

**On the Existence of
Short Admissible Pivot Sequences
for Feasibility Problems and
Linear Optimization (LO)**

Tamás Terlaky

K. Fukuda

Dept. Industrial and Systems Engineering
Lehigh University, Bethlehem, PA

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Consider

The Primal Feasibility Problem:

$$Ax = b, \quad x \geq 0$$

$A : m \times n$ matrix, $\text{rank}(A) = m$, $b \in \mathbb{R}^m$, $x \in \mathbb{R}^n$.

Problem: An arbitrary basis B and a feasible solution x^* is given.

Find a short admissible pivot sequence from B to a feasible basis.

- **History:**
 - Hirsch conjecture
 - Admissible pivot methods,
 - Criss-cross methods,
 - Lüthi, Fukuda, Namiki
existence under nondegeneracy
- **Admissible pivots**
- **Short admissible pivot sequence exists!**
- **Complexity: at most $n - m$ pivots**

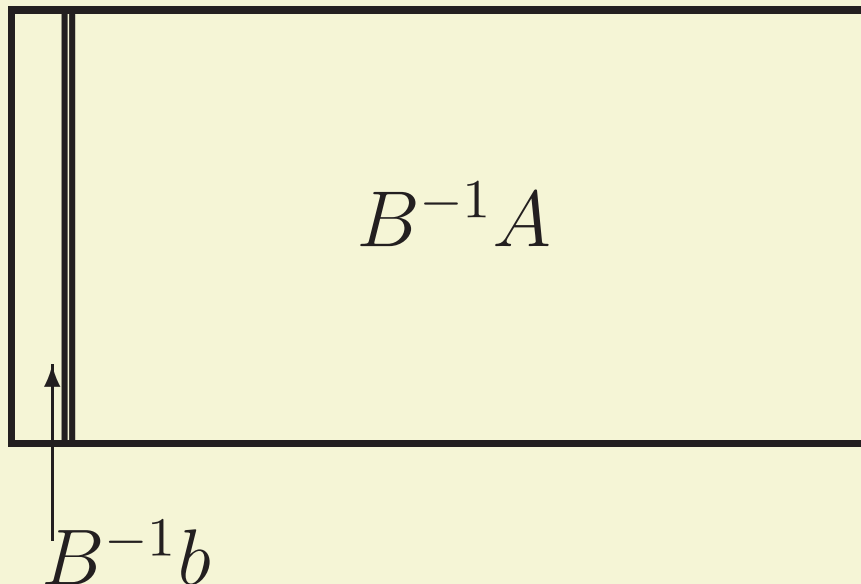
Dual Feasibility Problem: at most m pivots

Extension to LO: at most n pivots

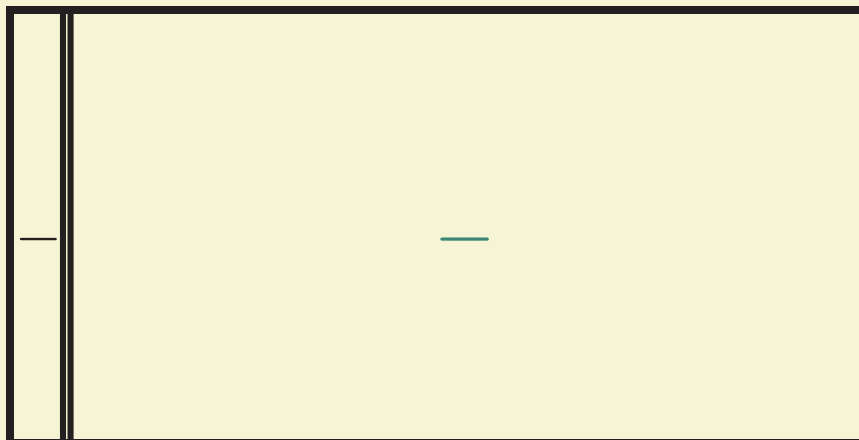
Basis B is given: $B^{-1}b = B^{-1}Ax$.

Basis solution: $x_B = B^{-1}b, x_N = 0$.

