Linear Programming

Lecture 07A - The Revised Simplex Algorithm

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02612 Constrained Optimization

Learning Objectives

- Describe different formulations of linear programs
- ► Convert a linear program to its standard form
- Describe active-set algorithms for linear programs
- Describe and implement the revised simplex algorithm

Reading material: Nocedal & Wright: Chap. 13

Literature:

- V. Chvatal: Linear Programming, W.H. Freeman and Company, 1983
- ▶ J.L. Nazareth: Computer Solutions of Linear Programs, Oxford Science Publications, 1987
- R.J. Vanderbei: Linear Programming. Foundations and Extensions, Springer, 2008
- M.C. Ferris, O.L. Mangaserian, S.J. Wright: Linear Programming with Matlab, SIAM, 2007
- D. Bertsimas, J.N. Tsitsiklis: Introduction to Linear Optimization, Athena Scientific, 1997
- ▶ H.P. Williams: Model Building in Mathematical Programming, 3rd Edition, Wiley, 1993

HISTORY of Linear Programming

- 1940's George Dantzig at Stanford invents the Simplex Algorithm. The simplex algorithm has exponential complexity.
- 2. **1984** Karmarkar publishes the interior-point algorithm with polynomial rather than exponential complexity.
- 1987-1991 Primal-dual interior-point algorithms developed. Initiated by Megido 1987.
- 4. **1992** Mehrotra describes the predictor-corrector method for primal-dual interior point algorithms.
- 1996 Stephen J. Wright publishes "Primal Dual Interior-Point Algorithms". Interior-point algorithms for LPs has become a matured technology.

Main classes of algorithms

- Active-set algorithms (simplex algorithms)
- ► Interior-point algorithms

Outline

Linear Programs / Linear Optimization Models

Conversion to the standard form LP

Theory

Optimality Conditions
Active-Set Treatment of the Complementarity Conditions

The Revised Simplex Algorithm

Presolving

Linear Program

► Example of a linear program (LP)

\min_{x_1,x_2,x_3}	$F(x_1, x_2, x_3) = -3x_1 + 2x_2 - 4x_3 + 5$	(1a)
s.t.	$x_1 + x_2 + x_3 = 1$	(1b)
	$x_1 + x_2 \ge 0.7$	(1c)
	$x_1 + x_3 \le 0.8$	(1d)
	$0 \le x_1 \le 1$	(1e)
	$0 \le x_2 \le 1$	(1f)
	$0 \le x_3 \le 1$	(1g)

Linear Program

► Affine objective function

$$F(x) = g'x + \gamma = \sum_{j=1}^{n} g_j x_j + \gamma$$
 (2)

► Linear equality constraints

$$a_i'x = \sum_{j=1}^{n} a_{ij}x_j = b_i$$
 (3)

► Linear inequality constraint

$$a_i'x = \sum_{j=1}^n a_{ij}x_j \ge b_i \tag{4}$$

$$a_i'x = \sum_{j=1}^{n} a_{ij}x_j \le b_i \tag{5}$$

▶ Bound constraints: $l_i \le x_i \le u_i$

Linear Program

Linear program

$$\min_{x \in \mathbb{R}^n} F(x) = g'x + \gamma \tag{6a}$$

$$s.t. a_i'x = b_i i \in \mathcal{E} (6b)$$

$$a_i'x \ge b_i \quad i \in \mathcal{I}_L$$
 (6c)

$$a_i'x \le b_i \quad i \in \mathcal{I}_U$$
 (6d)

