

MATH 312
LECTURE 23

: Linear Programming

Example: You have 1\$, and you have to invest it on assets A and B, satisfying:

- 1) You invest in A at least ~~twice~~ double than in B.
- 2) At least $1/2$ \$ is invested.

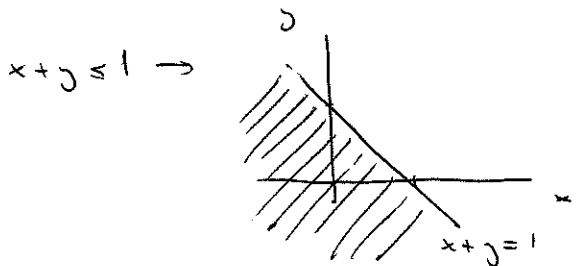
We can write this constraints as inequalities:

Denote $x =$ invested on A, then
 $y =$ " on B

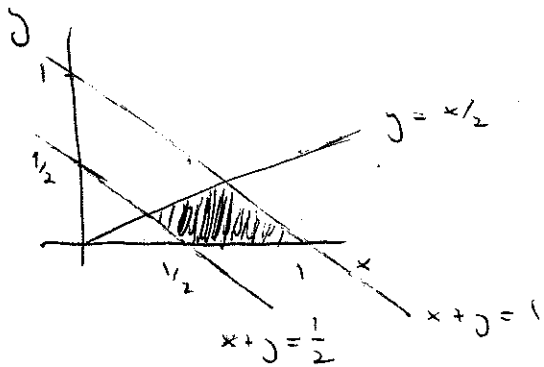
$$\left. \begin{array}{l} x + y \leq 1 \quad (\text{you only have } 1\$) \\ x \geq 2y \quad (\text{at least double in A than B}) \\ x + y \geq \frac{1}{2} \quad (\text{at least } \frac{1}{2} \$ \text{ invested}) \end{array} \right\} \text{and } \begin{cases} x \geq 0 \\ y \geq 0 \end{cases}$$

Q: What values are x, y allowed to take?

Note that each inequality represent a half plane:



Drawing all the constraints gives:



This is the feasible region.

We usually denote all the constraints in matrix form:


$$A\vec{x} \leq \vec{b}$$

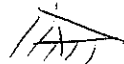
Q: What for $x + y \geq \frac{1}{2}$? $\rightarrow -x - y \leq -\frac{1}{2}$.

$$\text{So, } \begin{cases} x + y \leq 1 \\ -x + 2y \leq 0 \\ -x - y \leq -\frac{1}{2} \end{cases} \rightarrow \begin{bmatrix} 1 & 1 \\ -1 & 2 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \leq \begin{bmatrix} 1 \\ 0 \\ -\frac{1}{2} \end{bmatrix} \quad (\text{and } x \geq 0, y \geq 0)$$

Def: Feasible region

Intersection of the half-spaces arising from the inequalities.

Remark: It can be bounded: $x + y \leq 1, x \geq 0, y \geq 0$ 

• Unbounded: $x + y \leq 1$ 

• empty \emptyset : $x + y \geq 1, x + y \leq 0$.

Def. A linear Programming problem asks to maximize a linear profit $\vec{c}^T \vec{x}$ subject to linear constraints $A\vec{x} \leq \vec{b}$ and $\vec{x} \geq 0$, where A, \vec{c}, \vec{b} are data.

- In our example: Say that A has a twofold return while threefold in B.

Then, for an investment of x in A, y in B, we get back

$$2x + 3y = [2 \ 3] \begin{bmatrix} x \\ y \end{bmatrix} \rightarrow \vec{c}^T = [2 \ 3]$$

- So the lin. programming problem would be:

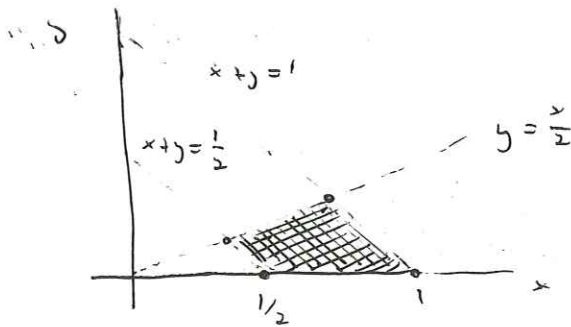
$$\left. \begin{array}{l} \max. \quad \vec{c}^T \vec{x} \\ \text{s.t.} \quad A\vec{x} \leq \vec{b} \\ \quad \quad \vec{x} \geq 0 \end{array} \right\} \text{where} \quad \vec{c} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 1 \\ -1 & 2 \\ -1 & -1 \end{bmatrix}$$

$$\vec{b} = \begin{bmatrix} 1 \\ 0 \\ 1/2 \end{bmatrix}, \quad \vec{x} = \begin{bmatrix} x \\ y \end{bmatrix} \text{ unknowns.}$$

~~Problem~~ \uparrow Standard form

- How to solve it? \rightarrow Geometric idea (we'll follow our example).

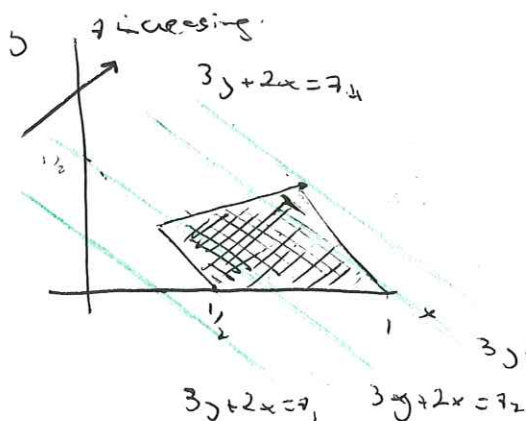
1) Draw the feasible region:



All possible solutions are in that region.

2) Maximize the profit: $\vec{z}^T \vec{x} = [2 \ 3] \begin{bmatrix} x \\ y \end{bmatrix} = 2x + 3y$

↳ Pick a profit z so the $z = 2x + 3y$ describes a line; each combination of x, y on that line gives same profit.



$$3y + 2x = z \rightarrow y = \frac{z}{3} - \frac{2}{3}x$$

$$\bullet z = 0 \rightarrow y = -\frac{2}{3}x$$

$$\bullet z = 1 \rightarrow y = \frac{1}{3} - \frac{2}{3}x$$

⋮

We "move" the line that gives the profit until we leave the feasible region. We can see that the maximum will always be obtained at a corner (*).

(*) Assuming the feasible region is bounded and that there is a unique solution.

3) Evaluate the profit function $\vec{c}^T \vec{x}$ at all corners and pick the maximum.

In our example,

$$\begin{aligned} (x = \frac{1}{2}, y = 0), & \quad x + y = \frac{1}{2} \\ (x = 1, y = 0), & \quad 2y - x = 0 \end{aligned} \left\{ \rightarrow (y = \frac{1}{6}, x = \frac{1}{3}) \right.$$

$$\begin{aligned} x + y = 1 \\ 2y - x = 0 \end{aligned} \left\{ \rightarrow (y = \frac{1}{3}, x = \frac{2}{3}) \right.$$

$$\hookrightarrow (\frac{1}{2}, 0) \rightarrow 2 \cdot \frac{1}{2} + 3 \cdot 0 = 1$$

$$(1, 0) \rightarrow 2$$

$$(\frac{1}{3}, \frac{1}{6}) \rightarrow 2 \cdot \frac{1}{3} + 3 \cdot \frac{1}{6} = \frac{7}{6}$$

$$(\frac{2}{3}, \frac{1}{3}) \rightarrow 2 \cdot \frac{2}{3} + 3 \cdot \frac{1}{3} = \frac{4}{3} + 1 = \frac{7}{3}$$

$$\left. \begin{aligned} & \rightarrow \text{maximum is } \frac{7}{3}, \text{ at} \\ & \text{point } x = \frac{2}{3}, y = \frac{1}{3} \end{aligned} \right\|$$

(We "saw" it on the picture).

Remark: In higher dimensions, finding all the corners and evaluating the profit is not an efficient method.

But the moral is true: the optimum is reached at a corner.

Dual problem

For every linear programming problem, there is a closely related linear programming problem, called its dual.

