

Outline

- 1 Motivation
- 2 Key Geometric Interpretation
- 3 LP Standard Form

For additional details, see Rardin (1998), Chapter 5.1-5.2

Linear Programming (LP): Principles and Concepts

Benoît Chachuat <benoit@mcmaster.ca>

McMaster University
Department of Chemical Engineering

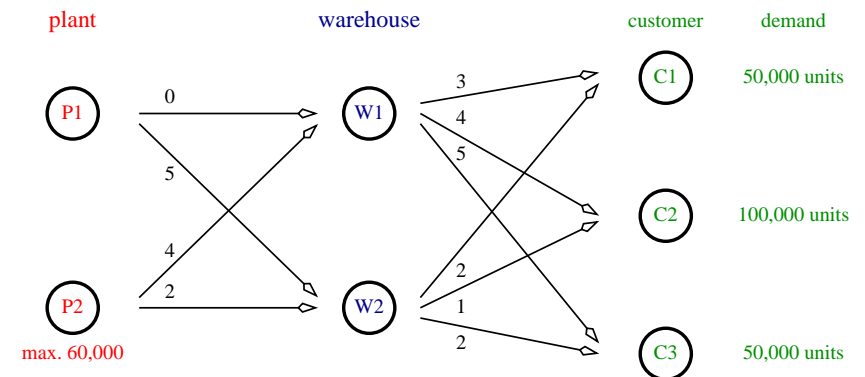
ChE 4G03: Optimization in Chemical Engineering

Linear Programming

Multiple Choice Question: Why Study Linear Programming?

- 1 The prof cannot formulate **nonlinear** models
- 2 Linear programming is the **most frequently** used optimization method
- 3 Linear programming was developed in the **1940's**
- 4 Formulation and solution knowledge will also **help** when addressing nonlinear models
- 5 Excellent **software** is available

Need for Optimization to Get the Best Solution



- The cost of manufacturing is the same
- The objective is thus to minimize the transportation cost (using values in the figure)

What is the Best Plant?

Learning Goals

Attitudes

- An optimal solution is much better than an answer
- Numbers without understanding are useless

Skills

- Translate a complex problem into a mathematical formulation
- Solve an optimization problem!
- Communicate optimization results in “engineering terms”

Knowledge

- When/how formulate LP models
- Analyze the LP solution sensitivity and diagnose “weird events”

Linear Programs

Definition

An optimization model is a **linear program** (or **LP**) if it has:

- 1 continuous variables;
- 2 a single linear objective function; and
- 3 any linear equality or inequality constraints

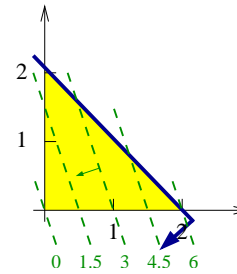
Standard notation for LP:

$$\begin{aligned} \min_{\mathbf{x}} z &\triangleq \mathbf{c}^T \mathbf{x} & x_j &\triangleq j\text{th decision variable} \\ \text{s.t. } \mathbf{A}_h \mathbf{x} &= \mathbf{b}_h & c_j &\triangleq \text{objective function coefficient of } x_j \\ \mathbf{A}_g \mathbf{x} &\leq \mathbf{b}_g & a_{i,j} &\triangleq \text{constraint coefficient of } x_j \text{ in the } i\text{th main constraint} \\ \mathbf{x}_{\min} &\leq \mathbf{x} \leq \mathbf{x}_{\max} & b_i &\triangleq \text{right-hand side (RHS) constant term of main constraint } i \\ & & m &\triangleq \text{number of main constraints} \\ & & n &\triangleq \text{number of decision variables} \end{aligned}$$

Learning Goals

Challenge

Understand both the **geometric interpretation** (excellent understanding) and the **matrix calculations** (for the quantitative results) – At the same time!



$$\begin{pmatrix} a_{1,1} & \cdots & a_{1,m} & a_{1,m+1} & \cdots & a_{1,n} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ a_{m,1} & \cdots & a_{m,m} & a_{m,m+1} & \cdots & a_{m,n} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$$

It Is Worth the Effort!

Interior, Boundary and Extreme Points

Classification of the Points in an LP Feasible Region

A **feasible** solution to a linear program is said to be:

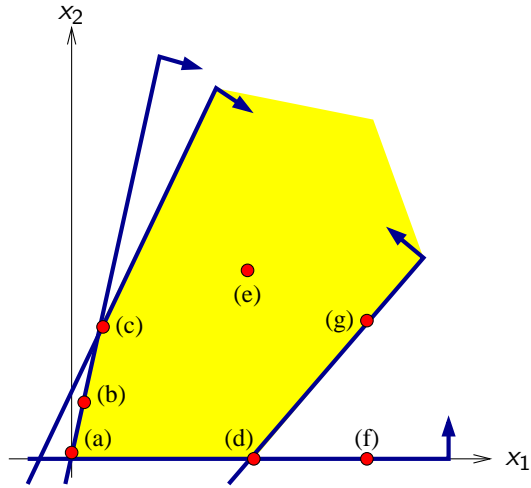
- a **boundary point** if at least one inequality constraint (that can be strict for some other feasible solutions) is satisfied as equality at the given point
- an **interior point** if no such inequalities are active
- an **extreme point** (or a **corner point** or a **vertex**) if every line segment in the feasible region containing it also has that point as an endpoint

