

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 $\bar{A}\mathbf{x} = \mathbf{b}$
- 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** $\hat{\mathbf{c}}_N = \mathbf{c}_N - N^T B^{-T} \mathbf{c}_B \leq 0$

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 $\hat{\mathbf{c}}_N \leq \mathbf{0}$ then **stop** (with $\hat{\mathbf{c}}_N = \mathbf{c}_N - N^T B^{-T} \mathbf{c}_B$)
The solution is optimal
2. Determine the index $q' \in \mathcal{N}$ of the variable $x_{q'}$ with the most positive reduced cost $\hat{c}_{q'}$
 q' is the q th entry in \mathcal{N} .
3. Let $\hat{\mathbf{a}}_q = B^{-1} \mathbf{a}_q$, where \mathbf{a}_q is column q of N
4. If $\hat{\mathbf{a}}_q \leq \mathbf{0}$ then **stop**
The LP is unbounded

5. Determine the index $p' \in \mathcal{B}$ of the variable $x_{p'}$ corresponding to $p = \operatorname{argmin}_{i=1, \hat{\mathbf{a}}_{iq} > 0}^m \frac{\hat{b}_i}{\hat{a}_{iq}}$ (with $\hat{\mathbf{b}} = B^{-1}\mathbf{b}$)
 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:

- $\hat{\mathbf{b}} = \mathbf{b}$
- $\hat{\mathbf{c}}_{\mathcal{N}} = \mathbf{c}$
- 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!