DM545 Linear and Integer Programming

Lecture 4 Exception Handling and Initialization

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Outline

1. Exception Handling

2. Initialization

Simplex: Exception Handling, Overview

Handling exceptions in the Simplex Method

- Unboundedness
- 2. More than one solution
- 3. Degeneracies
 - benign
 - cycling
- 4. Infeasible starting
 Phase I + Phase II

- a. $F = \emptyset$
- b. $F \neq \emptyset$ and \exists solution
 - i) one solution
 - ii) infinite solution
- c. $F \neq \emptyset$ and $\not\exists$ solution

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Exception Handling

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Unboundedness

$$\begin{array}{cccc} \max & 2x_1 & + & x_2 \\ & & x_2 & \leq & 5 \\ -x_1 & + & x_2 & \leq & 1 \\ & & x_1, x_2 & \geq & 0 \end{array}$$

Initial tableau

 $\theta = \min\{\frac{b_i}{a_i}: a_{is} > 0, i = 1, \ldots, n\}$

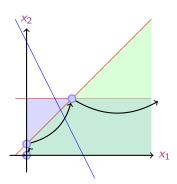
• x₂ entering, x₄ leaving

5

• x_1 entering, x_3 leaving

1		x1	1	x2	-	x3	1	x4	1	-z	1	ъl
	-+-		+-		+		+-		+-		+-	
I'=I	-1	1	1	0	-	1	1	-1	1	0	1	4
II'=II+I'	-1	0	1	1	-	1	1	0	1	0	1	5 I
	_+		+-		-+-		+-		-+-		+-	
III'=III-3I'	-	0	1	0	1	-3	1	2	1	1	1	-13

 x_4 was already in basis but for both I and II ($x_2 + 0x_4 = 5$), x_4 can increase arbitrarily



∞ solutions

$$\begin{array}{rll} \max & x_1 & + & x_2 \\ & 5x_1 & + & 10x_2 & \leq & 60 \\ & 4x_1 & + & 4x_2 & \leq & 40 \\ & & x_1, x_2 & \geq & 0 \end{array}$$

Initial tableau

• x₂ enters, x₃ leaves



• x_1 enters, x_4 leaves

$$\mathbf{x} = (8, 2, 0, 0), z = 10$$

nonbasic variables typically have reduced costs $\neq 0$. Here x_3 has r.c.

- = 0. Let's make it enter the basis
- x₃ enters, x₂ leaves

$$\mathbf{x} = (10, 0, 10, 0), z = 10$$

There are 2 optimal solutions \rightsquigarrow all their convex combinations are optimal solutions:

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$$\mathbf{x} = \sum_{i} \alpha_{i} \mathbf{x}_{i} \qquad \mathbf{x}_{1}^{T} = [8, 2, 0, 0]$$

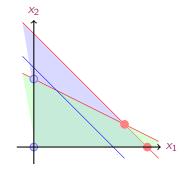
$$\alpha_{i} \geq 0 \qquad \mathbf{x}_{2}^{T} = [10, 0, 10, 0]$$

$$\alpha_{i} = \alpha$$

$$\sum \alpha_{i} = 1 \qquad \alpha_{2} = 1 - \alpha$$

$$\mathbf{x}_{1}^{T} = [8, 2, 0, 0] \\
\mathbf{x}_{2}^{T} = [10, 0, 10, 0] \\
\alpha_{1} = \alpha$$

$$\begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \\ x_{4} \end{bmatrix} = \alpha \begin{bmatrix} 8 \\ 2 \\ 0 \\ 0 \end{bmatrix} + (1-\alpha) \begin{bmatrix} 10 \\ 0 \\ 10 \\ 0 \end{bmatrix}$$



$$x_1 = 8\alpha + 10(1 - \alpha)$$

 $x_2 = 2\alpha$
 $x_3 = 10(1 - \alpha)$
 $x_4 = 0$

Degeneracy

Initial tableau

 $b_i = 0$ (one basic var. is zero) might lead to cycling

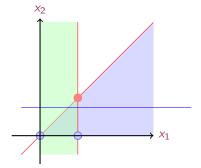
• degenerate pivot step: not improving, the entering variable stays at zero



• now nondegenerate:

											b	
												•
											2	
				•		•		•		•	2	
- 1	U	- 1	U	- 1	-1	- 1	-1	-	1	-	-2	П

$$x_1 = 2, x_2 = 2, z = 2$$



 $\geq n+1$ constraints meet at a vertex

Def: Improving variable, one with positive reduced cost

Under certain pivoting rules cycling can happen. So far we chose an arbitrary improving variable to enter.

Degenerate conditions may appear often in practice but cycling is rare and some pivoting rules prevent cycling. (Ex. 7 Sheet 3 shows the smallest possible example)

Theorem

If the simplex fails to terminate, then it must cycle.

