

We continue the examples from last time.

$$\Leftrightarrow x_2 = -\frac{2}{3}x_1 + \frac{z}{9}$$

2. • minimize $z = 6x_1 + 9x_2$ (slope $-\frac{2}{3}$)

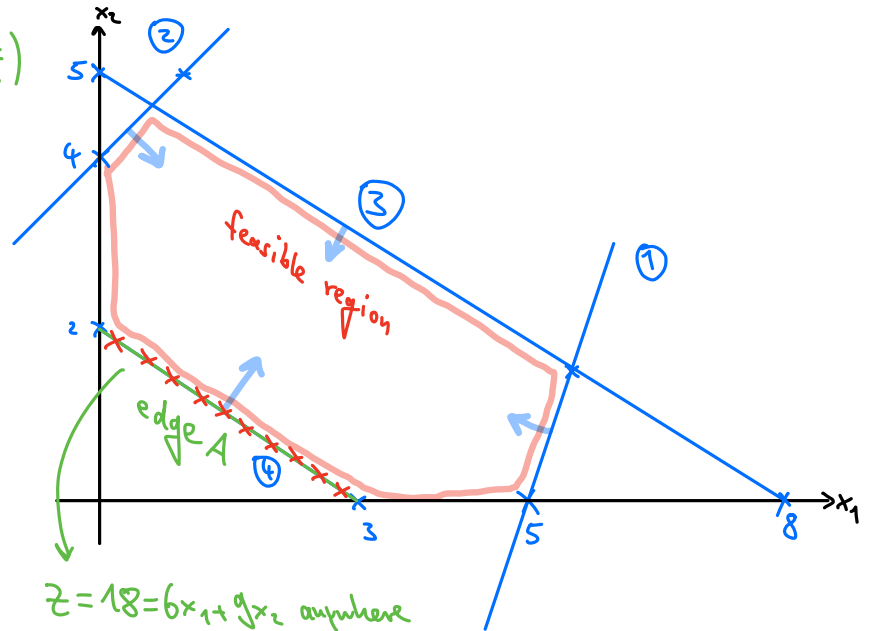
• constraints: $3x_1 - x_2 \leq 15$ ①

$-x_1 + x_2 \leq 4$ ②

$5x_1 + 8x_2 \leq 40$ ③

$x_2 \geq -\frac{2}{3}x_1 + 2 \Leftrightarrow 2x_1 + 3x_2 \geq 6$ ④

$x_1, x_2 \geq 0$



$z = 18 = 6x_1 + 9x_2$ anywhere on edge A.

Here, the slopes of objective fct. and constraint 4 are the same.

\Rightarrow Any point on edge A is an optimal solution, i.e., there are infinitely many.

We call such problems "degenerate".

meaning infinitely many points on the bounded line segment between points $(0, 2)$ and $(3, 0)$ (i.e., edge A)

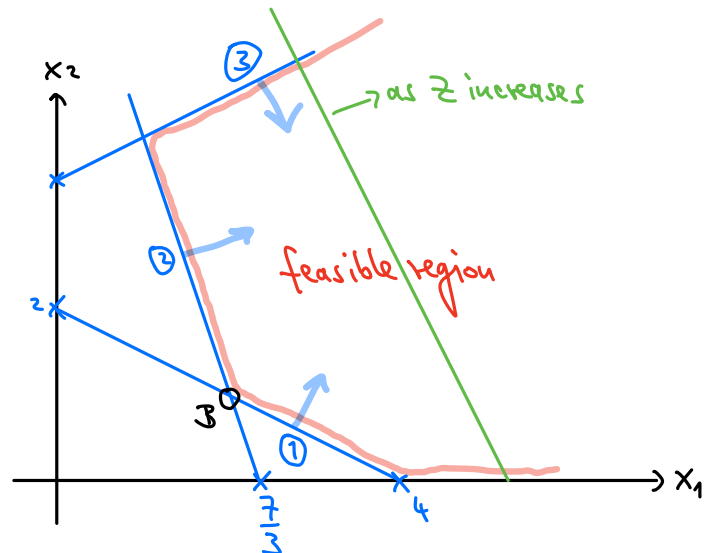
3. • maximize $z = 4x_1 + 2x_2$

• constraints: $x_1 + 2x_2 \geq 4$ ①

$3x_1 + x_2 \geq 7$ ②

$-x_1 + 2x_2 \leq 7$ ③

$x_1, x_2 \geq 0$



Here, feasible region is unbounded, and Z increases in unbounded direction.

