

60016 OPERATIONS RESEARCH

Introduction to Operations Research

Autumn Term - 2020-21

Netiquette

- ▶ The meeting is recorded, slides and videos will be shared
- ▶ Please mute your mic and switch off your webcam
- ▶ For questions, we will use live Q&A and Piazza, rather than the Teams chat.
- ▶ To ask a question, click on the "Raise your hand" button
 - ▶ Wait to be called by the lecturer
 - ▶ At times, it may take a few slides
 - ▶ Do not forget to "Lower hand" afterwards
- ▶ If the lecturer connection drops, please wait for him to rejoin.
- ▶ If uncomfortable to use mic/speak up, please use Piazza, we will come back later.

Course Information

- ▶ Lecturers:
 - ▶ [Giuliano Casale](mailto:g.casale@imperial.ac.uk) (g.casale@imperial.ac.uk)
 - ▶ [Dario Paccagnan](mailto:d.paccagnan@imperial.ac.uk) (d.paccagnan@imperial.ac.uk)
- ▶ Logistics for the first half:
 - ▶ Lectures: Mon 10:00-12:00
Fri 10:00-11:00
 - ▶ Tutorial: Fri 11:00-12:00
 - ▶ Changes of schedule will be announced in due course, if needed.
 - ▶ Tutorials will develop Q&A on tutorial sheets and exercises.
- ▶ Materials website:
 - ▶ [Basic Linear Algebra](#) refresher
 - ▶ Information for external students
- ▶ Other teaching aids:
 - ▶ Everything on Panopto
 - ▶ Course forum on Piazza
- ▶ One [assessed coursework](#) (CATE), please collaborate!

Some Books (Optional Readings)

- ▶ *F.Hillier & J.Lieberman:*
Introduction to Operations Research.
- ▶ *H.Taha:*
Operations Research.
- ▶ *D.G.Luenberger & Y.Ye:*
Linear and Nonlinear Programming.
- ▶ *W. Winston:*
Operations Research: Applications and Algorithms.

What is Operations Research?

- ▶ OR is a **multidisciplinary branch of mathematics** involving
 - ▶ mathematical modelling
 - ▶ mathematical optimisation
 - ▶ statistical analysisin order to find "good" solutions for **complex decision problems**.
- ▶ Typical **objectives** in OR are:
 - ▶ maximise **profit**
 - ▶ minimise **cost**
 - ▶ minimise **risk**
 - ▶ minimise **completion time**
 - ▶ maximise **efficiency**
 - ▶ etc.

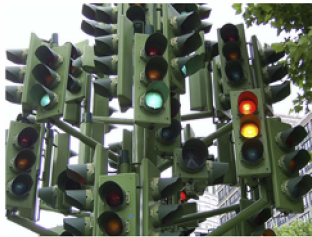
Scope of OR

OR techniques are ubiquitous in operations management, industrial engineering, economics and finance, ICT management, machine learning, health care, security, etc.

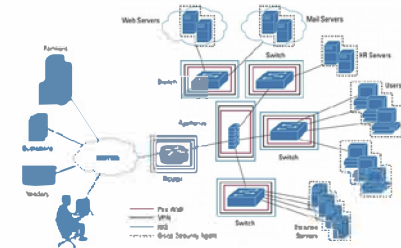
Designing the layout of a factory



road traffic management



Constructing a telecommunication network



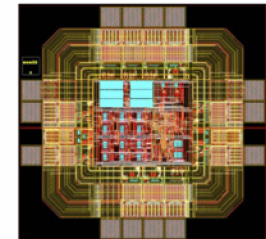
Determining the routes of city buses



scheduling project tasks



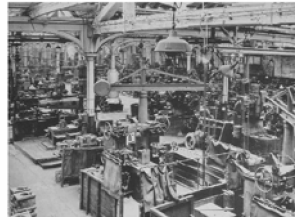
Designing the layout of a computer chip



Financial planning

History of OR

19th century:
Industrial Revolution



Efficiency of
production processes

World War II:
Birth of Modern OR



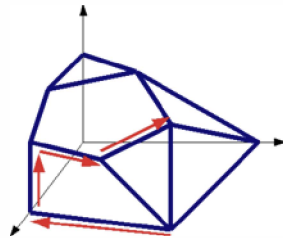
Optimal design of
convoy system



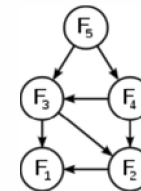
Where to add armour
in RAF bombers?

Allocate scarce resources to
various military operations
in an **effective** manner

1947
Simplex Algorithm



1953
Dynamic Programming



1970's
Personal Computers



Industry-size problems
can be solved efficiently

Phases of an OR Study

Hillier and Lieberman p. 8:

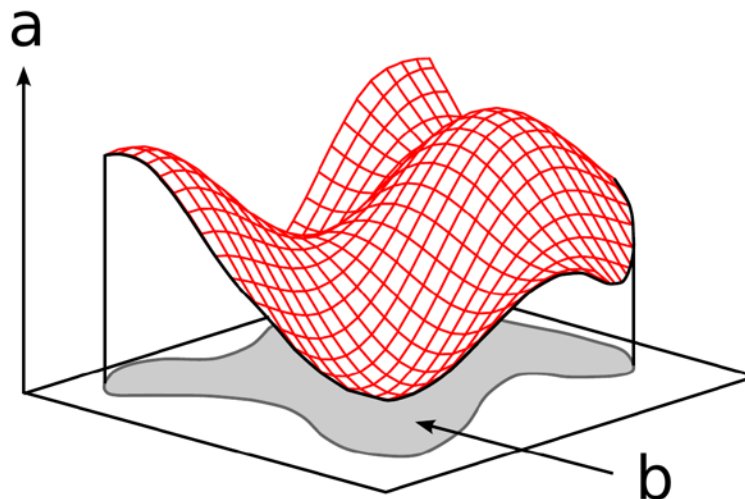
1. Define the problem of interest and **gather relevant data**.
2. Formulate a **mathematical model** to represent the problem.
3. Develop a **computer-based procedure** for deriving solutions to the problem from the model.
4. **Test** the model and **refine** as needed.
5. **Prepare for** the ongoing **application** of the model as prescribed by management.
6. **Implement**.