$$l \le x \le u \tag{6e}$$

Canonical form

$$\min_{x \in \mathbb{R}^n} \quad F(x) = g'x \tag{7a}$$

$$s.t. Ax > b (7b)$$

$$x \ge 0 \tag{7c}$$

Standard form

$$\min_{x \in \mathbb{R}^n} F(x) = g'x \tag{8a}$$

$$s.t. Ax = b (8b)$$

$$x \ge 0 \tag{8c}$$

► LP in standard from

$$\min_{x \in \mathbb{P}^n} F(x) = g'x \tag{9a}$$

$$s.t. Ax = b (9b)$$

$$x \ge 0 \tag{9c}$$

- ▶ Greater than inequality: $a_i'x \ge b_i$ \Rightarrow $a_i'x s = b_i$ $s \ge 0$
- ▶ Less than inequality: $a_i'x \le b_i \Rightarrow a_i'x + s = b_i \quad s \ge 0$
- ▶ Lower bound: $l_i \le x_i \implies x_i s = l_i \quad s \ge 0$
- ▶ Upper bound: $x_i \le u_i \implies x_i + s = u_i \quad s \ge 0$
- ▶ Free variable: $x = s_1 s_2, s_1 \ge 0, s_2 \ge 0.$

$$x - s_1 + s_2 = 0$$
 $s_1 \ge 0, s_2 \ge 0$

► Example of a linear program (LP)

\min_{x_1,x_2,x_3}	$F(x_1, x_2, x_3) = -3x_1 + 2x_2 - 4x_3 + 5$	(10a)
s.t.	$x_1 + x_2 + x_3 = 1$	(10b)
	$x_1 + x_2 \ge 0.7$	(10c)
	$x_1 + x_3 \le 0.8$	(10d)
	$0 \le x_1 \le 1$	(10e)
	$0 \le x_2 \le 1$	(10f)
	$0 \le x_3 \le 1$	(10g)

► Introduction of slack variables

$$\begin{array}{lllll} x_1 + x_2 + x_3 = 1 & \Rightarrow & x_1 + x_2 + x_3 = 1 & & \text{(11a)} \\ x_1 + x_2 \geq 0.7 & \Rightarrow & x_1 + x_2 - s_1 = 0.7 & s_1 \geq 0 & & \text{(11b)} \\ x_1 + x_3 \leq 0.8 & \Rightarrow & x_1 + x_3 + s_2 = 0.8 & s_2 \geq 0 & & \text{(11c)} \\ 0 \leq x_1 \leq 1 & \Rightarrow & x_1 \geq 0 & x_1 + s_3 = 1 & s_3 \geq 0 & & \text{(11d)} \\ 0 \leq x_2 \leq 1 & \Rightarrow & x_2 \geq 0 & x_2 + s_4 = 1 & s_4 \geq 0 & & \text{(11e)} \\ 0 \leq x_3 \leq 1 & \Rightarrow & x_3 \geq 0 & x_3 + s_5 = 1 & s_5 \geq 0 & & \text{(11f)} \\ \end{array}$$

► Matrix-vector formulation

$$x = \begin{bmatrix} x_1 & x_2 & x_3 & s_1 & s_2 & s_3 & s_4 & s_5 \end{bmatrix}'$$

$$A = \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & -1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \quad b = \begin{bmatrix} 1 \\ 0.7 \\ 0.8 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$
(12)

Consider the general LP

$$\min_{x \in \mathbb{R}^n} F(x) = g'x \tag{13a}$$

$$s.t. Ax = b$$
 (13b)

$$A_1 x > b_1 \tag{13c}$$

$$A_u x \le b_u \tag{13d}$$

$$l < x < u$$
 (13e)

