

# Modelling and Optimization

INF170

#4: General LP

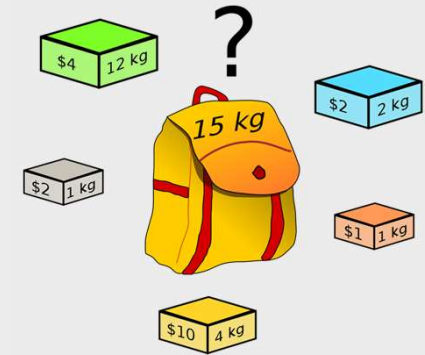
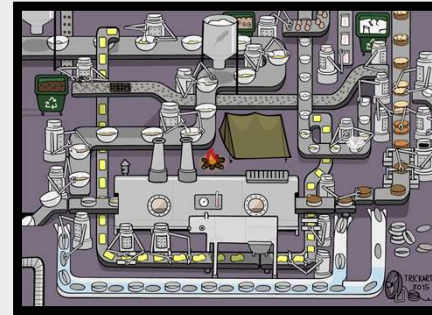
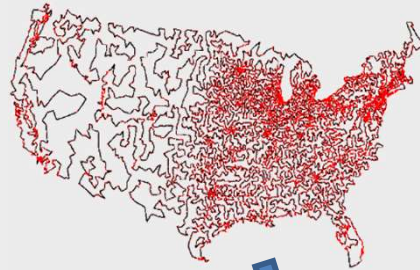
AHMAD HEMMATI

Optimization Group  
Dept. of Informatics  
University of Bergen

Fall Semester  
2018



# OPTIMIZATION MODEL



$$\begin{aligned} \min \quad & f(x) \\ \text{st} \quad & g(x) \leq b \\ & x \in X \end{aligned}$$

# MODEL STRUCTURE

- 1) Decision variables
- 2) Objective function
- 3) Constraints
- 4) Parameters
- 5) Sets

$$\begin{array}{ll}\min & f(x) \\st & g(x) \leq b \\ & x \in X\end{array}$$

# MODEL STRUCTURE

## 1) **Decision variables (or just variables)**

The unknown numerical value which is pursued to find through the resolution of the model.

Examples:

$x$  = Amount of crude oil to be processed this week

$y$  = Capacity of a new production plant

$z$  = "Yes" or "No" decisions

The decision maker has control on the value that the decision variables will take.

## 2) Objective function

Measure which allows to evaluate the performance of the different values that the variables can take.

Examples:

Minimize cost function  $C$

Maximize a profit function  $P$

Minimize carbon emissions  $E$

Maximize fairness  $F$

## 3) Constraints

Conditions that restrict the values of the decision variables. These conditions define relations ( $\leq$ ,  $=$ ,  $\geq$ ) the variables must satisfy.

Examples:

Total production  $\leq$  Capacity

Number of carbon emissions  $\leq$  Limit allowed

Output = Input \* Conversion factor

## 4) Parameters

The parameters are fixed values given beforehand in the problem (the decision maker has no control on them when solving the optimization problem).

Examples:

Capacity

Budget

Demand...

## 5) Sets

The sets define the domain of variables and parameters.

Examples:

Set of customers

Set of production plants

$\mathbb{R}$

$\{0,1\}$



# DIET MODEL

Variables:

$x_{BEEF}$  : the number of packages of beef to be purchased

...

Objective function:

$$\text{Minimize } 3.19 x_{BEEF} + 2.59 x_{CHK} + 2.29 x_{FISH} + 2.89 x_{HAM} + \\ 1.89 x_{MCH} + 1.99 x_{MTL} + 1.99 x_{SPG} + 2.49 x_{TUR}$$

Subject to (Constraints):

total percentage of vitamin “A” daily requirement met

$$60 x_{BEEF} + 8 x_{CHK} + 8 x_{FISH} + 40 x_{HAM} + \\ 15 x_{MCH} + 70 x_{MTL} + 25 x_{SPG} + 60 x_{TUR} \geq 700$$

...

# DIET MODEL

*Minimize*  $3.19 x_{beef} + 2.59 x_{chk} + 2.29 x_{fish} + 2.89 x_{ham} + 1.89 x_{mch} + 1.99 x_{mtl} + 1.99 x_{spg} + 2.49 x_{tur}$

*subject to A:*

$$60 x_{beef} + 8 x_{chk} + 8 x_{fish} + 40 x_{ham} + 15 x_{mch} + 70 x_{mtl} + 25 x_{spg} + 60 x_{tur} \geq 700$$

*subject to C:*

$$20 x_{beef} + 0 x_{chk} + 10 x_{fish} + 40 x_{ham} + 35 x_{mch} + 30 x_{mtl} + 50 x_{spg} + 20 x_{tur} \geq 700$$

*subject to B1:*

$$10 x_{beef} + 20 x_{chk} + 15 x_{fish} + 35 x_{ham} + 15 x_{mch} + 15 x_{mtl} + 25 x_{spg} + 15 x_{tur} \geq 700$$

*subject to B2:*

$$15 x_{beef} + 20 x_{chk} + 10 x_{fish} + 10 x_{ham} + 15 x_{mch} + 15 x_{mtl} + 15 x_{spg} + 10 x_{tur} \geq 700$$

$$x_{beef} \geq 0; x_{chk} \geq 0; x_{fish} \geq 0; x_{ham} \geq 0; x_{mch} \geq 0; x_{mtl} \geq 0; x_{spg} \geq 0; x_{tur} \geq 0;$$

# SUMMATION AND “FOR ...” NOTATION

## ➤ Summation notation

$$a_1x_1 + a_2x_2 + \dots + a_nx_n = \sum_{j=1}^n a_j x_j$$

## ➤ "for ..." notation

$$\left. \begin{array}{l} a_{1,1}x_1 + a_{1,2}x_2 \leq b_1 \\ a_{2,1}x_1 + a_{2,2}x_2 \leq b_2 \\ a_{3,1}x_1 + a_{3,2}x_2 \leq b_3 \end{array} \right\} \quad a_{i,1}x_1 + a_{i,2}x_2 \leq b_i \quad \text{for } i = 1, 2, 3$$

# DIET MODEL

## Sets:

$F$  = set of foods

$N$  = set of nutrients

## Parameters:

$a_{ij}$  = amount of nutrient  $j$  in food  $i$ ,  $\forall i \in F, \forall j \in N$

$c_i$  = cost per serving of food  $i$ ,  $\forall i \in F$

$Fmin_i$  = minimum number of required servings of food  $i$ ,  $\forall i \in F$

$Fmax_i$  = maximum allowable number of servings of food  $i$ ,  $\forall i \in F$

$Nmin_j$  = minimum required level of nutrient  $j$ ,  $\forall j \in N$

$Nmax_j$  = maximum allowable level of nutrient  $j$ ,  $\forall j \in N$

## Variables:

$x_i$  = number of servings of food  $i$  to purchase/consume,  $\forall i \in F$

## Objective Function:

Minimize the total cost of the food

# DIET MODEL

$$\text{Minimize } \sum_{i \in F} c_i x_i$$

#Constraint Set 1: For each nutrient  $j \in N$ , at least meet the minimum required level.

$$\sum_{i \in F} a_{ij} x_i \geq Nmin_j, \quad \forall j \in N$$

#Constraint Set 2: For each nutrient  $j \in N$ , do not exceed the maximum allowable level.

$$\sum_{i \in F} a_{ij} x_i \leq Nmax_j, \quad \forall j \in N$$

#Constraint Set 3: For each food  $i \in F$ , select at least the minimum required number of servings.

$$x_i \geq Fmin_i, \quad \forall i \in F$$

#Constraint Set 4: For each food  $i \in F$ , do not exceed the maximum allowable number of servings.

$$x_i \leq Fmax_i, \quad \forall i \in F$$

# MAXIMIZING PROFITS

```
var XB;  
var XC;  
maximize Profit: 25 * XB + 30 * XC;  
subject to Time: (1/200) * XB + (1/140) * XC <= 40;  
subject to B_limit: 0 <= XB <= 6000;  
subject to C_limit: 0 <= XC <= 4000;  
solve;  
display XB, XC;
```

# MAXIMIZING PROFITS

Set:

$P$ , a set of products

Parameters:

$a_j$  = tons per hour of product  $j$ , for each  $j \in P$

$b$  = hours available at the mill

$c_j$  = profit per ton of product  $j$ , for each  $j \in P$

$u_j$  = maximum tons of product  $j$ , for each  $j \in P$

variables:

$x_j$  = tons of product  $j$  to be made, for each  $j \in P$

Objective function:

$$\text{Maximize } \sum_{j \in P} c_j x_j$$

Subject to (Constraints):

$$\sum_{j \in P} (1/a_j) x_j \leq b$$

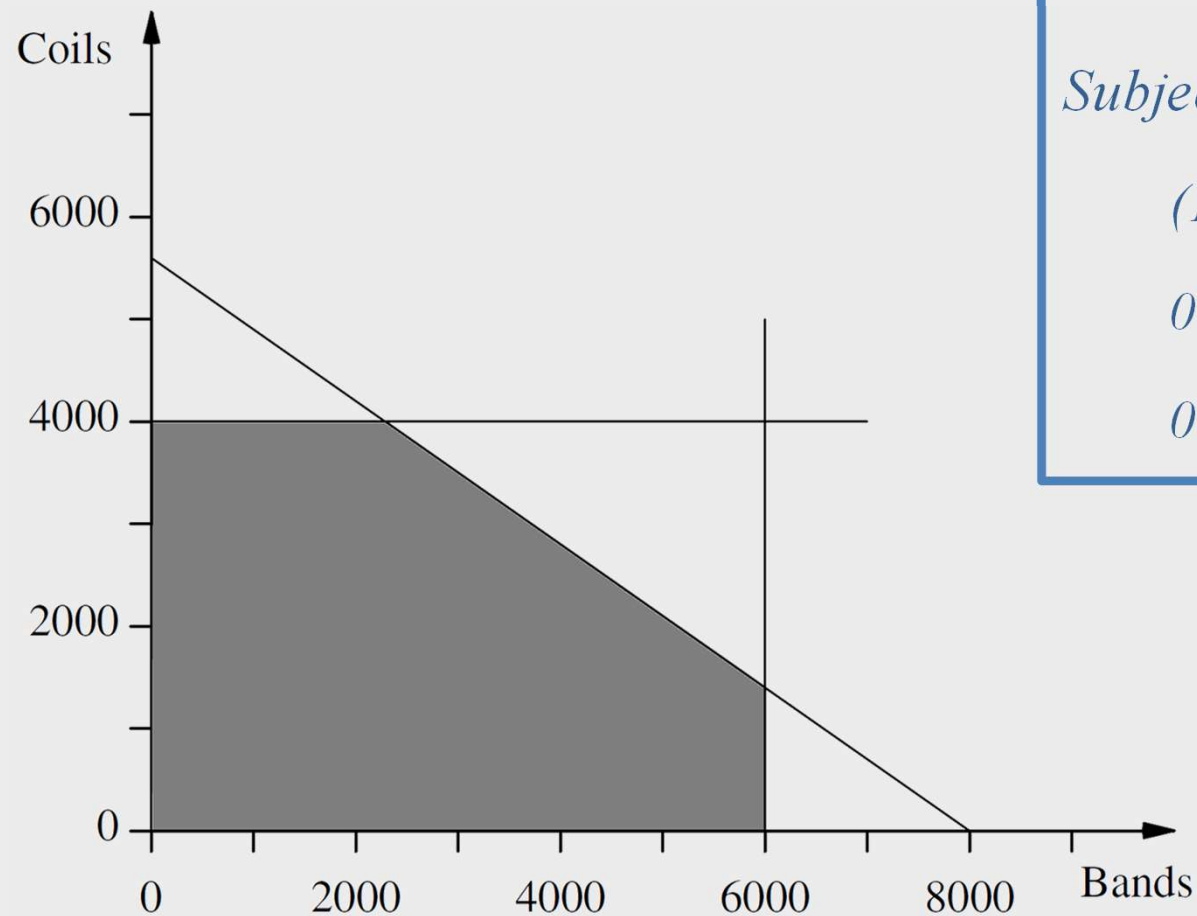
$$0 \leq x_j \leq u_j, \text{ for each } j \in P$$

# MAXIMIZING PROFITS

```
set P;  
  
param a {j in P};  
  
param b;  
  
param c {j in P};  
  
param u {j in P};  
  
var X {j in P};  
  
maximize Total_Profit: sum {j in P} c[j] * X[j];  
  
subject to Time: sum {j in P} (1/a[j]) * X[j] <= b;  
  
subject to Limit {j in P}: 0 <= X[j] <= u[j];
```



# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

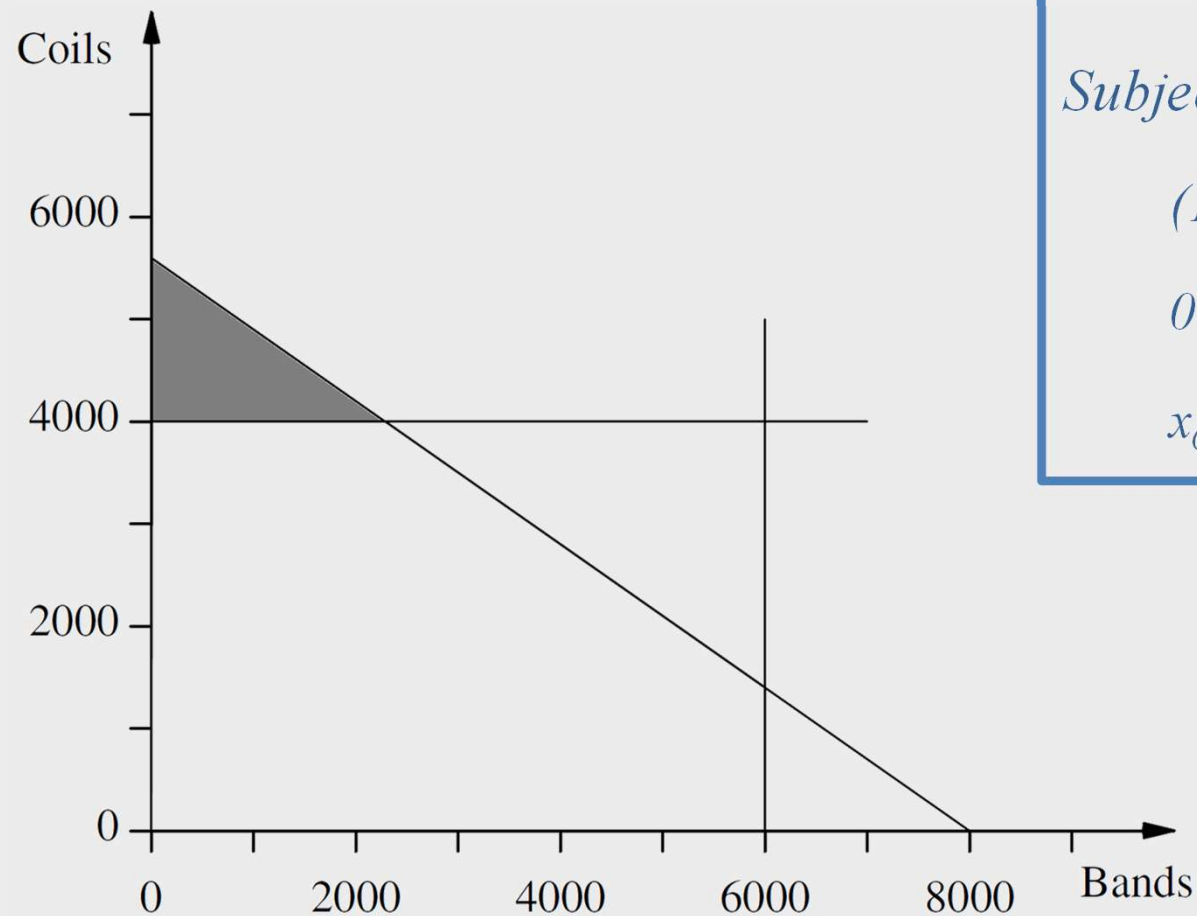
Subject to :

$$(1/200) x_B + (1/140) x_C \leq 40$$

$$0 \leq x_B \leq 6000$$

$$0 \leq x_C \leq 4000$$

# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

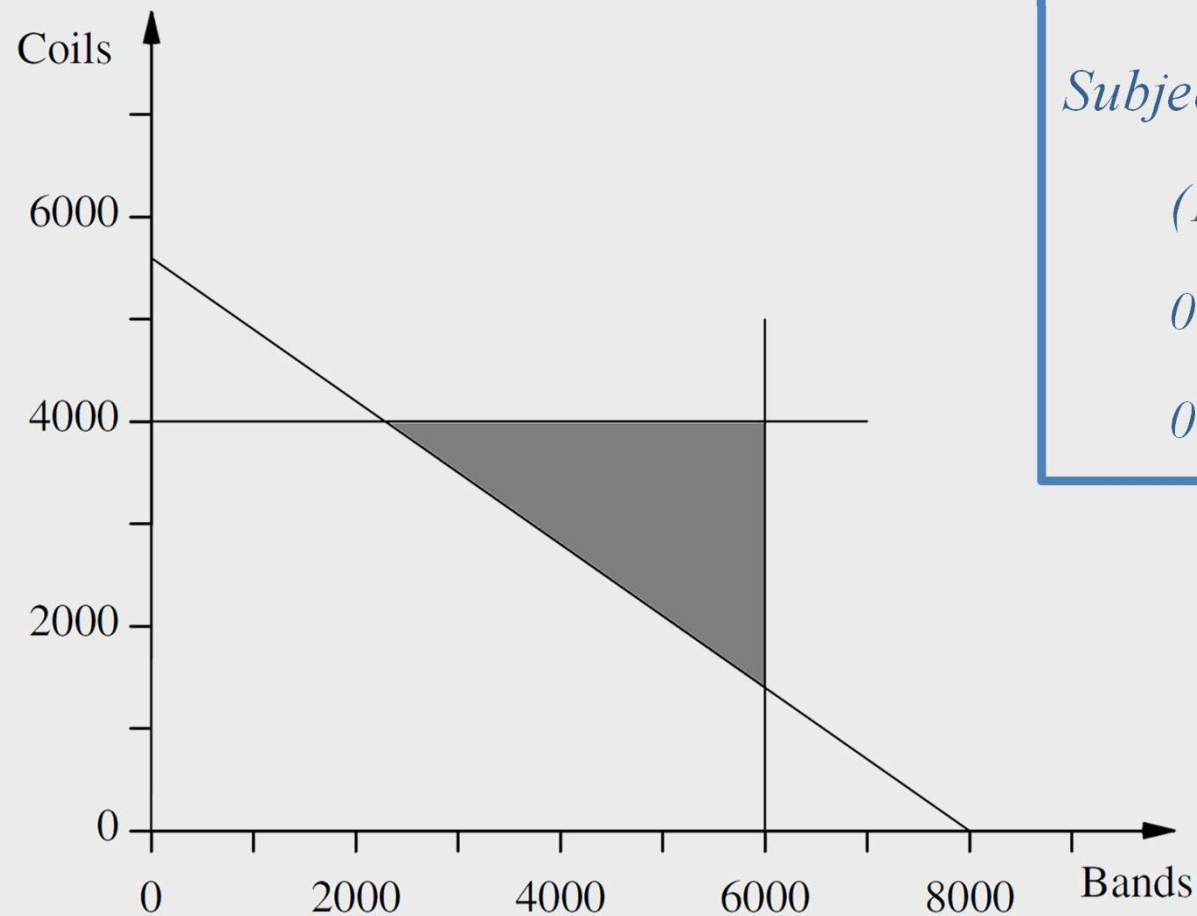
Subject to :

$$(1/200) x_B + (1/140) x_C \leq 40$$

$$0 \leq x_B \leq 6000$$

$$x_C \geq 4000$$

# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

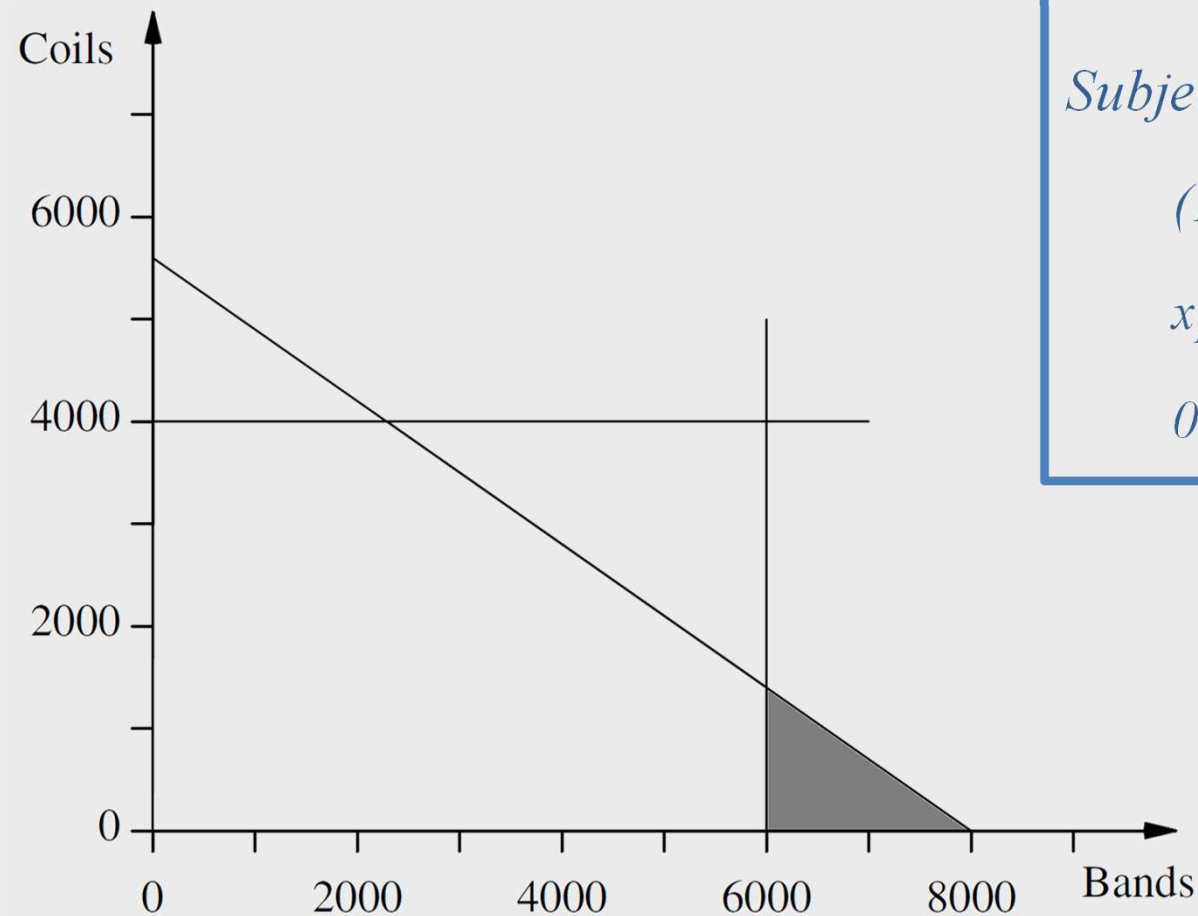
Subject to :

$$(1/200) x_B + (1/140) x_C \geq 40$$

$$0 \leq x_B \leq 6000$$

$$0 \leq x_C \leq 4000$$

# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

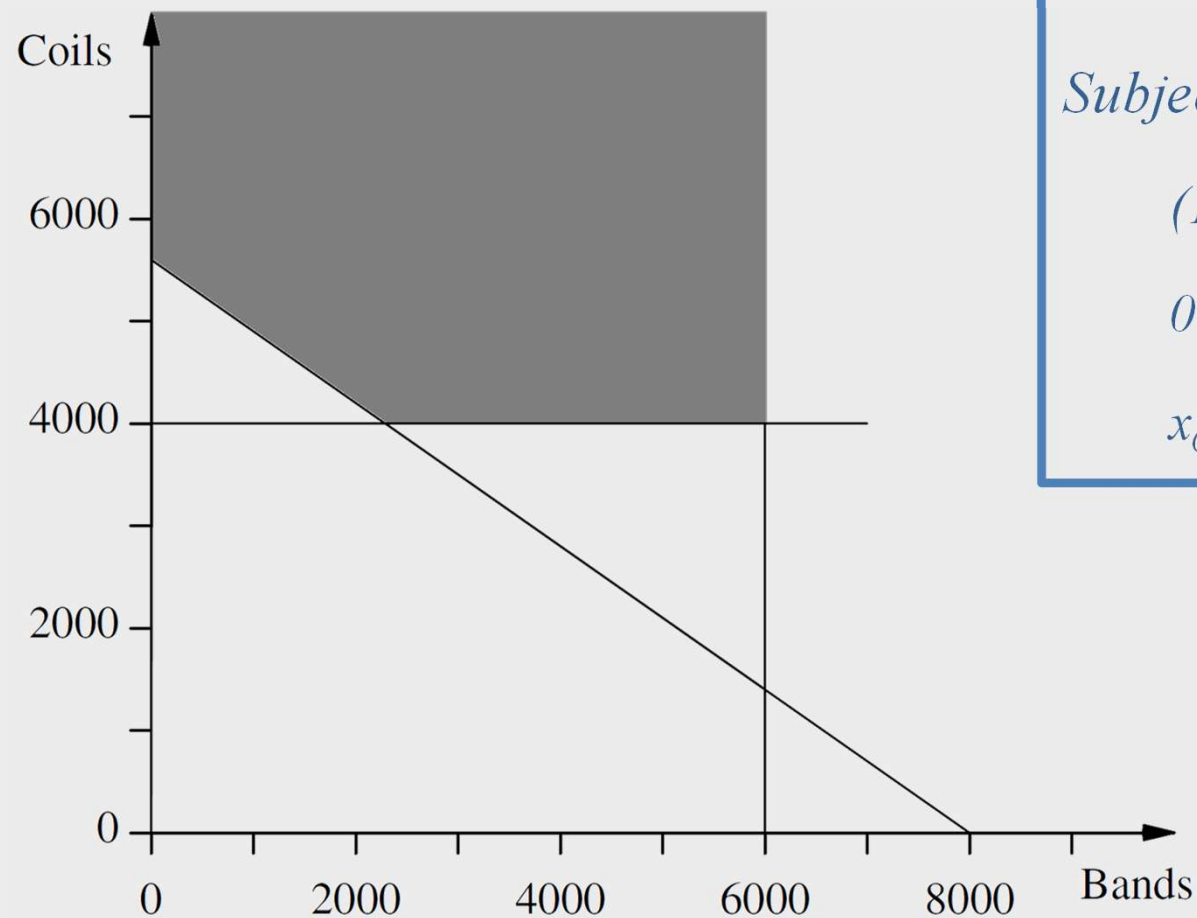
Subject to :