Course focuses on phases 2 and 3.

Mathematical Programming

OR solves mathematical programming models:

$$\begin{array}{ll}\underset{x}{\text{minimise}} & z = f(x) \\ \text{subject to} & x \in \mathcal{X},\end{array}$$



where

- ▶ $x \in \mathbb{R}^n$ are the **decision variables**
- ▶ $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is the **objective function** (e.g., cost)
- ▶ $\mathcal{X} \subseteq \mathbb{R}^n$ is the **feasible set** (set of admissible decisions)
- ▶ any vector x that minimises f is an **optimal solution** of the program and is denoted by x^*
- ▶ $z^* = f(x^*)$ is the **optimal value** achieved by x^*

Topics of this Course

- ▶ Linear Programming
- ▶ Integer Programming
- ▶ Duality and sensitivity analysis
- ▶ Game Theory
- ▶ Modelling, skill mostly acquired through:
 - ▶ Exercises: verify your solutions with the GNU GLPK solver.
 - ▶ Case studies: in tutorials (one exam question about this).

Linear Programming

- ▶ A **Linear program (LP)** is a mathematical program that
 - ▶ **optimises** (maximises or minimises) a **linear objective function**
 - ▶ over a **feasible set** described by **linear** equality and/or inequality constraints.
- ▶ Optimal decision tool
- ▶ Widely adopted, many success stories (see Hillier & Lieberman)
- ▶ LPs are much simpler to cope with than non-linear programs
 - ▶ *CO477 - Computational Optimisation* deals with the theory of non-linear optimization.

LP Running Case: Example 1

Resource Allocation Problem: A manufacturer produces A (acid) and C (caustic) and wants to decide a production plan.

Ingredients used for producing A and C are: X (e.g., a sulphate) and Y (e.g., sodium).

- ▶ Each ton of A requires: 2ton of X; 1ton of Y
- ▶ Each ton of C requires: 1ton of X ; 3ton of Y
- ▶ Supply of X limited to: 11ton/week
- ▶ Supply of Y limited to: 18ton/week
- ▶ A sells for: £1000/ton
- ▶ C sells for: £1000/ton
- ▶ Market research: max 4 tons of A/week can be sold.

Maximize weekly value of sales of A and C.

Example 1 (Modelling)

How much A and C to produce?

⇒ Formulate a mathematical programming model!

► Decision variables

- ▶ x_1 = weekly production of A (in tons)
- ▶ x_2 = weekly production of C (in tons)

► Objective function

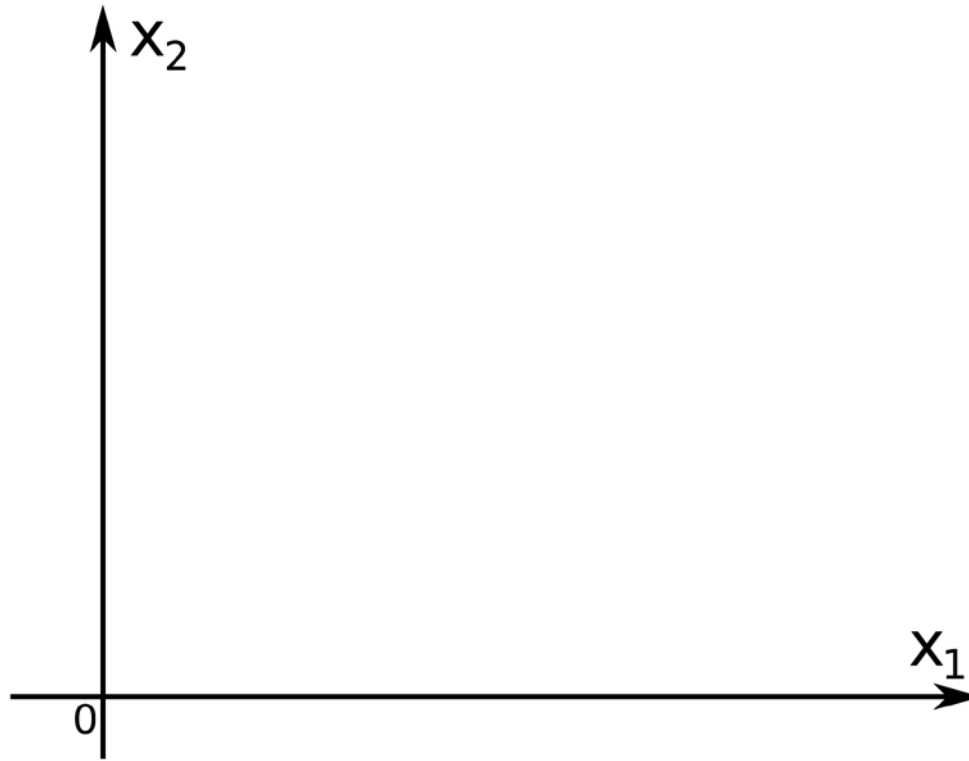
- ▶ $z = f(x_1, x_2)$ = weekly profit (in 1000 £)

► Feasible set

- ▶ \mathcal{X} = set of all implementable/admissible production plans
 $x = (x_1, x_2)$
- ▶ e.g., $x = (27, 2)$ is not possible (not enough supply!)