Introduce $x=x_u-x_l$ with $x_u\geq 0$ and $x_l\geq 0$ such that the general LP becomes

$$\min_{x_l, x_u, s_l, s_u, t_l, t_u} F(x_l, x_s, s_l, s_u, t_l, t_u) = g' x_u - g' x_l$$
 (14a)

$$s.t. Ax_u - Ax_l = b x_u \ge 0, x_l \ge 0 (14b)$$

$$A_l x_u - A_l x_l - s_l = b_l \qquad \qquad s_l \ge 0 \tag{14c}$$

$$A_{il}x_{il} - A_{il}x_{l} + s_{il} = b_{il} \qquad \qquad s_{il} > 0 \tag{14d}$$

$$u_{u} = u_{u} + v_{u} = v_{u}$$

$$x_u - x_l - t_l = l t_l \ge 0 (14e)$$

$$x_u - x_l + t_u = u t_u \ge 0 (14f)$$

Define

$$\begin{split} \bar{x} &= [x_l; x_u; s_l; s_u; t_l; t_u], \ \bar{g} = [-g; g; 0; 0; 0; 0; 0], \ \bar{b} = [b; b_l; b_u; l; u], \text{and} \\ \bar{A} &= [-A, A, 0, 0, 0, 0; -A_l, A_l, -I, 0, 0, 0; -A_u, A_u, 0, I, 0, 0; -I, I, 0, 0, -I, 0; -I, I, 0, 0, 0, I] \end{split}$$

such that the general LP can be expressed as the LP in standard form

$$\min_{\bar{z}} \quad \bar{g}'\bar{x} \tag{15a}$$

$$s.t.$$
 $\bar{A}\bar{x} = \bar{b}$ (15b)

$$\bar{x} > 0$$
 (15c)

Duality

Primal LP

$$\min_{x \in \mathbb{R}^n} F(x) = g'x$$

$$s.t. \quad Ax \ge b$$
(16a)

► Lagrange function

$$L(x,\lambda) = g'x - \lambda'(Ax - b) = (g - A'\lambda)'x + b'\lambda \tag{17}$$

First order necessary and sufficient optimality conditions

$$\nabla_x L(x,\lambda) = g - A'\lambda = 0 \tag{18a}$$

$$Ax \ge b \perp \lambda \ge 0 \tag{18b}$$

Dual (Lagrange Dual)

$$\max_{x,\lambda} \quad L(x,\lambda) = (g - A'\lambda)'x + b'\lambda$$

$$s.t. \quad \nabla_x L(x,\lambda) = g - A'\lambda = 0$$
(19a)

$$\lambda \ge 0 \tag{19c}$$

Duality

▶ Dual (Lagrange Dual)

$$\max_{x,\lambda} L(x,\lambda) = (g - A'\lambda)'x + b'\lambda$$
 (20a)

s.t.
$$\nabla_x L(x,\lambda) = g - A'\lambda = 0$$
 (20b)
 $\lambda \ge 0$ (20c)

▶ The dual reformulated

$$\max_{\lambda} b'\lambda \tag{21a}$$

$$s.t. A'\lambda = g (21b)$$

$$\lambda \ge 0$$
 (21c)

Dual of Standard LP

► Linear program in standard form

$$\min_{x \in \mathbb{R}^n} F(x) = g'x$$
 (22a)
s.t. $Ax = b$ (22b)
 $x > 0$ (22c)

► Lagrange function

$$L(x,\mu,\lambda) = g'x - \mu'(Ax - b) - \lambda'x \tag{23}$$

▶ Dual program

$$\max_{x,\mu,\lambda} L(x,\mu,\lambda) = (g - A'\mu - \lambda)'x + b'\mu$$
 (24a)

$$s.t. \quad \nabla_x L(x,\mu,\lambda) = g - A'\mu - \lambda = 0$$
 (24b)

$$\lambda \ge 0$$
 (24c)

Dual of Standard LP

Primal LP in standard form

$$\min_{x \in \mathbb{R}^n}$$

$$\min_{x \in \mathbb{R}^n} \quad F(x) = g'x$$
s.t.
$$Ax = b$$

 $x \ge 0$

 $L(x, \mu, \lambda) = (g - A'\mu - \lambda)'x + b'\mu$

 $b'\mu$

 $\lambda > 0$

s.t.
$$\nabla_x L(x, \mu, \lambda) = g - A'\mu - \lambda = 0$$

$$\lambda \ge 0$$

Dual program reformulated

$$\max_{\mu,\lambda}$$

$$s.t. A'\mu + \lambda = q$$

(25a)

(26a)

(26b)

(26c)

$$d = g$$

Optimality Conditions.

