

Consider a primal

$$\begin{aligned} \max \quad & \mathbf{c} \cdot \mathbf{x} \\ & A\mathbf{x} \leq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

If we have a dictionary with all the coefficients in the z row are negative (namely $\mathbf{c}_N - \mathbf{c}_B B^{-1} A_N \leq \mathbf{0}^T$) then we can call this *dual feasible* since $\mathbf{c}_B^T B^{-1}$ would be a feasible solution to the dual:

$$\begin{aligned} \min \quad & \mathbf{b} \cdot \mathbf{y} \\ & A^T \mathbf{y} \geq \mathbf{c} \\ & \mathbf{y} \geq \mathbf{0} \end{aligned}$$

If we start with a dictionary (for the primal) that is infeasible (namely $B^{-1}\mathbf{b} \not\geq \mathbf{0}$) which has all the coefficients in the z row being negative then we can proceed with the Dual Simplex algorithm. The following example gives one way that this could happen but you imagine that this could occur in a sensitivity analysis problem using the dual simplex.

$$\begin{aligned} \max \quad & -3x_1 \quad -x_2 \\ & 2x_1 \quad +2x_2 \leq 1 \\ & -2x_1 \quad -x_2 \leq -2 \\ & 4x_1 \quad +3x_2 \leq 1 \end{aligned} \quad x_1, x_2 \geq 0$$

We have our first dictionary

$$\begin{aligned} x_3 &= 1 - 2x_1 - 2x_2 \\ x_4 &= -2 + 2x_1 + x_2 \\ x_5 &= 1 - 4x_1 - 3x_2 \\ z &= -3x_1 - x_2 \end{aligned}$$

Rather than introduce x_0 and use our two phase method, we are able to embark directly on our dual simplex method. We choose x_4 to leave and then (in order to preserve dual feasibility) we choose x_2 as the entering variable. We obtain the following dictionary:

$$\begin{aligned} x_3 &= -3 + 2x_1 - 2x_4 \\ x_2 &= 2 - 2x_1 + x_4 \\ x_5 &= -5 + 2x_1 - 3x_4 \\ z &= -2 - x_1 - x_4 \end{aligned}$$

Note that we have made progress (we have a better dual solution with a smaller objective function value in the dual of -2 rather than 0). We choose x_5 to leave (greedily choosing the 'largest' negative coefficient) and then (in order to preserve dual feasibility) we choose x_1 as the entering variable. We obtain the following dictionary:

$$\begin{aligned} x_3 &= 2 + x_5 + x_4 \\ x_2 &= -3 - x_5 - 2x_4 \\ x_1 &= 5/2 + (1/2)x_5 + (3/2)x_4 \\ z &= -9/2 - (1/2)x_5 - (5/2)x_4 \end{aligned}$$

Again we have made progress finding a dual solution of value -9/2. We would choose x_2 to leave but we are unable to find an entering variable since $(-(1/2) \quad -(5/2)) + \lambda(-1 \quad -2) \leq \mathbf{0}^T$ for all

$\lambda \geq 0$). So we guess that the dual is unbounded but how can we see this? A solution which is somewhat wishful thinking is taking the current dual solution $\mathbf{y} = (0, 5/2, 1/2)$ (obtained as $\mathbf{c}_B^T B^{-1}$ which is readily obtained as the coefficients of the slack variables. Now why not add t times the same coefficients from the row for x_2 , namely $\mathbf{z} = (0, 2, 1)$ to obtain a solution $\mathbf{y} + t\mathbf{z} = (0, 5/2 + 2t, 1/2 + t)$ with objective function value $-9/2 - 3t$ which shows the dual is unbounded. This wishful thinking works and you can verify that I have a parametric set of feasible dual solutions whose objective function, in the dual, goes to $-\infty$. Below I make explicit the reason why this works.

Now we have reached a place where we have a potential leaving variable but no entering variable. Imagine in general that we are doing the dual simplex method and we have x_k leaving. Let $[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]$ denote the $m \times 1$ vector with a 1 in the column corresponding to x_k . Thus

$$[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}\mathbf{b} < 0$$

since the constant entry must be zero in the row corresponding to x_k .

If we are unable to determine an entering variable then that is because

$$[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}A_N \geq \mathbf{0}$$

namely the entries in the row corresponding to x_k must all be negative and the entries in that row are the row of $-B^{-1}A_N$.

Now we do the standard trickery (as done in the proof of Strong Duality). We have

$$[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}B \geq \mathbf{0}^T$$

and so for any variable x_i we have

$$[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}A_i \geq \mathbf{0}.$$

Now regroup the variables by original variables and slack variables and we obtain

$$[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}A \geq \mathbf{0}^T$$

and

$$[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}I \geq \mathbf{0}^T$$

If we set $\mathbf{z}^T = [0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}$ then we discover that $[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}A \geq \mathbf{0}$ implies $\mathbf{z}^T A \geq \mathbf{0}^T$ which is $A^T \mathbf{z} \geq \mathbf{0}$ and $[0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}I \geq \mathbf{0}^T$ yields $\mathbf{z}^T \geq \mathbf{0}^T$ and so $\mathbf{z} \geq \mathbf{0}$.

We also have that the i th entry of $B^{-1}\mathbf{b} = [0 \ 0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0]B^{-1}\mathbf{b} = \mathbf{z}^T \mathbf{b}$. Now the i th entry is less than zero, because that is why we are trying to do a dual simplex pivot. So $\mathbf{z}^T \mathbf{b} = \mathbf{b} \cdot \mathbf{z} < 0$. This is exactly what we need to have the dual be unbounded (towards $-\infty$). Assume \mathbf{y} is a dual solution. The $\mathbf{y} + t\mathbf{z}$ is also a dual feasible solution and, since $\mathbf{b} \cdot \mathbf{z} = \mathbf{z}^T \mathbf{b} < 0$, we have $\mathbf{b} \cdot (\mathbf{y} + t\mathbf{z}) = \mathbf{b} \cdot \mathbf{y} + t\mathbf{b} \cdot \mathbf{z}$ and so $\lim_{t \rightarrow \infty} \mathbf{b} \cdot (\mathbf{y} + t\mathbf{z}) = -\infty$.