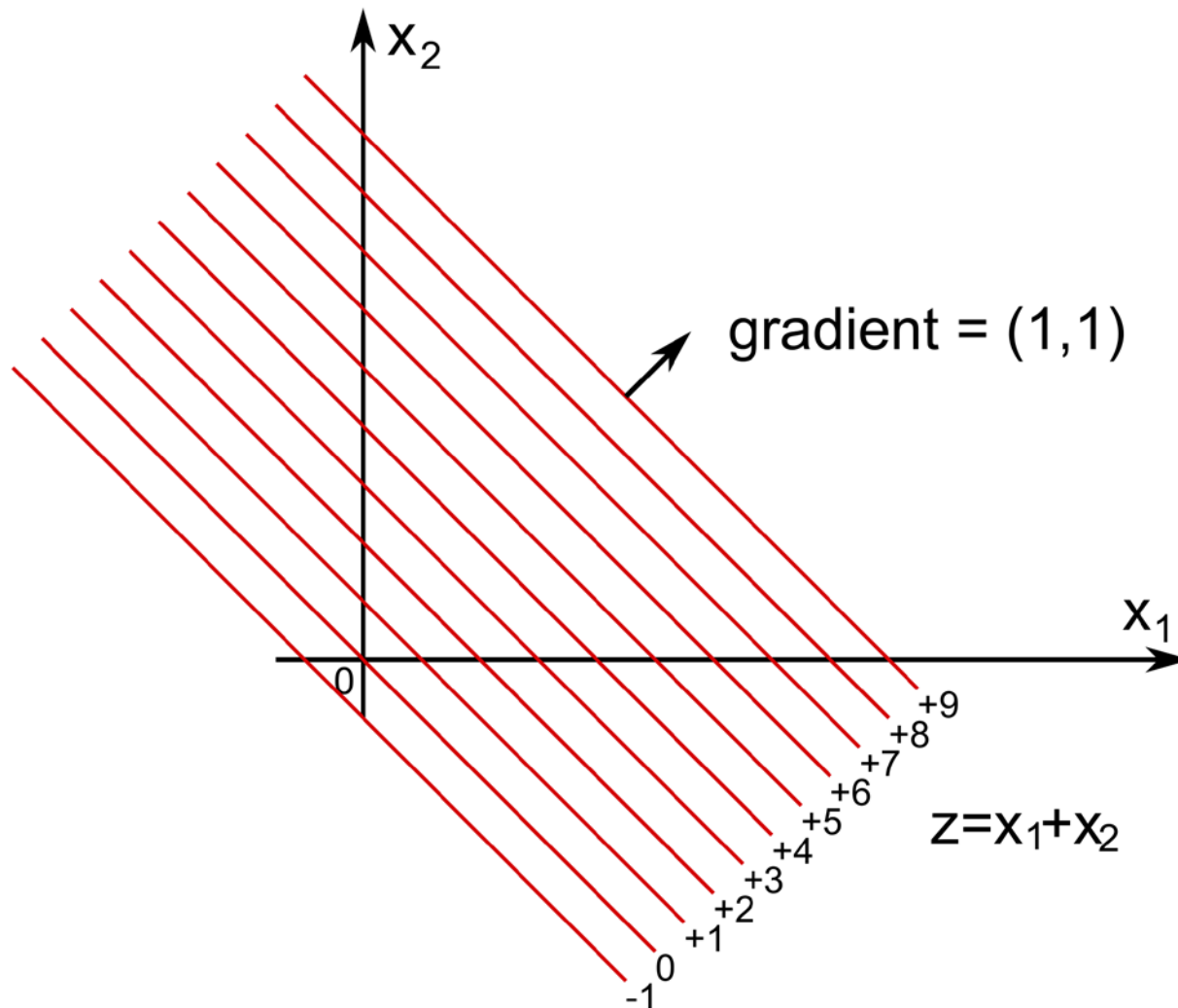
Example 1 (Decision Variables)

A **production plan** is representable as $x = (x_1, x_2)$



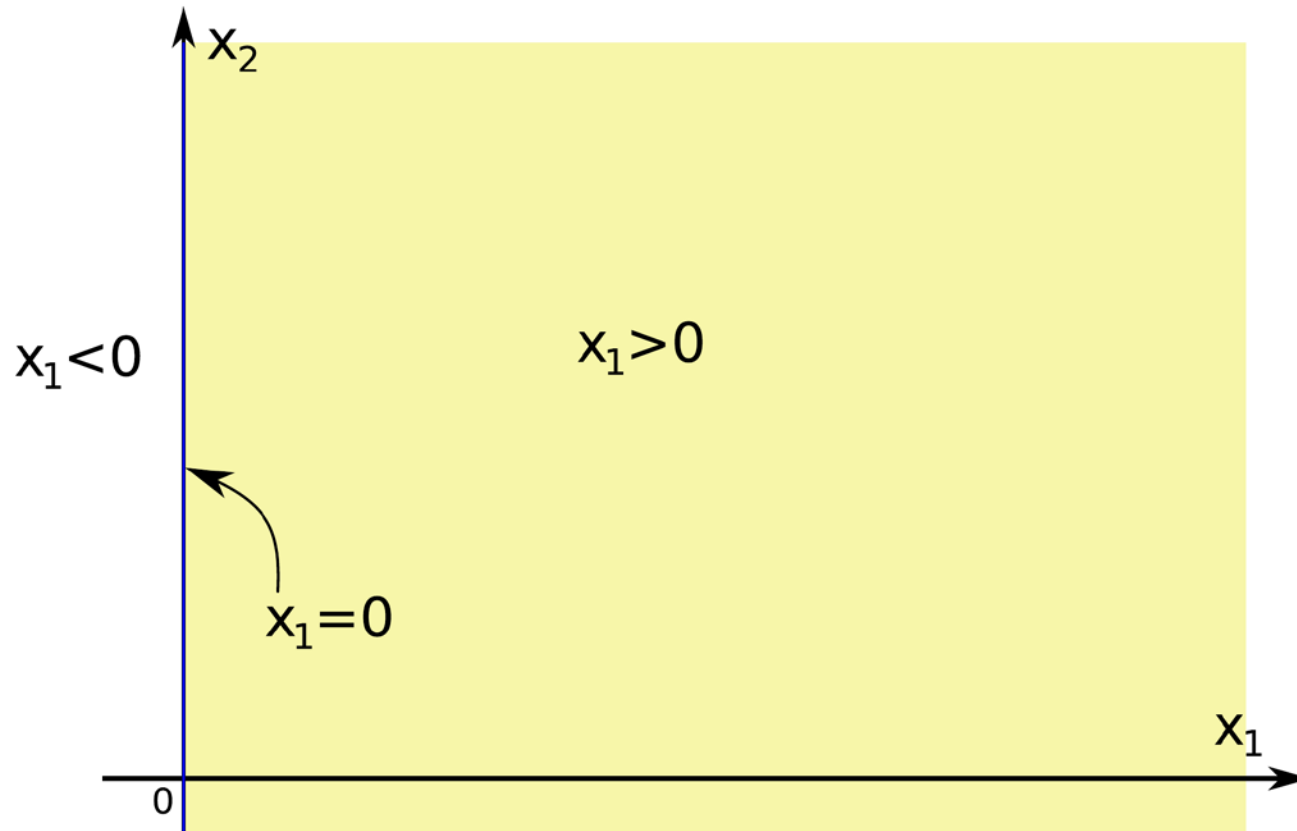
Example 1 (Objective Function)