Linear program (LP) in standard form $(A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m, g \in \mathbb{R}^n)$

$$\min_{x \in \mathbb{R}^n} F(x) = g'x \tag{28a}$$

$$s.t. \quad Ax = b \tag{28b}$$

$$x > 0 \tag{28c}$$

$$x \ge 0 \tag{28c}$$

Lagrange function

$$L(x,\mu,\lambda) = g'x - \mu'(Ax - b) - \lambda'x \tag{29}$$

Necessary and sufficient optimality conditions

$$\nabla_x L(x, \mu, \lambda) = g - A'\mu - \lambda = 0 \tag{30a}$$

$$Ax = b (30b)$$

$$x \ge 0 \perp \lambda \ge 0 \tag{30c}$$

Optimality Conditions.

Necessary and sufficient optimality conditions

$$\nabla_x L(x, \mu, \lambda) = g - A'\mu - \lambda = 0$$
 (31a)

$$Ax = b (31b)$$

$$x \ge 0 \perp \lambda \ge 0 \tag{31c}$$

► Formulation in matrix form

$$\begin{bmatrix} 0 & A' \\ A & 0 \end{bmatrix} \begin{bmatrix} x \\ \mu \end{bmatrix} = \begin{bmatrix} g - \lambda \\ b \end{bmatrix}$$
 (32a)

$$x \ge 0 \perp \lambda \ge 0 \tag{32b}$$

$$x \in \mathbb{R}^n \ \mu \in \mathbb{R}^m \ \lambda \in \mathbb{R}^n \ A \in \mathbb{R}^{m \times n}$$

Optimality Conditions.

Non-basic and basic sets

$$\mathcal{N} = \{i: x_i = 0\} \tag{33}$$

$$\mathcal{B} = \{1, 2, \dots, n\} \setminus \mathcal{N} \qquad x_i \ge 0 \, i \in \mathcal{B} \tag{34}$$

► Complementarity condition $(x \ge 0 \perp \lambda \ge 0)$

$$\lambda_i \ge 0 \qquad i \in \mathcal{N}$$

$$\lambda_i = 0 \qquad i \in \mathcal{B} \tag{36}$$

$$x = \begin{bmatrix} x_B \\ x_N \end{bmatrix} \quad \lambda = \begin{bmatrix} \lambda_B \\ \lambda_N \end{bmatrix} \quad g = \begin{bmatrix} g_B \\ g_N \end{bmatrix}$$
 (37)

$$A = \begin{bmatrix} B & N \end{bmatrix} \quad B \in \mathbb{R}^{m \times m} \text{ nonsingular}$$
 (38)

(35)

Optimality Conditions.

► Formulation in matrix form

$$\begin{bmatrix} 0 & A' \\ A & 0 \end{bmatrix} \begin{bmatrix} x \\ \mu \end{bmatrix} = \begin{bmatrix} g - \lambda \\ b \end{bmatrix}$$
 (39a)
$$x \ge 0 \perp \lambda \ge 0$$
 (39b)

▶ Use $A = \begin{bmatrix} B & N \end{bmatrix}$

$$\begin{bmatrix} 0 & 0 & B' \\ 0 & 0 & N' \\ B & N & 0 \end{bmatrix} \begin{bmatrix} x_B \\ x_N \\ \mu \end{bmatrix} = \begin{bmatrix} g_B - \lambda_B \\ g_N - \lambda_N \\ b \end{bmatrix}$$

• $x_N = 0$, $\lambda_B = 0$, $x_B \ge 0$, $\lambda_N \ge 0$

$$B'\mu = g_B - \lambda_B = g_B$$

$$\lambda_N = g_N - N'\mu$$
(41a)

$$Bx_B = b - Nx_N = b (41c)$$

(40)

Optimality Conditions.