There are (infinitely) many feasible solutions, but none of them is optimal.

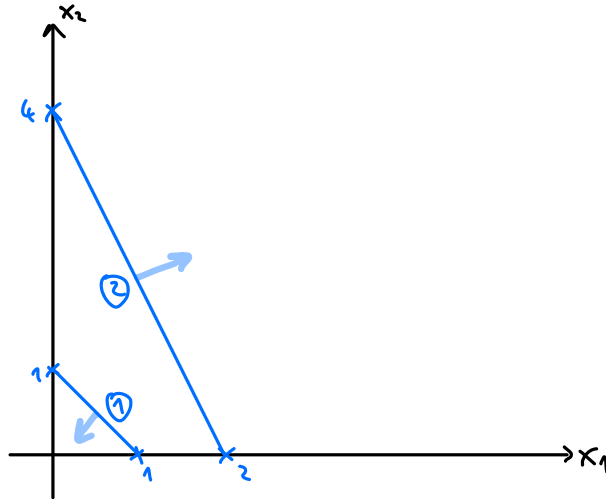
(Note: If Z would be minimized, the optimal solution would be at $B = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$, and $Z = 10$.)

4. • maximize $Z = 3x_1 + 4x_2$

• constraints: $x_1 + x_2 \leq 1$ ①

$2x_1 + x_2 \geq 4$ ②

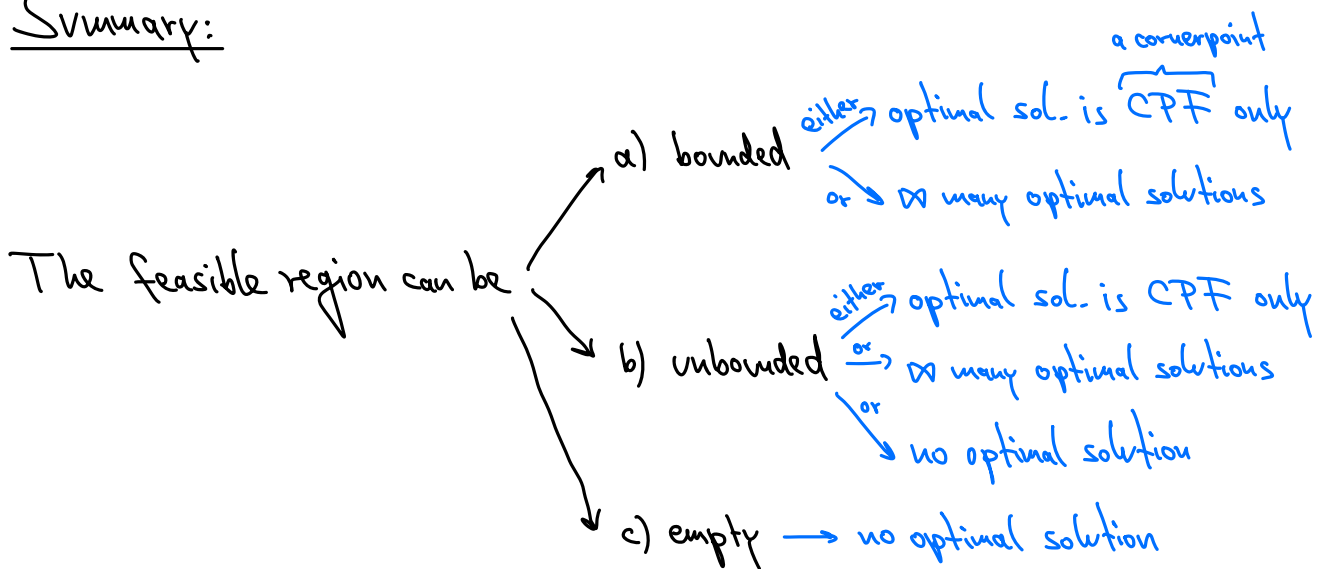
$x_1, x_2 \geq 0$



=> The feasible region is empty; there are no feasible solutions.