$$(1/200) x_B + (1/140) x_C \leq 40$$

$$x_B \geq 6000$$

$$0 \leq x_C \leq 4000$$

# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

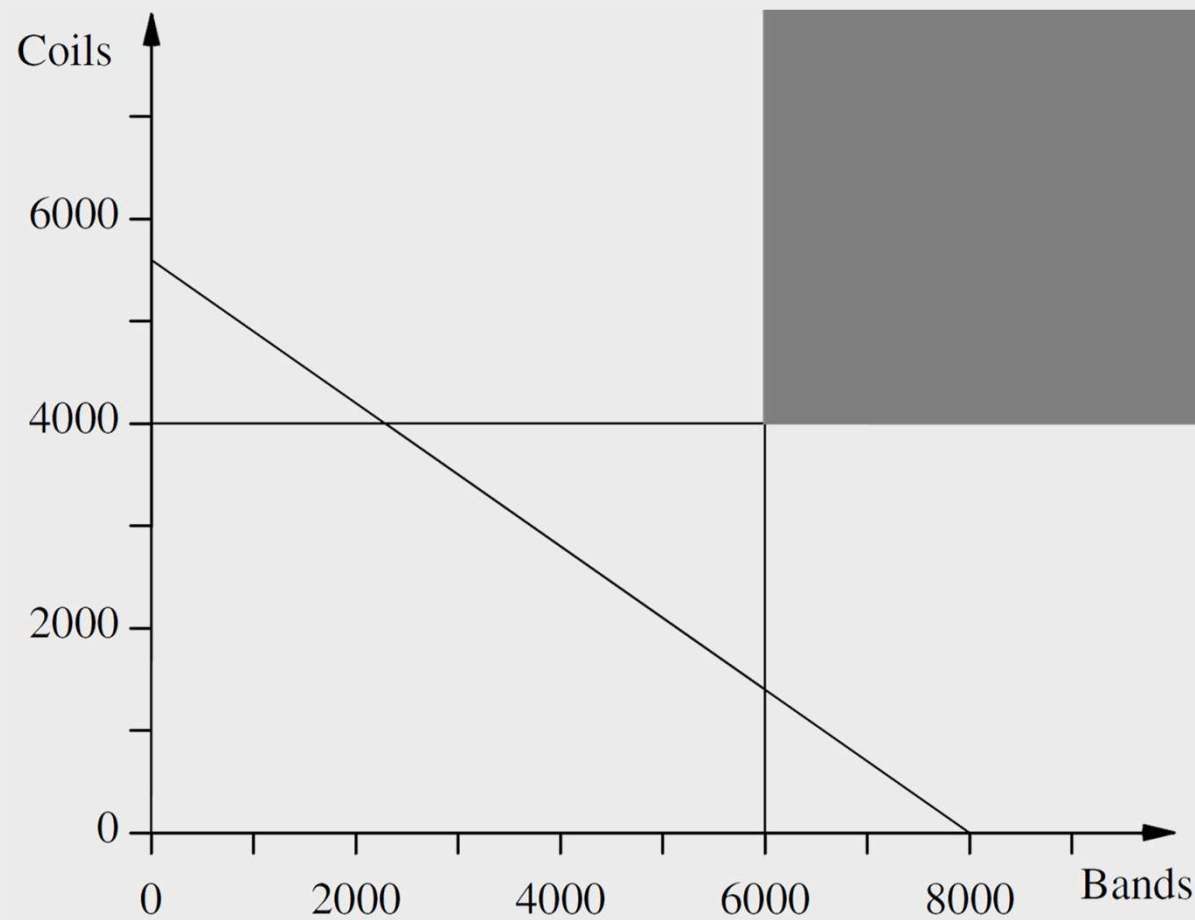
Subject to :

$$(1/200) x_B + (1/140) x_C \geq 40$$

$$0 \leq x_B \leq 6000$$

$$x_C \geq 4000$$

# FEASIBLE REGION



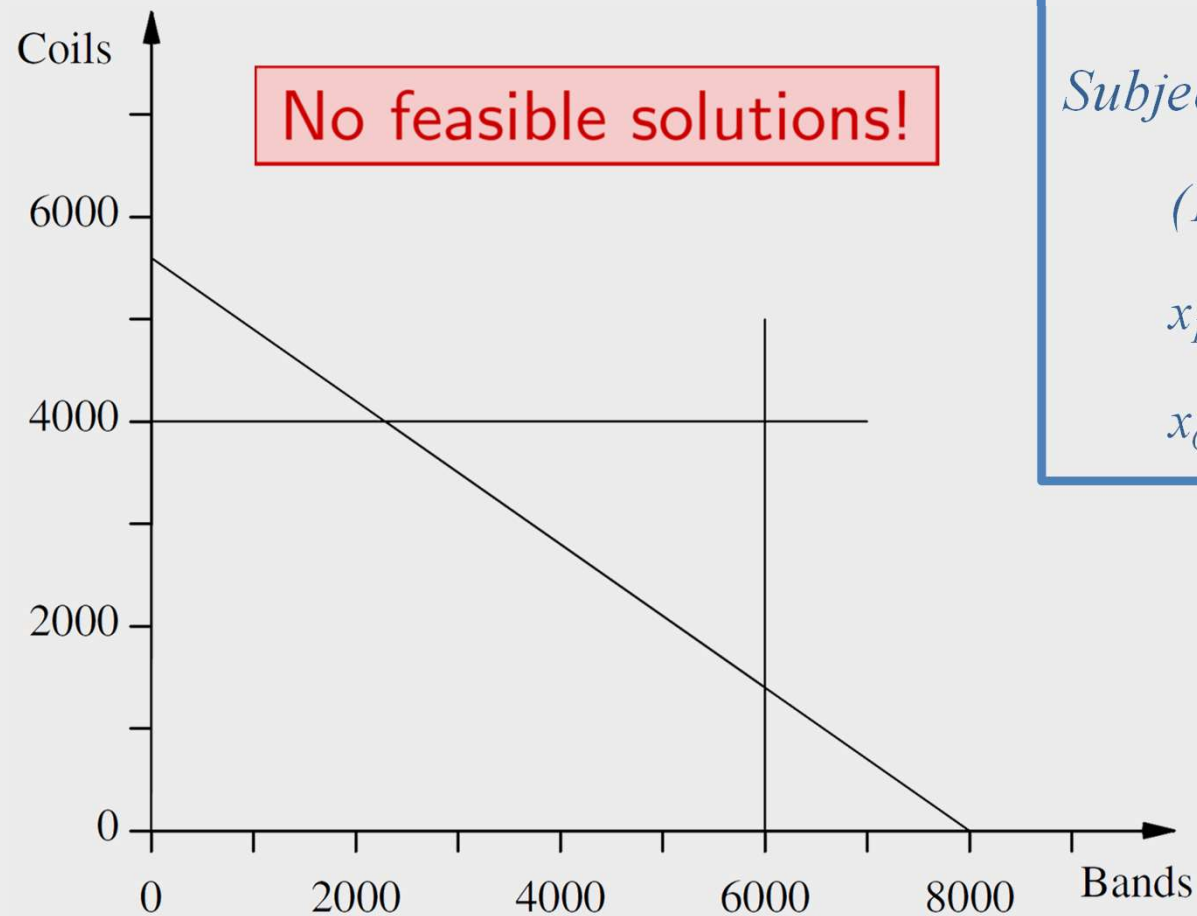
*Maximize*  $25 x_B + 30 x_C$

*Subject to :*

$$x_B \geq 6000$$

$$x_C \geq 4000$$

# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

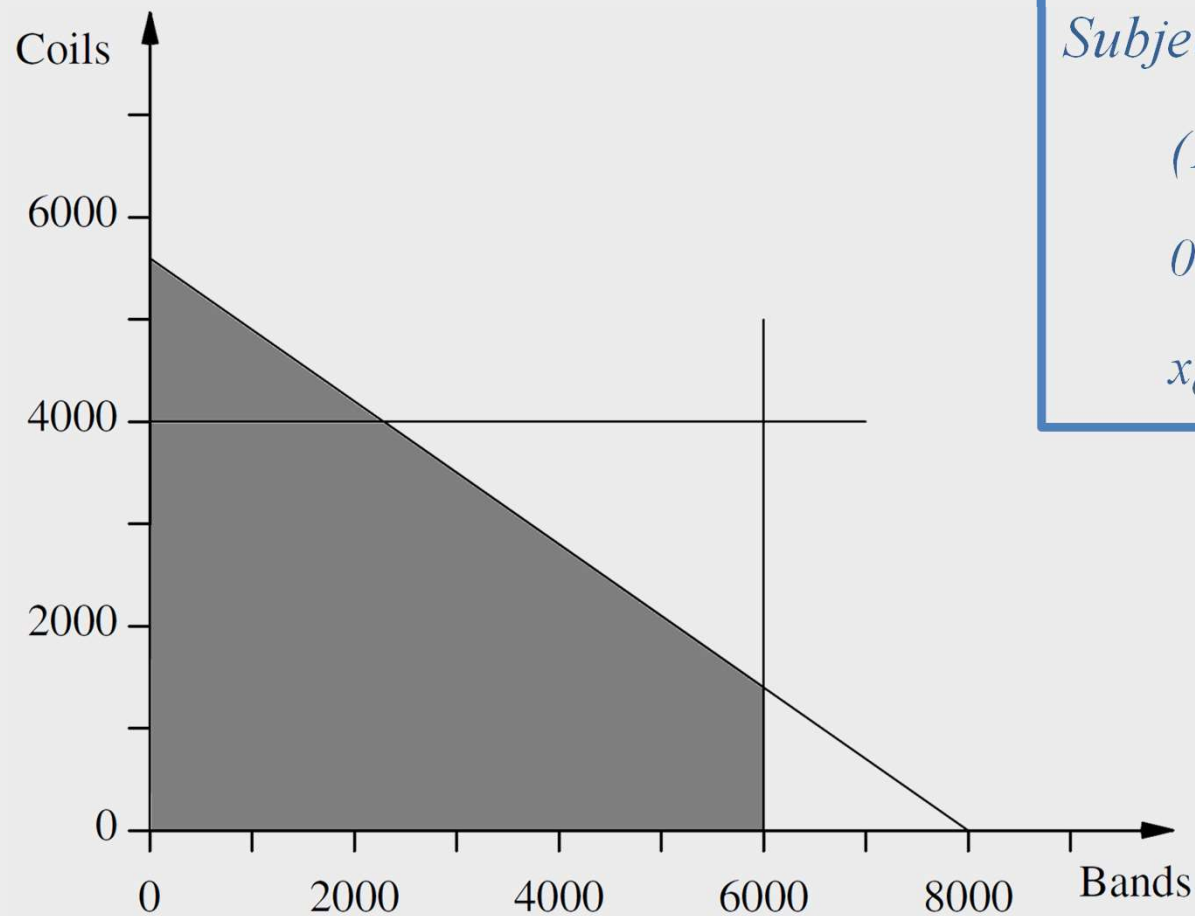
Subject to :

$$(1/200) x_B + (1/140) x_C \leq 40$$

$$x_B \geq 6000$$

$$x_C \geq 4000$$

# FEASIBLE REGION



Maximize  $25 x_B + 30 x_C$

Subject to :