Profit: $z = x_1 + x_2$ (in 1000 £)



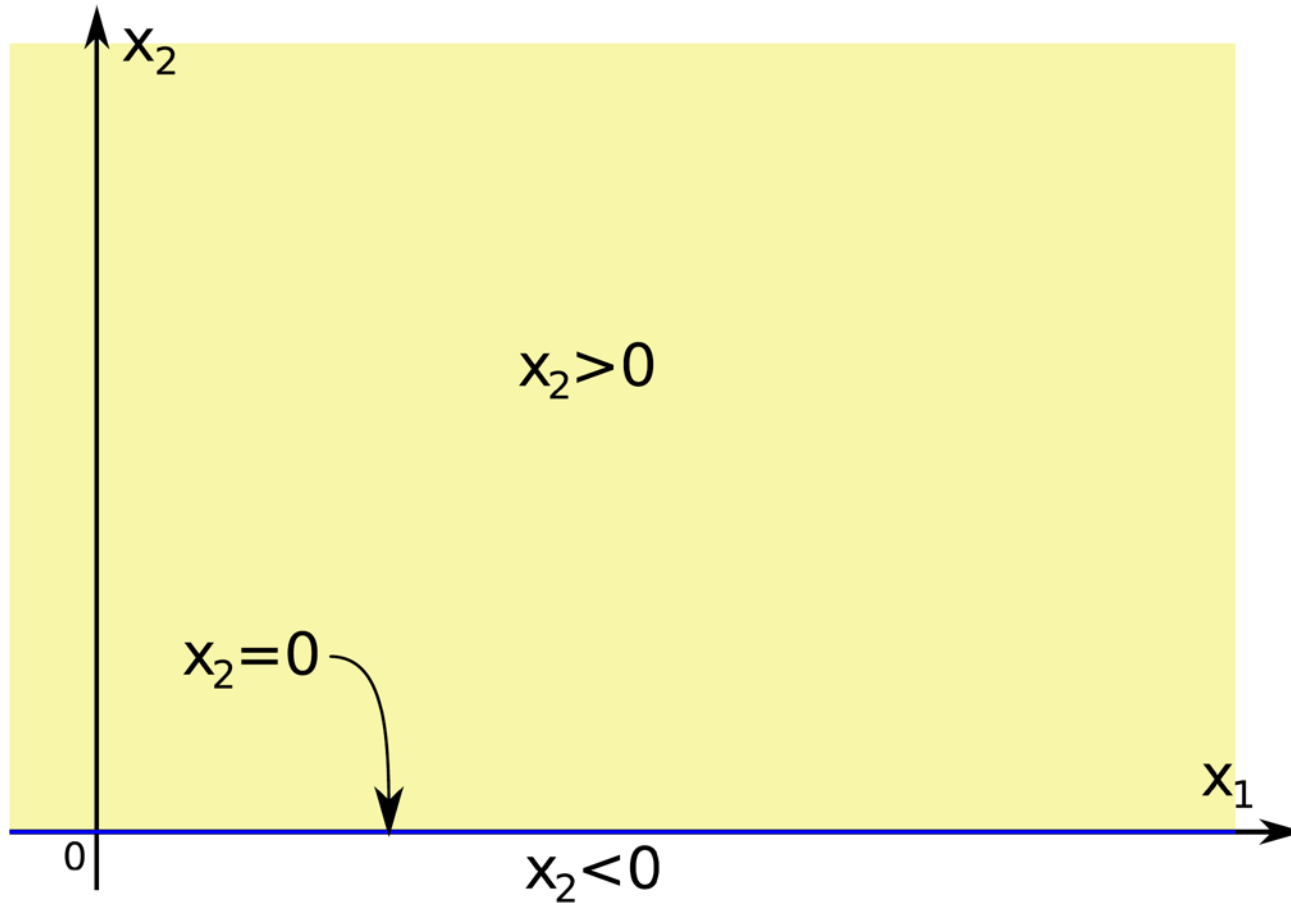
Example 1 (Feasible Set)

