

LP Duality

Diet Problem

Consumer has foods of type $j=1, \dots, n$ available on market at prices c_j per unit,

and nutritional requirements for nutrients $i=1, \dots, m$ (vitamins, minerals, etc.) b_i units required daily

a_{ij} = amount of i th nutrient present in one unit of j th food type

Solve:

$$\begin{aligned} \min \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.} \quad & \sum_{j=1}^n a_{ij} x_j \geq b_i \quad \forall i \\ & x_j \geq 0 \quad \forall j \end{aligned}$$

$\min \quad c^T x$
 $\text{s.t.} \quad Ax \geq b$
 $x \geq 0$

Find purchase amount x_j to minimize cost and meet nutritional requirement.

Related problem for producer of nutrient pills.

Produce pills of types $i=1, \dots, m$

Set unit prices y_i for pills so that consumer buys pills

$$y_i \geq 0, \quad i=1, \dots, m$$

Consider j th food type

- Consumer has 2 options:
- ① purchase 1 unit at cost c_j giving $\begin{bmatrix} a_{1j} \\ \vdots \\ a_{mj} \end{bmatrix}$ nutr
 - ② purchase nutrient for producer at cost $\sum_{i=1}^m y_i a_{ij} + \dots + y_m a_{mj}$

To make pill prices attractive, make option ② preferable to ①:

$$\text{i.e. } \sum_{i=1}^m y_i a_{ij} \leq c_j$$

Do this $\forall j$ to limit purchase to pills only.

So, in order to maximize profit from pill sales to consumer, producer

$$\begin{aligned} \text{solver} \quad \max \quad & \sum_i b_i y_i \\ \text{s.t.} \quad & \sum_{i=1}^m y_i a_{ij} \leq c_j, \quad 1 \leq j \leq n \\ & y_i \geq 0, \quad 1 \leq i \leq m. \end{aligned}$$

$$\text{i.e.} \quad \begin{aligned} \max \quad & b^T y \\ & A^T y \leq c \\ & y \geq 0 \end{aligned}$$

Primal duality correspondence:

<p style="text-align: center;">max</p> <p>rhs obj fun A</p> <p>$\left\{ \begin{array}{l} \text{eq.} \\ \leq \\ \geq \end{array} \right\}$ const. const.</p> <p>varbls $\left\{ \begin{array}{l} \text{unres} \\ \geq 0 \\ \leq 0 \end{array} \right\}$</p>	<p style="text-align: center;">min</p> <p>obj fun rhs A^T</p> <p>$\left\{ \begin{array}{l} \text{unres.} \\ \geq 0 \\ \leq 0 \end{array} \right\}$ varbls variable</p> <p>$\left\{ \begin{array}{l} \text{unres} \\ \geq \\ \leq \end{array} \right\}$ const.</p>
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eg: Prual of $\min c^T x \quad (P) \quad Ax = b \quad x \geq 0$ is $\max b^T y \quad (D) \quad A^T y \leq c$

Weak duality theorem

~~If x is feasible in the primal then~~

if x is primal feasible and y is dual feasible then

$$c^T x \geq b^T y.$$

Proof $b^T y = (Ax)^T y = x^T (Ay) \leq c^T x$ //

$\downarrow \geq 0$ $\downarrow \leq c$

Theorem
Sufficient Optimality Criterion:

if \bar{x} is (P)-feas and \bar{y} is (D)-feas and $c^T \bar{x} = b^T \bar{y}$ then
 \bar{x} is (P)-opt and \bar{y} is (D)-optimal.

Proof For any (P)-feas x , we have $c^T x \geq b^T \bar{y} = c^T \bar{x}$ by hypothesis
~~Then $c^T \bar{x}$:~~ //

Provides an easy check of optimality:

If someone gives you a \bar{x} and a \bar{y} and claims they are optimal for (P) and (D) respectively, you can check

- i) \bar{x} is (P)-feasible
- ii) \bar{y} is (D)-feasible
- iii) $c^T \bar{x} = b^T \bar{y}$.

