

15.081J/6.251J Introduction to Mathematical  
Programming

Lecture 6: The Simplex Method II

# 1 Outline

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- Revised Simplex method
- The full tableau implementation
- Anticycling

# 2 Revised Simplex

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Initial data:  $\mathbf{A}, \mathbf{b}, \mathbf{c}$

1. Start with basis  $\mathbf{B} = [\mathbf{A}_{B(1)}, \dots, \mathbf{A}_{B(m)}]$  and  $\mathbf{B}^{-1}$ .
2. Compute  $\mathbf{p}' = \mathbf{c}'_B \mathbf{B}^{-1}$   
 $\bar{c}_j = c_j - \mathbf{p}' \mathbf{A}_j$ 
  - If  $\bar{c}_j \geq 0$ ;  $x$  optimal; stop.
  - Else select  $j : \bar{c}_j < 0$ .

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3. Compute  $\mathbf{u} = \mathbf{B}^{-1} \mathbf{A}_j$ .
  - If  $\mathbf{u} \leq \mathbf{0} \Rightarrow$  cost unbounded; stop
  - Else
4.  $\theta^* = \min_{1 \leq i \leq m, u_i > 0} \frac{x_{B(i)}}{u_i} = \frac{u_{B(l)}}{u_l}$
5. Form a new basis  $\bar{\mathbf{B}}$  by replacing  $\mathbf{A}_{B(l)}$  with  $\mathbf{A}_j$ .
6.  $y_j = \theta^*$ ,  $y_{B(i)} = x_{B(i)} - \theta^* u_i$

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7. Form  $[\bar{\mathbf{B}}^{-1} | \mathbf{u}]$
8. Add to each one of its rows a multiple of the  $l$ th row in order to make the last column equal to the unit vector  $\mathbf{e}_l$ .  
The first  $m$  columns is  $\bar{\mathbf{B}}^{-1}$ .

## 2.1 Example

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$$\begin{array}{llllll} \min & x_1 + & 5x_2 & -2x_3 & & \\ \text{s.t.} & x_1 + & x_2 + & x_3 & \leq & 4 \\ & x_1 & & & \leq & 2 \\ & & & x_3 & \leq & 3 \\ & & 3x_2 + & x_3 & \leq & 6 \\ & x_1, & x_2, & x_3 & \geq & 0 \end{array}$$

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$$B = \{\mathbf{A}_1, \mathbf{A}_3, \mathbf{A}_6, \mathbf{A}_7\}, \quad \text{BFS: } \mathbf{x} = (2, 0, 2, 0, 0, 1, 4)'$$

$$\bar{\mathbf{c}}' = (0, 7, 0, 2, -3, 0, 0)$$

$$\mathbf{B} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix}, \quad \mathbf{B}^{-1} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ -1 & 1 & 1 & 0 \\ -1 & 1 & 0 & 1 \end{bmatrix}$$

$$(u_1, u_3, u_6, u_7)' = \mathbf{B}^{-1}\mathbf{A}_5 = (1, -1, 1, 1)'$$

$$\theta^* = \min\left(\frac{2}{1}, \frac{1}{1}, \frac{4}{1}\right) = 1, \quad l = 6$$

$l = 6$  ( $\mathbf{A}_6$  exits the basis).

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$$[\mathbf{B}^{-1}|\mathbf{u}] = \begin{bmatrix} 0 & 1 & 0 & 0 & 1 \\ 1 & -1 & 0 & 0 & -1 \\ -1 & 1 & 1 & 0 & 1 \\ -1 & 1 & 0 & 1 & 1 \end{bmatrix}$$

$$\Rightarrow \bar{\mathbf{B}}^{-1} = \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 0 & 1 & 0 \\ -1 & 1 & 1 & 0 \\ 0 & 0 & -1 & 1 \end{bmatrix}$$

## 2.2 Practical issues

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- **Numerical Stability**

$\mathbf{B}^{-1}$  needs to be computed from scratch once in a while, as errors accumulate

