DM559/DM545 Linear and Integer Programming

Lecture 3 The Simplex Method

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1. Simplex Method

Standard Form Basic Feasible Solutions Algorithm Tableaux and Dictionaries

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Simplex Method

A Numerical Example

$$\begin{array}{ll} \max & \sum\limits_{j=1}^n c_j x_j \\ & \sum\limits_{j=1}^n a_{ij} x_j \ \leq b_i, \ i=1,\ldots,m \\ & x_j \ \geq 0, \ j=1,\ldots,n \end{array}$$

 $\begin{array}{ll} \max \ \mathbf{c}^{T} \mathbf{x} & \max & \begin{bmatrix} 6 & 8 \end{bmatrix} \begin{bmatrix} x_{1} \\ x_{2} \end{bmatrix} \\ \mathbf{x} \geq \mathbf{0} & \begin{bmatrix} 5 & 10 \\ 4 & 4 \end{bmatrix} \begin{bmatrix} x_{1} \\ x_{2} \end{bmatrix} \leq \begin{bmatrix} 60 \\ 40 \end{bmatrix} \\ \mathbf{x} \in \mathbb{R}^{n}, \mathbf{c} \in \mathbb{R}^{n}, A \in \mathbb{R}^{m \times n}, \mathbf{b} \in \mathbb{R}^{m} \\ & x_{1}, x_{2} > 0 \end{array}$

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1. Simplex Method Standard Form

Basic Feasible Solutions Algorithm Tableaux and Dictionaries

Standard Form

Every LP problem can be converted in the form:

$$\begin{array}{l} \max \, \mathbf{c}^{\mathsf{T}} \mathbf{x} \\ A \mathbf{x} \, \leq \, \mathbf{b} \\ \mathbf{x} \, \in \, \mathbb{R}^{n} \end{array}$$
$$\mathbf{c} \in \mathbb{R}^{n}, A \in \mathbb{R}^{m \times n}, \mathbf{b} \in \mathbb{R}^{m} \end{array}$$

and then be put in equational standard form

 $\begin{array}{l} \max \ \mathbf{c}^{\mathsf{T}} \mathbf{x} \\ A \mathbf{x} \ = \ \mathbf{b} \\ \mathbf{x} \ \ge \ \mathbf{0} \\ \mathbf{x} \in \mathbb{R}^{n}, \mathbf{c} \in \mathbb{R}^{n}, A \in \mathbb{R}^{m \times n}, \mathbf{b} \in \mathbb{R}^{m} \end{array}$

- if equations, then put two constraints, $ax \le b$ and $ax \ge b$
- if $\mathbf{a} x \ge b$ then $-\mathbf{a} x \le -b$
- if min $\mathbf{c}^T \mathbf{x}$ then max $(-\mathbf{c}^T \mathbf{x})$

- 1. "=" constraints
- 2. $\textbf{x} \geq 0$ nonnegativity constraints
- 3. ($\mathbf{b} \ge 0$)
- 4. max

Transformation to Std Form

Every LP problem can be transformed in eq. std. form

1. introduce slack variables (or surplus)

 $5x_1 + 10x_2 + x_3 = 60$ $4x_1 + 4x_2 + x_4 = 40$

2. if
$$x_1 \gtrsim 0$$
 then $\begin{array}{l} x_1 = x_1' - x_1'' \\ x_1' \geq 0 \\ x_1'' \geq 0 \end{array}$
3. $(b \geq 0)$
4. $\min c^T x \equiv \max(-c^T x)$

LP in $m \times n$ converted into LP with at most (m + 2n) variables and m equations (n # original variables, m # constraints)

Geometry of LP in Eq. Std. Form

$$\max\{\mathbf{c}^T\mathbf{x} \mid A\mathbf{x} = \mathbf{b}, \mathbf{x} \ge \mathbf{0}\}$$

From linear algebra:

- the set of solutions of $A\mathbf{x} = \mathbf{b}$ is an affine space (hyperplane not passing through the origin).
- $x \ge 0$ nonegative orthant (octant in \mathbb{R}^3)