Amount of A produced is **non-negative**: $x_1 \geq 0$



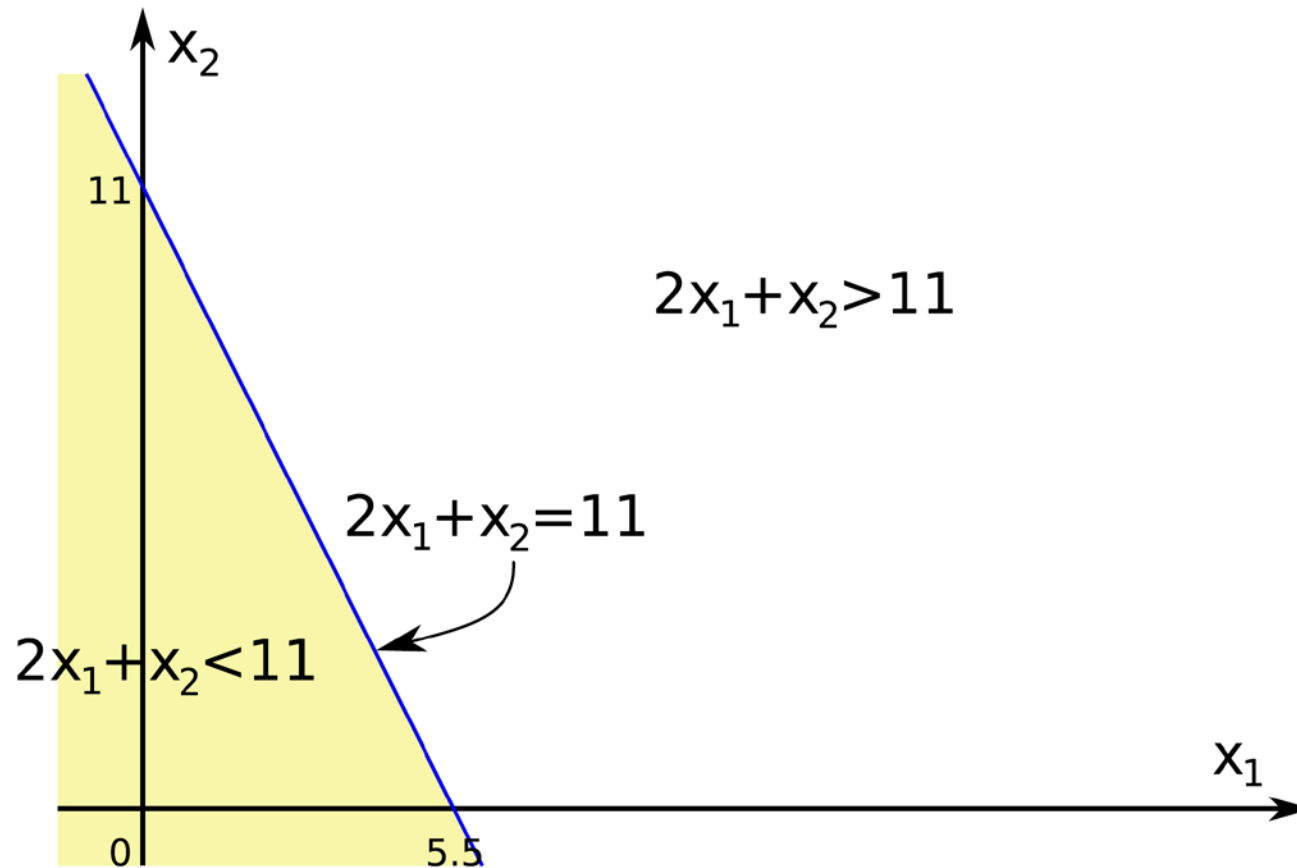
Example 1 (Feasible Set)

Amount of C produced is **non-negative**: $x_2 \geq 0$



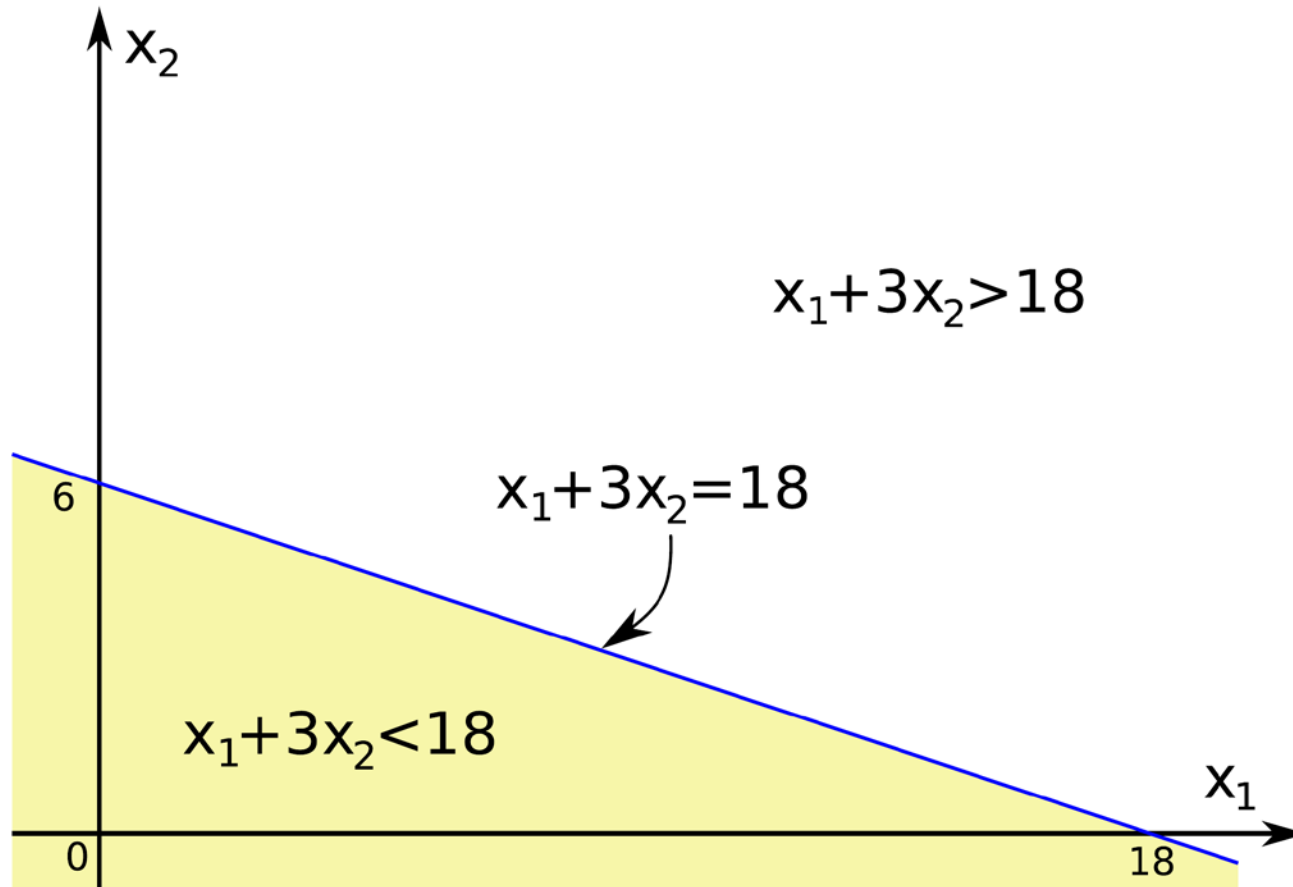
Example 1 (Feasible Set)

x_1 tons of A & x_2 tons of C require $2x_1 + x_2$ ton of X
X is limited to 11ton/week: $2x_1 + x_2 \leq 11$