Important Remarks About Extreme Points:

- Extreme points are special boundary points so-named because they “stick out” (due to convexity of the feasible region of an LP)
- Every extreme point is determined by a set of constraints that are active *only* at that solution

Classifying Feasible Points

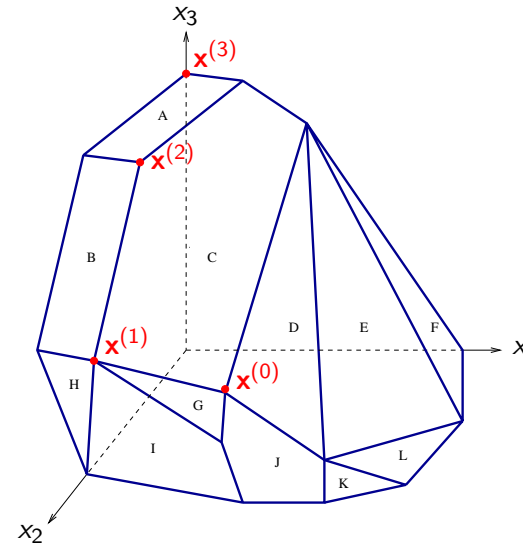
Class Exercise: Classify the labeled solutions as **interior**, **boundary**, and/or **extreme** points of the following LP feasible region:



- (a)
- (b)
- (c)
- (d)
- (e)
- (f)
- (g)

Extreme Points

Class Exercise: List all sets of 3 constraints determining points $x^{(0)}$, $x^{(1)}$, $x^{(2)}$ and $x^{(3)}$



- $x^{(0)}$:
- $x^{(1)}$:
- $x^{(2)}$:
- $x^{(3)}$:

Adjacent Extreme Points and Edges

Adjacent Extreme Points

Two extreme points of an LP feasible region are **adjacent** if they are determined by active constraint sets differing in **only one** element.

Edge

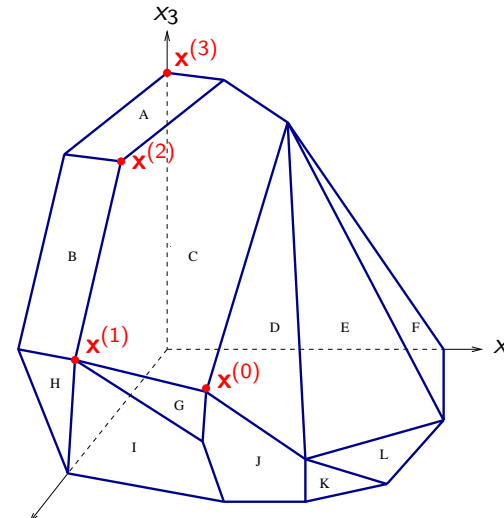
An **edge** of an LP feasible region is a 1-dimensional set of feasible points along a line determined by a collection of active constraints.

Remark:

- Adjacent extreme points are joined by an edge determined by the active constraints the extreme points have in common

Adjacent Extreme Points

Class Exercise: Which points among $x^{(0)}$, $x^{(1)}$, $x^{(2)}$ and $x^{(3)}$ are adjacent extreme points? Which active constraints do these points have in common?

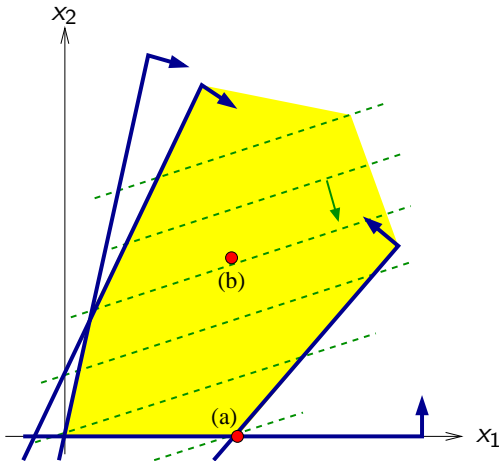


- Adjacent Extreme Points:
- Common Active Constraints:

Optimal Points in LP

Fundamental Result

Every optimal solution to an LP (that has a non-constant objective function) will be a **boundary point** of its feasible region



Class Exercise: Explain why point (b) cannot be optimal? How about point (a)?

