

Linear Programming

Chongnan Li
16271221@bjtu.edu.cn

Lecture 4 Simplex Method

Outline:

1.Simplex method

2.Phase one method

3.Big M method

● Key steps of Simplex Method

Step 1: Find a bfs \mathbf{x} with $\mathbf{A} = [\mathbf{B} | \mathbf{N}]$.

Step 2: Check for n.b.v.'s

$$r_q = \mathbf{c}^T \mathbf{d}_q = c_q - \mathbf{c}_q \mathbf{B}^{-1} \mathbf{A}_q$$

If $r_q \geq 0, \forall$ nonbasic x_q , then the current bfs is optimal.

Otherwise, pick one (Brand's Rule to prevent cycling) $r_q < 0$. Go to next step.

Step 3: If $\mathbf{d}_q \geq 0$, then LP is unbounded below.

Otherwise, find

$$\alpha = \min_{i:\text{basic}} \left\{ \frac{x_i}{-d_{qi}} \mid d_{qi} < 0 \right\}$$

Then $\mathbf{x} := \mathbf{x} + \alpha \mathbf{d}_q$ is a new bfs.

Update \mathbf{B} and \mathbf{N} . Go to Step 2.

● More details

Q: How to judge when LP has infinitely many optimal solutions? Of course in this case the optimal objective value is a finite number.

A: When we find one optimal solution, consider the reduced costs for the n.b.v.'s. If $r_q = 0$ for some n.b.v. x_q , then LP has infinitely many optimal solutions.

● How to start the simplex method?

Two-phase method (Phase I problem) or Big-M method.

● **Two-phase method**

Step 1: Make the right hand side vector nonnegative:

$$\begin{aligned} \min \quad & \mathbf{c}^T \mathbf{x} \\ \text{(LP)} \quad & \text{s.t. } \mathbf{Ax} = \mathbf{b} (\geq \mathbf{0}) \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

Step 2: Add m artificial variables for Phase I problem:

$$\mathbf{u} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_m \end{bmatrix} \quad \begin{aligned} \min \quad & \sum_{i=1}^m u_i \\ \text{(PhI)} \quad & \text{s.t. } \mathbf{Ax} + \mathbf{Iu} = \mathbf{b} (\geq \mathbf{0}) \\ & \mathbf{x}, \mathbf{u} \geq \mathbf{0} \end{aligned}$$

● **Observations for Phase I problem**

1. $\mathbf{u} = \mathbf{b}, \mathbf{x} = \mathbf{0}$ is a bfs of (PhI).
2. (PhI) is bounded below by 0.
3. (LP) is feasible $\iff z_{\text{PhI}}^* = 0$.
4. Underdegeneracy, if $z_{\text{PhI}}^* = 0$, then the optimal solution of (PhI) is a bfs of (LP).

How about degenerate cases?

5. If $z_{\text{PhI}}^* = 0$ at an optimal bfs which is degenerate with at least one artificial variable u_j in the basis.

Suppose that $u_i = 0$ is the k th basic variable in the current basis, then:

- (1) if $\mathbf{e}_k^T \mathbf{B}^{-1} \mathbf{A}_q \neq 0$ for a nonbasic variable x_q , then u_i can be replaced by x_q to form a starting basis.
- (2) if $\mathbf{e}_k^T \mathbf{B}^{-1} \mathbf{A}_q = 0, \forall \text{n.b.v. } x_q$, then the k th row of $\mathbf{Ax} = \mathbf{b}$ is redundant. We remove it and start again.

(A little hard to comprehend? Please see the following explanation)

Reduced cost: $r_q = c_q - \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{A}_q$.

$\mathbf{B} = [\mathbf{A}_{q1} | \mathbf{A}_{q2} | \cdots | \mathbf{A}_{qk} | \cdots | \mathbf{A}_{qm}]$, $\mathbf{c}_B^T = [c_{q1}, c_{q2}, \cdots, c_{qk}, \cdots, c_{qm}]$,

$\mathbf{c}_B^T = [0, 0, \cdots, 1, \cdots, 0]$.

For n.b.v. x_q , $c_q = 0$, $\mathbf{c}_B^T = \mathbf{e}_k^T$. Thus $r_q = -\mathbf{e}_k^T \mathbf{B}^{-1} \mathbf{A}_q$.

● **Big-M method (no further discussion)**

$$\begin{aligned} \min \quad & \sum_{j=1}^n c_j x_j + \sum_{i=1}^m M u_i \\ \text{(PhI)} \quad & \text{s.t. } \mathbf{Ax} + \mathbf{Iu} = \mathbf{b} (\geq \mathbf{0}) \\ & \mathbf{x}, \mathbf{u} \geq \mathbf{0} \end{aligned}$$