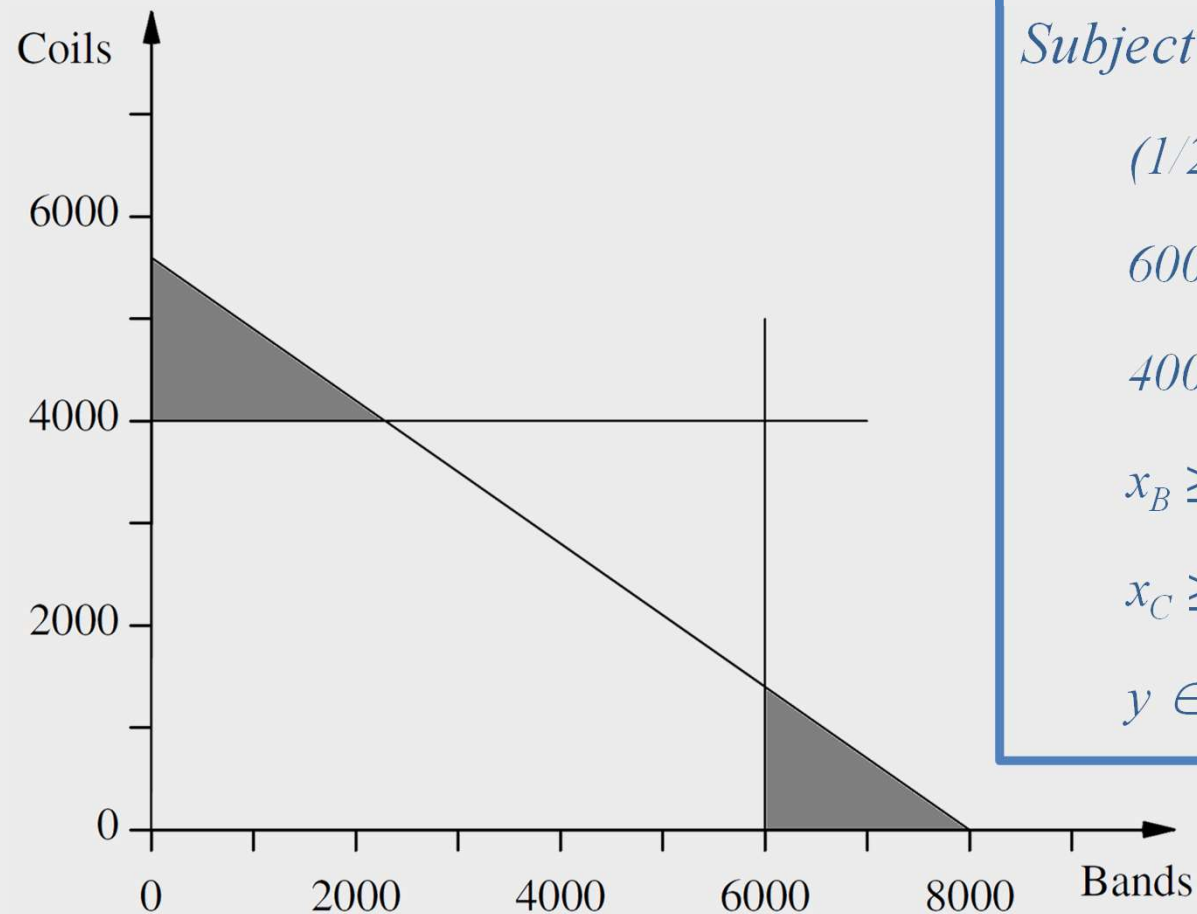
$$(1/200) x_B + (1/140) x_C \leq 40$$

$$0 \leq x_B \leq 6000$$

$$x_C \geq 0$$



# FEASIBLE REGION



*Maximize*  $25 x_B + 30 x_C$

*Subject to :*

$$(1/200) x_B + (1/140) x_C \leq 40$$

$$6000 - M(1 - y) \leq x_B \leq 6000 + My$$

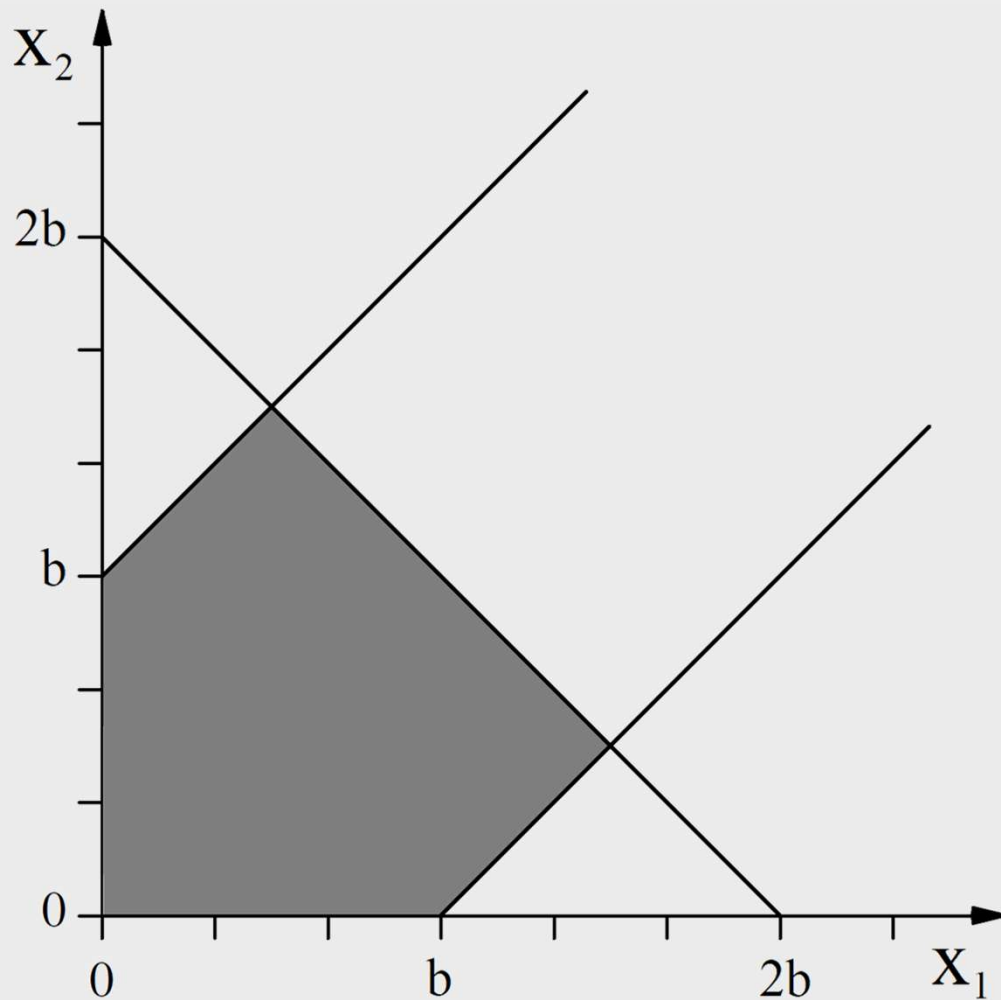
$$4000 - My \leq x_C \leq 4000 + M(1 - y)$$

$$x_B \geq 0$$

$$x_C \geq 0$$

$$y \in \{0, 1\}$$

# FEASIBLE REGION



Maximize  $c_1 x_1 + c_2 x_2$

Subject to :  $x_1 + x_2 \leq 2b$

$|x_1 - x_2| \leq b$

$x_1, x_2 \geq 0$

Maximize  $c_1 x_1 + c_2 x_2$

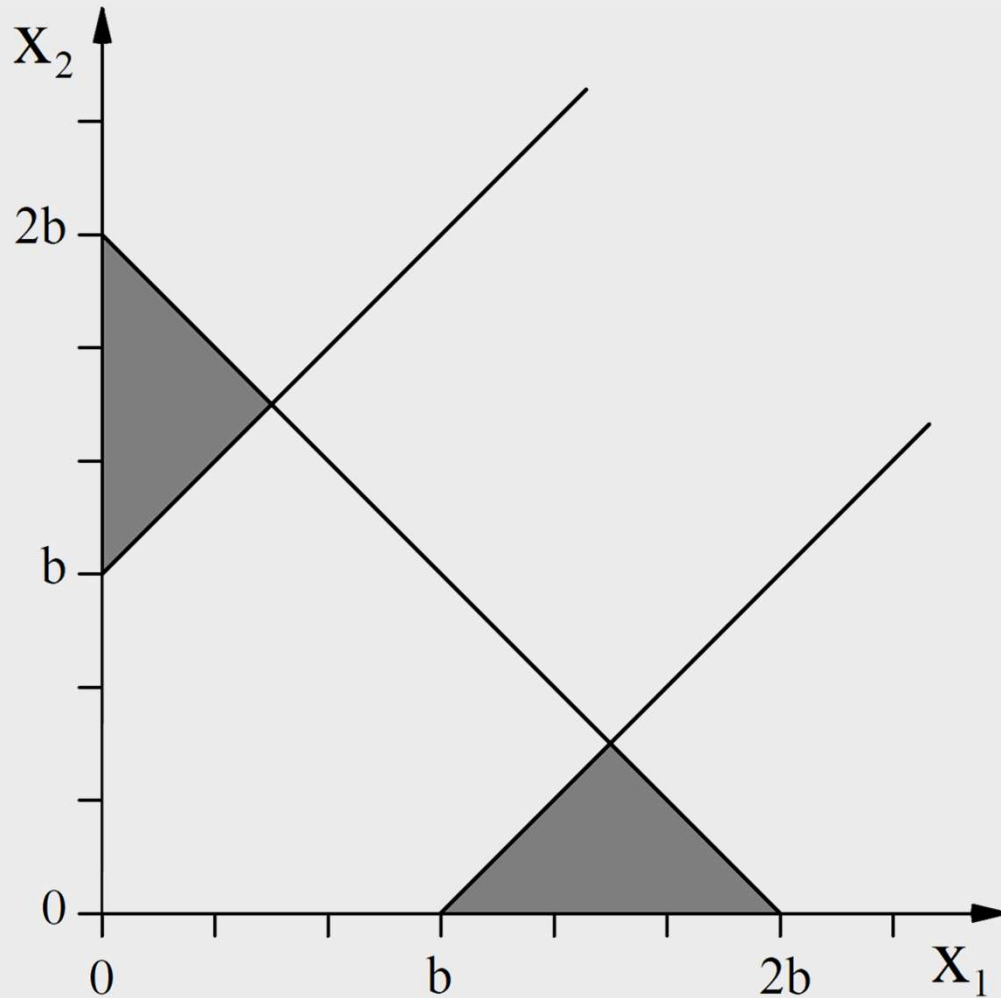
Subject to :  $x_1 + x_2 \leq 2b$

$x_1 - x_2 \leq b$

$x_1 - x_2 \geq -b$

$x_1, x_2 \geq 0$

# FEASIBLE REGION



$$\text{Maximize} \quad c_1 x_1 + c_2 x_2$$

$$\text{Subject to :} \quad x_1 + x_2 \leq 2b$$

$$|x_1 - x_2| \geq b$$

$$x_1, x_2 \geq 0$$

$$\text{Maximize} \quad c_1 x_1 + c_2 x_2$$

$$\text{Subject to :} \quad x_1 + x_2 \leq 2b$$

$$x_1 - x_2 \geq b - My$$

$$x_1 - x_2 \leq -b + M(1 - y)$$

$$x_1, x_2 \geq 0$$

# IF-THEN CONDITION

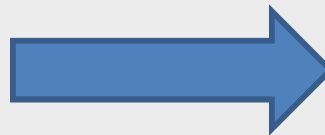
Minimize  $c_1 x + c_2 y$

st:  $g(x) \geq b$

$x \geq 0$  and  $y \in \{0, 1\}$

If  $x > 0 \leftrightarrow y = 1$

If  $x = 0 \leftrightarrow y = 0$



$x \leq M y$

	$y = 0$	$y = 1$
$x = 0$	✓	✗
$x > 0$	✗	✓

# FEASIBLE REGION

Feasible region:

$$2x_1 + 3x_2 \leq 10 \quad (1)$$

$$x_1 + 3x_2 \leq 8 \quad (2)$$

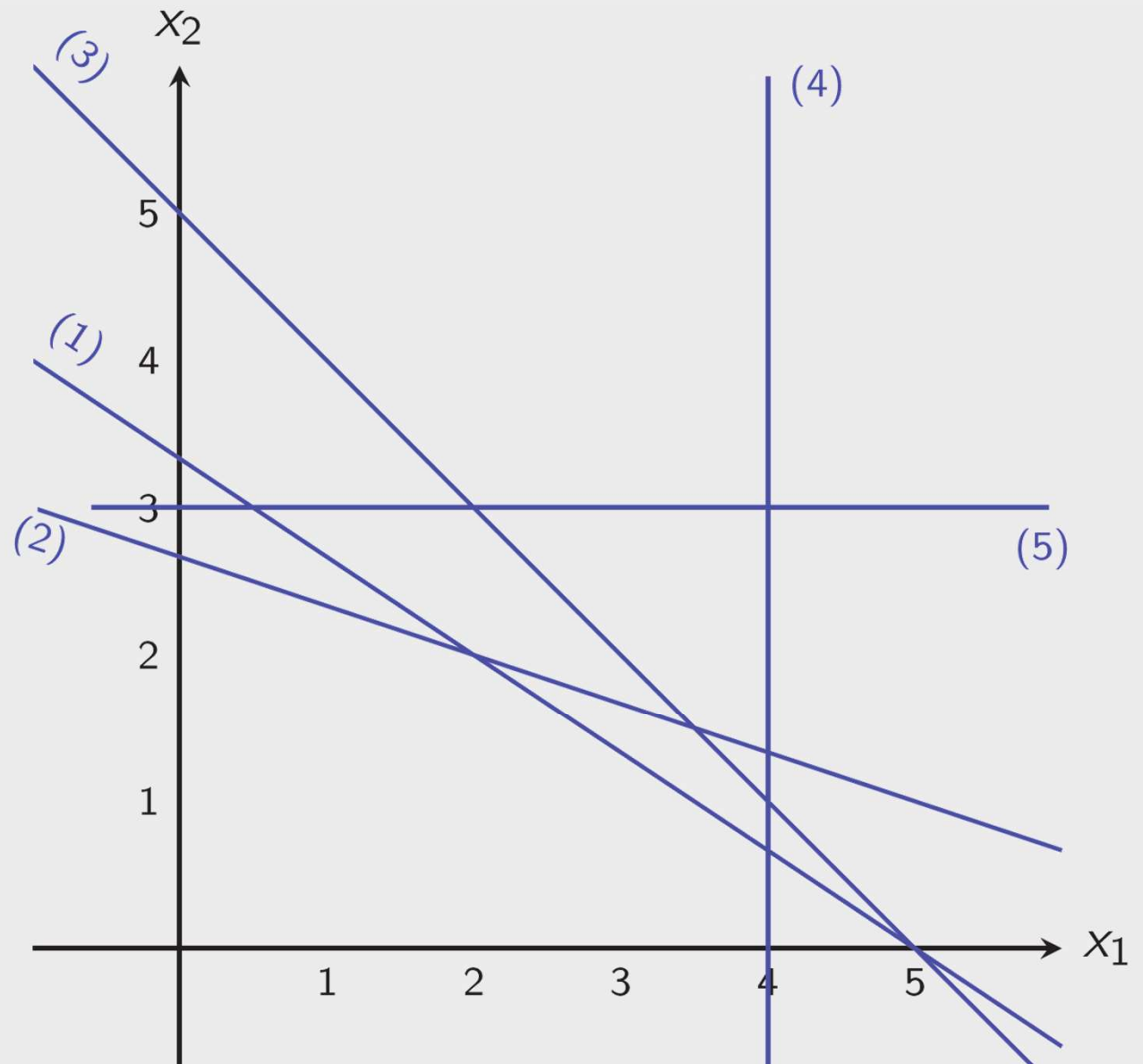
$$x_1 + x_2 \leq 5 \quad (3)$$

$$x_1 \leq 4 \quad (4)$$

$$x_2 \geq 3 \quad (5)$$

$$x_1 \geq 0 \quad (6)$$

$$x_2 \geq 0 \quad (7)$$



# FEASIBLE REGION

No feasible solutions!