Pivot tableau



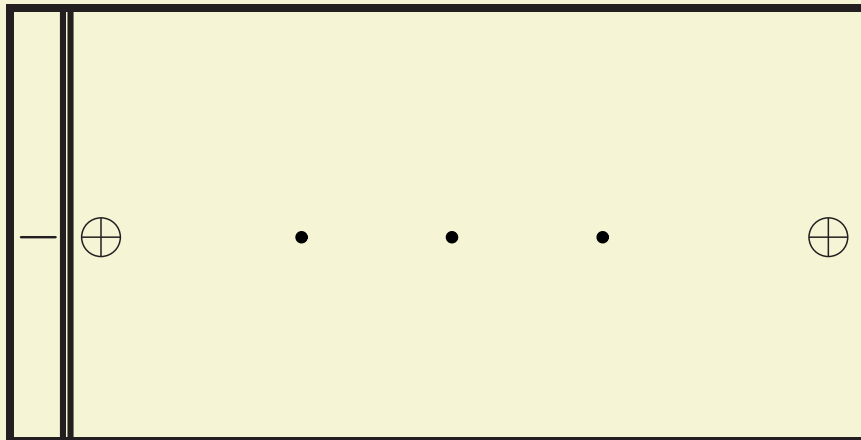
Admissible pivot



Feasible Tableau



Infeasible Tableau



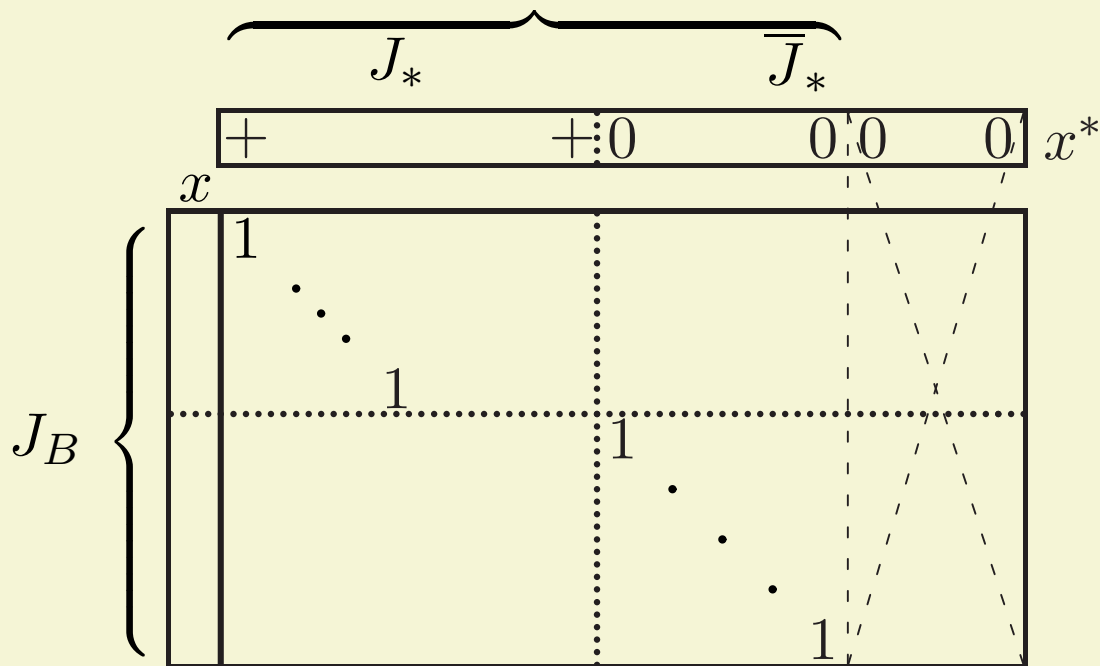
If we assume that a feasible solution x^* is given, then the latter tableau cannot occur.

Let $J = \{1, \dots, n\}$, $J_* = \{i \mid x_i^* > 0\}$,
 $J_B = \{i \mid i \in \text{basis}\}$, and $I = J_* \cup J_B$.

Remove the variables not in I . Then

$$A_I x_I = b, \quad x_I \geq 0 \quad \text{is feasible.}$$

$$I := J_* \cup J_B$$



We have the following cases:

If $x_B \geq 0$, then a
 feasible basis solution is found.

If $x_i \geq 0$, for all $i \in J_B \setminus J_*$,
 then reduce J_* ;
 else reduce I .