We call such problems "over-constrained".

Summary:



More generally, two prototypical examples (but mixtures are also possible) of linear Programming (LP) models are:

I) Activity analysis problem: (e.g., Wyndor)

- A = set of activities (or products)
- R = set of resources (or production facilities)
- w_{ij} = workload required from activity $i \in A$ on resource $j \in R$
- c_j = available capacity of resource $j \in R$
- p_i = profit from performing one unit of activity $i \in A$
- decision variables x_i : # of units of activity $i \in A$ to perform

LP problem: • maximize $Z = \sum_{i \in A} p_i x_i$ (total profit)

• constraints: $\sum_{i \in A} w_{ij} x_i \leq c_j$ for all $j \in R$, and $x_i \geq 0$ for all $i \in A$

II) Diet-type problem:

- F = set of foods
- N = set of nutrients
- c_i = unit cost of food $i \in F$
- r_j = minimum requirement for nutrient $j \in N$
- a_{ij} = amount of nutrient $j \in N$ from eating one unit of food $i \in F$
- decision variables x_i = # of units of food $i \in F$ to consume

LP problem: • minimize $Z = \sum_{i \in F} c_i x_i$ (total cost)

• constraint: $\sum_{i \in F} a_{ij} x_i \geq r_j$ for all $j \in N$, and $x_i \geq 0$ for all $i \in F$

2.2 Standard Form of LP Problems

Goal: Bring all LP problems into a standardized form. Then later we can easier develop a general algorithm to solve them.

Goal: Write LP problems in the following **standard form**: (note: some books might use other very similar standards)

- Minimize $z = c^T x$, with $c \in \mathbb{R}^m$, $x \in \mathbb{R}^m$
- Constraints: $Ax = b$, with A an $n \times m$ matrix, $b \in \mathbb{R}^n$
and $x \geq 0$ (meaning $x_j \geq 0$ for all $j = 1, \dots, m$)

Explanation of notation:

- $c = \begin{pmatrix} c_1 \\ \vdots \\ c_m \end{pmatrix}$, $x = \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix}$, $b = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix}$ are column vectors

$$c^T = (c_1, \dots, c_m) = c \text{ transpose} = \text{row vector}$$

$$\Rightarrow c^T x = (c_1, \dots, c_m) \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix} = \sum_{i=1}^m c_i x_i$$

↖ multiplication of a $(1 \times m)$ matrix with an $(m \times 1)$ matrix

- $A = \begin{pmatrix} A_{11} & \dots & A_{1m} \\ \vdots & \ddots & \vdots \\ A_{n1} & \dots & A_{nm} \end{pmatrix} = n \times m$ matrix, or $A \in \underbrace{\text{Mat}(n, m)}_{\text{set of } n \times m \text{ matrices}}$

$$\text{Recall: } Ax = \begin{pmatrix} A_{11} & \dots & A_{1m} \\ \vdots & \ddots & \vdots \\ A_{n1} & \dots & A_{nm} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix} = \begin{pmatrix} A_{11}x_1 + \dots + A_{1m}x_m \\ A_{21}x_1 + \dots + A_{2m}x_m \\ \vdots \\ A_{n1}x_1 + \dots + A_{nm}x_m \end{pmatrix}$$

$$\text{i.e., } (Ax)_i = \sum_{j=1}^m A_{ij} x_j$$

$\Rightarrow Ax = b$ means: $\sum_{j=1}^m A_{ij}x_j = b_i$ for all $i = 1, \dots, n$

Claim: Every LP problem can be written in standard form.

We illustrate this with the following example ("proof by example") next time:

• Maximize $z = x_1 + 2x_2 + 3x_3$

• Constraints: $x_1 + x_2 - x_3 = 1$

$$-2x_1 + x_2 + 2x_3 \geq -5$$

$$x_1 - x_2 \leq 4$$

$$x_2 + x_3 \leq 5$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$