Feasible region:

$$2x_1 + 3x_2 \leq 10 \quad (1)$$

$$x_1 + 3x_2 \leq 8 \quad (2)$$

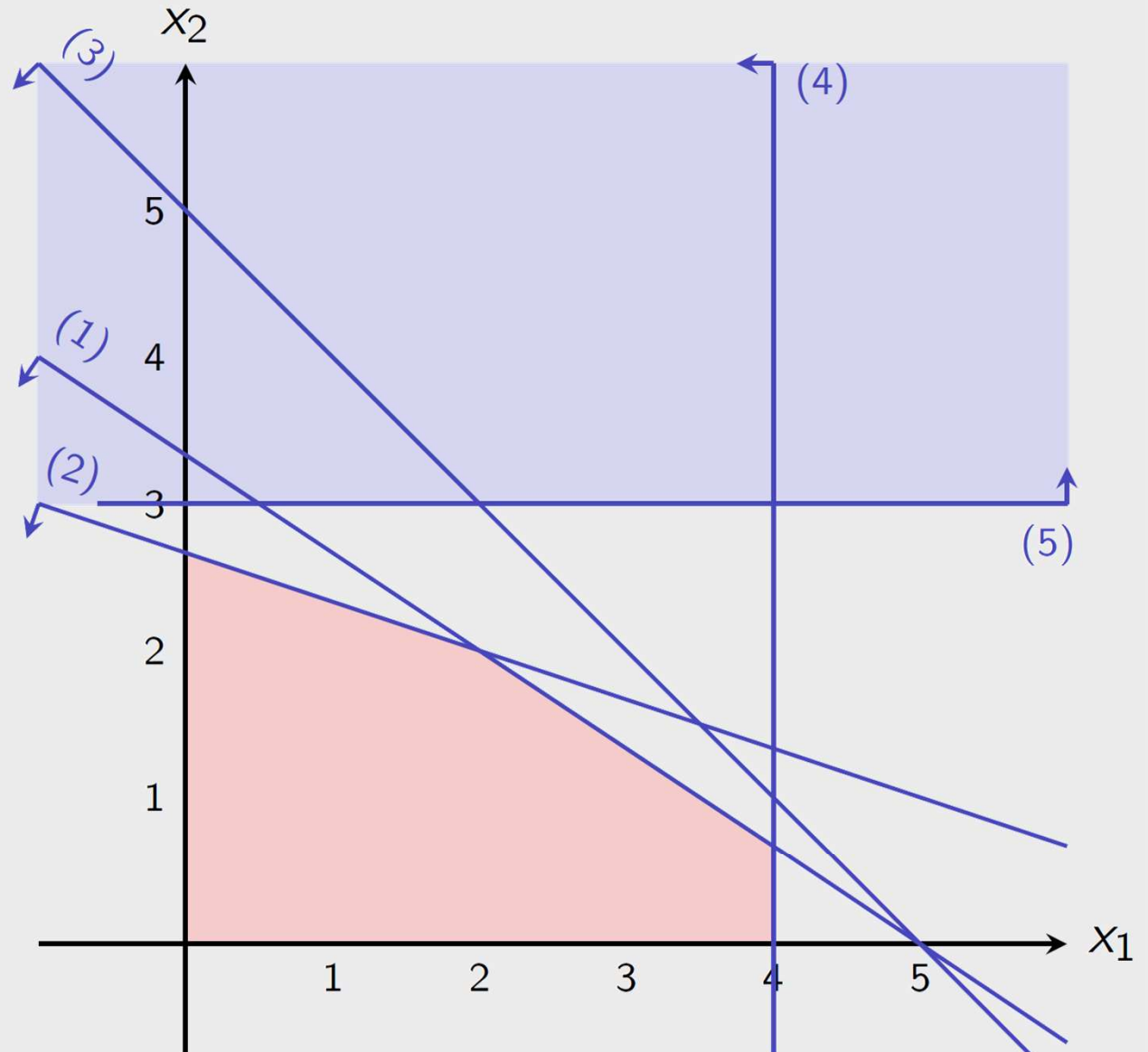
$$x_1 + x_2 \leq 5 \quad (3)$$

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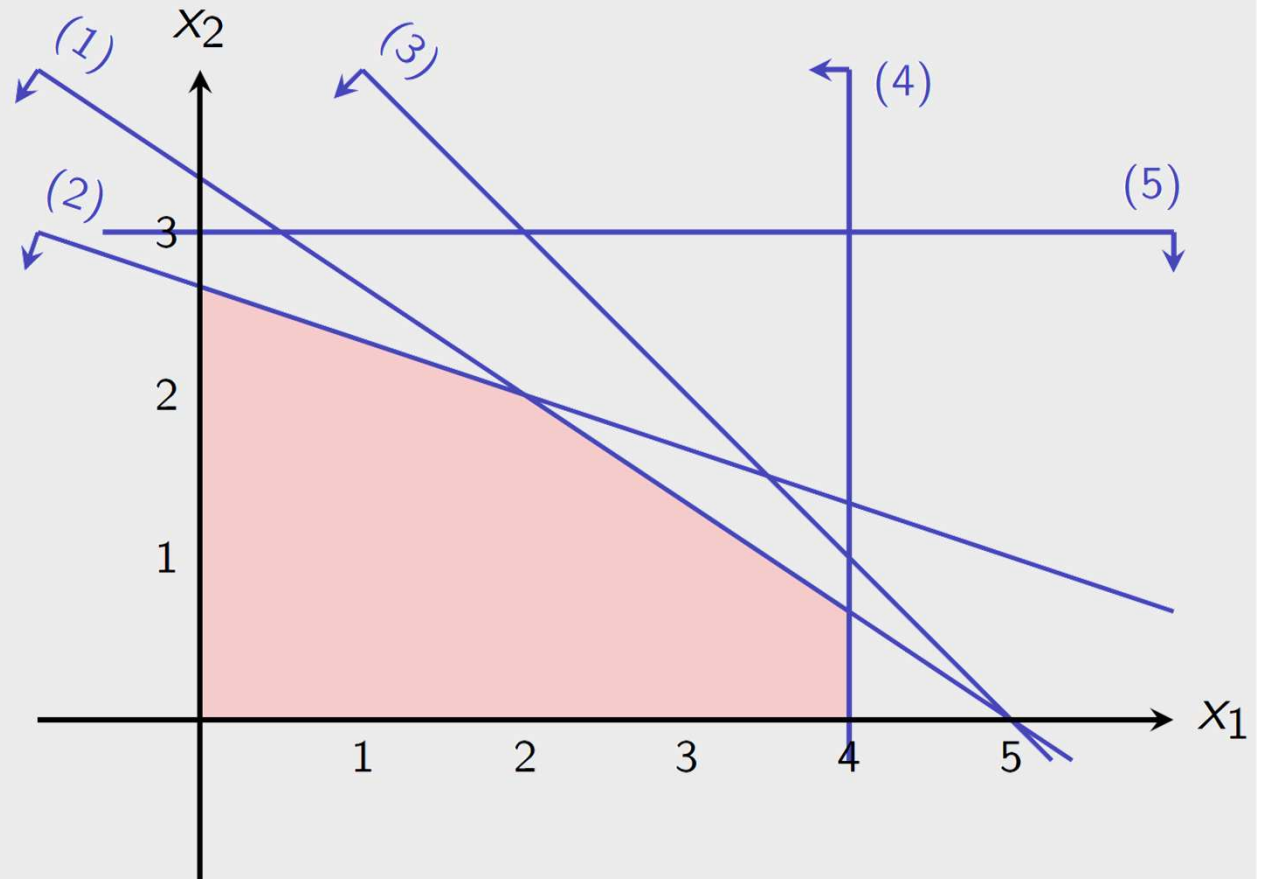
$$x_1 \geq 0 \quad (6)$$

$$x_2 \geq 0 \quad (7)$$



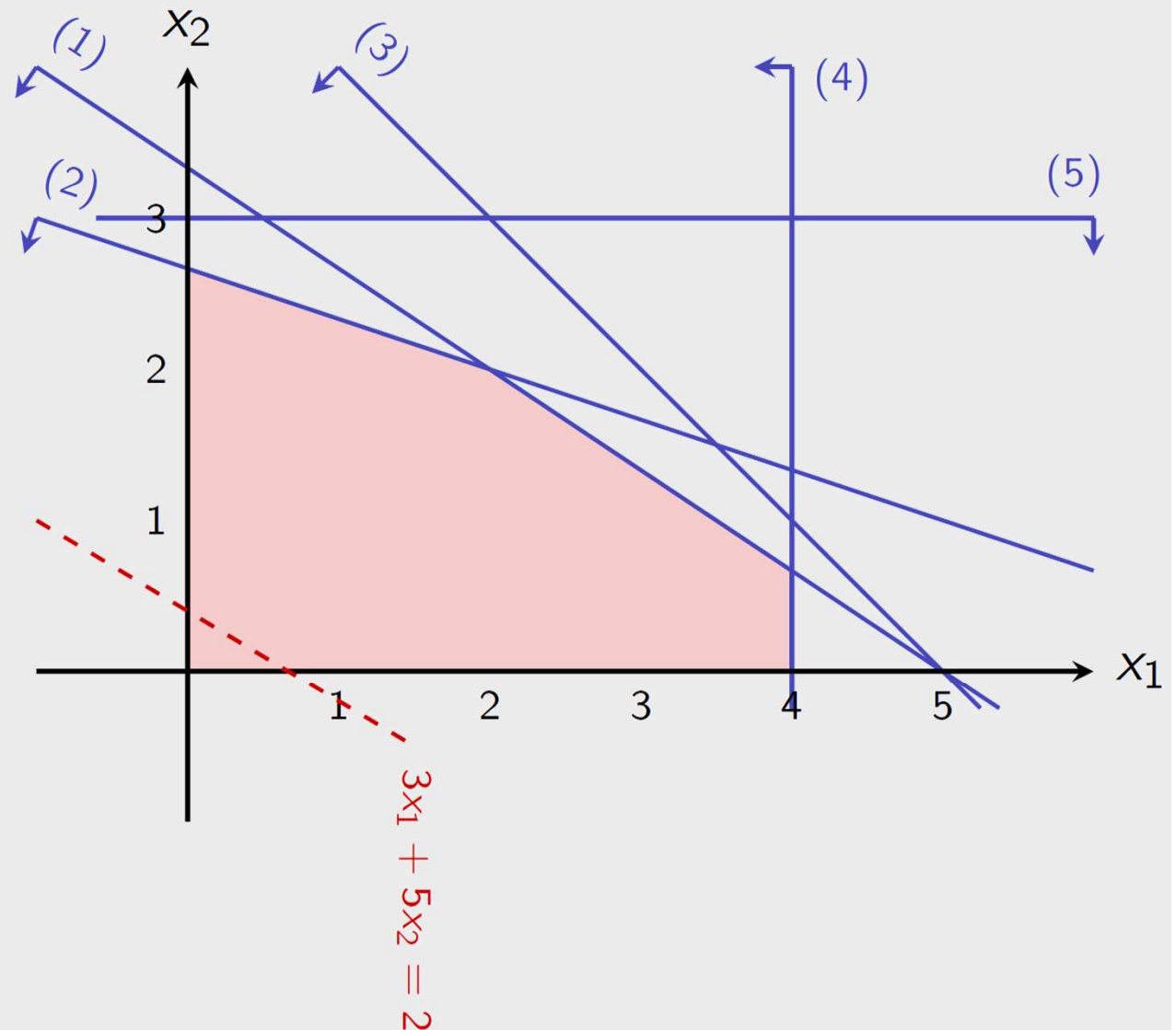
# OBJECTIVE FUNCTION

$$\begin{array}{ll} \max & 3x_1 + 5x_2 \\ \text{s.t.} & 2x_1 + 3x_2 \leq 10 \quad (1) \\ & x_1 + 3x_2 \leq 8 \quad (2) \\ & x_1 + x_2 \leq 5 \quad (3) \\ & x_1 \leq 4 \quad (4) \\ & x_2 \leq 3 \quad (5) \\ & x_1 \geq 0 \quad (6) \\ & x_2 \geq 0 \quad (7) \end{array}$$