▶ Select the active set, i.e. the non-basic set \mathcal{N} , and thereby the basic set \mathcal{B} . Then

$$x_N = 0 (42a)$$

$$\lambda_B = 0 \tag{42b}$$

▶ Compute x_B , μ , and λ_N :

$$Bx_B = b (42c)$$

$$B'\mu = g_B \tag{42d}$$

$$\lambda_N = g_N - N'\mu \tag{42e}$$

► Verify the complementarity conditions

$$x_B \ge 0 \tag{42f}$$

$$\lambda_N \ge 0 \tag{42g}$$

Revised Simplex Algorithm

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Require: The sets \mathcal{B} and \mathcal{N} such that x_B = B^{-1}b > 0 and x_N = 0 in which B = [a_i]_{i \in \mathcal{B}},
    N = [a_i]_{i \in \mathcal{N}}, x_B = (x_i)_{i \in \mathcal{B}}, x_N = (x_i)_{i \in \mathcal{N}}, \text{ and } A = [a_1 \quad a_2 \quad \dots \quad a_n].
    while not STOP do
          Set B = [a_i]_{i \in \mathcal{B}} and N = [a_i]_{i \in \mathcal{N}}.
          Solve for \mu:
                                                                       B'\mu = g_B, \quad g_B = (g_i)_{i \in \mathcal{B}}
                                                                                                                                                                                 (43)
          Compute: \lambda_N = g_N - N'\mu in which g_N = (g_i)_{i \in \mathcal{N}} and \lambda_N = (\lambda_i)_{i \in \mathcal{N}}.
          if (\lambda_N)_i > 0 \forall i then
                 STOP = true (optimal solution found)
          else
                 Select s:(\lambda_N)_s<0. Let i_s=\mathcal{N}(s) be the corresponding entry in \mathcal{N}.
                 Solve for h:
                                                                                           Bh = a_i
                                                                                                                                                                                 (44)
                \mathcal{J} = \left\{ \arg\min_{i:h_i > 0} \frac{(x_B)_i}{h_i} \right\}
                if \mathcal{J} = \emptyset then
                       STOP = true (unbounded problem, no solution)
                else
                       Select j \in \mathcal{J} and \alpha = \frac{(x_B)_j}{h_j}
                       x_B \leftarrow x_B - \alpha h, ((x_B)_i \leftarrow 0), (x_N)_s \leftarrow \alpha
                \begin{array}{ccc} i_{j} \leftarrow \mathcal{B}(j), \mathcal{B} \leftarrow (\mathcal{B} \setminus \left\{i_{j}\right\}) \cup \left\{i_{s}\right\}, \mathcal{N} \leftarrow (\mathcal{N} \setminus \left\{i_{s}\right\}) \cup \left\{i_{j}\right\} \\ \text{end if} \end{array}
          end if
   end while
    \lambda_B \leftarrow 0 in which \lambda_B = (\lambda_i)_{i \in B}
```

FINDING A FEASIBLE POINT

► LP in standard form

$$\min_{x \in \mathbb{R}^n} F(x) = g'x \tag{45a}$$

$$s.t. \quad Ax = b \tag{45b}$$

$$x \ge 0 \tag{45c}$$

▶ Constrained l_1 - or l_∞ -regression

s.t.
$$x \ge 0$$
 (46b)
$$\min_{x \in \mathbb{R}^n} F(x) = \|Ax - b\|_{\infty}$$
 (47a)

