Linear Programming: Chapter 3 Degeneracy

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Degeneracy

Definitions.

A dictionary is degenerate if one or more "rhs"-value vanishes.

Example:

A *pivot* is degenerate if the objective function value does not change.

Examples (based on above dictionary):

- 1. If x_2 enters, then w_5 must leave, pivot is degenerate.
- 2. If w_1 enters, then w_2 must leave, pivot is *not* degenerate.

Cycling

Definition.

A *cycle* is a sequence of pivots that returns to the dictionary from which the cycle began.

Note: Every pivot in a cycle must be degenerate. Why?

Pivot Rules.

Definition.

Explicit statement for how one chooses entering and leaving variables (when a choice exists).

Largest-Coefficient Rule.

A common pivot rule for entering variable:

Choose the variable with the largest coefficient in the objective function.

Hope.

Some pivot rule, such as the largest coefficient rule, will be proven never to cycle.

Hope Fades

An example that cycles using the following pivot rules:

- entering variable: largest-coefficient rule.
- leaving variable: smallest-index rule.

$$\frac{\zeta = 10x_1 - 57x_2 - 9x_3 - 24x_4}{w_1 = -0.5x_1 + 5.5x_2 + 2.5x_3 - 9x_4}
w_2 = -0.5x_1 + 1.5x_2 + 0.5x_3 - x_4}
w_3 = 1 - x_1.$$

Here's a demo of cycling...

							_				
obj	=	0.0	+	10	x1 +	-57	x2 +	-9	x3 +	-24	×4
w1	=	0.0	-	0.5	x1 -	-5.5	x2 -	-2.5	x 3 -	9	x4
w2	=	0.0	-	0.5	x1 -	-1.5	x2 -	-0.5	x 3 -	1	x4
w3	=	1	-	1	x1 -	0.0	x2 -	0.0	x 3 -	0.0	x4
obj	=	0	+	-20	w1 +	53	x2 +	41	x3 +	-204	×4
x1	=	0	-	2	w1 -	-11	x2 -	-5	x 3 -	18	×4
w2	=	0	-	-1	w1 -	4	×2 -	2	x 3 -	-8	×4
w3	=	1	-	-2	w1 -	11	×2 -	5	x3 -	-18	×4
obj	=	0	+	-6.75	w1 +	-13.25	w2 +	14.5	x3 +	-98	x4
x1	=	0	-	-0.75	w1 -	2.75	w2 -	0.5	x 3 -	-4	x4
x 2	=	0	-	-0.25	w1 -	0.25	w2 -	0.5	x 3 -	-2	x4
w3	=	1	-	0.75	w1 -	-2.75	w2 -	-0.5	x3 -	4	x4
obj	=	0	+	15	w1 +	-93	w2 +	-29	x1 +	18	x4
x 3	=	0	-	-1.5	w1 -	5.5	w2 -	2	x1 -	-8	×4
x 2	=	0	-	0.5	w1 -	-2.5	w2 -	-1	x1 -	2	×4
w3	=	1	-	0	w1 -	0	w2 -	1	x1 -	0	×4

-1				40 F	w1 +	B0 F	w2 +	-00	x1 +		x 2
obj	=	0	_ +	10.5	w1 +	-70.5	w2 +	-20	x1 +	-9	**
x 3	=	0	-	0.5	w1 -	-4.5	w2 -	-2	x1 -	4	×2
x4	=	0	-	0.25	w1 -	-1.25	w2 -	-0.5	x1 -	0.5	x2
w3	=	1	-	0	w1 -	0	w2 -	1	x1 -	0	x2
obj	=	0	+	-21	x 3 +	24	w2 +	22	x1 +	-93	x 2
w1	=	0	-	2	x 3 -	-9	w2 -	-4	x1 -	8	x2
x4	=	0	-	-0.5	x3 -	1	w2 -	0.5	x1 -	-1.5	x2
w3	=	1	-	0	x3 -	0	w2 -	1	x1 -	0	x2
obj	=	0	+	-9	x 3 +	-24	x4 +	10	x1 +	-57	x 2
w1	=	0	-	-2.5	x3 -	9	x4 -	0.5	x1 -	-5.5	x2
w2	=	0	-	-0.5	x3 -	1	x4 -	0.5	x1 -	-1.5	x2
w3	=	1	-	0	x3 -	0	x4 -	1	x1 -	0	x2