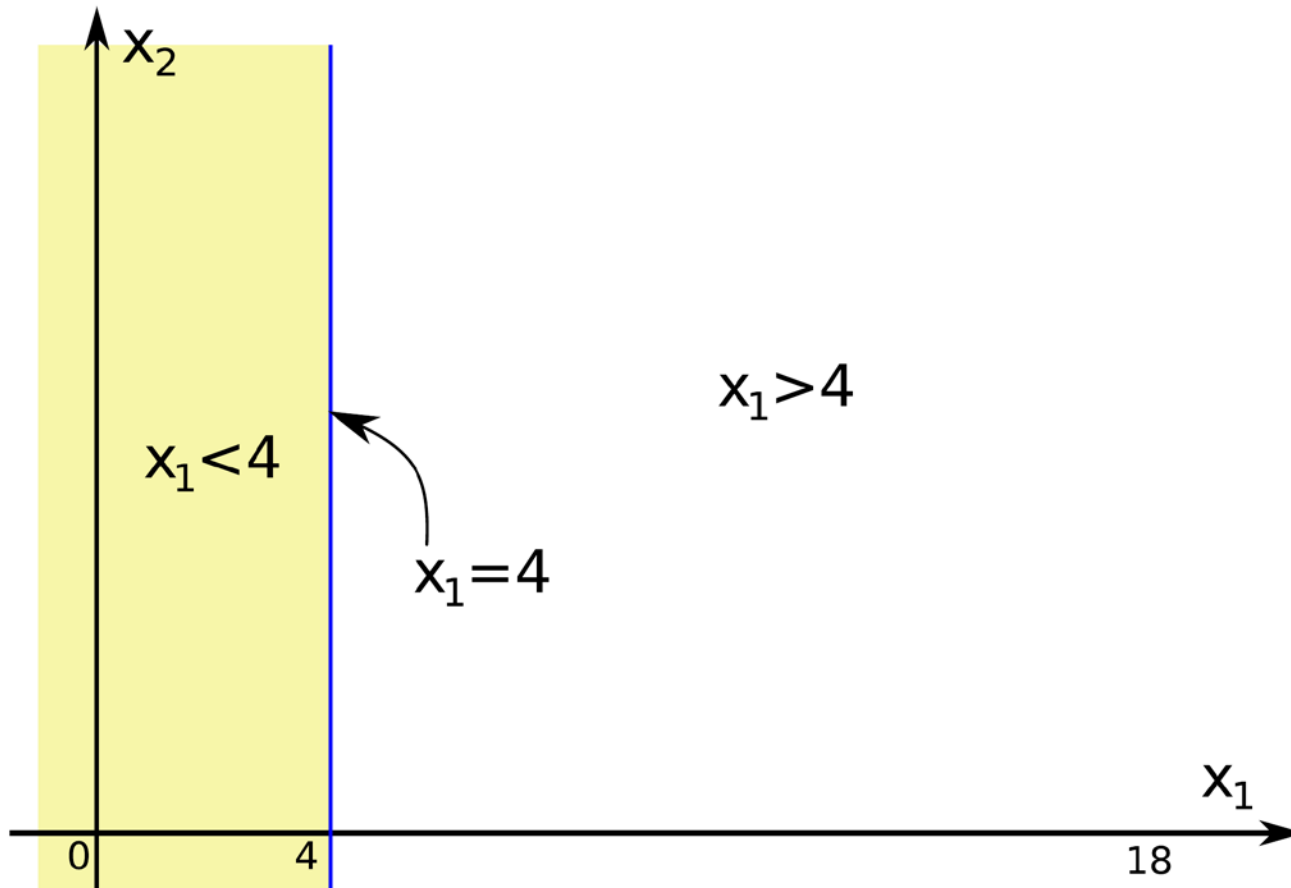
Example 1 (Feasible Set)

x_1 tons of A & x_2 tons of C require $x_1 + 3x_2$ ton of Y
Y is limited to 18ton/week: $x_1 + 3x_2 \leq 18$