 $\min_{x \in \mathbb{R}^n} \quad F(x) = \|Ax - b\|_1$

s.t. x > 0

▶ Initial feasible point:
$$x = 0$$
.

▶ Feasible problem if: $F(x^*) = 0$

(46a)

(47b)

FINDING A FEASIBLE POINT

► l_{∞} -regression problem

$$\min_{x \in \mathbb{R}^n} F(x) = \|Ax - b\|_{\infty}$$
 (48a)

$$s.t. x \ge 0 (48b)$$

▶ The l_{∞} -regression problem as an LP

$$\min_{x \in \mathbb{R}^n, t \in \mathbb{R}} \quad \begin{bmatrix} 0 \\ 1 \end{bmatrix}' \begin{bmatrix} x \\ t \end{bmatrix} \tag{49a}$$

s.t.
$$\begin{bmatrix} A & e \\ -A & e \end{bmatrix} \begin{bmatrix} x \\ t \end{bmatrix} \ge \begin{bmatrix} b \\ -b \end{bmatrix}$$
 (49b)

Next: Convert this LP to standard form and solve using the simplex method. Easy to find a feasible point of this program.

FINDING A FEASIBLE POINT

► Introduce slack variables to have an LP in standard form

$$\min_{x \in \mathbb{R}^{n}, t \in \mathbb{R}, s_{1} \in \mathbb{R}^{m}, s_{2} \in \mathbb{R}^{m}} \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} x \\ t \\ s_{1} \\ s_{2} \end{bmatrix}$$

$$(50a)$$

$$s.t.$$

$$\begin{bmatrix} A & e & -I & 0 \\ -A & e & 0 & -I \end{bmatrix} \begin{bmatrix} x \\ t \\ s_{1} \\ s_{2} \end{bmatrix} = \begin{bmatrix} b \\ -b \end{bmatrix}$$

$$(50b)$$

Initial feasible point: x = 0, $t = \max |b_i|$, $s_1 = te - b$, $s_2 = te + b$

REVISED SIMPLEX PROCEDURE

1. Convert the stated LP problem to an LP in standard form:

$$\min_{x \in \mathbb{R}^n} \quad F(x) = g'x \tag{51a}$$

$$s.t. Ax = b (51b)$$

$$x \ge 0 \tag{51c}$$

2. Phase I simplex: Find a feasible point or detect infeasibility by solution of a l_{∞} -regression problem

$$\min_{x \in \mathbb{R}^n} F(x) = ||Ax - b||_{\infty}$$
 (52a)

$$s.t. x \ge 0 (52b)$$

using the revised simplex algorithm.

- 3. **Phase II simplex**: Use the solution, x, of the l_{∞} -regression problem in phase I as an initial feasible solution for solution of (51) by the revised simplex algorithm.
- 4. Report the solution in terms of the original problem.

Revised Simplex Procedure

► LP in standard form

$$\min_{x \in \mathbb{R}^n} F(x) = g'x \tag{53a}$$

$$s.t. Ax = b (53b)$$

$$x \ge 0 \tag{53c}$$

▶ Phase 1: LP for finding a feasible solution [standard LP]

$$\min_{x,s,t} F(x,s,t) = e's + e't$$
 (54a)

$$s.t. \quad Ax + s - t = b \tag{54b}$$

$$x \ge 0, \ s \ge 0, \ t \ge 0$$
 (54c)

▶ Phase 2: If $s^* = 0$ and $t^* = 0$ for the phase 1 LP, set the initial point to $x = x^*$ and solve the standard LP

$$\min_{x \in \mathbb{R}^n} \quad F(x) = g'x \tag{55a}$$

$$s.t. Ax = b (55b)$$

$$x \ge 0 \tag{55c}$$

Presolving

The purpose of presolving is to reduce the size of the user-defined LP before passing it to the solver.

Andersen and Andersen (1995): Presolving in linear programming. Mathematical Programming, 71, pp. 221-245

$$\min_{x} c'x \tag{56a}$$

$$s.t. \quad Ax = b \tag{56b}$$

$$l \le x \le u \tag{56c}$$

- ► Row singleton
- ► Free column singleton
- ► Zero rows and columns
- ► Forcing or dominated constraints