variables f_e for every $e \in E$

objective function: $\max \sum_{\vec{sz} \in E} f_{sz}$

constraints:

for every $e \in E$, $0 \leq f_e \leq c_e$

for every $v \in V - \{s, t\}$, $\sum_{\vec{wr} \in E} f_{wr} = \sum_{\vec{vz} \in E} f_{vz}$

Standard form for LPs:

Variables $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$ m constraints $A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix}$

objective $c = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$

$b = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$

Max $c^T x$

s.t. $Ax \leq b$

$x \geq 0$

LP example from last class:

$$\max x_1 + 6x_2 + 10x_3$$

$$\text{s.t. } x_1 \leq 300 \quad \textcircled{1}$$

$$x_2 \leq 200 \quad \textcircled{2}$$

$$x_1 + 3x_2 + 2x_3 \leq 1000 \quad \textcircled{3}$$

$$x_2 + 3x_3 \leq 500 \quad \textcircled{4}$$

$$x_1, x_2, x_3 \geq 0 \quad \textcircled{5a, 5b, 5c}$$

Optimal is at $(200, 200, 100)$ which has profit = 2400

How do we know it's optimal?

$$\text{Let } y = (y_1, y_2, y_3, y_4) = (0, \frac{1}{3}, 1, \frac{8}{3})$$

$$\text{Then } y_1 x_1 + y_2 x_2 + y_3 x_3 + y_4 x_4$$

$$\iff x_1 + 6x_2 + 10x_3 \leq 2400$$

$$\text{So profit} \leq 2400. \checkmark$$

How do we find this y ?

$$y_1 \times \textcircled{1} + y_2 \times \textcircled{2} + y_3 \times \textcircled{3} + y_4 \times \textcircled{4}$$

$$\Leftrightarrow x_1 y_1 + x_2 y_2 + x_1 y_3 + 3x_2 y_3 + 2x_3 y_3 + x_2 y_4 + 3x_3 y_4 \leq 300 y_1 + 200 y_2 + 1000 y_3 + 500 y_4$$

$$\Leftrightarrow x_1(y_1 + y_3) + x_2(y_2 + 3y_3 + y_4) + x_3(2y_3 + 3y_4) \leq 300 y_1 + 200 y_2 + 1000 y_3 + 500 y_4$$

goal: this is $\geq x_1 + 6x_2 + 10x_3$

goal: minimize this quantity

which means

$$\begin{aligned} y_1 + y_3 &\geq 1 \\ y_2 + 3y_3 + y_4 &\geq 6 \\ 2y_3 + 3y_4 &\geq 10 \end{aligned}$$

Dual LP:

$$\min 300 y_1 + 200 y_2 + 1000 y_3 + 500 y_4$$

$$\begin{aligned} \text{s.t. } y_1 + y_3 &\geq 1 \\ y_2 + 3y_3 + y_4 &\geq 6 \\ 2y_3 + 3y_4 &\geq 10 \\ y_1, y_2, y_3, y_4 &\geq 0 \end{aligned}$$

Recall standard form for LPs:

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

Earlier example:

$$c = \begin{pmatrix} 1 \\ 6 \\ 10 \end{pmatrix}$$

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

$$b = \begin{pmatrix} 300 \\ 200 \\ 1000 \\ 500 \end{pmatrix}$$

$$\text{Variables } \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

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

$$\begin{aligned} \text{Dual LP:} \\ & \min b^T y \\ & \text{s.t. } A^T y \geq c \\ & \quad y \geq 0 \end{aligned}$$

$$\text{Variables } y = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

one variable per constraint

n variables \rightarrow m variables
 m constraints \rightarrow n constraints

Primal LP:

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

Dual LP:

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

$$\begin{array}{ccc} \uparrow \text{standard form} & & \downarrow \text{standard form} \\ \min \quad -c^T x & \xrightarrow{\text{dual}} & \max \quad -b^T y \\ \text{s.t.} \quad -Ax \geq -b & & \text{s.t.} \quad -A^T y \leq -c \\ & & y \geq 0 \\ & & x \geq 0 \end{array}$$

So, $\text{dual}(\text{dual}) = \text{Primal}$.

For a LP, optimum is achieved at a vertex of the feasible region except if:

a) it's infeasible

(example: $x_1 \leq 100$ & $x_1 \geq 200$
So feasible region is empty)

b) it's unbounded

(example: $\max x_1 + x_2$
s.t. $x_1, x_2 \geq 0$)

Weak Duality Theorem:

(6)

For a LP whose primal & dual are feasible,
Then their optimum x^* & y^* satisfy:

$$C^T x^* \leq b^T y^*$$

Consequence:

if we have a feasible x of primal LP
& a feasible y of dual LP

and $C^T x = b^T y$

then x & y are optimal solutions.

Strong Duality Theorem:

If primal LP has optimal x^* & dual LP has optimal y^*

then: $C^T x^* = b^T y^*$

If primal LP is unbounded then dual LP is infeasible.

If dual LP is unbounded then primal LP is infeasible.

(could be that both primal & dual LPs are infeasible.)

Another example:

(7)

Line fitting:

Given points $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

find the line $y = ax + b$ which minimizes the max distance.

Variables: a, b, e

Goal: minimize e where:

$$\text{for all } i, \quad |y_i - (ax_i + b)| \leq e$$

Y-coordinate
of i^{th} point

Y-coordinate of line
at x_i

Example: $(x_1, y_1) = (1, 1), (x_2, y_2) = (2, 1), (x_3, y_3) = (3, 2)$

LP: Variables a, b, e

min e

$$\text{s.t. } |1 - a - b| \leq e \rightarrow \begin{array}{l} 1 - a - b \leq e \quad \textcircled{1} \\ 1 - a - b \geq -e \quad \textcircled{2} \end{array}$$

$$|1 - 2a - b| \leq e \rightarrow \begin{array}{l} 1 - 2a - b \leq e \quad \textcircled{3} \\ 1 - 2a - b \geq -e \quad \textcircled{4} \end{array}$$

$$|2 - 3a - b| \leq e \rightarrow \begin{array}{l} 2 - 3a - b \leq e \quad \textcircled{5} \\ 2 - 3a - b \geq -e \quad \textcircled{6} \end{array}$$

In standard form:

$$\max -e$$

$$\text{s.t. } -a - b - e \leq -1$$

$$a + b - e \leq 1$$

$$-2a - b - e \leq -1$$

$$2a + b - e \leq 1$$

$$-3a - b - e \leq -2$$

$$3a + b - e \leq 2$$

~~but~~ but how do we get nonnegative constraints?

make variables $a^+, a^-, b^+, b^-, e^+, e^-$

replace $a = a^+ - a^-$
 $b = b^+ - b^-$
 $e = e^+ - e^-$

New LP: $\max -e^+ + e^-$

$$\text{s.t. } -a^+ + a^- - b^+ - b^- - e^+ + e^- \leq -1$$
$$a^+ - a^- + b^+ - b^- - e^+ + e^- \leq 1$$

$$a^+, a^-, b^+, b^-, e^+, e^- \geq 0$$