If the "primal problem" is given in standard form:

$$\left. \begin{array}{l} \max \quad \vec{c}^T \vec{x} \\ \text{s.t.} \quad A \vec{x} \leq \vec{b} \\ \vec{x} \geq \vec{0} \end{array} \right\} \text{its dual is} \left. \begin{array}{l} \min \quad \vec{b}^T \vec{p} \\ \text{s.t.} \quad A^T \vec{p} \geq \vec{c} \\ \vec{p} \geq \vec{0} \end{array} \right\}$$

Theorem: Weak duality

Given any solutions \vec{x}_* to primal and \vec{p}_* to dual, it holds that

$$\vec{c}^T \vec{x}_* \leq \vec{b}^T \vec{p}_*$$

$$\left. \begin{array}{l} \text{Proof } \vec{x}_* \text{ solution} \Rightarrow \vec{x}_* \text{ feasible} \Rightarrow A \vec{x}_* \leq \vec{b} \\ \vec{p}_* \text{ to dual} \Rightarrow \vec{p}_* \text{ feasible in dual} \Rightarrow A^T \vec{p}_* \geq \vec{c} \end{array} \right\}$$

$$\text{Also, } \vec{p}_* \geq \vec{0}, \text{ so } \vec{p}_*^T A \vec{x}_* \leq \vec{p}_*^T \vec{b} \Rightarrow$$

$$\Rightarrow \vec{x}_*^T A^T \vec{p}_* \leq \vec{b}^T \vec{p}_* \Rightarrow \vec{x}_*^T \vec{c} \leq \vec{x}_*^T A^T \vec{p}_* \leq \vec{b}^T \vec{p}_* \Rightarrow$$

$$\Rightarrow \vec{c}^T \vec{x}_* \leq \vec{b}^T \vec{p}_* //$$

Theorem: Strong duality.

If \vec{x}_* optimum solution to primal, with $\vec{c}^T \vec{x}_*$ finite, then,

$$\vec{c}^T \vec{x}_* = \vec{b}^T \vec{p}_* .$$

• Example:

$$\left. \begin{array}{l} \max 3x + 2y \\ \text{s.t. } x + y \leq 5 \\ x, y \geq 0 \end{array} \right\} \xrightarrow{\text{dual}} \left. \begin{array}{l} \min 5p \\ \text{s.t. } p \geq 3 \\ \quad \quad p \geq 2 \\ p \geq 0 \end{array} \right\}$$

Trivially, solution to dual is $p_* = 3$, $\vec{b}^T \vec{p}_* = 15$, so

we have that $\vec{c}^T \vec{x}_* = 15$.

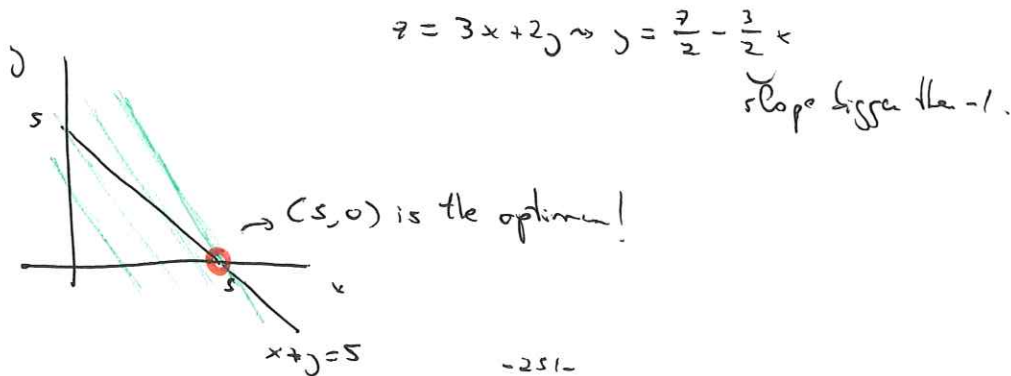
That is, the optimal solution to primal satisfies $3x + 2y = 15$.