In \mathbb{R}^3 :

- $A\mathbf{x} = \mathbf{b}$ is a system of equations that we can solve by Gaussian elimination
- Elementary row operations of $\begin{bmatrix} A & b \end{bmatrix}$ do not affect set of feasible solutions
 - multiplying all entries in some row of $\begin{bmatrix} A & b \end{bmatrix}$ by a nonzero real number λ
 - replacing the *i*th row of $\begin{bmatrix} A & | & b \end{bmatrix}$ by the sum of the *i*th row and *j*th row for some $i \neq j$
- Let n' be the number of vars in eq. std. form.

we assume $n' \ge m$ and $\operatorname{rank}([A \mid \mathbf{b}]) = \operatorname{rank}(A) = m$

ie, rows of A are linearly independent otherwise, remove linear dependent rows

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Tableaux and Dictionaries

Basic Feasible Solutions

Basic feasible solutions are the vertices of the feasible region:



More formally:

Let $B = \{1 \dots m\}$, $N = \{m + 1 \dots n + m = n'\}$ be subsets partitioning the columns of A: A_B be made of columns of A indexed by B:

Definition

 $\mathbf{x} \in \mathbb{R}^n$ is a basic feasible solution of the linear program $\max{\{\mathbf{c}^T \mathbf{x} \mid A\mathbf{x} = \mathbf{b}, \mathbf{x} \ge \mathbf{0}\}}$ for an index set *B* if:

- $x_j = 0 \ \forall j \notin B$
- the square matrix A_B is nonsingular, ie, all columns indexed by B are lin. indep.
- $\mathbf{x}_B = A_B^{-1} \mathbf{b}$ is nonnegative, ie, $\mathbf{x}_B \ge 0$ (feasibility)

We call x_j for $j \in B$ basic variables and remaining variables nonbasic variables.

Theorem

A basic feasible solution is uniquely determined by the set B.

Proof:

$$A\mathbf{x} = A_B \mathbf{x}_B + A_N \mathbf{x}_N = b$$
$$\mathbf{x}_B + A_B^{-1} A_N \mathbf{x}_N = A_B^{-1} b$$
$$\mathbf{x}_B = A_B^{-1} b$$

 A_B is nonsingular hence one solution

Note: we call B a (feasible) basis

Extreme points and basic feasible solutions are geometric and algebraic manifestations of the same concept:

Theorem

Let P be a (convex) polyhedron from LP in eq. std. form. For a point $v \in P$ the following are equivalent:

- (i) v is an extreme point (vertex) of P
- (ii) v is a basic feasible solution of LP

Proof: by recognizing that vertices of P are linear independent and such are the columns in A_B

Theorem

Let $LP = \max{c^T \mathbf{x} \mid A\mathbf{x} = \mathbf{b}, \mathbf{x} \ge \mathbf{0}}$ be feasible and bounded, then the optimal solution is a basic feasible solution.

Proof. consequence of previous theorem and fundamental theorem of linear programming

Note, a similar theorem is valid for arbitrary linear programs (not in eq. form)

Definition

A basic feasible solution of a linear program with n variables is a feasible solution for which some n linearly independent constraints hold with equality.

However, an optimal solution does not need to be basic:

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\max x_1 + x_2 subject to x_1 + x_2 \le 1
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- Idea for solution method:
- examine all basic solutions.
- There are finitely many: $\binom{m+n}{m}$.
- However, if n = m then $\binom{2m}{m} \approx 4^m$.

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Standard Form Basic Feasible Solutions Algorithm Tableaux and Dictionarie

Simplex Method

$$\max \quad z = \begin{bmatrix} 6 & 8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$
$$\begin{bmatrix} 5 & 10 & 1 & 0 \\ 4 & 4 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 60 \\ 40 \end{bmatrix}$$
$$x_1, x_2, x_3, x_4 \ge 0$$

Canonical eq. std. form: one decision variable is isolated in each constraint with coefficient 1 and does not appear in the other constraints nor in the obj. func. and b terms are positive

It gives immediately a basic feasible solution:

 $x_1 = 0, x_2 = 0, x_3 = 60, x_4 = 40$

Is it optimal? Look at signs in $z \rightsquigarrow$ if positive then an increase would improve.