# OBJECTIVE FUNCTION

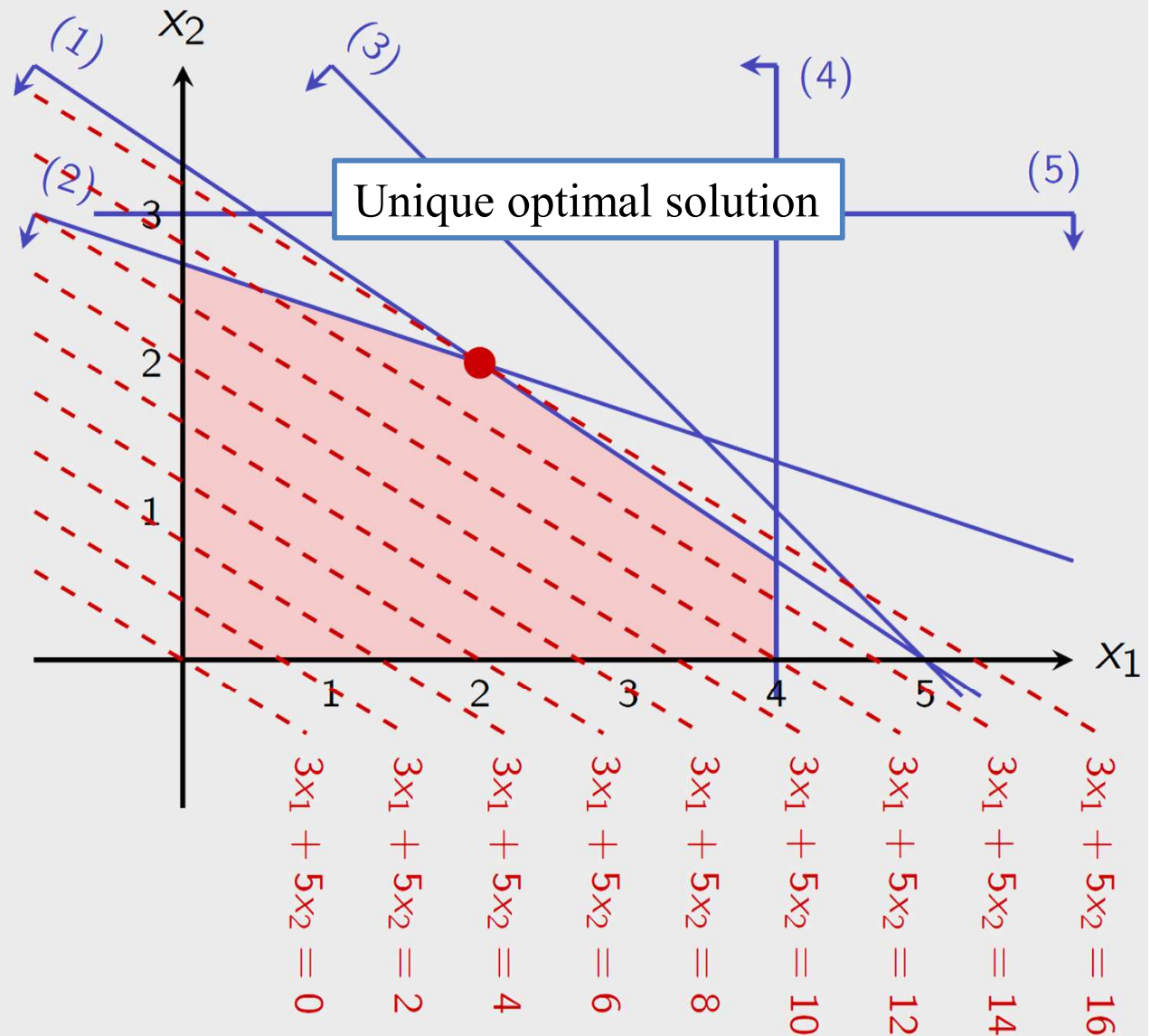
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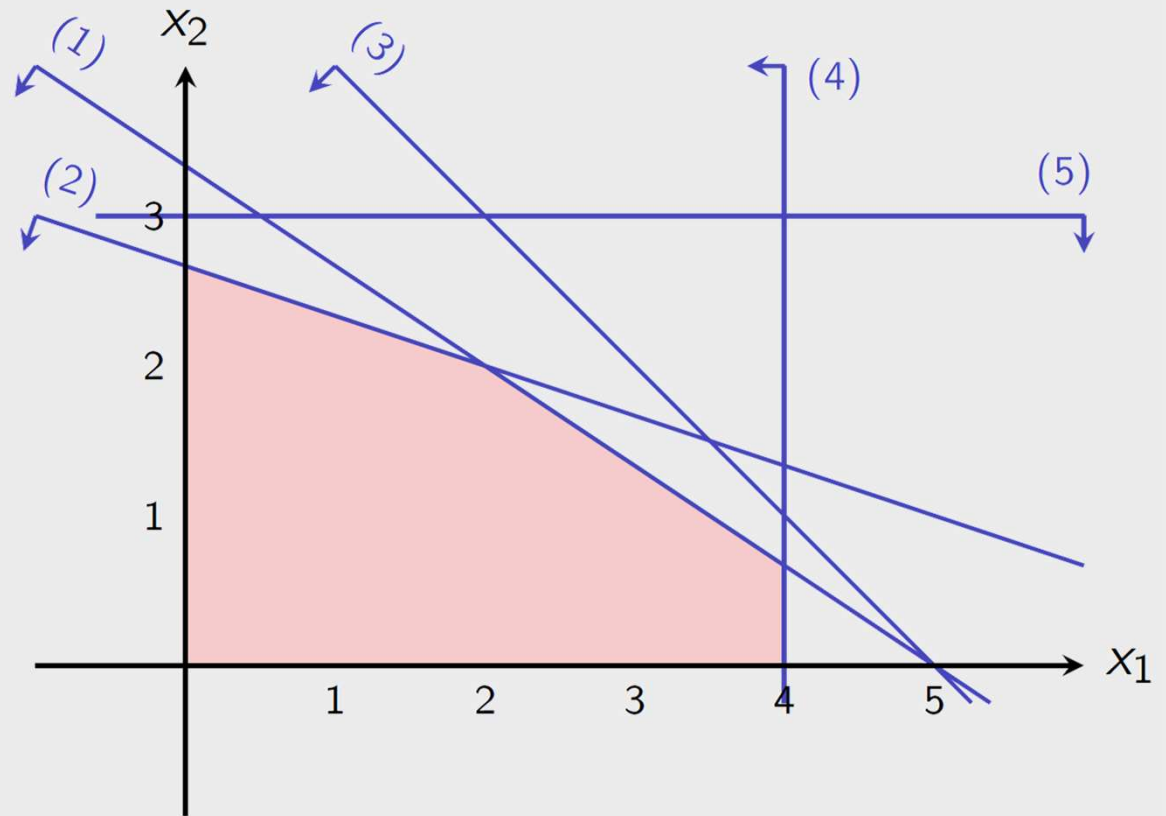
# OBJECTIVE FUNCTION

$$\begin{aligned}
 \max \quad & 3x_1 + 5x_2 \\
 \text{s.t.} \quad & 2x_1 + 3x_2 \leq 10 \quad (1) \\
 & x_1 + 3x_2 \leq 8 \quad (2) \\
 & x_1 + x_2 \leq 5 \quad (3) \\
 & x_1 \leq 4 \quad (4) \\
 & x_2 \leq 3 \quad (5) \\
 & x_1 \geq 0 \quad (6) \\
 & x_2 \geq 0 \quad (7)
 \end{aligned}$$



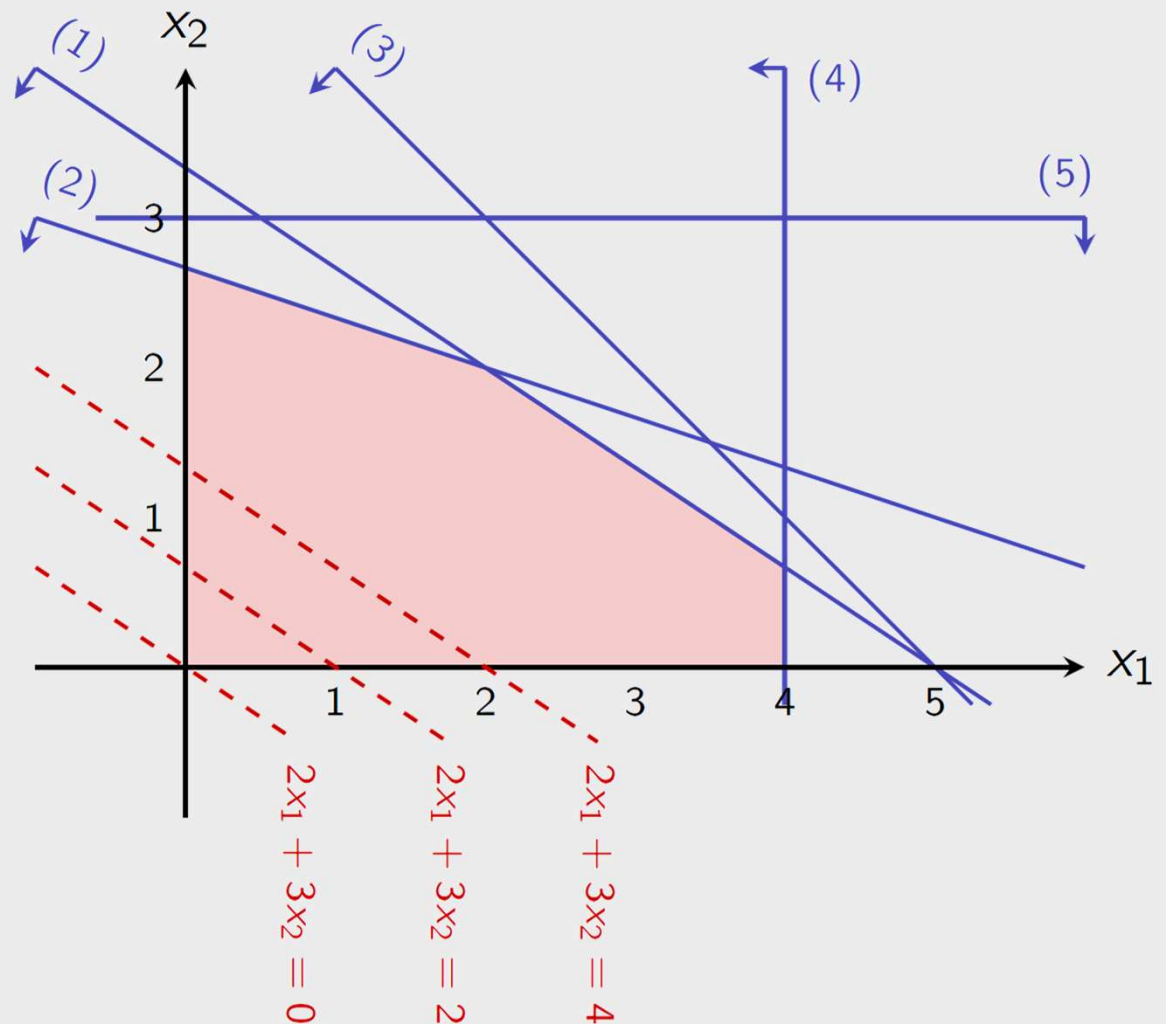
# OBJECTIVE FUNCTION

$$\begin{array}{ll}\max & 2x_1 + 3x_2 \\ \text{s.t.} & 2x_1 + 3x_2 \leq 10 \quad (1) \\ & x_1 + 3x_2 \leq 8 \quad (2) \\ & x_1 + x_2 \leq 5 \quad (3) \\ & x_1 \leq 4 \quad (4) \\ & x_2 \leq 3 \quad (5) \\ & x_1 \geq 0 \quad (6) \\ & x_2 \geq 0 \quad (7)\end{array}$$



