

# Linear Programming, Modelling & Solution

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## Derivation

We have

$$\begin{aligned} f &= \bar{\mathbf{c}}^T \mathbf{x} = \mathbf{c}_B^T \mathbf{x}_B + \mathbf{c}_N^T \mathbf{x}_N \\ \bar{\mathbf{A}}\mathbf{x} &= \mathbf{b} = B\mathbf{x}_B + N\mathbf{x}_N \end{aligned}$$

combining the two gives us:

$$\begin{aligned} f &= \hat{f} + \hat{\mathbf{c}}_N^T \mathbf{x}_N \\ &= \{\mathbf{c}_B^T \hat{\mathbf{b}}\} + \{\mathbf{c}_N - N^T B^{-T} \mathbf{c}_B\}^T \mathbf{x}_N \\ &= \{\mathbf{c}_B^T B^{-1} \mathbf{b}\} + \{\mathbf{c}_N - N^T B^{-T} \mathbf{c}_B\}^T \mathbf{x}_N \end{aligned}$$

where

$$\begin{aligned} \hat{\mathbf{c}}_N &= \mathbf{c}_N - N^T B^{-T} \mathbf{c}_B \\ \hat{f} &= \mathbf{c}_B^T \hat{\mathbf{b}} \end{aligned}$$

$\hat{f}$  is the objective value when  $\mathbf{x}_N = \mathbf{0}$  (so  $\mathbf{x}_B = \hat{\mathbf{b}}$ )

## LP Theory: Analysis of LPs in standard form

### Definition: Feasible Vertex I

The vertex of the feasible region  $K$  is a point  $\mathbf{x} \in K$  which does not lie strictly within any line segment joining two points in  $K$ .

### Theorem 1: A unique optimal solution is a vertex

If an LP has a unique optimal solution then it is a vertex.

### Theorem 2: Non-unique optimal solution at a vertex

If an LP has a non-unique optimal solution then there is an optimal solution at a vertex.

### Definition: Feasible Vertex II

A vertex of the feasible region  $K$  is a point  $\mathbf{x} \in K$  with

- $n$  zero components
- $m$  non-negative components uniquely defined by  $\bar{\mathbf{A}}\mathbf{x} = \mathbf{b}$

## LP Theory: Basic feasible solutions and optimality conditions for LP problems

### Definition: A basic solution

The point  $\mathbf{x} \in \mathbb{R}^{n+m}$  is a **basic solution** of an LP problem in standard form if there is a **partition** of  $\{1, 2, \dots, n+m\}$  into

- A set  $\mathcal{N}$  of  $n$  indices of **non-basic variables** with value zero
- A set  $\mathcal{B}$  of  $m$  indices of **basic variables** whose values are then uniquely defined by the  $m$  equations.

## Why can't we just work with vertices?

- A vertex is also a point with  $n$  zero components and  $m$  non-negative components uniquely defined by  $\overline{A}\mathbf{x} = \mathbf{b}$
- A **degenerate** vertex has more than  $n$  zero components
  - More than one partition of  $\{1, 2, \dots, n + m\}$  into sets  $\mathcal{N}$  and  $\mathcal{B}$  is possible
  - There may be more than one basic solution at a degenerate vertex.

## Theorem 3: A sufficient optimality condition for LP problems

A point  $\mathbf{x} \in K$  is an optimal solution of an LP problem if it is a basic feasible solution with non-positive **reduced costs**  $\widehat{\mathbf{c}}_N = \mathbf{c}_N - N^T B^{-T} \mathbf{c}_B \leq 0$

## The simplex algorithm

### Description of the simplex algorithm

1. If the reduced costs are non-positive then **stop**  
**The solution is optimal**
2. Determine the non-basic variable  $x_{q'}$  with the most positive reduced cost
3. Determine the feasible direction  $\mathbf{d}$  when  $x_{q'}$  is increased from zero
4. If no basic variable is zeroed on  $\mathbf{x} + \alpha \mathbf{d}$  then **stop**  
**The LP is unbounded**
5. Determine the first basic variable  $x_{p'}$  to be zeroed on  $\mathbf{x} + \alpha \mathbf{d}$
6. Make  $x_{p'}$  non-basic and  $x_{q'}$  basic
7. Go to 1

