

From:

"Chapter 2. The Simplex
Algorithm",

Combinatorial Optimization:
Algorithms and Complexity.

by C. H. Papadimitriou and
K. Steiglitz.

Consider:

$$\min c^T x$$

$$Ax = b$$

$$x \geq 0$$

 A is $m \times n$, $m < n$. $\sim \otimes$
Assume A is of rank m .Let $B = \{A_{j_1}, \dots, A_{j_m}\}$ be a basis for A .Eg. If $A = \begin{bmatrix} 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 2 \\ 0 & 1 & 1 & 1 & 3 \end{bmatrix}$, then A is of rank $= 3$.
 $B = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$ is a basis for A .
A basic solution x is one for which:

$$x_j = 0, \text{ for } A_j \in B$$

$$x_{j_k} = k\text{th component of } B^{-1}b,$$

$$k = 1, \dots, m, A_{j_k} \notin B.$$

To find a solution to $Ax = b$, do:

1. Choose a set B of linearly independent columns of A .
2. Set all components of x corresponding to columns not in B equal to zero.
3. Solve the m resulting equations to determine the remaining components of x . These are the basic variables.

In our example.

$\frac{2}{8}$

1. $B = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

2. $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ 0 \\ 0 \end{bmatrix} \left\{ \begin{array}{l} \text{corresponding to} \\ \text{columns 1, 2, and 3.} \\ \text{remaining two.} \end{array} \right.$

3. Solve.

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = b$$

bfs it also
a solution
to (*)
for x_1, x_2, x_3 .
basic variables.

(p.42) 2.4 Moving from bfs to bfs

Let x_0 be a bfs of an instance of LP with matrix A , with:

$$B = \{A_{B(i)} : i = 1, \dots, m\}$$

If the basic components of x_0 are:

$$x_{i_0}, i_0 = 1, 2, \dots, m,$$

then:

$$\sum_{i=1}^m x_{i_0} A_{B(i)} = b, \quad x_{i_0} \geq 0.$$

(*)

Note that for the non-basic columns, we have to be able to express them in-terms of the basis: $\frac{3}{8}$

$$\sum_{i=1}^m x_{ij} A_{B(i)} = A_j \quad \text{---} \textcircled{**}$$

Multiply $\textcircled{**}$ by $\theta > 0$ and subtract from $\textcircled{*}$ to get:

$$\sum_{i=1}^m \underbrace{(x_{i0} - \theta x_{ij})}_{\Delta} A_{B(i)} + \theta A_j = b.$$

To change x and remain feasible, we must have the new x still remains positive. Thus, the maximum value for θ must be:

$$\theta_0 = \min_i \frac{x_{i0}}{x_{ij}} \quad \text{such that } x_{ij} > 0$$

$$(\text{from } x_{i0} - \theta x_{ij} = 0)$$

P.40

In term of a bfs, we have the cost:

$$z_0 = \sum_{i=1}^m x_{i0} C_{B(i)}$$

As we know, other columns can be expressed in terms of the basis columns:

$$A_j = \sum_{i=1}^m x_{ij} A_{B(i)}$$

Thus, when x_j enters, x_{ij} from $x_{B(i)}$ must leave, reducing

cost to $\underbrace{C_j - \sum_{i=1}^m x_{ij} C_{B(i)}}_{\text{definition of } z_j} = \bar{C}_j$

(cost of entering $x_j=1$)

where \bar{C}_j is hereby defined as the relative cost of column j .

It is thus profitable (since it will reduce cost) to bring-in column j , provided $\bar{C}_j < 0$.

If all $\bar{C}_j \geq 0$, then there is $\frac{5}{8}$
no hope of reducing cost;

\Rightarrow We are at global minimum.