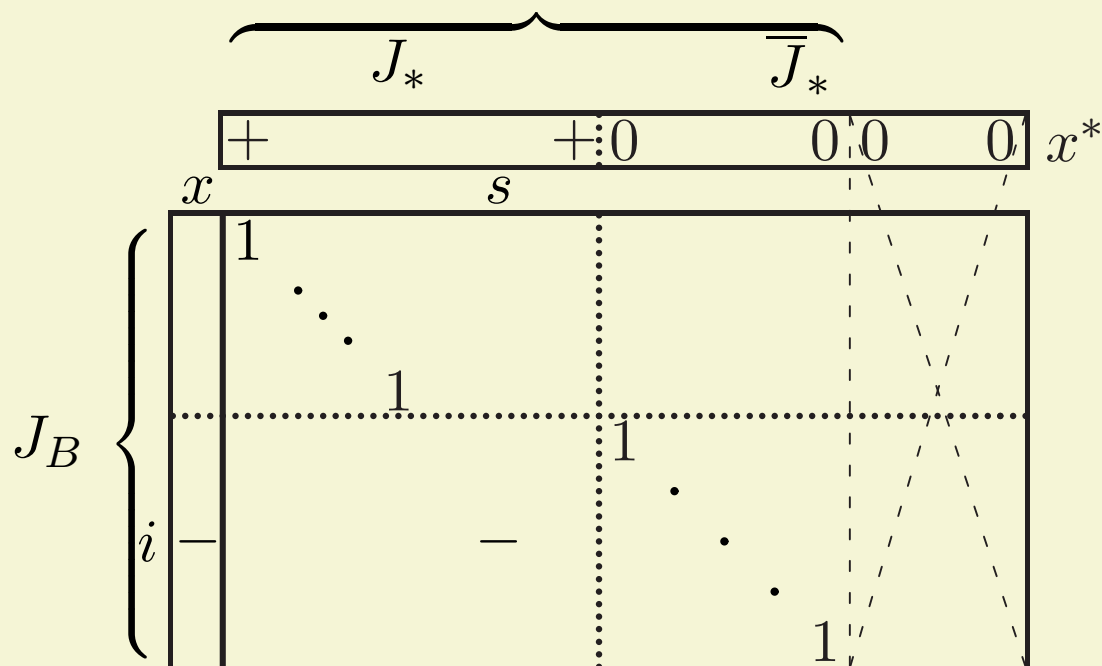
There is an $i \in J_B \setminus J_*$ such that $x_i < 0$.

Then there is an $s \in J_* \setminus J_B$ such that

(i, s) is admissible.

(True, because $A_I x_I = b$, $x_I \geq 0$ is feasible.)

$$I := J_* \cup J_B$$



Do pivot at (i, s) ;

let $J_B := J_B \cup \{s\} \setminus \{i\}$; $I := I \setminus \{i\}$.

Restart the algorithm.

Note: By one pivot $|I|$ decreases by one.

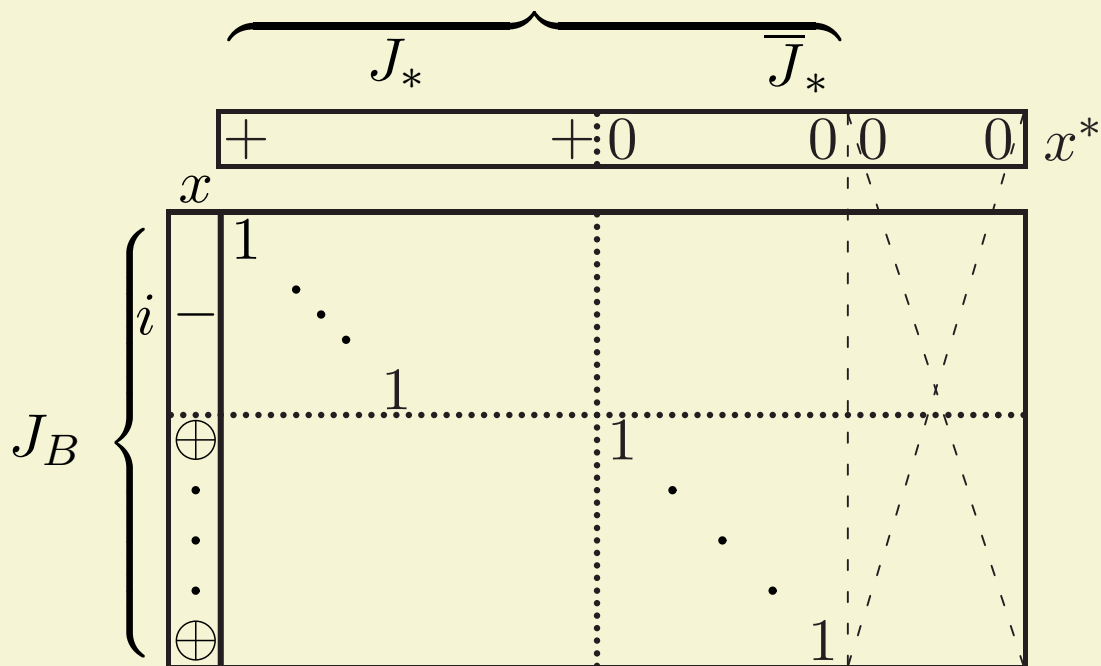
Here x^* does not change, only the basis.