### Definition of the simplex algorithm

Given a basic feasible solution  $\mathbf{x}$  with  $\mathcal{B}$  and  $\mathcal{N}$

1. If  $\widehat{\mathbf{c}}_N \leq \mathbf{0}$  then **stop** (with  $\widehat{\mathbf{c}}_N = \mathbf{c}_N - N^T B^{-T} \mathbf{c}_B$ )  
**The solution is optimal**
2. Determine the index  $q' \in \mathcal{N}$  of the variable  $x_{q'}$  with the most positive reduced cost  $\widehat{c}_q$   
 $q'$  is the  $q$ th entry in  $\mathcal{N}$ .
3. Let  $\widehat{\mathbf{a}}_q = B^{-1} \mathbf{a}_q$ , where  $\mathbf{a}_q$  is column  $q$  of  $N$
4. If  $\widehat{\mathbf{a}}_q \leq \mathbf{0}$  then **stop**  
**The LP is unbounded**
5. Determine the index  $p' \in \mathcal{B}$  of the variable  $x_{p'}$  corresponding to  $p = \operatorname{argmin}_{i=1, \widehat{a}_{iq} > 0}^m \frac{\widehat{b}_i}{\widehat{a}_{iq}}$  (with  $\widehat{\mathbf{b}} = B^{-1} \mathbf{b}$ )  
 $p'$  is the  $p$ th entry in  $\mathcal{B}$
6. Exchange indices  $p'$  and  $q'$  between  $\mathcal{B}$  and  $\mathcal{N}$  to yield a new basic feasible solution
7. Go to 1

## Obtaining the initial basic feasible solution

As the initial basic feasible solution, try the “all slack” basis (i.e. starting at the origin)

$$\mathcal{B} = \{n+1, \dots, n+m\} \text{ and } \mathcal{N} = \{1, \dots, n\}$$

So we have:

- $\hat{\mathbf{b}} = \mathbf{b}$
- $\hat{\mathbf{c}}_N = \mathbf{c}$
- Basis is feasible iff  $\mathbf{b} \geq \mathbf{0}$

## How to start if $\mathbf{b} \not\geq \mathbf{0}$

Can't use the “all-slack” basis (because the origin is not in the feasible region)

- If  $\mathbf{b} \not\geq \mathbf{0}$  then, for each constraint  $i$ , subtract an **artificial variable**  $x_{n+m+i} \geq 0$
- Replace the objective  $f = \bar{\mathbf{c}}^T \mathbf{x}$  with the **Phase I** objective  $f = -\sum_{i=1}^m x_{n+m+i}$  (i.e. the negated sum of infeasibilities)

## The Phase I problem

Construct an initial basic feasible solution as follows: For  $i = 1, \dots, m$

If  $b_i \geq 0$

- Slack  $x_{n+i} = b_i \geq 0$  is basic
- Artificial  $x_{n+m+i} = 0$  is non-basic
- Column  $i$  of  $B$  is  $\mathbf{e}_i$

If  $b_i < 0$

- Slack  $x_{n+i}$  is non-basic
- Artificial  $x_{n+m+i} = -b_i > 0$  is basic
- Column  $i$  of  $B$  is  $-\mathbf{e}_i$

Basis matrix  $B$  is non-singular and, by construction,  $\hat{\mathbf{b}} \geq \mathbf{0}$

## At an optimal solution of the Phase I problem

The simplex algorithm drives  $f$  up towards zero.

At an optimal basic feasible solution  $x$  of the Phase I problem:

If  $f = 0$

- The values of the original and slack variables at  $x$  yield a basic feasible solution for the original LP

If  $f < 0$

- The artificial variables cannot all be driven to zero
- The original LP is **infeasible**

If the Phase I problem is solved with  $f = 0$  (and all artificial variables being in  $\mathcal{N}$ )

1. Remove the artificial variables from the problem (they are now zero)
2. Revert to the original objective function
3. Solve the original **Phase II** problem

**Does the algorithm terminate?**

- If  $\hat{\mathbf{b}}$  has any zero components then  $\mathbf{x}$  is a **degenerate** vertex
- There may be several basic feasible solutions at  $\mathbf{x}$
- If  $\hat{b}_p = 0$  then  $\bar{\alpha} = 0$  so the simplex algorithm does not move to a new vertex
- It may never leave!