Optimal Points in LP

Fundamental Result

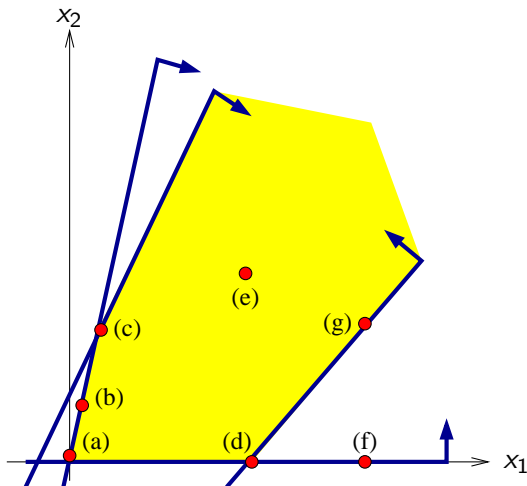
Every optimal solution to an LP (that has a non-constant objective function) will be a **boundary point** of its feasible region

Other Fundamental Properties:

- 1 If an LP has *any* extreme point, then one such point must be an optimal solution
Why?
- 2 If an LP has a *unique* optimal solution, that optimum must occur at an extreme point of its feasible region
Why?
- 3 *Every* local optimum for an LP is a global optimum
Why?

Identifying Optimal Points

Class Exercise: Indicate which of the labeled points can be optimal or uniquely optimal for some objective function:



- (a)
- (b)
- (c)
- (d)
- (e)
- (f)
- (g)

Naive Algorithm

Exploiting the extreme point concept as the foundation for a solution method:

Algorithm:

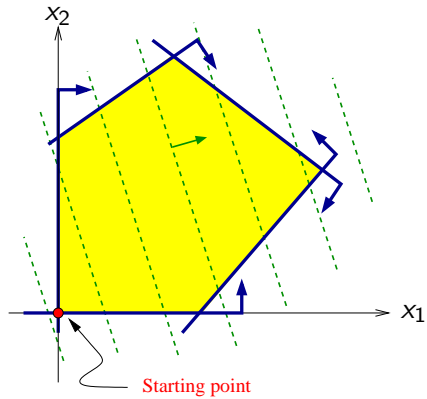
Calculate the objective function value at *all* extreme points, and keep whichever is best!

Application: Consider an LP comprising **10 variables** and **20 inequality constraints**.

Up to how many extreme points exist for this (rather small) LP problem?

Back to the Drawing Board!

Class Exercise: Develop an algorithm that combines the basic numerical approach with the key geometric insight



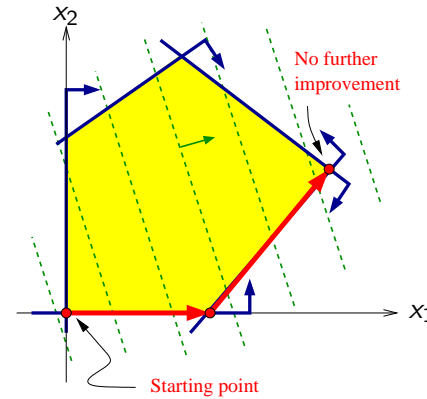
Basic Numerical Optimization:

- Use local information *only*, in an iterative scheme
- Determine a feasible next point that improves the objective value
- check if further improvement is possible: if 'yes', continue; else, stop

Basis of the Simplex Algorithm!

Back to the Drawing Board!

Class Exercise: Develop an algorithm that combines the basic numerical approach with the key geometric insight



One possible approach:

- Consider *only adjacent* extreme points for improvement direction
- Move **along the edge** that yields the greatest rate of improvement
- Move until another extreme point has been reached
- Check if further improvement is possible: if 'yes', continue; else, stop

Basis of the Simplex Algorithm!

LP Standard Form

Definition:

LP in **standard form** have:

- only *equality* main constraints
- only *nonnegative* variables
- objective function and main constraints simplified so that each variable appears at most once (on the left-hand side), and any constant term (possibly zero) appears on the right-hand side

$$\min_{\mathbf{x}} z \triangleq \mathbf{c}^T \mathbf{x}$$

$$\text{s.t. } \mathbf{Ax} = \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

$x_j \triangleq j$ th decision variable
 $c_j \triangleq$ objective function coefficient of x_j
 $a_{i,j} \triangleq$ constraint coefficient of x_j in the i th main constraint
 $b_i \triangleq$ right-hand side (RHS) constant term of main constraint i
 $m \triangleq$ number of main constraints
 $n \triangleq$ number of decision variables

LP Standard Form: Why and How?

Why LP in Standard Form?

-
-

How to Yield LP Standard Form?

- 1 Convert \leq and \geq inequalities to equalities
- 2 Convert nonpositive variables to nonnegative
- 3 Convert unrestricted (URS) variables to nonnegative

LP Standard Form: Some Practice!

Class Exercise:

- 1 Place each of the following LP in standard form
- 2 Identify the m , n , \mathbf{A} , \mathbf{b} and \mathbf{c} of standard matrix representation

$$\begin{array}{ll} \min & 9w_1 + 6w_2 \\ \text{s.t.} & 2w_1 + w_2 \geq 10 \\ & w_1 \leq 50 \\ & w_1 + w_2 = 40 \\ & 100 \geq w_1 + 2w_2 \geq 15 \\ & w_1 \geq 0, w_2 \leq 0 \end{array} \qquad \begin{array}{ll} \max & 15(2x_1 + 8x_2) - 4x_3 \\ \text{s.t.} & 2(10 - x_1) + x_2 + 5(9 - x_3) \geq 10 \\ & x_1 + 2x_2 \leq x_3 \\ & x_1, x_2 \geq 0, x_3 \text{ URS} \end{array}$$