- **Sparsity**

$\mathbf{B}^{-1}$  is represented in terms of sparse triangular matrices

## 3 Full tableau implementation

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$-\mathbf{c}'_B \mathbf{B}^{-1} \mathbf{b}$	$\mathbf{c}' - \mathbf{c}'_B \mathbf{B}^{-1} \mathbf{A}$
$\mathbf{B}^{-1} \mathbf{b}$	$\mathbf{B}^{-1} \mathbf{A}$

or, in more detail,

$-\mathbf{c}'_B \mathbf{x}_B$	$\bar{c}_1$	...	$\bar{c}_n$
$x_{B(1)}$			
$\vdots$	$\mathbf{B}^{-1} \mathbf{A}_1$	...	$\mathbf{B}^{-1} \mathbf{A}_n$
$x_{B(m)}$			

### 3.1 Example

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$$\begin{aligned} \min \quad & -10x_1 - 12x_2 - 12x_3 \\ \text{s.t.} \quad & x_1 + 2x_2 + 2x_3 \leq 20 \\ & 2x_1 + x_2 + 2x_3 \leq 20 \\ & 2x_1 + 2x_2 + x_3 \leq 20 \\ & x_1, x_2, x_3 \geq 0 \end{aligned}$$

$$\begin{aligned} \min \quad & -10x_1 - 12x_2 - 12x_3 \\ \text{s.t.} \quad & x_1 + 2x_2 + 2x_3 + x_4 = 20 \\ & 2x_1 + x_2 + 2x_3 + x_5 = 20 \\ & 2x_1 + 2x_2 + x_3 + x_6 = 20 \\ & x_1, \dots, x_6 \geq 0 \end{aligned}$$

BFS:  $\mathbf{x} = (0, 0, 0, 20, 20, 20)'$

$\mathbf{B} = [\mathbf{A}_4, \mathbf{A}_5, \mathbf{A}_6]$

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	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$
	-10	-12	-12	0	0	0
$x_4 =$	20	1	2	2	1	0
$x_5 =$	20	2*	1	2	0	1
$x_6 =$	20	2	2	1	0	0

$$\bar{\mathbf{c}}' = \mathbf{c}' - \mathbf{c}'_B \mathbf{B}^{-1} \mathbf{A} = \mathbf{c}' = (-10, -12, -12, 0, 0, 0)$$

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	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$
	100	0	-7	-2	0	5
$x_4 =$	10	0	1.5	1*	1	-0.5
$x_1 =$	10	1	0.5	1	0	0.5
$x_6 =$	0	0	1	-1	0	-1

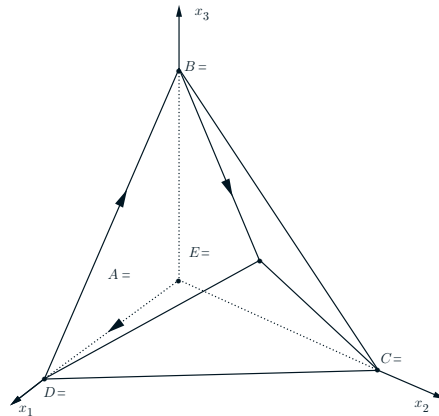
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	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$
	120	0	-4	0	2	4
$x_3 =$	10	0	1.5	1	1	-0.5
$x_1 =$	0	1	-1	0	-1	1
$x_6 =$	10	0	2.5*	0	1	-1.5

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	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$
136	0	0	0	3.6	1.6	1.6
$x_3 =$	4	0	0	1	0.4	0.4
$x_1 =$	4	1	0	0	-0.6	0.4
$x_2 =$	4	0	1	0	0.4	-0.6

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## 4 Comparison of implementations

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	Full tableau	Revised simplex
Memory	$O(mn)$	$O(m^2)$
Worst-case time	$O(mn)$	$O(mn)$
Best-case time	$O(mn)$	$O(m^2)$

## 5 Anticycling

### 5.1 Degeneracy in Practice

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Does degeneracy really happen in practice?

$$\sum_{j=1}^n x_{ij} = 1$$

$$\sum_{i=1}^n x_{ij} = 1$$

$$x_{ij} \geq 0$$

$n!$  vertices:

For each vertex  $\exists 2^{n-1}n^{n-2}$  different bases ( $n = 8$ ) for each vertex  $\exists 33, 554, 432$  bases.