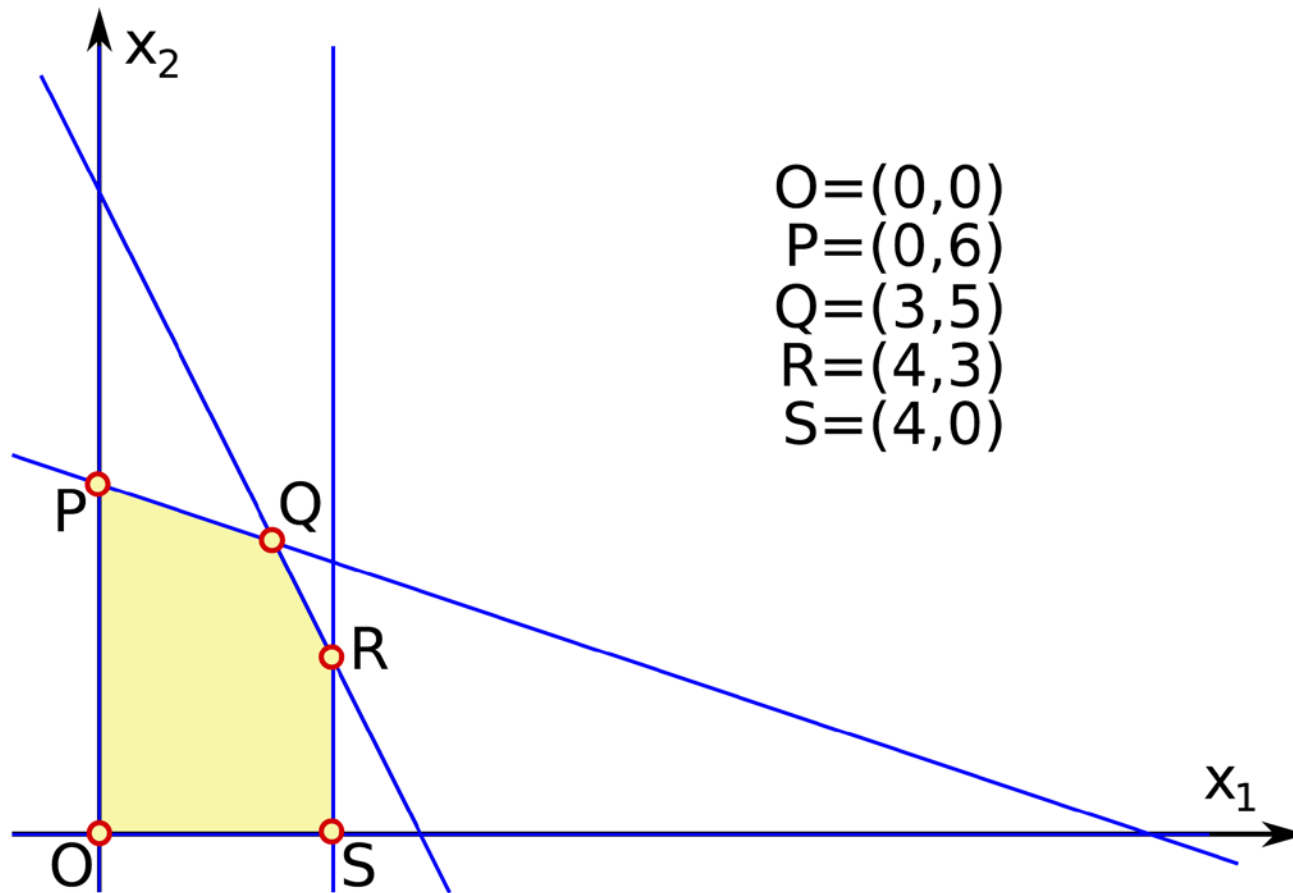
Example 1 (Feasible Set)

Cannot sell more than 4 tons of A/week: $x_1 \leq 4$



Example 1 (Feasible Set)

To obtain the overall feasible set,
intersect the feasible sets of all individual constraints



Example 1 (Feasible Set)

- ▶ The feasible set is a **convex polygon**
- ▶ The corner points O,P,Q,R,S of the feasible set are termed **vertices**
- ▶ Each vertex is given by the **intersection of two blue lines**
 - ▶ its coordinates can be computed by jointly solving the two linear equations defining the blue lines
- ▶ We obtain $O=(0,0)$, $P=(0,6)$, $Q=(3,5)$, $R=(4,3)$, $S=(4,0)$

Example 1 (Summary)

The **best production plan** is obtained by solving the following mathematical problem:

maximise $z = x_1 + x_2$: **objective function**

subject to $2x_1 + x_2 \leq 11$: **constraint on availability of X**

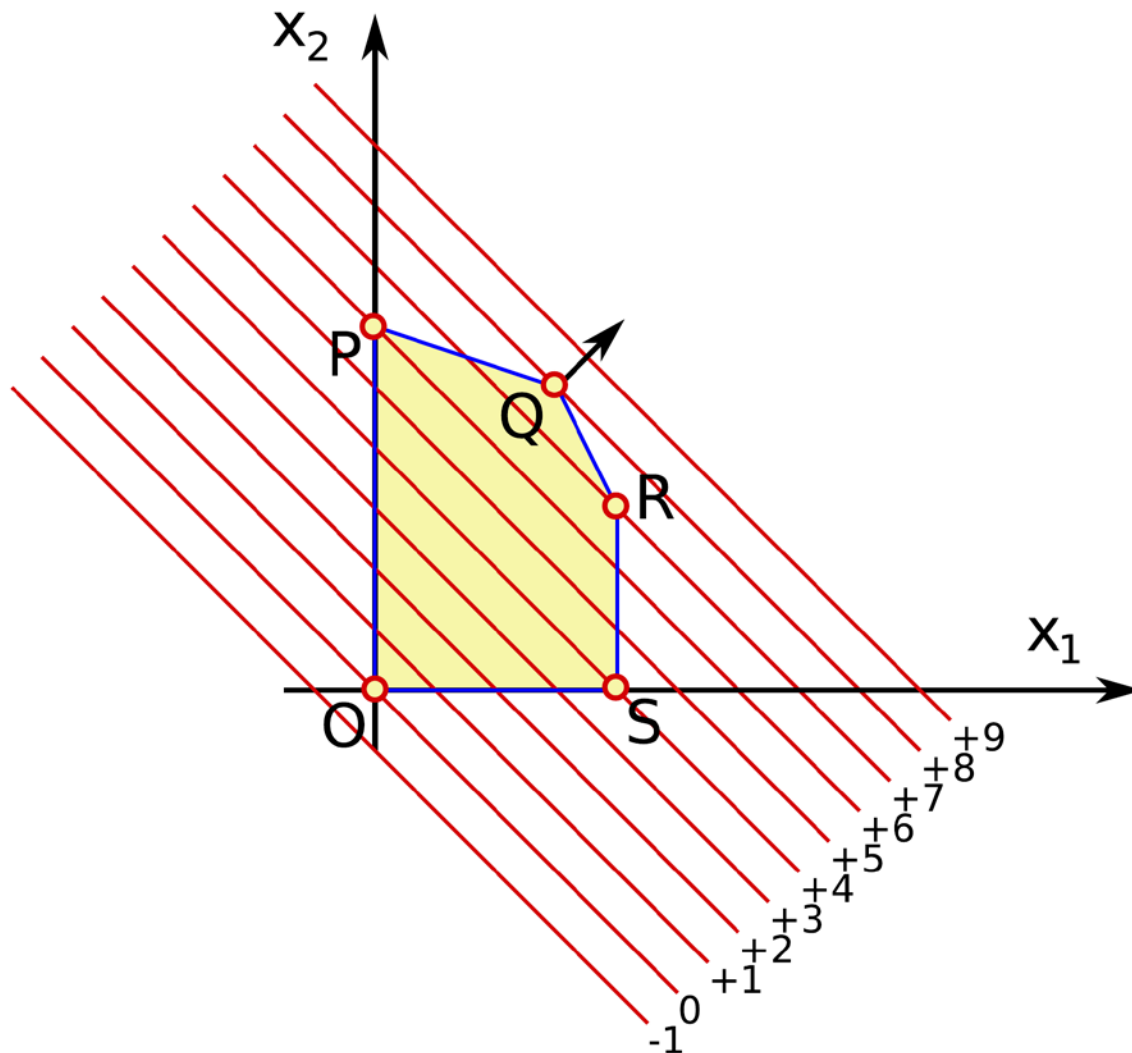
$x_1 + 3x_2 \leq 18$: **constraint on availability of Y**

