

15.053/8

February 21, 2013

Simplex Method Continued

Quote of the Day

“Everyone designs who devises courses of action aimed at changing existing situations into preferred ones.”

-- Herbert Simon

Today's Lecture

- **Very quick review of the simplex algorithm.**
- **Phase 1: How to obtain the initial bfs**
- **Finiteness (assuming bases do not repeat)**
 - Degeneracy
 - Anti-cycling rule(s)
- **Alternative optima**

A very quick review

-z	x_1	x_2	x_3	s_1	s_2	s_3		RHS
1	5	4.5	6	0	0	0		0
0	6	5	8	1	0	0		60
0	10	20	10	0	1	0		150
0	1	0	0	0	0	1		8

The **basic variables** here are $-z$, s_1 , s_2 , s_3 .

It is optional whether to call $-z$ basic.

The **basic feasible solution (bfs)** is

$$z = 0; x_1 = 0, x_2 = 0, x_3 = 0, s_1 = 60; s_2 = 150; s_3 = 8$$

A very quick review

-z	x_1	x_2	x_3	s_1	s_2	s_3		RHS	Ratio
1	5	4.5	6	0	0	0		0	
0	6	5	8	1	0	0		60	60/6
0	10	20	10	0	1	0		150	150/10
0	1	0	0	0	0	1		8	8/1

If all reduced costs are ≤ 0 , then you are optimal.
 Otherwise, choose a reduced cost that is positive.

We could have chosen the 5 or the 4.5 or the 6.

Use the min ratio rule to determine the pivot element
 (and the exiting variable).

Ending conditions: Optimality

If all coefficients in the z-row are nonpositive ($\bar{c}_i \leq 0$ for all i), then the current basic solution is optimal.

Basic Variable	-z	x_1	x_2	x_3	x_4	x_5		RHS
-z	1	0	-5	0	0	-1	=	-1
x_3	0	0	2	1	0	-2	=	1
x_4	0	0	-1	0	1	-2	=	7
x_1	0	1	6	0	0	0	=	3

Ending conditions: Unboundedness

If the z-row coefficient of x_s is positive for some s , and if all (other) coefficients in the column for x_s are nonpositive, then the optimal objective value is unbounded from above.

BV	-z	x_1	x_2	x_3	x_4	x_5		RHS
-z	1	0	-2	0	0	+1	=	-6
x_3	0	0	2	1	0	-2	=	4
x_4	0	0	-1	0	1	-2	=	2
x_1	0	1	6	0	0	0	=	3

$$\begin{aligned}
 z &= 6 + \infty \\
 x_1 &= 3 \\
 x_2 &= 0 \\
 x_3 &= 4 + 2\infty \\
 x_4 &= 2 + 2\infty \\
 x_5 &= \infty
 \end{aligned}$$

The pivot rule (min ratio version)

Basic Var	-z	x_1	x_2	x_3	x_4	x_5	=	RHS
-z	1	0	-2	0	0	6	=	-11
x_3	0	0	2	1	0	2	=	4
x_4	0	0	-1	0	1	-2	=	1
x_1	0	1	6	0	0	3	=	9

$= -z_0$

Choose a variable x_s (column) for which the z-row coefficient is positive.

Determine the constraint for which the following ratio is minimum. $\{\text{RHS coeff} / \text{Col coeff} : \text{Col coeff} > 0\}$

Constraint	(1)	(2)	(3)
Ratio	4/2	-2 < 0	9/3

The pivot

Basic Var	-z	x_1	x_2	x_3	x_4	x_5		RHS
-z	1	0	-2	0	0	6	=	-11
x_3	0	0	2	1	0	2	=	4
x_4	0	0	-1	0	1	-2	=	1
x_1	0	1	6	0	0	3	=	9

Basic Var	-z	x_1	x_2	x_3	x_4	x_5		RHS
-z	1	0	-8	-3	0	0	=	-23
x_5	0	0	1	0.5	0	1	=	2
x_4	0	0	1	1	1	0	=	5
x_1	0	1	3	-1.5	0	0	=	3

How do we find the first bfs?