Thus, $x = 5 - \frac{2}{3}y$

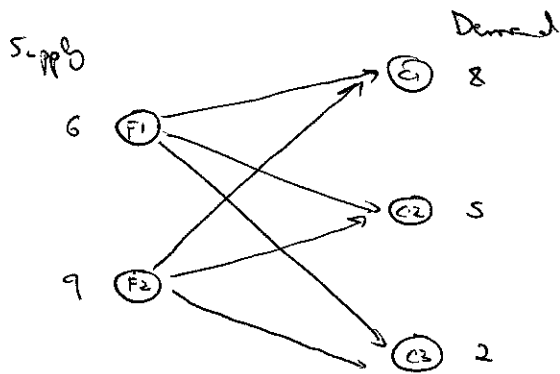
$$\hookrightarrow x + y \leq 5 \rightarrow 5 - \frac{2}{3}y + y \leq 5 \rightarrow y \leq 0 \Rightarrow y = 0 \Rightarrow x = 5 .$$

Optim $\begin{array}{l} x_* = 5 \\ y_* = 0 \end{array} \parallel$

→ Graphically without dual:



Example: Write the primal standard form and its dual for the following transport problem:



Costs of transport

	C1	C2	C3	} c_{ij}
F1	5	5	3	
F2	6	4	1	

Goal: Minimize cost of satisfying all the demand.

Define x_{ij} = product transported from factory i to center j ($i=1,2$; $j=1,2,3$).

Then, we want to

$$\min \sum_{i,j} c_{ij} x_{ij} = 5x_{11} + 5x_{12} + 3x_{13} + 6x_{21} + 4x_{22} + x_{23}$$

Constraints:

$$\left. \begin{aligned} x_{11} + x_{21} &= 8 \\ x_{12} + x_{22} &= 5 \\ x_{13} + x_{23} &= 2 \end{aligned} \right\} \text{satisfying the demand.}$$

$$\left. \begin{aligned} x_{11} + x_{12} + x_{13} &\leq 6 \\ x_{21} + x_{22} + x_{23} &\leq 9 \end{aligned} \right\} \text{the production of each factory is limited,}$$

~~Constraints:~~

$$x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23} \geq 0$$

This is not in standard form: we have min instead of max, and some = instead of \leq .

$$D \quad \min \vec{c}^T \vec{x} = -\max (-\vec{c}^T \vec{x})$$

$$\Rightarrow \text{Ex: } x_{11} + x_{21} = 8 \rightarrow \text{equivalent to } \left. \begin{array}{l} x_{11} + x_{21} \leq 8 \\ x_{11} + x_{21} \geq 8 \end{array} \right\}$$

and $x_{11} + x_{21} \geq 8$ is same as $-x_{11} - x_{21} \leq -8$.

$$\text{So, replace } x_{11} + x_{21} = 8 \text{ by } \left. \begin{array}{l} x_{11} + x_{21} \leq 8 \\ -x_{11} - x_{21} \leq -8 \end{array} \right\}.$$

By doing so, the problem in standard form is

$$\begin{array}{l} -\max (-\vec{c}^T \vec{x}) \\ \text{s.t. } A\vec{x} \leq \vec{b} \\ \vec{x} \geq \vec{0} \end{array} \quad \text{with } \vec{c}^T = [5 \quad 5 \quad 3 \quad 6 \quad 4 \quad 1] \\ A = \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 & -1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & -1 & 0 & 0 & -1 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \end{bmatrix}, \quad \vec{b} = \begin{bmatrix} 8 \\ -8 \\ 5 \\ -5 \\ 2 \\ -2 \\ 6 \\ 9 \end{bmatrix}$$

Then, its dual is

$$\begin{cases} \min \vec{b}^T \vec{p} \\ \text{s.t. } A^T \vec{p} \geq \vec{c} \\ \vec{p} \geq \vec{0} \end{cases}$$

Introduction to Simplex Method

$$\begin{cases} \max \vec{c}^T \vec{x} \\ \text{s.t. } A\vec{x} \leq \vec{b} \\ \vec{x} \geq \vec{0} \end{cases}$$

We'll follow a particular example:

$$\begin{cases} \max 3x - 4y \\ \text{s.t. } x + y \leq 10 \\ 3x + 2y \leq 6 \\ x, y \geq 0 \end{cases}$$

1) Step 1): Write the problem in standard form.