$x_1 \leq 4$: **constraint on demand of A**

$x_1, x_2 \geq 0$: **non-negativity constraints**

This is a **linear program**.

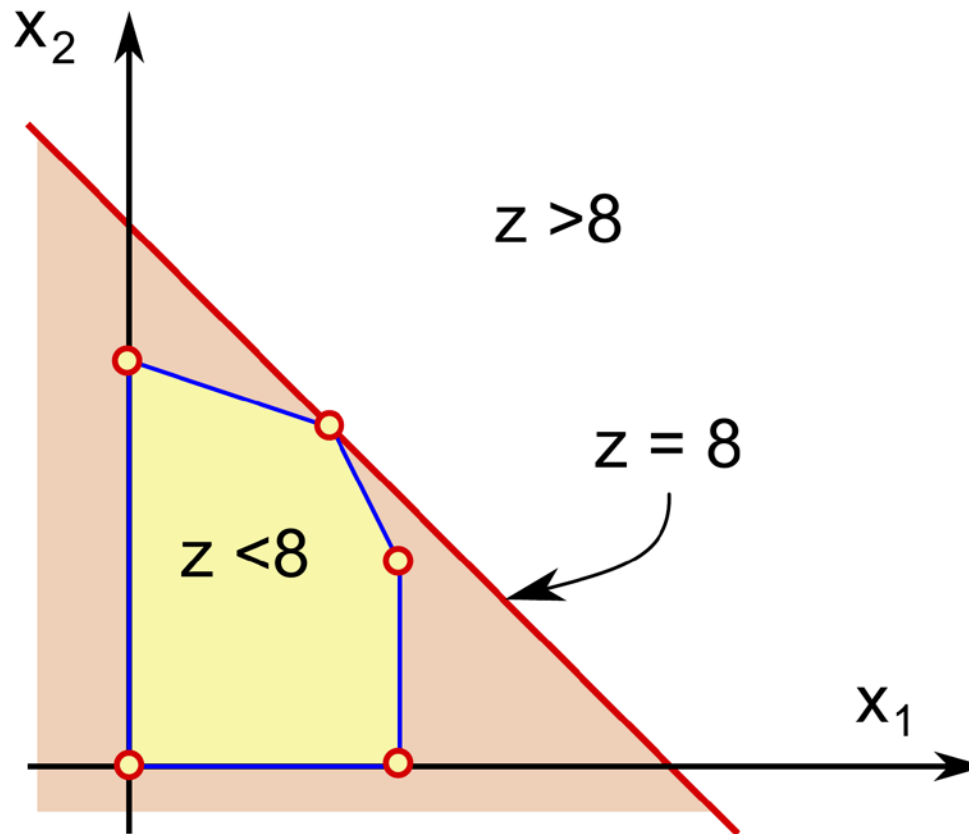
Example 1 (Graphical Solution)



Example 1 (Graphical Solution)

All feasible points satisfy $z \leq 8$

Q is the only feasible point (x_1, x_2) with $z = x_1 + x_2 = 8$



Linear Programming

- ▶ Assume feasible set \mathcal{X} bounded and nonempty
- ▶ We can prove that LPs have an optimal vertex solution
 - ▶ LPs may be solved by inspecting all vertices, but ...
 - ▶ The number of vertices grows exponentially with the number of constraints and variables in the LP
- ▶ How to program a computer to efficiently solve LPs?
 - ▶ Simplex Algorithm finds an optimal vertex
 - ▶ Vertices inspected by the Simplex algorithm are often a small subset of the total
 - ▶ What made the Simplex algorithm famous is that it works well on most instances

Variants of Example 1

- ▶ Minimise $z = 3x_1 - x_2$ over feasible set of Example 1

Examine the objective function at all vertices:

O=(0,0)	P=(0,6)	Q=(3,5)	R=(4,3)	S=(4,0)
0	-6	4	9	12

\Rightarrow P: $x_1 = 0, x_2 = 6$ is optimal.

- ▶ Maximise $z = 2x_1 + x_2$ over feasible set of Example 1:

Any point on the line segment QR is optimal.

\Rightarrow points other than vertices can be optimal, but there is at least one optimal vertex

Variants of Example 1

- ▶ Set a minimum weekly production goal of 7 tons of C

We add a new constraint $x_1 \geq 7$. Then the feasible set becomes **empty**, because we previously imposed $x_1 \leq 4$

\Rightarrow the LP is **infeasible**

- ▶ Remove constraints on availability of X and Y

Objective function can now grow to $+\infty$ on the feasible set.
There is no maximum!

\Rightarrow the LP is **unbounded**