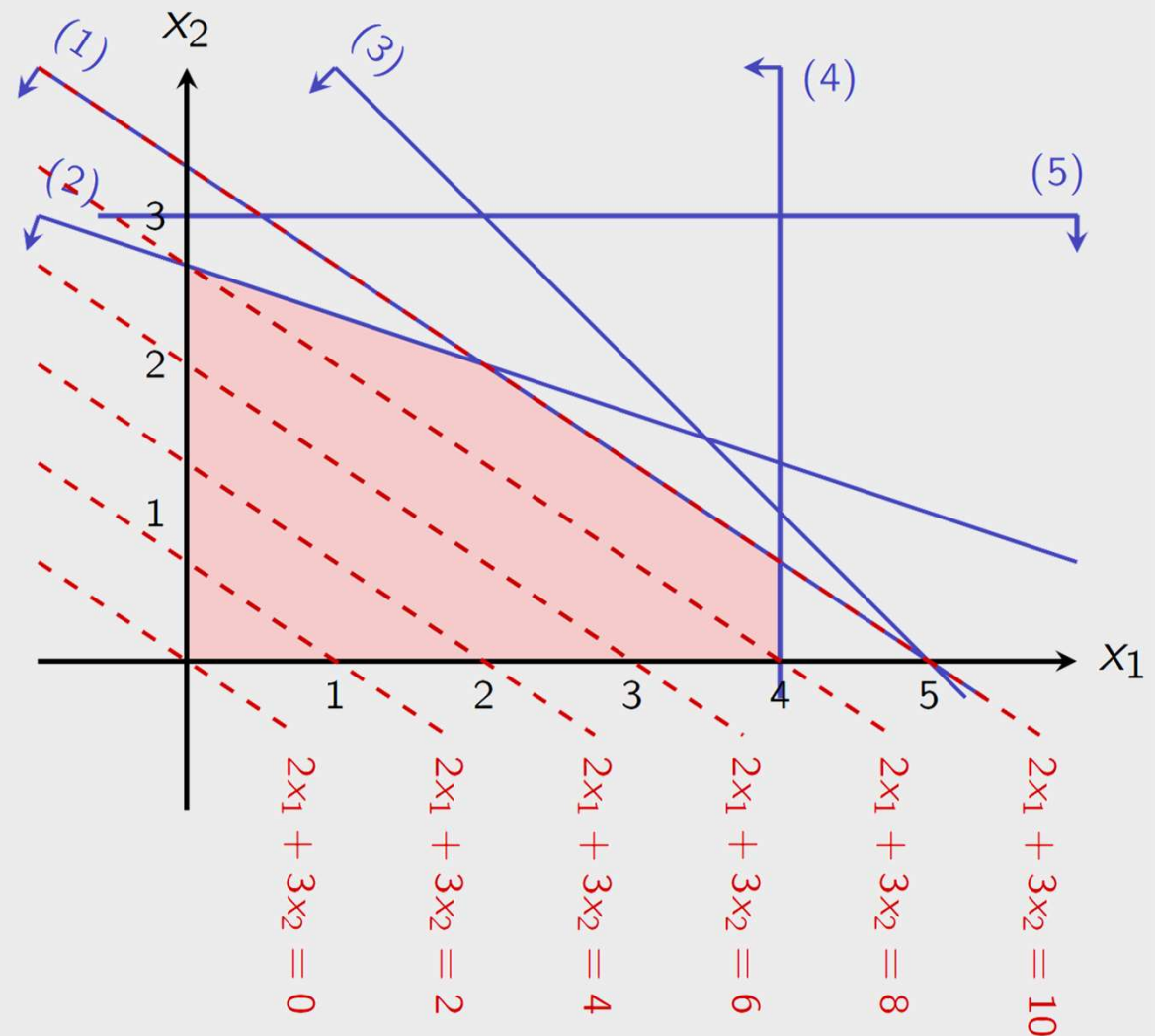
# OBJECTIVE FUNCTION

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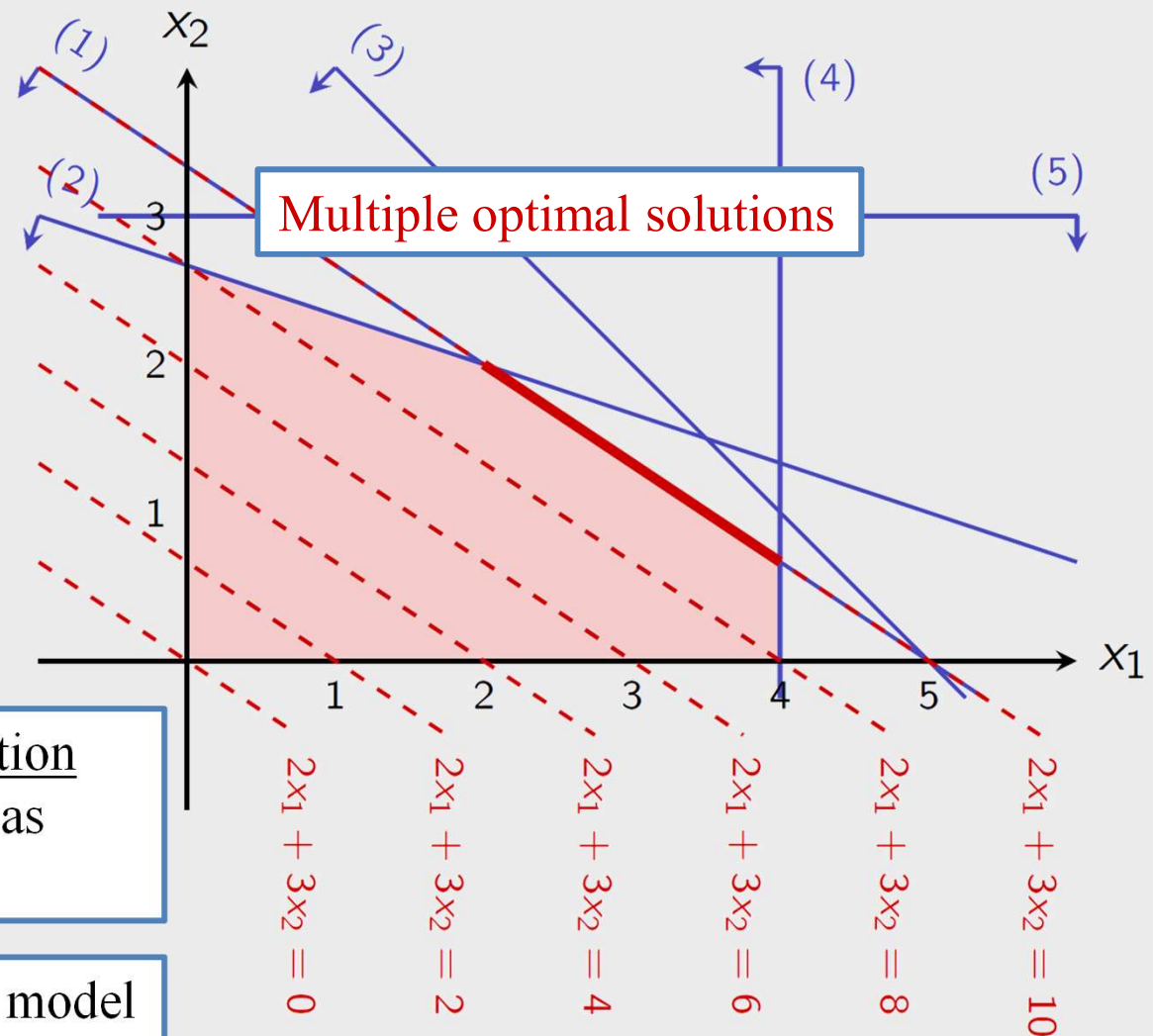
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# OBJECTIVE FUNCTION

$$\begin{array}{ll}\max & 2x_1 + 3x_2 \\ \text{s.t.} & 2x_1 + 3x_2 \leq 10 \quad (1) \\ & x_1 + 3x_2 \leq 8 \quad (2) \\ & x_1 + x_2 \leq 5 \quad (3) \\ & x_1 \leq 4 \quad (4) \\ & x_2 \leq 3 \quad (5) \\ & x_1 \geq 0 \quad (6) \\ & x_2 \geq 0 \quad (7)\end{array}$$

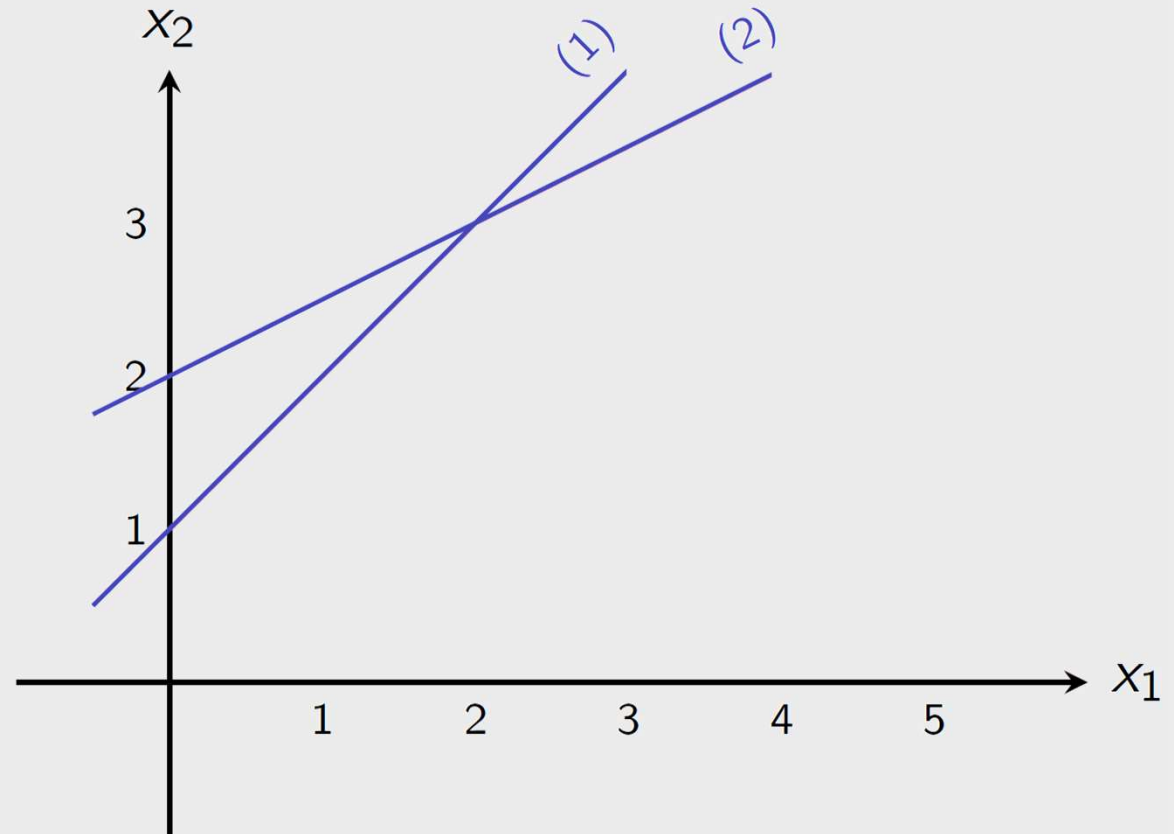


An **optimal solution** is a feasible solution with objective function value at least as good as any other feasible solution

The **optimal value** of an optimization model is the objective function value of any optimal solution

# OBJECTIVE FUNCTION

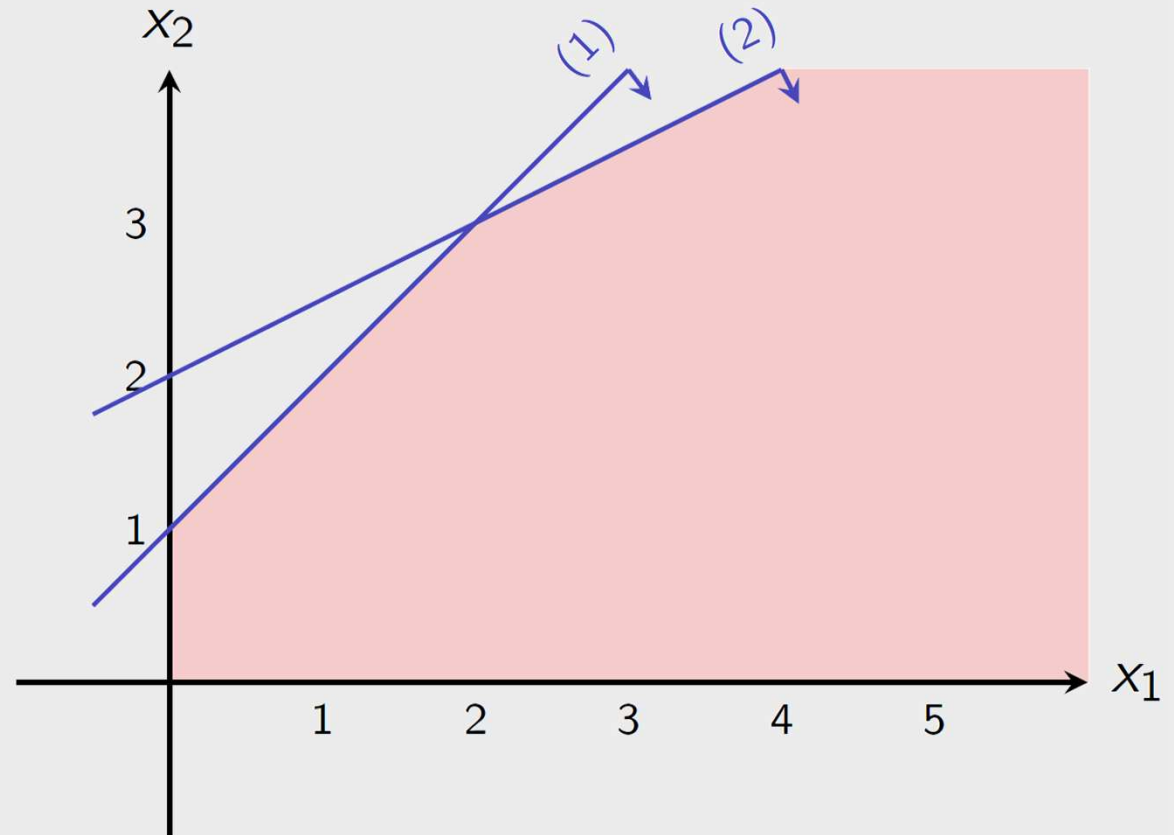
$$\begin{array}{ll}\max & 3x_1 + x_2 \\ \text{s.t.} & -x_1 + x_2 \leq 1 \quad (1) \\ & x_1 - 2x_2 \geq -4 \quad (2) \\ & x_1 \geq 0 \quad (3) \\ & x_2 \geq 0 \quad (4)\end{array}$$





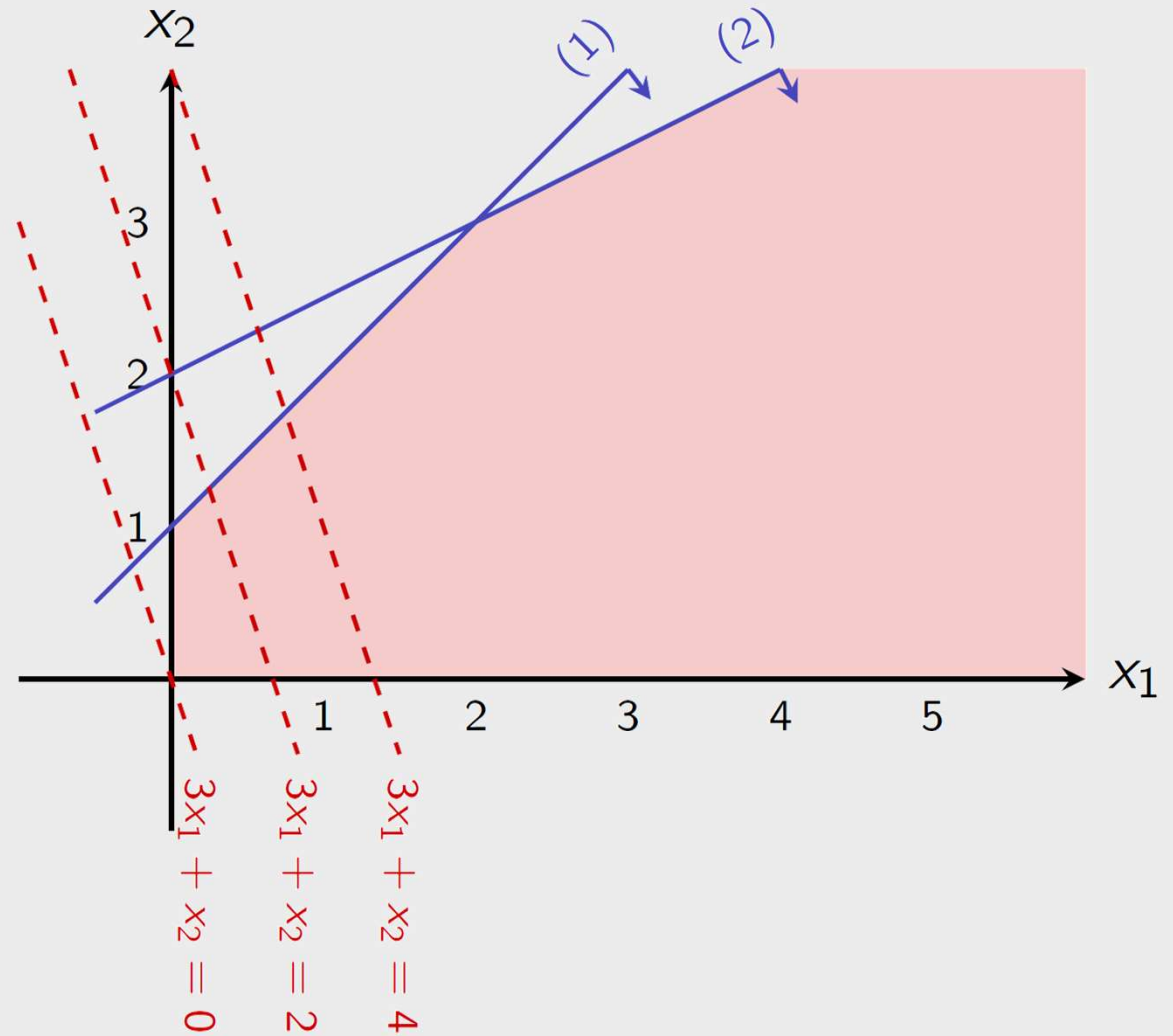
# OBJECTIVE FUNCTION

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# OBJECTIVE FUNCTION

$$\begin{array}{ll}\max & 3x_1 + x_2 \\ \text{s.t.} & -x_1 + x_2 \leq 1 \quad (1) \\ & x_1 - 2x_2 \geq -4 \quad (2) \\ & x_1 \geq 0 \quad (3) \\ & x_2 \geq 0 \quad (4)\end{array}$$



