

**Polynomial expected behavior of a pivoting algorithm:**

- Class of problems;
- Variant algorithm;
- Probability distribution;
- Analysis.

–The general linear complementarity problem.

–The lexicographic Lemke (LL) algorithm:

This is Lemke's algorithm (complementary pivot algorithm 2) with  $d = \begin{pmatrix} \delta^n \\ \vdots \\ \delta \end{pmatrix}$ ,  $0 < \delta \ll 1$ .

Consider this for two cases:

- $M$  is a  $P$ -matrix; then recall the LCP has a unique complementary solution for all  $q$ . Think of  $z_0 = \delta^{-n+j-0.5}$ ,  $0 \leq j \leq n$ . As  $j$  goes from 0 to  $n$ , we get the complementary solutions to all the LCP subproblems defined by the first  $j$  rows and  $j$  columns of  $M$  (also giving a  $P$ -matrix) and the first  $j$  rows of  $q$ . We are (almost) not perturbing the first  $j$  entries of  $q$  and making the others  $+\infty$ , so  $w_k = +\infty$ ,  $z_k = 0$ , for  $k > j$ . This doesn't work for LP, since the corresponding LCP has a block of zeros in the top left. So it has no complementary solution to such subproblems in general.
- For LP, consider the linear complementarity problem associated with the LP problems:

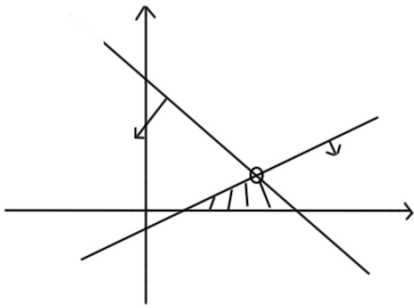
$$(P) \quad \begin{aligned} \min \quad & c^T x \\ & Ax \geq b \\ & x \geq 0, \end{aligned}$$

$$(D) \quad \begin{aligned} \max \quad & b^T y \\ & A^T y \leq c \\ & y \geq 0. \end{aligned}$$

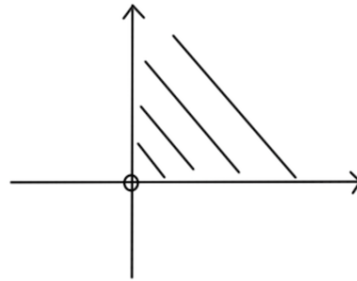
Then, we have  $M = \begin{pmatrix} 0 & -A^T \\ A & 0 \end{pmatrix}$  and  $q = \begin{pmatrix} c \\ -b \end{pmatrix}$ . (We assume  $A$  is  $m \times p$ , with  $n = m + p$ , so the LCP is of dimension  $n$  as usual.) As the algorithm progresses, we get solutions to parametrized LPs with  $b$  replaced by  $b - \begin{pmatrix} \delta^m \\ \vdots \\ \delta \end{pmatrix} z_0$ , and  $c$  replaced by  $c + \begin{pmatrix} \delta^n \\ \vdots \\ \delta^{m+1} \end{pmatrix} z_0$ . So we can view the algorithm (assuming  $z_0$  decreases steadily) as working on  $(D)$ , imposing the constraints one by one (with the perturbed  $b_i \ll -1$  for all  $i$ ), and then restoring the true values of the  $b_i$ 's one by one.

Example:

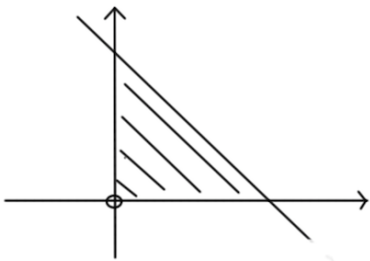
$$(D) \quad \begin{aligned} \max \quad & y_1 + 3y_2 \\ & y_1 + y_2 \leq 4 \\ & -y_1 + 2y_2 \leq -1 \\ & y \geq 0. \end{aligned}$$



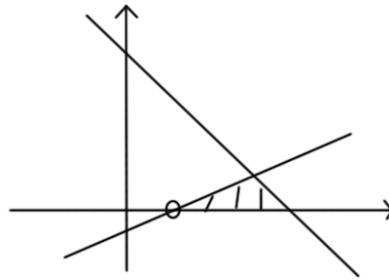
Optimal solution (3;1) for original LP



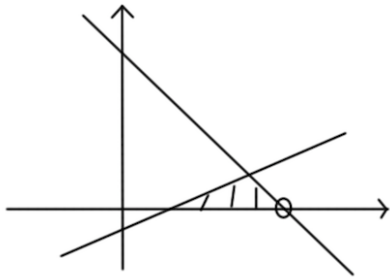
$z_0 = \delta^{-4.5}$ , no constraints imposed



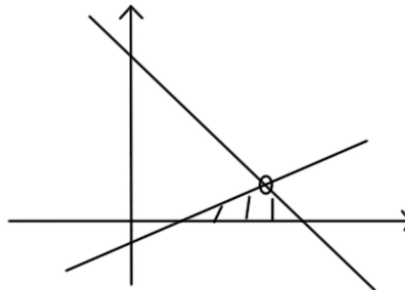
$z_0 = \delta^{-3.5}$ , first constraint imposed



$z_0 = \delta^{-2.5}$ , both constraints imposed



$z_0 = \delta^{-1.5}$ , true  $b_1$  restored



$z_0 = \delta^{-0.5}$ , true  $b_2$  restored

This describes the algorithm, but not in a useful way. The column  $-d$  associated with  $z_0$  will be basic until the algorithm stops, and having  $\begin{pmatrix} -\delta^n \\ \vdots \\ -\delta \end{pmatrix}$ ,  $0 < \delta \ll 1$ , where  $\delta$  is a parameter sufficient close to 0, in the basis matrix is horrible.