Here $x_i \geq 0$ for all $i \in J_B \setminus J_*$,

but there is an $i \in J_B \setminus J_*$ such that $x_i < 0$.

We eliminate a coordinate of x^* .

$$I := J_* \cup J_B$$



$$\text{Let } \lambda := \min \left\{ \frac{x_i^*}{x_i^* - x_i} \mid x_i < 0 \right\} = \frac{x_r^*}{x_r^* - x_r}.$$

Then we have

$$x^* := \lambda x + (1 - \lambda)x^* \geq 0 \text{ and } A_I x_I^* = b.$$

Restart the algorithm

with the new x^* , $I \subseteq I \setminus \{r\}$.

Note: This way $|I|$ decreases by one.

We can always reduce I by using $x_i < 0$.

Theorem 1 *Let us consider the problem*

$$Ax = b, \quad x \geq 0.$$

Let a feasible solution x^ and an arbitrary basis B be given. Then there is*

- *a path of basis solutions initiated at B ,*
- *leading to a feasible basis and*
- *connected by admissible pivots.*

The length of path is at most $n - m$.

Proof: The above presented algorithm produces such an admissible pivot path.

At the step **Reduce J_*** no pivots, just elimination is made.

At the step **Reduce I** the cardinality of I is reduced by one.

At initialization the cardinality of I is at most n , at termination I contains at least m elements, thus we need at most $n - m$ pivots to find a feasible basis. ●

Note: The information that a feasible solution is known was heavily used. **Basis identification in IPMs.**

The algorithm does not give a strongly polynomial algorithm to solve the feasibility problem.

