

Notes on the Revised Simplex Method

Motivation: Why do we want to use the Revised Simplex Method?

- Because when we are trying to find the optimal (basic) solution, it suffices to find out the values of the basis variables, since non-basis variables are evaluated to be zeros.

However, if we work with the entire tableau using the simplex method, especially when there are ~~way more~~ too many variables, ~~compared~~ we end up spending most of our time computing coefficients of non-basis variables which are not necessary.

So the idea is that we only maintain a mini-tableau, while using the more convenient matrices to keep track of other necessary numbers for the simplex method.

Example (same as the lecture notes)

$$\begin{aligned} \text{Max: } Z &= x_1 + x_2 + x_3 + 2x_4 \\ \left\{ \begin{array}{l} 3x_1 - x_2 + 4x_3 - x_4 \leq 4 \\ 2x_2 + x_4 \leq 5 \\ \vec{x} \geq 0 \end{array} \right. \end{aligned}$$

The whole tableau is :

	x_1	x_2	x_3	x_4	x_5	x_6	
Z	-1	-1	-1	-2	0	0	0
x_5	3	-1	4	-1	1	0	4
x_6	0	2	0	1	0	1	5

Choose x_5, x_6 as basis variables, and we only keep track of the basis variables and their coefficients as well as the entering variable in the mini-tableau:

B	A_B^{-1}	x_1	\vec{p}
x_5	1 0	(3)	4
x_6	0 1	0	5

Notes: B : basis ; A_B^{-1} : the coefficient matrix of the new (basis & non-basis) variables. \vec{p} : constants ; x_1 : entering, since $-1 < 0$ and we are doing in Bland's fashion.

B	A_B^{-1}	x_1	\vec{p}
$\star \triangle x_1$	$\frac{1}{3} 0$	(1)	$\frac{4}{3}$
x_6	0 1	0	5

Note: normalizing to be 1 ; x_5 is replaced with x_1

Now we need to decide the next entering variable.

Compute the vector: $\vec{r} = \vec{c}_B^T A_B^{-1} A_N - \vec{c}_N^T$, where \vec{c}_B^T denotes

the coefficient vector of the basis variables at this step;

\vec{c}_N^T denotes the coefficient vector of the non-basis variables at this step;
in our case, it is

A_B^{-1} as usual, the coefficient matrix of our mini-tableau at this step;

A_N denotes the coefficient matrix of the non-basis variables in the ~~initial~~ tableau;

\vec{c}_B^T is the coefficient vector of .

In our case, the initial tableau is

	x_1	x_2	x_3	x_4	x_5	x_6	
x_1	2	-1				0	0
x_5	3	-1	4	1	0	0	4
x_6	0	2	0	1	0	1	5

basis variables

The mini-tableau is :

	B	A_B^{-1}	entering	\vec{P}
x_1	$\frac{1}{3} 0$	1	$\frac{4}{3}$	
x_6	$0 1$	0	5	

$$\vec{c}_N^T = (1 \ 1 \ 2 \ 0)$$

So $\vec{c}_B^T = (1 \ 0)$, $A_B^{-1} = \left(\begin{smallmatrix} \frac{1}{3} & 0 \\ 0 & 1 \end{smallmatrix} \right)$, $A_N = \left(\begin{smallmatrix} 1 & 4 & 1 & 1 \\ 2 & 0 & 1 & 0 \end{smallmatrix} \right)$, and we have

$$\vec{r} = (1 \ 0) \left(\begin{smallmatrix} \frac{1}{3} & 0 \\ 0 & 1 \end{smallmatrix} \right) \left(\begin{smallmatrix} 1 & 4 & 1 & 1 \\ 2 & 0 & 1 & 0 \end{smallmatrix} \right) - (1 \ 1 \ 2 \ 0) = \left(-\frac{4}{3} \ \cancel{\frac{1}{3}} \ -\frac{7}{3} \ \frac{1}{3} \right)$$

Since the first entry of \vec{r} corresponds to x_2 and is non-positive, we can choose x_2 as the next entering variable.