Proof:

- there is a finite number of basis and simplex chooses to always increase the cost
- hence the only situation for not terminating is that a basis must appear again. Two dictionaries with the same basis are the same (related to uniqueness of basic solutions)

Pivot Rules

Rules for breaking ties in selecting entering improving variables (more important than selecting leaving variables)

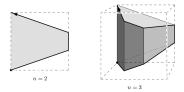
- Largest Coefficient: the improving var with largest coefficient in last row of the tableau.
 Original Dantzig's rule, can cycle
- Largest increase: absolute improvement: $argmax_j\{c_j\theta_j\}$ computationally more costly
- Steepest edge the improving var that if entering in the basis moves the
 current basic feasible sol in a direction closest to the direction of the
 vector c (ie, maximizes the cosine of the angle between the two vectors):

$$\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos \theta \implies \max \frac{\mathbf{c}^T (\mathbf{x}_{\text{new}} - \mathbf{x}_{\text{old}})}{\|\mathbf{x}_{\text{new}} - \mathbf{x}_{\text{old}}\|}$$

- Bland's rule chooses the improving var with the lowest index and, if there are more than one leaving variable, the one with the lowest index Prevents cycling but is slow
- Random edge select var uniformly at random among the improving ones
- Perturbation method perturb values of b_i terms to avoid $b_i = 0$, which must occur for cycling.
 - To avoid cancellations: $0<\epsilon_m\ll\epsilon_{m-1}\ll\cdots\ll\epsilon_1\ll 1$ can be shown to be the same as lexicographic method, which prevents cycling

Efficiency of Simplex Method

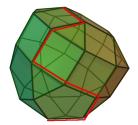
- Trying all points is $\approx 4^m$
- In practice between 2m and 3m iterations
- Klee and Minty 1978 constructed an example that requires 2ⁿ 1 iterations:



• random shuffle of indexes + lowest index for entering + lexicographic for leaving: expected iterations $< e^{C\sqrt{n \ln n}}$

Efficiency of Simplex Method

- unknown if there exists a pivot rule that leads to polynomial time.
- Clairvoyant's rule: shortest possible sequence of steps Hirsh conjecture O(n) but best known $n^{1+\ln n}$



• smoothed complexity: slight random perturbations of worst-case inputs D. Spielman and S. Teng (2001), Smoothed analysis of algorithms: why the simplex algorithm usually takes polynomial time $O(max(n^5 \log^2 m, n^9 \log^4 n, n^3 \sigma^{-4}))$

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Initial Infeasibility

Initial tableau

→ we do not have an initial basic feasible solution!!

In general finding any feasible solution is difficult as finding an optimal solution, otherwise we could do binary search

Auxiliary Problem (I Phase of Simplex)

We introduce auxiliary variables:

$$w^* = \max -x_5 \equiv \min x_5$$

$$x_1 + x_2 + x_3 = 2$$

$$2x_1 + 2x_2 - x_4 + x_5 = 5$$

$$x_1, x_2, x_3, x_4, x_5 > 0$$

if $w^* = 0$ then $x_5 = 0$ and the two problems are equivalent if $w^* > 0$ then not possible to set x_5 to zero.

Initial tableau

Keep z always in basis

• we reach a canonical form simply by letting x_5 enter the basis:

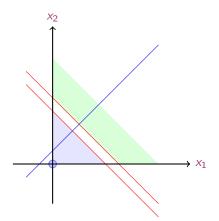
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	-+-		+.		-+-		-+-		+-		٠+٠		+-		+.		-
1	- [1	1	1	1	1	1	0	1	0	1	0	ı	0	1	2	1
1		2	1	2	1	0	1	-1	1	1	1	0		0	1	5	1
z		1	1	-1	1	0	1	0	1	0	1	1		0	1	0	1
	-+-		+		+-		+-		+-		+.		+-		+.		- [
IV+II	1	2	ī	2	Ι	0	Τ	-1	Ι	0	Τ	0	ī	1	Ι	5	Ĺ

now we have a basic feasible solution!

• x₁ enters, x₃ leaves

 $w^*=-1$ then no solution with $X_5=0$ exists then no feasible solution to initial problem

$$\begin{array}{rll} \max & x_1 & - & x_2 \\ & x_1 & + & x_2 & \leq 2 \\ & 2x_1 & + & 2x_2 & \geq 5 \\ & & x_1, x_2 & \geq 0 \end{array}$$



Initial Infeasibility - Another Example

Auxiliary problem (I phase):

- - → we do not have an initial basic feasible solution.
- set in canonical form:

• x_1 enters, x_5 leaves

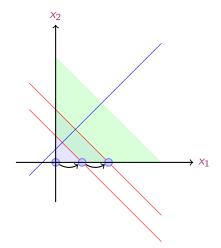
 $w^* = 0$ hence $x_5 = 0$ we have a starting feasible solution for the initial problem.

• (II phase) We keep only what we need:

								x4					
1	 +.		-+-		+-		+-		+-		+-		-
	l	0	-	0	1	1	1	1/2	-	0	1	1	1
	l	1	-	1	1	0		-1/2	-	0		1	1
1	 +.		-+-		+-		+-		+		+.		- [
z	ı	0	١	-2	1	0	1	1/2	١	1	Ι	-1	Ī

Optimal solution: $x_1 = 2, x_2 = 0, x_3 = 0, x_4 = 2, z = 2.$

$$\begin{array}{rrrr} \max & x_1 & - & x_2 \\ & x_1 & + & x_2 & \leq 2 \\ & 2x_1 & + & 2x_2 & \geq 2 \\ & & x_1, x_2 & \geq 0 \end{array}$$



In Dictionary Form

$$\begin{array}{rll} \max & x_1 & - & x_2 \\ & x_1 & + & x_2 & \leq 2 \\ & 2x_1 & + & 2x_2 & \geq 5 \\ & & x_1, x_2 & \geq 0 \end{array}$$

$$x_3 = 2 - x_1 - x_2$$

 $x_4 = -5 + 2x_1 + 2x_2$
 $z = x_1 + x_2$

sol. infeasible

We introduce corrections of infeasibility

$$x_3 = 2 - x_1 - x_2$$

 $x_4 = -5 + 2x_1 + 2x_2 + x_0$
 $z = -x_0$

It is still infeasible but it can be made feasible by letting x_0 enter the basis which variable should leave?

the most infeasible: the var with the \emph{b} term whose negative value has the largest magnitude

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