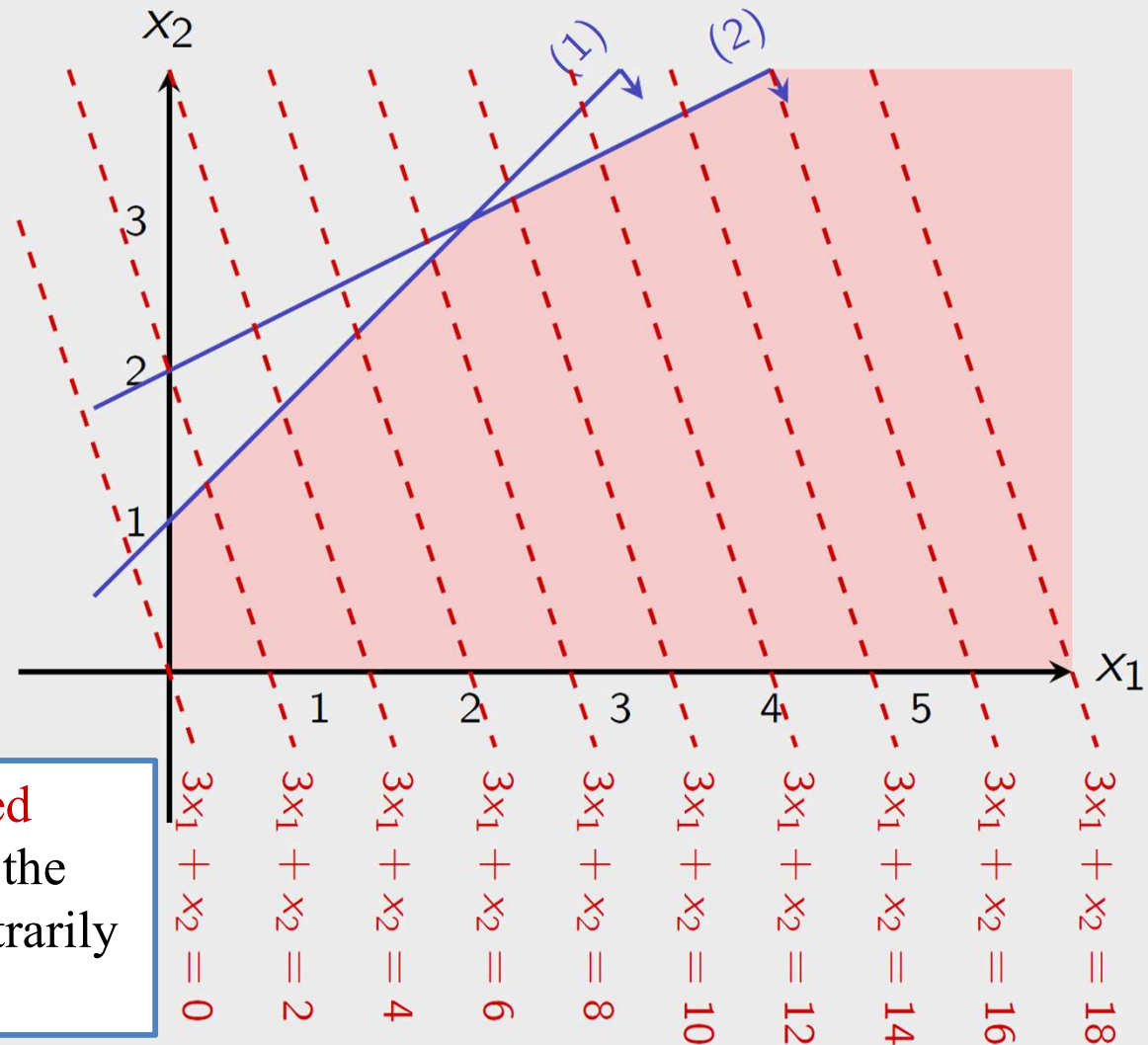


# OBJECTIVE FUNCTION

$$\begin{array}{ll}\max & 3x_1 + x_2 \\ \text{s.t.} & -x_1 + x_2 \leq 1 \quad (1) \\ & x_1 - 2x_2 \geq -4 \quad (2) \\ & x_1 \geq 0 \quad (3) \\ & x_2 \geq 0 \quad (4)\end{array}$$

The model is "**unbounded**"

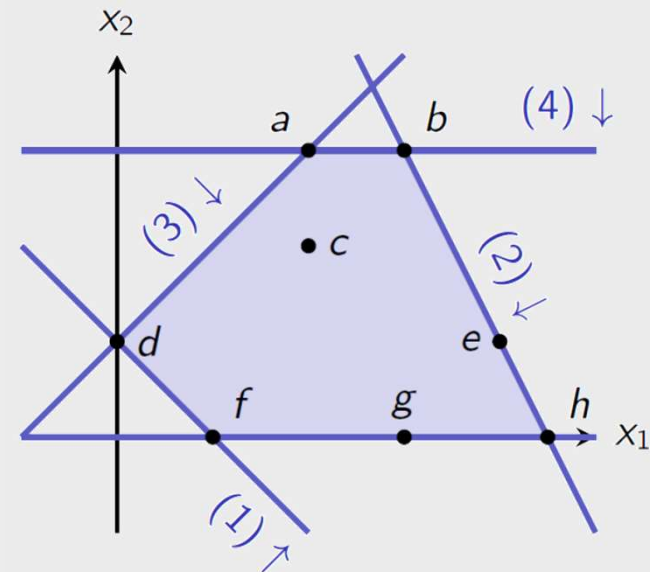
An optimization model is **unbounded** when feasible choices of values for the decision variables can produce arbitrarily good objective function values



# OBJECTIVE FUNCTION

Consider the following LP. Note that  $c_1$  and  $c_2$  are unspecified.

$$\begin{array}{ll}\min & c_1x_1 + c_2x_2 \\ \text{s.t.} & x_1 + x_2 \geq 1 \quad (1) \\ & 2x_1 + x_2 \leq 9 \quad (2) \\ & x_1 - x_2 \geq -1 \quad (3) \\ & x_2 \leq 3 \quad (4) \\ & x_1 \geq 0, x_2 \geq 0\end{array}$$



- Suppose we have a unique optimal solution. Which points could it be, if any?
- Suppose “e” is an optimal solution. Which other points must also be optimal, if any?
- Suppose “f” and “h” are both optimal solutions. Which other points must also be optimal, if any?

Based on characteristics of:

- decision variables
- constraints
- objective function

# CLASSIFICATION OF OPTIMIZATION MODELS

- A decision variable is **continuous** if it can take on any value in a specified interval
  - Example: a nonnegative variable is continuous, since it can take on any value in  $[0, +\infty)$
- A decision variable is **integral** if it is restricted to a specified interval of integers
  - Sometimes referred to as **discrete**
  - Special example: **binary** (0 or 1) variables are integral

# CLASSIFICATION OF OPTIMIZATION MODELS

- A function  $f(x_1, \dots, x_n)$  is **linear** (in  $x_1, \dots, x_n$ ) if it is a constant-weighted sum of  $x_1, \dots, x_n$

- Put another way,  $f$  can be written in the form

$$f(x_1, \dots, x_n) = c_1x_1 + c_2x_2 + \dots + c_nx_n$$

where  $c_1, \dots, c_n$  are constants

- Otherwise, a function is **nonlinear**
- An objective function can be linear or nonlinear

# CLASSIFICATION OF OPTIMIZATION MODELS

$$f(x_1, x_2, x_3) = 9x_1 - 17x_3$$

linear

$$f(x_1, x_2, x_3) = \frac{5}{x_1} + 3x_2 - 6x_3$$

nonlinear

$$f(x_1, x_2, x_3) = \sum_{j=1}^3 x_j$$

linear

$$f(x_1, x_2, x_3) = \frac{x_1 - x_2}{x_2 + x_3}$$

nonlinear

$$f(x_1, x_2, x_3) = x_1x_2 + 3x_3$$

nonlinear

# CLASSIFICATION OF OPTIMIZATION MODELS

- A constraint can be written in the form

$$g(x_1, \dots, x_n) \left\{ \begin{array}{c} \leq \\ = \\ \geq \end{array} \right\} b$$

where  $g(x_1, \dots, x_n)$  is a function of the decision variables  $x_1, \dots, x_n$  and  $b$  is a specified constant

- The Constraint is
  - **linear** if  $g(x_1, \dots, x_n)$  is linear in  $x_1, \dots, x_n$
  - **nonlinear** if  $g(x_1, \dots, x_n)$  is nonlinear in  $x_1, \dots, x_n$
- We don't allow strict inequalities ( $<$  or  $>$ ) in the optimization models we study

# CLASSIFICATION OF OPTIMIZATION MODELS

- An optimization model is a **linear program (LP)** if
  - the decision variables are continuous,
  - the objective function is linear, and
  - the constraints are linear
- An optimization model is a **nonlinear program (NLP)** if
  - the decision variables are continuous
  - the objective function is nonlinear or at least one of the constraints is nonlinear



# CLASSIFICATION OF OPTIMIZATION MODELS

$$\min \quad 12z_1 + 4z_3$$

$$\text{s.t.} \quad z_1 z_2 z_3 = 1$$

$$z_1, z_2 \geq 0$$

nonlinear

$$\max \quad 3z_1 + 14z_2 + 7z_3$$

$$\text{s.t.} \quad 10z_1 + 5z_2 \leq 25 - 18z_3$$

$$z_j \geq 0 \quad \text{for } j = 1, \dots, 3$$

linear

$$\min \quad 7z_1 z_2 + 4z_1 z_3 + 2z_1 z_3$$

$$\text{s.t.} \quad \sum_{j=1}^3 z_j = 2$$

$$z_j \leq 0 \quad \text{for } j = 1, \dots, 3$$

nonlinear

# CLASSIFICATION OF OPTIMIZATION MODELS

- An optimization model is an **integer program** if at least one of its decision variables is integral
- Finer classification:
  - If all variables are integral  $\rightarrow$  **pure integer program**
  - Otherwise  $\rightarrow$  **mixed integer program**

# CLASSIFICATION OF OPTIMIZATION MODELS

- An optimization model is an (mixed) integer linear program (MILP/ILP) if
  - (At least one/All) of its decision variables is integral,
  - the objective function is linear, and
  - the constraints are linear
- An integer program is an integer nonlinear program (INLP) if
  - at least one of its decision variables is integral,
  - the objective function is nonlinear *or* at least one of the constraints are nonlinear

# CLASSIFICATION OF OPTIMIZATION MODELS

$$\begin{array}{ll}\min & 3w_1 + 14w_2 - w_3 \\ \text{s.t.} & w_1 w_2 \leq 1 \\ & w_1 + w_2 + w_3 = 10 \\ & w_j \geq 0 \quad \text{for } j = 1, \dots, 3 \\ & w_1 \text{ integer}\end{array}$$

(mixed) integer nonlinear

$$\begin{array}{ll}\max & 19w_1 \\ \text{s.t.} & w_1 \leq w_2 \\ & w_2 + w_3 = 10 + w_1 \\ & w_2 \geq 1, w_3 \geq 1 \\ & w_1 \geq 0\end{array}$$

linear

# CLASSIFICATION OF OPTIMIZATION MODELS

- In general:
  - linear constraints and objective functions are preferred to nonlinear ones
  - continuous variables are preferred to integral variables
- Rule of thumb: when the optimal solution has values that are likely to be large enough so that fractions have no practical importance, using continuous variables is generally preferred.
- For example:
  - 2,000,000 pounds of rice vs 2,000,001 pounds of rice
  - 12 aircraft vs 13 aircraft
- Do **not** follow this rule in the exam/mandatory assignment unless it is clearly mentioned!

# CLASSIFICATION OF OPTIMIZATION MODELS

- Sometimes, "tricks of the trade" can be used to convert certain nonlinear programs into linear programs
- Some nonlinear programs can be solved "as quickly" as a linear program
- Some integer linear programs can be solved "as quickly" as a linear program
- Linear programs are "the holy grail"

# CLASSIFICATION OF OPTIMIZATION MODELS

- Classification of optimization models via characteristics of variables, constraints, and objective functions
  - linear programs
  - nonlinear programs
  - (pure and mixed) integer linear programs
  - (pure and mixed) integer nonlinear programs
- Linear better than nonlinear
- Continuous better than integral

# MODEL COMPLEXITY

- Model size
  - Number of variables
  - Number of constraints
- Mathematical formulations
  - Linear/ Nonlinear formulations
- Variables
  - Continuous/ Discrete
- Data
  - Deterministic/ Stochastic
  - Aggregated / Disaggregated



# A GOOD MODEL

- Ease of understanding the model
- Ease of detecting errors
- Ease of computing the solution
- Units of measurement / scaling
- Modelling languages/ Solvers

**There is no recipe...**

**ASSIGNMENT #2:**  
**AMPL BOOK**  
**CHAPTER 2. EXERCISES (1-8)**

# LECTURE #5:

## LP MODELS