Perturbation Method

Whenever a vanishing "rhs" appears perturb it. If there are lots of them, say k, perturb them all. Make the perturbations at different *scales*:

other nonzero data $\gg \epsilon_1 \gg \epsilon_2 \gg \cdots \gg \epsilon_k > 0$.

An Example.

	Current Dictionary															
obj	=	0.0	+	0.0	e1	+	0.0	e2	+	0.0	е3	+	2.0	x1 +	4.0	×2
w1	=	0.0	+	1.0	e1	+	0.0	e2	+	0.0	е3	-	-1.0	x1 -	1.0	x 2
w2	=	0.0	+	0.0	e1	+	1.0	e2	+	0.0	е3	-	-3.0	x1 -	1.0	×2
w3	=	0.0	+	0.0	e1	+	0.0	e2	+	1.0	е3	-	4.0	x1 -	-1.0	×2

Entering variable: x_2 Leaving variable: w_2

	Current Dictionary															
obj	=	0.0	+	0.0	e1	+	4.0	e2	+	0.0	e3	+	14.0	x1 +	-4.0	w2
w1	=	0.0	+	1.0	e1	+	-1.0	e2	+	0.0	e3	-	2.0	x1 -	-1.0	w2
x 2	=	0.0	+	0.0	e1	+	1.0	e2	+	0.0	e3	-	-3.0	x1 -	1.0	w2
w3	=	0.0	+	0.0	e1	+	1.0	e2	+	1.0	e3	-	1.0	x1 -	1.0	w2

Perturbation Method—Example Con't.

Recall current dictionary:

	Current Dictionary														
obj	=	0.0	+	0.0	e1	+	4.0	e2	+	0.0	e3	+	14.0	x1 +	-4.0 w2
w1	=	0.0	+	1.0	e1	+	-1.0	e2	+	0.0	e3	-	2.0	x1 -	-1.0 w2
x 2	=	0.0	+	0.0	e1	+	1.0	e2	+	0.0	e3	-	-3.0	x1 -	1.0 w2
w3	=	0.0	•	0.0	e1	+	1.0	e2	+	1.0	е3	-	1.0	x1 -	1.0 w2

Entering variable: x_1 Leaving variable: w_3

	Current Dictionary															
obj	=	0.0	+	0.0	e1	+	18.0	e2	+	14.0	е3	+	-14.0	w3 +	-18.0	w2
w1	=	0.0	+	1.0	e1	+	-3.0	e2	+	-2.0	е3	-	-2.0	w3 -	-3.0	w2
x2	=	0.0	•	0.0	e1	+	4.0	e2	+	3.0	е3	-	3.0	w3 -	4.0	w2
x1	=	0.0	•	0.0	e1	+	1.0	e2	+	1.0	е3	-	1.0	w3 -	1.0	w2

Other Pivot Rules

Smallest Index Rule.

Choose the variable with the smallest index (the x variables are assumed to be "before" the w variables).

Note: Also known as Bland's rule.

Random Selection Rule.

Select at random from the set of possibilities.

Greatest Increase Rule.

Pick the entering/leaving pair so as to maximize the increase of the objective function over all other possibilities.

Note: Too much computation.

Theoretical Results

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

Why?

Fundamental Theorem of Linear Programming. For an arbitrary linear program in standard form, the following statements are true:

- 1. If there is no optimal solution, then the problem is either infeasible or unbounded.
- 2. If a feasible solution exists, then a basic feasible solution exists.
- 3. If an optimal solution exists, then a basic optimal solution exists.

Geometry