- **Fact 1:** If start with a basic feasible solution, we can use the simplex algorithm to find an optimal basic feasible solution.
- **Fact 2:** If we start with an LP with “ \leq ” constraints and non-negative RHS, it is easy to find an initial bfs.
- **How can we use these facts to find the first bfs for problem P?**

$$\begin{array}{ll} \max & z = -3x_1 + x_2 + x_3 \\ \text{s.t.} & x_1 + x_2 + x_3 = 4 \\ & -2x_1 + x_2 - x_3 = 1 \\ & x_1 \geq 0, x_2 \geq 0, x_3 \geq 0 \end{array}$$

**Example of
Problem P**

How do we find the first bfs?

We will create a new problem P^* such that

1. It is easy to find a bfs for P^*
2. An optimal solution for P^* is feasible for P .

Choose a solution x

$$\begin{aligned} \text{s.t.} \quad & x_1 + x_2 + x_3 \leq 4 \\ & -2x_1 + x_2 - x_3 \leq 1 \\ & x_1 \geq 0, x_2 \geq 0, x_3 \geq 0 \end{aligned} \quad \text{Problem } P^*$$

and so that $x_1 + x_2 + x_3$ is as close to 4 as possible
and $-2x_1 + x_2 - x_3$ is as close to 1 as possible.

The Phase 1 Problem

minimize $y_1 + y_2$

maximize $v = -y_1 - y_2$

s.t. $x_1 + x_2 + x_3 + y_1 = 4$

$-2x_1 + x_2 - x_3 + y_2 = 1$

Problem P*

$x_1 \geq 0, x_2 \geq 0, x_3 \geq 0, y_1 \geq 0, y_2 \geq 0$

-v	x_1	x_2	x_3	y_1	y_2	RHS
1	0	0	0	-1	-1	0
0	1	1	1	1	0	4
0	-2	1	-1	0	1	1

Rules for creating Problem P*

Assume we start with equality constraints and RHS ≥ 0 .

Change the equality constraints to " \leq constraints".

Add "**artificial variables**" y as slack variables.

Minimize $y_1 + y_2 + \dots$

$$\text{minimize } y_1 + y_2$$

$$\text{s.t. } x_1 + x_2 + x_3 + y_1 = 4$$

$$-2x_1 + x_2 - x_3 + y_2 = 1$$

$$x_1 \geq 0, x_2 \geq 0, x_3 \geq 0, y_1 \geq 0, y_2 \geq 0$$

Problem P*

The Phase 1 Problem in canonical form

-v	x₁	x₂	x₃	y₁	y₂	RHS
1	0	0	0	-1	-1	0
0	1	1	1	1	0	4
0	-2	1	-1	0	1	1

Add constraints 1 and 2 to the objective in order to get into canonical form.

-v	x₁	x₂	x₃	y₁	y₂	RHS
1	-1	2	0	0	0	5
0	1	1	1	1	0	4
0	-2	1	-1	0	1	1

Time for a mental break

Even smart people get it wrong occasionally.

“Even considering the improvements possible... the gas turbine could hardly be considered a feasible application to airplanes because of the difficulties of complying with the stringent weight requirements.”

-- US National Academy Of Science, 1940

“People have been talking about a 3,000 mile high-angle rocket shot from one continent to another, carrying an atomic bomb and so directed as to be a precise weapon... I think we can leave that out of our thinking.”

