

15.081J/6.251J Introduction to Mathematical  
Programming

Lecture 7: The Simplex Method III

# 1 Outline

SLIDE 1

- Finding an initial BFS
- The complete algorithm
- The column geometry
- Computational efficiency
- The diameter of polyhedra and the Hirsch conjecture

# 2 Finding an initial BFS

SLIDE 2

- **Goal:** Obtain a BFS of  $\mathbf{Ax} = \mathbf{b}$ ,  $\mathbf{x} \geq \mathbf{0}$   
or decide that LOP is infeasible.
- Special case:  $\mathbf{b} \geq \mathbf{0}$

$$\begin{aligned}\mathbf{Ax} &\leq \mathbf{b}, \quad \mathbf{x} \geq \mathbf{0} \\ \Rightarrow \mathbf{Ax} + \mathbf{s} &= \mathbf{b}, \quad \mathbf{x}, \mathbf{s} \geq \mathbf{0} \\ \mathbf{s} &= \mathbf{b}, \quad \mathbf{x} = \mathbf{0}\end{aligned}$$

## 2.1 Artificial variables

SLIDE 3

$$\mathbf{Ax} = \mathbf{b}, \quad \mathbf{x} \geq \mathbf{0}$$

1. Multiply rows with  $-1$  to get  $\mathbf{b} \geq \mathbf{0}$ .
2. Introduce artificial variables  $\mathbf{y}$ , start with initial BFS  $\mathbf{y} = \mathbf{b}$ ,  $\mathbf{x} = \mathbf{0}$ , and apply simplex to auxiliary problem

$$\begin{aligned}\min \quad & y_1 + y_2 + \dots + y_m \\ \text{s.t.} \quad & \mathbf{Ax} + \mathbf{y} = \mathbf{b} \\ & \mathbf{x}, \mathbf{y} \geq \mathbf{0}\end{aligned}$$

SLIDE 4

3. If cost  $> 0 \Rightarrow$  **LOP infeasible**; stop.
4. If cost = 0 and no artificial variable is in the basis, then a BFS was found.
5. Else, all  $y_i^* = 0$ , but some are still in the basis. Say we have  $\mathbf{A}_{B(1)}, \dots, \mathbf{A}_{B(k)}$  in basis  $k < m$ . There are  $m - k$  additional columns of  $\mathbf{A}$  to form a basis.

SLIDE 5

6. Drive artificial variables out of the basis: If  $l$ th basic variable is artificial examine  $l$ th row of  $\mathbf{B}^{-1}\mathbf{A}$ . If all elements = 0  $\Rightarrow$  row redundant. Otherwise pivot with  $\neq 0$  element.

## 2.2 Example

SLIDE 6

$$\begin{aligned}
 \min \quad & x_1 + x_2 + x_3 \\
 \text{s.t.} \quad & x_1 + 2x_2 + 3x_3 = 3 \\
 & -x_1 + 2x_2 + 6x_3 = 2 \\
 & 4x_2 + 9x_3 = 5 \\
 & 3x_3 + x_4 = 1 \\
 & x_1, \dots, x_4 \geq 0.
 \end{aligned}$$

$$\begin{aligned}
 \min \quad & x_5 + x_6 + x_7 + x_8 \\
 \text{s.t.} \quad & x_1 + 2x_2 + 3x_3 + x_5 = 3 \\
 & -x_1 + 2x_2 + 6x_3 + x_6 = 2 \\
 & 4x_2 + 9x_3 + x_7 = 5 \\
 & 3x_3 + x_4 + x_8 = 1 \\
 & x_1, \dots, x_8 \geq 0.
 \end{aligned}$$

SLIDE 7

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$
	-11	0	-8	-21	-1	0	0	0
$x_5 =$	3	1	2	3	0	1	0	0
$x_6 =$	2	-1	2	6	0	0	1	0
$x_7 =$	5	0	4	9	0	0	0	1
$x_8 =$	1	0	0	3	1*	0	0	0

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$
	-10	0	-8	-18	0	0	0	1
$x_5 =$	3	1	2	3	0	1	0	0
$x_6 =$	2	-1	2	6	0	0	1	0
$x_7 =$	5	0	4	9	0	0	0	1
$x_4 =$	1	0	0	3*	1	0	0	0

SLIDE 8

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$
	-4	0	-8	0	6	0	0	7
$x_5 =$	2	1	2	0	-1	1	0	-1
$x_6 =$	0	-1	2*	0	-2	0	1	-2
$x_7 =$	2	0	4	0	-3	0	0	-3
$x_3 =$	1/3	0	0	1	1/3	0	0	1/3

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$
-4	-4	0	0	-2	0	4	0	-1
$x_5 =$	2	2*	0	0	1	1	-1	0
$x_2 =$	0	-1/2	1	0	-1	0	1/2	0
$x_7 =$	2	2	0	0	1	0	-2	1
$x_3 =$	1/3	0	0	1	1/3	0	0	1/3

SLIDE 9

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$
0	0	0	0	0	2	2	0	1
$x_1 =$	1	1	0	0	1/2	1/2	-1/2	0
$x_2 =$	1/2	0	1	0	-3/4	1/4	1/4	0
$x_7 =$	0	0	0	0	-1	-1	1	0
$x_3 =$	1/3	0	0	1	1/3	0	0	1/3

SLIDE 10

	$x_1$	$x_2$	$x_3$	$x_4$
*	*	*	*	*
$x_1 =$	1	1	0	0
$x_2 =$	1/2	0	1	0
$x_3 =$	1/3	0	0	1

### 3 A complete Algorithm for LO

SLIDE 11

#### Phase I:

1. By multiplying some of the constraints by  $-1$ , change the problem so that  $\mathbf{b} \geq \mathbf{0}$ .
2. Introduce  $y_1, \dots, y_m$ , if necessary, and apply the simplex method to  $\min \sum_{i=1}^m y_i$ .
3. If  $\text{cost} > 0$ , original problem is infeasible; STOP.
4. If  $\text{cost} = 0$ , a feasible solution to the original problem has been found.
5. Drive artificial variables out of the basis, potentially eliminating redundant rows.