## 5.2 Perturbations

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$$\begin{array}{ll} (P) \min & \mathbf{c}'\mathbf{x} \\ \text{s.t.} & \mathbf{Ax} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array} \quad \begin{array}{ll} (P_\epsilon) \min & \mathbf{c}'\mathbf{x} \\ \text{s.t.} & \mathbf{Ax} = \mathbf{b} + \begin{pmatrix} \epsilon \\ \epsilon^2 \\ \vdots \\ \epsilon^m \end{pmatrix} \\ & \mathbf{x} \geq \mathbf{0}. \end{array}$$

### 5.2.1 Theorem

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$\exists \epsilon_1 > 0$ : for all  $0 < \epsilon < \epsilon_1$

$$\begin{array}{ll} \mathbf{Ax} = \mathbf{b} + \begin{pmatrix} \epsilon \\ \vdots \\ \epsilon^m \end{pmatrix} \\ \mathbf{x} \geq \mathbf{0} \end{array}$$

is non-degenerate.

### 5.2.2 Proof

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Let  $\mathbf{B}_1, \dots, \mathbf{B}_r$  be all the bases.

$$\mathbf{B}_r^{-1} \left[ \mathbf{b} + \begin{pmatrix} \epsilon \\ \vdots \\ \epsilon^m \end{pmatrix} \right] = \begin{bmatrix} \bar{b}_1^r + \mathbf{B}_{11}^r \epsilon + \dots + \mathbf{B}_{1m}^r \epsilon^m \\ \vdots \\ \bar{b}_m^r + \mathbf{B}_{m1}^r \epsilon + \dots + \mathbf{B}_{mm}^r \epsilon^m \end{bmatrix}$$

where:

$$\mathbf{B}_r^{-1} = \begin{bmatrix} \mathbf{B}_{11}^r & \dots & \mathbf{B}_{1m}^r \\ \vdots & & \vdots \\ \mathbf{B}_{m1}^r & \dots & \mathbf{B}_{mm}^r \end{bmatrix}, \mathbf{B}_r^{-1} \mathbf{b} = \begin{bmatrix} \bar{b}_1^r \\ \vdots \\ \bar{b}_m^r \end{bmatrix}$$

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- $\bar{b}_i^r + \mathbf{B}_{i1}^r \theta + \dots + \mathbf{B}_{im}^r \theta^m$  is a polynomial in  $\theta$
- Roots  $\theta_{i,1}^r, \theta_{i,2}^r, \dots, \theta_{i,m}^r$
- If  $\epsilon \neq \theta_{i,1}^r, \dots, \theta_{i,m}^r \Rightarrow \bar{b}_i^r + \mathbf{B}_{i1}^r \epsilon + \dots + \mathbf{B}_{im}^r \epsilon^m \neq 0$ .
- Let  $\epsilon_1$  the smallest positive root  $\Rightarrow 0 < \epsilon < \epsilon_1$  all RHS are  $\neq 0 \Rightarrow$  non-degeneracy.

### 5.3 Lexicography

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- $\mathbf{u}$  is lexicographically larger than  $\mathbf{v}$ ,  $\mathbf{u} \stackrel{L}{>} \mathbf{v}$ , if  $\mathbf{u} \neq \mathbf{v}$  and the first nonzero component of  $\mathbf{u} - \mathbf{v}$  is positive.
- Example:

$$(0, 2, 3, 0) \stackrel{L}{>} (0, 2, 1, 4),$$

$$(0, 4, 5, 0) \stackrel{L}{<} (1, 2, 1, 2).$$

### 5.4 Lexicography-Perturbation

#### 5.4.1 Theorem

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Let  $\mathbf{B}$  be a basis of  $\mathbf{Ax} = \mathbf{b}$ ,  $\mathbf{x} \geq \mathbf{0}$ . Then  $\mathbf{B}$  is feasible for  $\mathbf{Ax} = \mathbf{b} + (\epsilon, \dots, \epsilon^m)'$ ,  $\mathbf{x} \geq \mathbf{0}$  for sufficiently small  $\epsilon$  if and only if

$$\mathbf{u}_i = (\bar{b}_i, B_{i1}, \dots, B_{im}) \stackrel{L}{>} \mathbf{0}, \forall i$$

$$\begin{aligned} \mathbf{B}^{-1} &= (B_{ij}) \\ (\mathbf{B}^{-1}\mathbf{b})_i &= (\bar{b}_i) \end{aligned}$$