Old view:

$$Iw + (-d)z_0 + (-M)z = q. \quad (1)$$

New view: If we have a solution with  $z_0 > 0$ , we get

$$I\left(\frac{w}{z_0}\right) + (-q)\left(\frac{1}{z_0}\right) + (-M)\left(\frac{z}{z_0}\right) = d \quad (2)$$

and from now on, we let  $v = \frac{w}{z_0}$ ,  $y_0 = \frac{1}{z_0}$  and  $y = \frac{z}{z_0}$ . So now the basis consists of some columns of  $[I, -q, -M]$  and  $d$  is the RHS. Suppose we have a non-degenerate a.c. but not complementary basic solution to (1). The basic variable will be  $z_0$ ,  $w_J$  for some  $J \subseteq N := \{1, 2, \dots, n\}$ ,  $z_K$ ,  $K \subseteq N$  with both  $w_l$  and  $z_l$  nonbasic, and  $J \cup K \cup \{l\}$  a disjoint partition of  $N$ . Then  $v_J$ ,  $y_0$  and  $y_K$  will be the basic variables in an a.c. basic solution to (2). Recall  $d = \begin{pmatrix} \delta^n \\ \vdots \\ \delta \end{pmatrix}$ , for  $0 < \delta \ll 1$  and this does not lie in any hyperplane formed by  $n - 1$  columns of  $[I, -q, -M]$ , so we get a non-degenerate basic solution. In terms of  $v$ ,  $y$  and (2), the LL algorithm operates as follows:

**Algorithm:**

- Step 0: If  $q \geq 0$ , we have a complementary solution: stop. Otherwise, let  $B = I$  be the initial basis in (2). Introduce  $y_0$  with column  $-q$  into the basis: since  $q$  isn't greater than or equal to 0, its increase will be blocked, so some  $v_l$  goes to 0. Set  $J = N \setminus \{l\}$ ,  $K = \emptyset$ .
- Step 1: We have a feasible basis  $B$ ,  $J$ ,  $K$  and  $l$ , giving an a.c. but not complementary solution to (2). Just one of  $v_l$  and  $y_l$  has just hit zero. Increase its complement. If its increase is blocked by  $y_0$  hitting 0, then by rescaling we have a secondary ray of (1); go to step 2. If its increase is blocked by  $v_j$  or  $y_k$ , make the pivot, update  $B$ ,  $J$ ,  $K$ , and  $l$  and go to step 1. If its increase is unblocked, we have a ray of solutions:

$$(v_J; y_0; y_K; 0) + \lambda(\bar{v}_J; \bar{y}_0; \bar{y}_K; 1),$$

where the last component corresponds to the variable we are increasing,  $v_l$  or  $y_l$ . If  $\bar{y}_0=0$ , then  $y_0$  (and hence  $z_0$ ) stays the same, and by scaling, we find a secondary ray of (1); go to step 2. If  $\bar{y}_0 > 0$ , then by rescaling, we get a complementary solution; go to step 3.

- Step 2(Failure): We have (by scaling) a secondary ray of (1). Stop!
- Step 3(Success): Suppose we were increasing  $y_l$ , then ray termination implies

$$I\bar{v} + (-q)\bar{y}_0 - M\bar{y} - m_l \cdot 1 = 0,$$

where  $\bar{v}_K = 0$ ,  $\bar{y}_J = 0$ . Then  $I(\frac{\bar{v}}{\bar{y}_0}) - M(\frac{\bar{y}}{\bar{y}_0} + \frac{1}{\bar{y}_0}e_l) = q$ , which gives a complementary solution to (1). Similarly if we were increasing  $v_l$ .

**Question:** When is a particular partition  $J \cup K \cup l = N$  feasible?

The basis matrix  $B$  has the  $J$  columns of  $I$ , the column  $-q$  and the  $K$  columns of  $-M$ :

$$B = \begin{bmatrix} I_{JJ} & -q_J & -M_{JK} \\ 0 & -q_L & -M_{LK} \end{bmatrix}.$$

with  $L := K \cup \{l\}$ . For simplicity, let  $F := [q, M]$ , with columns indexed 0 through  $n$ , and let  $H := \{0\} \cup K$ : then

$$B = \begin{bmatrix} I_{JJ} & -F_{JH} \\ 0 & -F_{LH} \end{bmatrix},$$

with inverse:

$$C := B^{-1} = \begin{bmatrix} I_{JJ} & -F_{JH}F_{LH}^{-1} \\ 0 & -F_{LH}^{-1} \end{bmatrix}.$$

For feasibility, we want  $B^{-1}d = C \begin{pmatrix} \delta^n \\ \delta \end{pmatrix} \geq 0$ . So feasibility “depends on” the signs of the column of  $\begin{pmatrix} -F_{JH}F_{LH}^{-1} \\ -F_{LH}^{-1} \end{pmatrix}$  corresponding to the last index in  $L$ .