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The presentation of the simplex algorithm in "A Study Manual For Operations Research" (edited by Eugene A. Narragon) is difficult to follow. The purpose of this note is to show that a better understanding can be obtained by means of a geometric interpretation and the exchange method. The following ideas are not original, see for example E. Stiefel's text "An introduction to numerical mathematics", Academic Press, 1963.

We consider a linear programming problem of the following form:

Maximize $z=c_{1} x_{1}+\cdots+c_{n} x_{n}$ subject to
(1)

$$
x_{1} \geq 0, \cdots, x_{n} \geq 0, \eta_{1} \geq 0, \cdots, n_{m} \geq 0
$$

where
$\eta_{i}=a_{i}-\sum_{j=1}^{n} a_{i j} x_{j}$. It is assumed that $c_{j}>0, a_{j}>0, a_{i j}>0$. The set of feasible points (i.e., where the inequalities (1) are satisfied) is an intersection of $n+m$ halfspaces and is therefore a convex polyhedron in $R^{n}$. Thus the maximum of $z$ will be assumed on one of its vertices.

A combinatorial method to find the solution is as follows. Set $n$ of the variables $x_{1}, \cdots, x_{n}, n_{1}, \cdots, \eta_{m}$ equal to zero; thus we are examining $\binom{n+m}{n}$ points. If one of the other variables has
a negative value, the point is not feasible and can be ignored. If all the other variables are nonnegative, the point is a vertex and we compute the z-value. Finally, the optimum is found by comparison.

The simplex algorithm is a more economical method: with it, the optimal vertex can be found without examination of all the other vertixes.

We start with the vertex corresponding to $x_{1}=x_{2}=\cdots=x_{n}=0$. In a first step we exchange one of the $x$ 's, say $x_{j}$, against one of the $\eta^{\prime} s$, say $\eta_{i}$. This means that we proceed to the vertex that is given by $x_{\ell}=0(\ell \neq j), \eta_{i}=0$. In the second step we exchange one of the remaining $x$ 's against one of the remaining $\eta$ 's, etc. At each step, the choice of the pivot (i.e. the decision which $x$ should be exchanged against which $\eta$ ) is governed by two criteria:
(a) The new point should again be feasible, i.e. a vertex.
(b) The z-value should increase.

When it is not possible to perform another step, i.e. to find a pivot such that (a) and (b) are satisfied, we have arrived at the optimal vertex.

The remaining question is how the pivot should be chosen to satisfy (a) and (b). Before we look at it, we consider an
example: Maximize

$$
z=100 x_{1}+200 x_{2}+50 x_{3}
$$

subject to

$$
\begin{aligned}
5 x_{1}+5 x_{2}+10 x_{3} & \leq 1000 \\
10 x_{1}+8 x_{2}+5 x_{3} & \leq 2000 \\
10 x_{1}+5 x_{2} & \leq 500
\end{aligned}
$$

(This is the example discussed in the OR-Manual). We start out with the following table, corresponding to $x_{1}=x_{2}=x_{3}=0$ :

|  | $\mathrm{x}_{1}$ | $\mathrm{x}_{2}$ | $\mathrm{x}_{3}$ | 1 |
| :---: | :---: | :---: | :---: | :---: |
| $\eta_{1}$ | -5 | -5 | -10 | 1000 |
| $\eta_{2}=$ | -10 | -8 | -5 | 2000 |
| $n_{3}=$ | -10 | -5 | 0 | 500 |
| $\mathrm{z}=$ | 100 | 200 | 50 | 0 |

Then we exchange $x_{2}$ with $n_{3}$, and obtain the table corresponding to the vertex $x_{1}=x_{3}=\eta_{3}=0$;

|  | $\mathrm{x}_{1}$ | $\eta_{3}$ | $\mathrm{x}_{3}$ | 1 |
| :---: | :---: | :---: | :---: | :---: |
| $\eta_{1}=$ | 5 | 1 | -10) | 500 |
| $\mathrm{n}_{2}=$ | 6 | 1.6 | -. 2 | 1200 |
| $\mathrm{x}_{2}=$ | -2 | -. 2 | 0 | 100 |
| $\mathrm{z}=$ | -300 | -40 | 50 | 20000 |

The corresponding $z$-value is in the lower right-hand-corner and is 20000. Now we exchange $x_{3}$ with $\eta_{1}$ and obtain the table corresponding to the vertex $\mathrm{x}_{1}=\eta_{1}=\eta_{3}=0$ :


The last line shows that the maximal value of $z$ is obtained at this vertex and is 22500. The last column shows the values of $x_{3}, \eta_{2}, x_{2}$ at the optimal vertex. The remaining entries are not of interest and have not been computed.

Finally we discuss the choice of the pivot. Suppose that after a number of steps we have arrived at the following table:


We are dealing with a vertex, i.e. $a_{1} \geq 0, \cdots, a_{m} \geq 0$, and the corresponding $z$-value is $c>0$. If we choose $a_{i j}$ as the new pivot, the coefficients in the last column of the table will change as follows:
(i) $a_{i} \rightarrow-\frac{a_{i}}{a_{i j}}$
(ii) $a_{k}+a_{k}-\frac{a_{i} a_{k j}}{a_{i j}}, k \neq i$
(iii) $c \rightarrow c-\frac{a_{i} c_{j}}{a_{i j}}$

Since the new value of $a_{i}$ must be positive, it follows from (i) that the pivot $a_{i j}$ must be negative. Then it follows from postulate (b) and (iii) that $c_{j}$ must be positive. Now consider the condition that the new value of $a_{k}$ is nonnegative. This can be written as follows:

$$
\begin{equation*}
\frac{a_{k}}{-a_{k j}} \geq \frac{a_{i}}{-a_{i j}} \tag{2}
\end{equation*}
$$

Thus the pivot should be chosen according to the following rules:

- Pivot Column: Column $j$ can be chosen iff $c_{j}>0$ and if at least one of the coefficients in this column is negative.
- Pivot Row: Suppose that column j has been chosen as the pivot column. Now we determine the pivot row $i$ such that (2) holds. Thus whenever $a_{k j}$ is negative, we compute the ratio $a_{k} /-a_{k j}$. Then $i$ is the value of $k$ for which this ratio is the smallest. Let us revisit the numerical example. In the first table all $c_{\ell}^{\prime} s$ are positive and all $a_{k \ell}$ are negative. Thus all threee columns qualify as pivot column. Quite arbitrarily, we choose j $=2$. Then we compute

$$
\begin{aligned}
& a_{1} /-a_{12}=1000 / 5=200 \\
& a_{2} /-a_{22}=2000 / 8=250 \\
& a_{3} /-a_{32}=500 / 5=100
\end{aligned}
$$

Hence $i=3$. In the second table only the third column qualifies.
Hence $j=3$. We compute

$$
\begin{aligned}
& a_{1} /-a_{13}=500 / 10=50 \\
& a_{2} /-a_{23}=1200 / .2=6000
\end{aligned}
$$

Hence $i=1$. Finally, in the last table all $c_{\ell}$ 's are negative: The choice of a new pivot to satisfy (a) and (b) is not possible, and we have arrived at the optimal vertex.