#### 5.4.2 Proof

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$\mathbf{B}$  is feasible for perturbed problem “ $\Leftrightarrow \mathbf{B}^{-1}(\mathbf{b} + (\epsilon, \dots, \epsilon^m)') \geq \mathbf{0} \Leftrightarrow \bar{b}_i + \mathbf{B}_{i1}\epsilon + \dots + \mathbf{B}_{im}\epsilon^m \geq 0 \forall i$   
 $\Leftrightarrow$  First non-zero component of  $\mathbf{u}_i = (\bar{b}_i, B_{i1}, \dots, B_{im})$  is positive  $\forall i$ .

### 5.5 Summary

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1. We start with:  $(P) : \mathbf{Ax} = \mathbf{b}$ ,  $\mathbf{x} \geq \mathbf{0}$
2. We introduce  $(P_\epsilon) : \mathbf{Ax} = \mathbf{b} + (\epsilon, \dots, \epsilon^m)'$ ,  $\mathbf{x} \geq \mathbf{0}$
3. A basis is feasible + non-degenerate in  $(P_\epsilon) \Leftrightarrow \mathbf{u}_i \stackrel{L}{>} \mathbf{0}$  in  $(P)$ .
4. If we maintain  $\mathbf{u}_i \stackrel{L}{>} \mathbf{0}$  in  $(P) \Rightarrow (P_\epsilon)$  is non-degenerate  $\Rightarrow$  Simplex is finite in  $(P_\epsilon)$  for sufficiently small  $\epsilon$ .

### 5.6 Lexicographic pivoting rule

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1. Choose an entering column  $\mathbf{A}_j$  arbitrarily, as long as  $\bar{c}_j < 0$ ;  $\mathbf{u} = \mathbf{B}^{-1}\mathbf{A}_j$ .
2. For each  $i$  with  $u_i > 0$ , divide the  $i$ th row of the tableau (including the entry in the zeroth column) by  $u_i$  and choose the lexicographically smallest row. If row  $l$  is lexicographically smallest, then the  $l$ th basic variable  $x_{B(l)}$  exits the basis.

### 5.6.1 Example

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- $j = 3$

1	0	5	3	...
2	4	6	-1	...
3	0	7	9	...

- $x_{B(1)}/u_1 = 1/3$  and  $x_{B(3)}/u_3 = 3/9 = 1/3$ .
- We divide the first and third rows of the tableau by  $u_1 = 3$  and  $u_3 = 9$ , respectively, to obtain:

1/3	0	5/3	1	...
*	*	*	*	...
1/3	0	7/9	1	...

- Since  $7/9 < 5/3$ , the third row is chosen to be the pivot row, and the variable  $x_{B(3)}$  exits the basis.

### 5.6.2 Uniqueness

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- Why lexicographic pivoting rule always leads to a unique choice for the exiting variable?
- Otherwise, two rows in tableau proportional  $\Rightarrow \text{rank}(\mathbf{B}^{-1}\mathbf{A}) < m \Rightarrow \text{rank}(\mathbf{A}) < m$

### 5.7 Theorem

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If simplex starts with all the rows in the simplex tableau, other than the zeroth row, lexicographically positive and the lexicographic pivoting rule is followed, then

- Every row of the simplex tableau, other than the zeroth row, remains lexicographically positive throughout the algorithm.
- The zeroth row strictly increases lexicographically at each iteration.
- The simplex method terminates after a finite number of iterations.

### 5.8 Smallest subscript pivoting rule

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1. Find the smallest  $j$  for which the reduced cost  $\bar{c}_j$  is negative and have the column  $\mathbf{A}_j$  enter the basis.
2. Out of all variables  $x_i$  that are tied in the test for choosing an exiting variable, select the one with the smallest value of  $i$ .



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