And the column for x_2 is given by

$$A_B^{-1} A_2 = \begin{pmatrix} \frac{1}{3} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \end{pmatrix} = \begin{pmatrix} -\frac{1}{3} \\ 2 \end{pmatrix}$$

where A_2 is the column in A_N that corresponds to x_2 , so it is the first column.

The mini-tableau becomes:

B	A_B^{-1}	x_2	\vec{P}
x_1	$\frac{1}{3} \ 0$	$-\frac{1}{3}$	$\frac{4}{3}$
leaving $\leftarrow x_6$	$0 \ 1$	2	5

Now we determine the leaving variable. Since in x_2 column above, 2 is the only one that > 0 , so only x_6 can leave.

We have :

B	A_B^{-1}	x_2	\vec{P}
x_1	$\frac{1}{3} \ \frac{1}{6}$	0	$\frac{13}{5}$
x_2	$0 \ \frac{1}{2}$	1	$\frac{5}{2}$

Draft:
 $0 + \frac{1}{2} \times \frac{1}{3} = \frac{1}{6}$
 ~~$\frac{1}{2}$~~
 $\frac{4}{3} + \frac{1}{3} \times \frac{5}{2} = \frac{13}{6}$

Iterate the above process:

$$\vec{r} = \vec{c}_B^T A_B^{-1} A_N - \vec{c}_N^T = (1 \ 1) \begin{pmatrix} \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 4 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{pmatrix} - (1 \ 2 \ 0 \ 0)$$

$$= \left(\frac{1}{3} - \frac{5}{3} \quad \frac{1}{3} \quad \frac{2}{3} \right)$$

Only $-\frac{5}{3} < 0$ and it corresponds to x_4 . So x_4 will be the new entering variable. Compute the column of x_4 : $A_B^{-1} A_4 = \begin{pmatrix} \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} -\frac{1}{6} \\ \frac{1}{2} \end{pmatrix}$

B	A_B^{-1}	\downarrow x_4	\vec{p}
x_1	$\frac{1}{3} \quad \frac{1}{6}$	$-\frac{1}{6}$	$\frac{13}{6}$
leaving $\leftarrow x_2$	$0 \quad \frac{1}{2}$	$\frac{1}{2}$	$\frac{5}{2}$

Since $\frac{1}{2}$ is the only entry in the x_4 column that is > 0 , only x_2 can leave.

The updated mini-tableau:

B	A_B^{-1}	x_4	\vec{p}
x_1	$\frac{1}{3} \quad \frac{1}{3}$	0	3
x_4	0 1	1	5

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Draft:

$$\begin{cases} \frac{1}{6}x_1 + \frac{1}{6}x_4 = 3 \\ \frac{1}{6}x_1 + \frac{1}{6} = \frac{1}{3} \end{cases}$$

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Finally, compute $\vec{r} = \vec{c}_B^T A_B^{-1} A_N - \vec{c}_N^T$ in this step:

$$\begin{aligned}\vec{r} &= \vec{c}_B^T A_B^{-1} A_N - \vec{c}_N^T = (1 \ 2) \begin{pmatrix} \frac{1}{3} & \frac{1}{3} \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 4 & 1 & 0 \\ 2 & 0 & 0 & 1 \end{pmatrix} - (1 \ 1 \ 0 \ 0) \\ &= \left(\frac{10}{3} \quad \frac{1}{3} \quad \frac{1}{3} \quad \frac{7}{3} \right) > 0\end{aligned}$$

Thus we are done since $\vec{r} > 0$.

And when the non-basis variables take zeros, i.e., $x_2 = x_3 = x_5 = x_6 = 0$, we look at \vec{p} in the last mini-tableau. This \vec{p} is exactly equal to $(x_1 \ x_4)^T$. So the optimal solution is given by $\vec{x} = (3 \ 0 \ 0 \ 5 \ 0 \ 0)^T$. \square