SLIDE 12

#### Phase II:

1. Let the final basis and tableau obtained from Phase I be the initial basis and tableau for Phase II.
2. Compute the reduced costs of all variables for this initial basis, using the cost coefficients of the original problem.
3. Apply the simplex method to the original problem.

### 3.1 Possible outcomes

SLIDE 13

1. Infeasible: Detected at Phase I.
2.  $\mathbf{A}$  has linearly dependent rows: Detected at Phase I, eliminate redundant rows.
3. Unbounded (cost =  $-\infty$ ): detected at Phase II.
4. Optimal solution: Terminate at Phase II in optimality check.

## 4 The big- $M$ method

SLIDE 14

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

## 5 The Column Geometry

SLIDE 15

$$\begin{aligned} \min \quad & \mathbf{c}'\mathbf{x} \\ \text{s.t.} \quad & \mathbf{Ax} = \mathbf{b} \\ & \mathbf{e}'\mathbf{x} = 1 \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

$$x_1 \begin{bmatrix} \mathbf{A}_1 \\ c_1 \end{bmatrix} + x_2 \begin{bmatrix} \mathbf{A}_2 \\ c_2 \end{bmatrix} + \cdots + x_n \begin{bmatrix} \mathbf{A}_n \\ c_n \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ z \end{bmatrix}$$

SLIDE 16

SLIDE 17

## 6 Computational efficiency

SLIDE 18

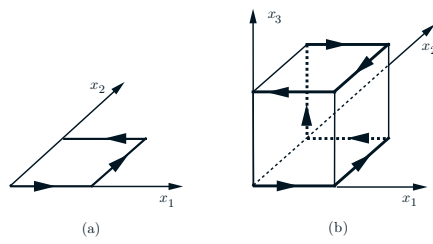
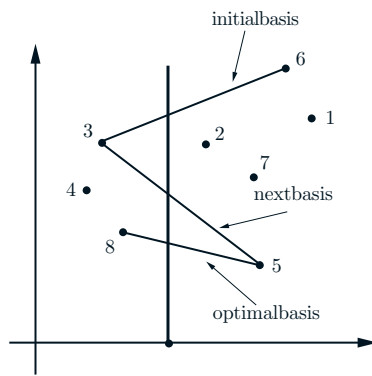
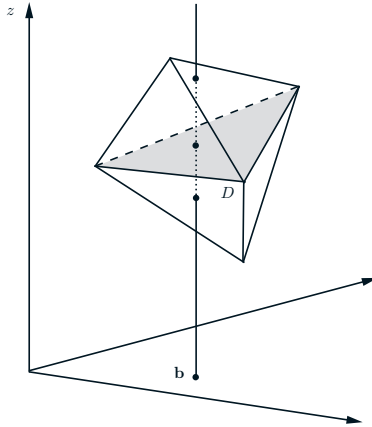
Exceptional practical behavior: linear in  $n$   
Worst case

$$\begin{aligned} \max \quad & x_n \\ \text{s.t.} \quad & \epsilon \leq x_1 \leq 1 \\ & \epsilon x_{i-1} \leq x_i \leq 1 - \epsilon x_{i-1}, \quad i = 2, \dots, n \end{aligned}$$

SLIDE 19

SLIDE 20

Theorem



- The feasible set has  $2^n$  vertices
- The vertices can be ordered so that each one is adjacent to and has lower cost than the previous one.
- There exists a pivoting rule under which the simplex method requires  $2^n - 1$  changes of basis before it terminates.

## 7 The Diameter of polyhedra

SLIDE 21

- Given a polyhedron  $P$ , and  $\mathbf{x}, \mathbf{y}$  vertices of  $P$ , the distance  $d(\mathbf{x}, \mathbf{y})$  is the minimum number of jumps from one vertex to an adjacent one to reach  $\mathbf{y}$  starting from  $\mathbf{x}$ .
- The diameter  $D(P)$  is the maximum of  $d(\mathbf{x}, \mathbf{y}) \forall \mathbf{x}, \mathbf{y}$ .

SLIDE 22

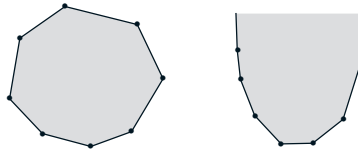
- $\Delta(n, m)$  as the maximum of  $D(P)$  over all *bounded* polyhedra in  $\mathbb{R}^n$  that are represented in terms of  $m$  inequality constraints.
- $\Delta_u(n, m)$  is like  $\Delta(n, m)$  but for possibly unbounded polyhedra.

### 7.1 The Hirsch Conjecture

SLIDE 23

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$$\Delta(2, m) = \left\lfloor \frac{m}{2} \right\rfloor, \quad \Delta_u(2, m) = m - 2$$



- **Hirsch Conjecture:**  $\Delta(n, m) \leq m - n$ .

SLIDE 24

- We know that

$$\Delta_u(n, m) \geq m - n + \left\lfloor \frac{n}{5} \right\rfloor$$

$$\Delta(n, m) \leq \Delta_u(n, m) < m^{1+\log_2 n} = (2n)^{\log_2 m}$$

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