-- Dr. Vannevar Bush, 1945

**Fooling around with alternating current is a waste of time.
Nobody will use it, ever.**

-- Thomas Edison

**There is not the slightest indication that nuclear energy
will be obtainable.**

-- Albert Einstein 1932

**Rail travel at high speed is not possible because
passengers, unable to breathe, would die of asphyxia.**

-- Dr. Dionysus Lardner, 1793-1859

**Inventions have long since reached their limit, and I see no
hope for future improvements.**

-- Julius Frontenus, 10 AD

The Phase 1 Problem

$-v$	x_1	x_2	x_3	y_1	y_2	
1	-1	2	0	0	0	5
0	1	1	1	1	0	4
0	-2	1	-1	0	1	1

The variables y_1, y_2, y_3 are called **artificial variables**.

Theorem. *There is a feasible solution for P if and only if the optimal objective value for P^* is 0.*

The next pivot

-v	x₁	x₂	x₃
1	-1	2	0
0	1	1	1
0	-2	1	-1

y₁	y₂
0	0
1	0
0	1

5
4
1

-v	x₁	x₂	x₃
1	3	0	2
0	3	0	2
0	-2	1	-1

y₁	y₂
0	-2
1	-1
0	1

3
3
1

One more pivot till the optimum for Phase 1

$-v$	x_1	x_2	x_3	y_1	y_2	
1	3	0	2	0	-2	3
0	3	0	2	1	-1	3
0	-2	1	-1	0	1	1

$-v$	x_1	x_2	x_3	y_1	y_2	
1	0	0	0	-1	-1	0
0	$3/2$	0	1	$1/2$	$-1/2$	$3/2$
0	$-1/2$	1	0	$1/2$	$1/2$	$5/2$

Let P be the original linear program. Let P^* be the LP after adding artificial variables. Suppose $y_j > 0$ in the optimal solution for P^* , where y_j is artificial. Then

- 1. The problem P has no feasible solution.**
- 2. The problem P is unbounded from above.**
- 3. If we ignore y_j , the solution is feasible for P .**
- 4. Either (1) or (2) is true.**

Phase 1, Phase 2

- **If there is a feasible solution for P, then Phase 1 ends with a feasible basis.**
- **To start Phase 2, put back the original objective function. Then put the tableau in canonical form. (The basis is almost in canonical form. But the z-row is not yet right.)**
- **Then pivot until optimal (or until there is proof of unboundedness.)**

End of Phase 1.

-v	x₁	x₂	x₃
1	0	0	0
0	3/2	0	1
0	-1/2	1	0

y₁	y₂
-1	-1
1/2	-1/2
1/2	1/2

0
3/2
5/2

-z	x₁	x₂	x₃
1	-3	1	1

If the RHS is greater than 0, then the next bfs has greater objective value.

-z	x_1	x_2	x_3	s_1	s_2	s_3		RHS
1	5	4.5	6	0	0	0		0
0	6	5	8	1	0	0		60
0	10	20	10	0	1	0		150
0	1	0	0	0	0	1		8

1	0.5	0.75	0	-0.75	0	0		-45
	0.75	0.625	1	0.125				7.5
	2.5	13.75		-1.25	1			75
	1					1		8

Is the Simplex Method Finite?

Theorem. *If the objective value improves at every iteration, then every basic feasible solution is different, and the simplex method is finite.*

Proof. Each canonical tableau is uniquely determined by choosing n basic variables out of n variables. The number of bases is at most:

$$\binom{n}{m} = \frac{n!}{m! (n - m)!}$$

If the RHS is 0, it is possible that the solution stays the same after a pivot.

-Z	x_1	x_2	x_3	s_1	s_2	s_3		RHS
1	5	4.5	6	0	0	0		0
0	6	5	8	1	0	0		0
0	10	20	10	0	1	0		150
0	1	0	0	0	0	1		8

								RHS
1	0.5	0.75	0	-0.75	0	0		0
	0.75	0.625	1	0.125				0
	2.5	13.75		-1.25	1			150
	1					1		8

If one of the basic variables is 0 (RHS is 0), we say that the tableau is degenerate.

If the RHS is 0, it is possible that the objective increases.