2) Step 2): Add slack variable so that inequalities become equalities.

$$\begin{cases} \max 3x - 4y \\ \text{s.t. } x + y + s_1 = 10 \\ 3x + 2y + s_2 = 6 \\ x, y, s_1, s_2 \geq 0 \end{cases}$$

3) Create the following augmented matrix:

$$\left[\begin{array}{ccc|c} 1 & -\vec{c}^T & 0 & 0 \\ 0 & A & I_n & \vec{b} \end{array} \right] = B \quad \rightsquigarrow \begin{cases} z = \vec{c}^T \vec{x} \\ A\vec{x} + \vec{s}I = \vec{b} \end{cases}$$

In the ex,

$$B = \begin{array}{c|ccccc|c} & x & y & s_1 & s_2 & \\ \hline 1 & -3 & 4 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 10 \\ 0 & 3 & 2 & 0 & 1 & 6 \end{array}$$

What we are doing is: fix x, y to zero, and slack variables to 0.
(they are pivots)

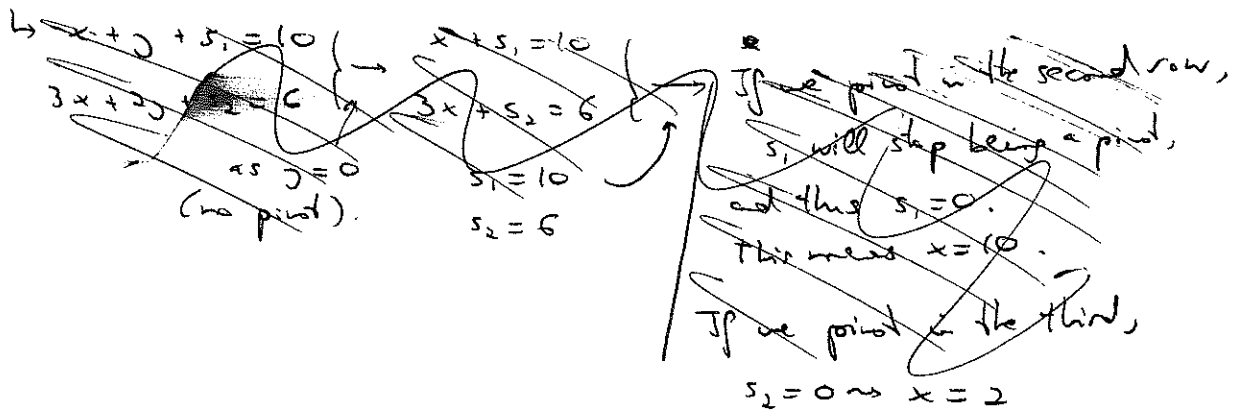
Remark: If we start with the standard form, we will always have this structure!

4) Pivoting: Look at the first row.

If one entry is negative, it means that by increasing the corresponding variable we increase the profit Z .

→ So we want x to be a new pivot.

→ Q: Should we choose 2nd or 3rd row?



Notice that the second equation $x + y + s_1 = 10$ tell us that $x \leq 10$,

while the third $3x + 2y + s_2 = 6 \Rightarrow x \leq 2$.

So the third is more restrictive.

→ We always choose the row with minimum quotient b_i/a_{ij}
(when introducing pivot in col j).

So,

$$\left[\begin{array}{cccc|c} 1 & -3 & 4 & 0 & 0 & 10 \\ 0 & 1 & 1 & 1 & 0 & 6 \\ 0 & 3 & 2 & 0 & 1 & 6 \end{array} \right] \rightarrow \left[\begin{array}{cccc|c} 1 & 0 & 6 & 0 & 1 & 5 \\ 0 & 1 & 1/3 & 1 & -1/3 & 8 \\ 0 & 1 & 2/3 & 0 & 1/3 & 2 \end{array} \right]$$

Now we have, $x = 2, s_1 = 8, y = 0 = s_2$ // optimum! (first row of $B \geq 0$).
 $z = 6 \cdot 0 - 6y - s_2 = 6$ //

↑ ↑
this tell us any modification will not be \rightarrow smaller.