Strong Duality Theorem

The following are mutually exclusive and exhaustive possibilities

for (P) and (D):

- (I) (P) and (D) infeas
- (II) (D) infeas, (P) feasible (& unbd'd in obj. value)
- (III) (P) infeas, (D) feas (& unbd'd in obj. value)
- (IV) (P) & (D) feas and $\max = \min$.

Proof: If optimal value of (P) is finite: let $\bar{x} = B^{-1}b$ be optimal soln. (Prove strong duality)

$$\bar{y} = c_B B^{-1}. \quad \text{Assume } A \text{ has full rank}$$

$$\bar{r} = c_{B^c} - c_B B^{-1} A \geq 0, \text{ so } \bar{y} \text{ is dual feasible.}$$

$$\text{Also, } b^T \bar{y} = c_B^T B^{-1} b = c^T \bar{x}.$$



Complementary Slackness

A pair of primal and dual feasible solutions are optimal to the respective problems in a primal-dual pair of LPs, iff, whenever these feasible solutions make a slack variable in one problem strictly positive, the value of the associated non-negative variable of the other problem is zero.

Proof

we prove for

$$\begin{aligned} \max \quad & c^T x \\ \text{subject to} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

$$\begin{aligned} \max \quad & b^T y \\ \text{subject to} \quad & A^T y \leq c \end{aligned}$$

Test for optimality

Then states:

$$\hat{x}, \hat{y} \text{ optimal} \iff (\hat{x}_i > 0 \iff a_i^T \hat{y} = c_i)$$

i) $\hat{x}_i > 0 \implies a_i^T \hat{y} = c_i$

and ii) $a_i^T \hat{y} < c_i \implies \hat{x}_i = 0$.

ie \hat{x}, \hat{y} optimal $\iff \hat{x}_i (c_i - a_i^T \hat{y}) = 0$

$$\implies c_i - a_i^T \hat{y} \geq 0, \hat{x}_i \geq 0 \text{ by feasibility.}$$

Also, ~~$\hat{x}^T (c - A^T \hat{y}) = 0$~~
 $\hat{x}_i (c_i - a_i^T \hat{y}) \geq 0 \quad \forall i$

$$\begin{aligned} 0 \leq \sum_{i=1}^n \hat{x}_i (c_i - a_i^T \hat{y}) &= \hat{x}^T (c - A^T \hat{y}) = c^T \hat{x} - (A\hat{x})^T \hat{y} \\ &= c^T \hat{x} - b^T \hat{y} = 0 \text{ since } \hat{x}, \hat{y} \text{ optimal pair} \end{aligned}$$

$$\Leftarrow : \hat{x}_i (c_i - a_i^T \hat{y}) = 0 \quad \forall i$$

$$\therefore \hat{x}^T (c - A^T \hat{y}) = 0$$

$$\therefore c^T \hat{x} = (A \hat{x})^T \hat{y} = b^T \hat{y} \quad //$$

Farkas' Lemma

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

Exactly one of the following systems is feasible:

$$(I) : \begin{aligned} Ax &= b \\ x &\geq 0 \end{aligned}$$

$$(II) : \begin{aligned} A^T y &\leq 0 \\ b^T y &> 0 \end{aligned}$$

Proof Consider the primal LP

$$\begin{aligned} \max \quad & 0 \cdot x \\ & Ax = b \\ & x \geq 0 \end{aligned} \quad (\hat{P})$$

Its dual is

$$\begin{aligned} \max \quad & b^T y \\ & A^T y \leq 0 \end{aligned} \quad (\hat{D})$$