-z	x_1	x_2	x_3	s_1	s_2	s_3		RHS
1	5	4.5	6	0	0	0		0
0	6	5	8	1	0	0		60
0	10	20	10	0	1	0		150
0	1	0	0	0	0	1		0

1	0.5	0.75	0	-0.75	0	0		-45
	0.75	0.625	1	0.125				7.5
	2.5	13.75		-1.25	1			75
	1					1		0

If many bases are degenerate, it is possible for the simplex algorithm to **cycle**, that is, repeat a sequence of basic feasible solutions.

-Z	X₁	X₂	X₃	X₄	S₁	S₂	S₃		RHS
1	0.75	-20	0.5	-6	0	0	0		-3
0	0.25	-8	-1	9	1	0	0		0
0	0.5	-12	-0.5	3	0	1	0		0
0	0	0	1	0	0	0	1		1

1	0	4	3.5	-33	-3	0	0		-3
	1	-32	-4	36	4				0
		4	1.5	-15	-2	1			0
			1				1		1

The Klee and Minty example,
which can cycle.



Bland's Rule

- There are several ways of guaranteeing that no set of basic variables repeats.
- The simplest way of avoiding “cycling” is Bland's rule.

Bland's Rule:

1. Among variables that have a positive coefficient in the z-row, choose the one with least index.
2. Among rows that satisfy the min ratio rule, choose the one with least index.

Theorem. The simplex method with Bland's rule is finite.

Non-degeneracy and finiteness.

Lemma. *If the RHS of a tableau is positive, then the next pivot will lead to an improved objective function value.*

If a coefficient of the RHS of a tableau is 0, the tableau is ***degenerate*** (and the bfs is ***degenerate***). If a bfs is degenerate, it is possible that the next pivot will lead to a different basis, but the same solution.

Theorem. *If no basis is degenerate, then the simplex method is finite.*

Alternative Optima

	-z	x_1	x_2	x_3	x_4	x_5		RHS
A_0	1	0	0	0	0	-1	=	-2
A_1	0	0	2	1	0	-1	=	4
A_2	0	0	-1	0	1	2	=	1
A_3	0	1	6	0	0	3	=	3

Let $x_2 = \Delta$;

$$x_1 = 3 - 6\Delta$$

$$x_2 = \Delta$$

$$x_3 = 4 - 2\Delta$$

$$x_4 = 1 + \Delta$$

$$x_5 = 0$$

$$z = 2$$

This tableau satisfies the optimality conditions.

If a tableau satisfies the optimality conditions, and if $\bar{c}_j = 0$ for a nonbasic variable, then there may be multiple alternative optima solutions.

Non-degeneracy guarantees that we can choose $\Delta > 0$.

Alternative Optima and Pivoting

	-z	x_1	x_2	x_3	x_4	x_5		RHS
A_0	1	0	0	0	0	-1	=	-2
A_1	0	0	2	1	0	-1	=	4
A_2	0	0	-1	0	1	2	=	1
A_3	0	1	6	0	0	3	=	3

If a tableau satisfies the optimality conditions, and if $\bar{c}_j = 0$ for a nonbasic variable, we can pivot to get an alternative optimal bfs. (or prove that there is a ray along which the objective stays the same).

	-z	x_1	x_2	x_3	x_4	x_5		RHS
$B_0 = A_0$	1	0	0	0	0	-1	=	-2
$B_1 = A_1 - 2 B_3$	0	-1/3	0	1	0	-2	=	3
$B_2 = A_2 + B_3$	0	1/6	0	0	1	2.5	=	1.5
$B_3 = A_3/6$	0	1/6	1	0	0	.5	=	.5

Overview

- **The simplex method has been a huge success in optimization.**
 - It solves linear programs efficiently
 - We can solve problems with millions of variables
 - It can be a starting point for problems that are not linear
- **The simplex method requires some simple techniques to get started**
 - Transformation into standard form
 - Phase 1 of the simplex algorithm
 - In practice, it requires lots of implementation care
- **Degeneracy and techniques to avoid “cycling”.**
- **Alternative optima**

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