Let's try to increase a promising variable, ie, x_1 , one with positive coefficient in z

 $5x_1 + x_3 = 60$ $x_1 = \frac{60}{5} - \frac{x_3}{5}$ $x_3 = 60 - 5x_1 \ge 0$

If $x_1 > 12$ then $x_3 < 0$

 $4x_1 + x_4 = 40$ $x_1 = \frac{40}{4} - \frac{x_4}{4}$ $x_4 = 40 - 4x_1 \ge 0$

If $x_1 > 10$ then $x_4 < 0$

we can take the minimum of the two $\rightsquigarrow x_1$ increased to 10 x_4 exits the basis and x_1 enters



Simplex Method

Simplex Tableau

First simplex tableau:

	x_1	<i>x</i> ₂	<i>x</i> 3	<i>x</i> ₄	-z	Ь
<i>x</i> ₃	5	10	1	0	0	60
<i>x</i> 4	4	4	0	1	0	40
	6	8	0	0	1	0

we want to reach this new tableau

Pivot operation:

1. Choose pivot:

column: one *s* with positive coefficient in obj. func. row: ratio between coefficient *b* and pivot column: choose the one with smallest ratio:

$$\theta = \min_{i} \left\{ \frac{b_i}{a_{is}} : a_{is} > 0 \right\}, \qquad \theta$$

 θ increase value of entering var.

2. elementary row operations to update the tableau

- x_4 leaves the basis, x_1 enters the basis
 - Divide pivot row by pivot
 - Send to zero the coefficient in the pivot column of the first row
 - Send to zero the coefficient of the pivot column in the third (cost) row

From the last row we read: $2x_2 - 3/2x_4 - z = -60$, that is: $z = 60 + 2x_2 - 3/2x_4$. Since x_2 and x_4 are nonbasic we have z = 60 and $x_1 = 10, x_2 = 0, x_3 = 10, x_4 = 0$.

• Done? No! Let x₂ enter the basis

Definition (Reduced costs)

We call reduced costs the coefficients in the objective function of the nonbasic variables, \bar{c}_N

Proposition (Optimality Condition)

The basic feasible solution is optimal when the reduced costs in the corresponding simplex tableau are nonpositive, ie, such that:

 $\bar{c}_N \leq 0$

Proof: Let z_0 be the obj value when $\bar{c}_N \leq 0$. For any other feasible solution $\tilde{\mathbf{x}}$ we have:

$$\tilde{\mathbf{x}}_N \ge 0$$
 and $\mathbf{c}^T \tilde{\mathbf{x}} = z_0 + \bar{\mathbf{c}}_N^T \tilde{\mathbf{x}}_N \le z_0$

Simplex Method

Graphical Representation



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Tableaux and Dictionaries

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$$x_{n+i} = b_i - \sum_{j=1}^n a_{ij}x_j, \quad i = 1, \dots, m$$
$$z = \sum_{j=1}^n c_j x_j$$

Dictionary

$$\begin{aligned} x_r &= \bar{b}_r - \sum_{\substack{s \notin B}} \bar{a}_{rs} x_s, \quad r \in B \\ z &= \bar{d} + \sum_{\substack{s \notin B}} \bar{c}_s x_s \end{aligned}$$

pivot operations in dictionary form: choose col *s* with r.c. > 0 choose row with min $\{-\bar{b}_i/\bar{a}_{is} \mid a_{is} < 0, i = 1, ..., m\}$ update: express entering variable and substitute in other rows

Tableau

 $\begin{bmatrix} I & \bar{A}_N & 0 & \bar{b} \\ \hline 0 & \bar{c}_N & 1 & -\bar{d} \end{bmatrix}$

Example

$$\max \begin{array}{l} 6x_1 + 8x_2 \\ 5x_1 + 10x_2 \le 60 \\ 4x_1 + 4x_2 \le 40 \\ x_1, x_2 \ge 0 \end{array}$$

After 2 iterations:

Summary

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