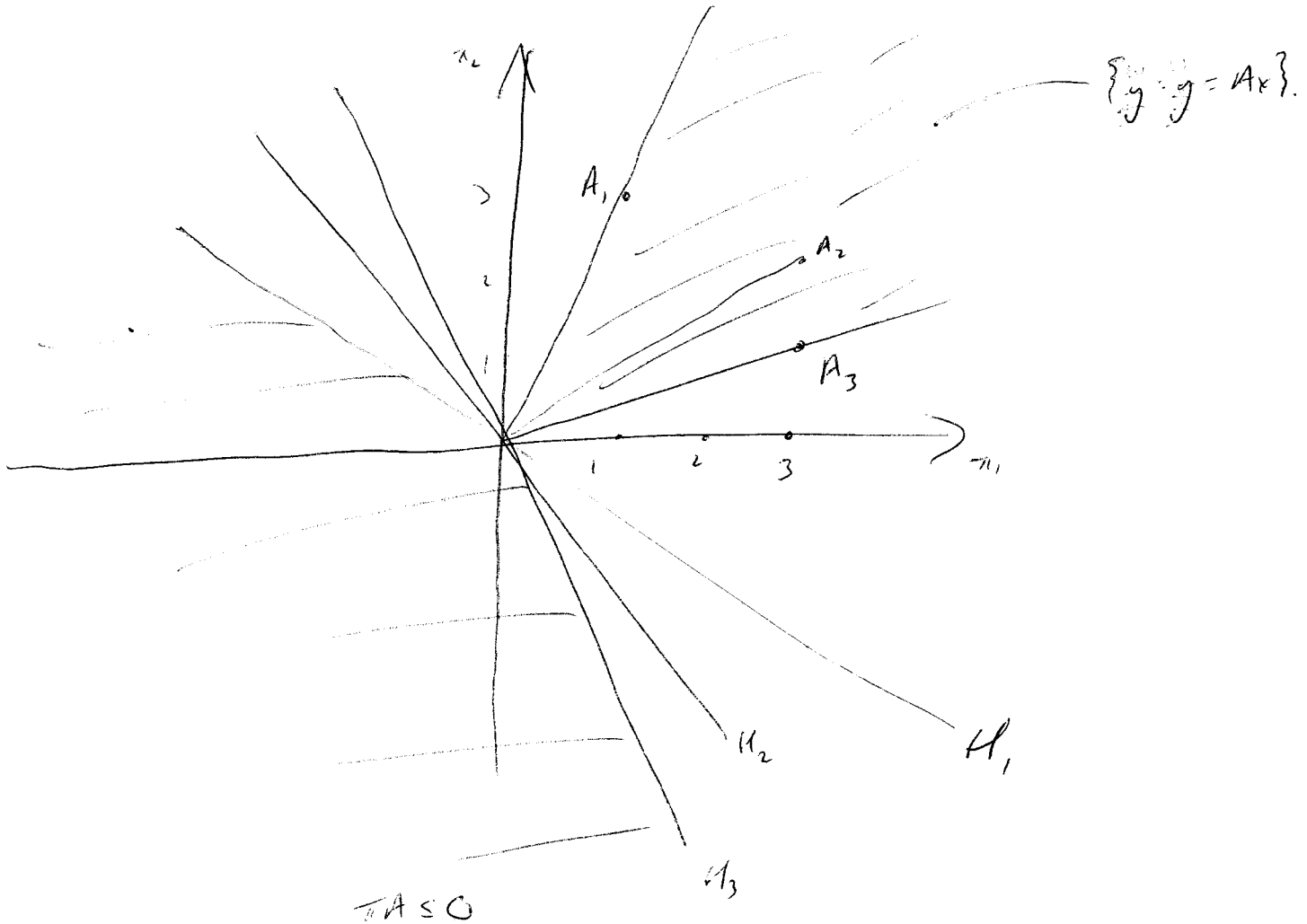
Assume (\hat{P}) has a feasible solution, it has optimal value 0.

optimal value of (\hat{D}) is $\leq 0 \Rightarrow$ if $A^T y \leq 0$ then $b^T y \leq 0$.

Now also, (\hat{D}) always feasible (take $y = 0$) //

Geometrical illustration:

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



If $b \in \text{col}(A, A_1, A_2, A_3)$ then
 b can be written as $y = Ax$
 (\Rightarrow)
 every row of A is a linear combination of the other two rows.