Now return to (for unboundedness):

$$\sum_{i=1}^m x_{ij} A_{B(i)} = A_j.$$

If all the $x_{ij} \leq 0$, all i , some j , then
clearly if we also assume:

$$C_j - \sum_{i=1}^m x_{ij} C_{B(i)} < 0.$$

$$\Rightarrow C_j + \underbrace{(-x_{1j}) C_{B(1)}}_{\text{all of them are positive.}} + \dots + \underbrace{(-x_{mj}) C_{B(m)}}_{\text{all of them are positive.}} < 0$$

\Rightarrow At least one of:

$C_{B(1)}, \dots, C_{B(m)}, C_j$ is negative,

and setting that component $x_k = \infty$
grows unbounded to $-\infty$.

Pseudocode for the Simplex algorithm is given on p.49: $\frac{6}{8}$

Procedure simplex

begin

opt := 'no', unbounded := 'no';

(if either becomes 'yes', we terminate)

while opt == 'no' and
unbounded = 'no' do

if $\bar{c}_j \geq 0$ for all j ,

then opt := 'yes' (terminate)

else

begin

Choose any j with $c_j < 0$;

if $x_{ij} \leq 0$ for all i

then unbounded := 'yes'

else

find $\theta_0 = \min_{\substack{i \\ x_{ij} > 0}} \left[\frac{x_{i0}}{x_{ij}} \right]$

$= \frac{x_{k0}}{x_{kj}}$

and pivot on x_{kj}

end

end

p49. Example 2.6

$\frac{7}{8}$

Set $C = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$, and:

$$3x_1 + 2x_2 + x_3 = 1$$

$$5x_1 + x_2 + x_3 = 3$$

$$2x_1 + 5x_2 + x_3 + x_5 = 4$$

In "tableau form":

| | x_1 | x_2 | x_3 | x_4 | x_5 | |
|---|-------|-------|-------|-------|-------|-------|
| 0 | 1 | 1 | 1 | 1 | 1 | row 0 |
| 1 | 3 | 2 | 1 | 0 | 0 | row 1 |
| 3 | 5 | 1 | 1 | 1 | 0 | row 2 |
| 4 | 2 | 5 | 1 | 0 | 1 | row 3 |

1. Identify the basis for columns 3, 4, 5.

2. To compute the values for x_3, x_4, x_5 , "convert to identity":

1. Subtract row 1 from rows 2+3:

2. Subtract them from row 0 to get the relative costs of the non-basic variables (see book for proof)

After these two steps, we are left with:

$\frac{9}{8}$

$$\begin{array}{c|ccccc} -z = -6 & -3 & -3 & 0 & 0 & 0 \\ x_3 = 1 & 3 & (2) & 1 & 0 & 0 \\ x_4 = 2 & 2 & -1 & 0 & 1 & 0 \\ x_5 = 3 & -1 & 3 & 0 & 0 & 1 \end{array}$$

So, it is profitable to consider x_1, x_2
($\bar{c}_1 = -3, \bar{c}_2 = -3$).

For column 2:

$$\Theta_0 = \min_i \left[\frac{x_{i0}}{x_{i2}} \right]_{x_{i2} > 0} = \min \left\{ \frac{1}{2}, \frac{3}{3} \right\} = \frac{1}{2}$$

\uparrow
 $j=2$

Next, pivot on (2), making -3 zero:

$$\begin{array}{c|ccccc} & x_1 & x_2 & x_3 & x_4 & x_5 \\ \hline -z = -9/2 & 3/2 & 0 & 3/2 & 0 & 0 \\ \text{New var } x_2 = 1/2 & 3/2 & 1 & 1/2 & 0 & 0 \\ x_4 = 5/2 & 7/2 & 0 & 1/2 & 1 & 0 \\ x_5 = 3/2 & -11/2 & 0 & -3/2 & 0 & 1 \end{array}$$

Final answer: $\frac{1}{2} + \frac{5}{2} + \frac{3}{2} = \frac{9}{2}$

$\bar{c}_i \geq 0 \Rightarrow \text{stop}$