$$A^T y + s = c, \quad s \geq 0$$

Basis (B) solution: $y = B^{-T} c_B, \quad s = c - A^T y.$

Pivot tableau

$-s = c_B^T B^{-1} A - c$	1
$B^{-1} A$	

Dual admissible pivot

+	1
+	

Dual feasible tableau

\ominus		\cdot	\cdot	\cdot	\ominus	1

Dual infeasible tableau

		$+$				1
		\ominus				
		\cdot				
		\cdot				
		\cdot				
		\ominus				

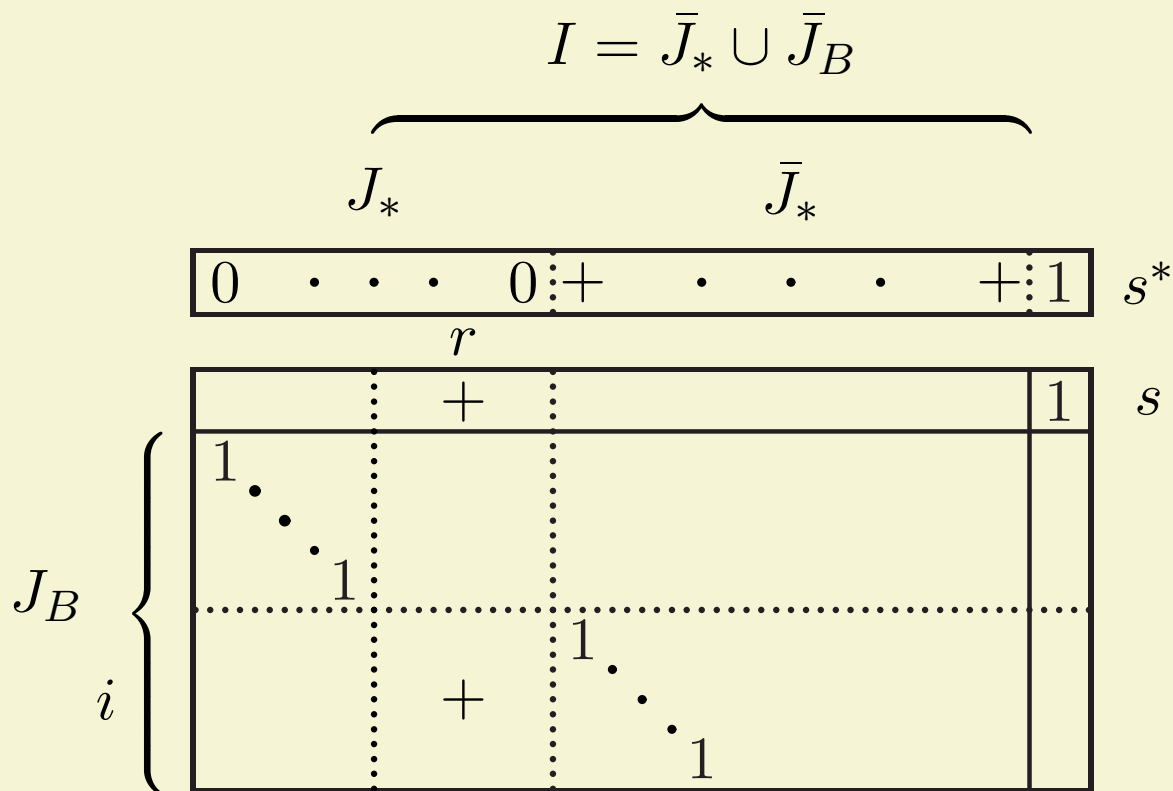
If we assume that a feasible solution (y^*, s^*) is given, then the latter tableau cannot occur.

There is an $r \in \bar{J}_B \cap J_*$ such that $s_r < 0$.

Then there is an $i \in J_B \setminus J_*$ such that

(i, r) is admissible.

(True because $A^T y + s = c, s \geq 0$ is feasible.)



Do pivot at (i, r) ;

let $J_B := J_B \cup \{r\} \setminus \{i\}; I := I \setminus \{r\}$.

Restart the algorithm.

Note: By one pivot $|I|$ decreases, $|J_B \cap J_*|$ increases by one. Here s^* does not change.

Theorem 2 *Let us consider the problem*

$$A^T y + s = c, \quad s \geq 0.$$

Let a feasible solution (y^, s^*) and an arbitrary basis B be given. Then there is*

- *a path of basis solutions initiated at B ,*
- *leading to a (dual) feasible basis and*
- *connected by admissible pivots.*

The length of path is at most m .

Proof: The algorithm presented above produces such an admissible pivot path.

At the step **Reduce \bar{J}_*** no pivots, just elimination is made.

At the step **Reduce I** the $|J_B \cap J_*|$ is increased by one.

At initialization $|J_B \cap J_*|$ is at least 0,

at termination $|J_B \cap J_*|$ is at most m , thus

we need at most m pivots to find a feasible basis. ●

Note: The information that a feasible solution is known was heavily used. **Basis identification in IPMs.**

The algorithm does not give a strongly polynomial algorithm to solve the dual feasibility problem.

$$\begin{aligned} \min \{ c^T x \mid Ax = b, \quad x \geq 0 \\ \max \{ b^T y \mid A^T y + s = c, \quad s \geq 0 \end{aligned}$$

Basis (B) solution:

$$x = B^{-1}b, \quad Ax_B = b; \quad y = B^{-T}c_B, \quad s = c - A^T y.$$

Pivot tableau

	$B^{-1}b$	$-s = c_B^T B^{-1}A - c$	1
↓		$B^{-1}A$	0
			•
			•
			•
			0

Admissible Pivots

		+	1
			0
-		-	•
		+	•
			•
			0

Optimal tableau

	\ominus	.	.	.	\ominus	1
\oplus						0
						.
						.
						.
\oplus						0

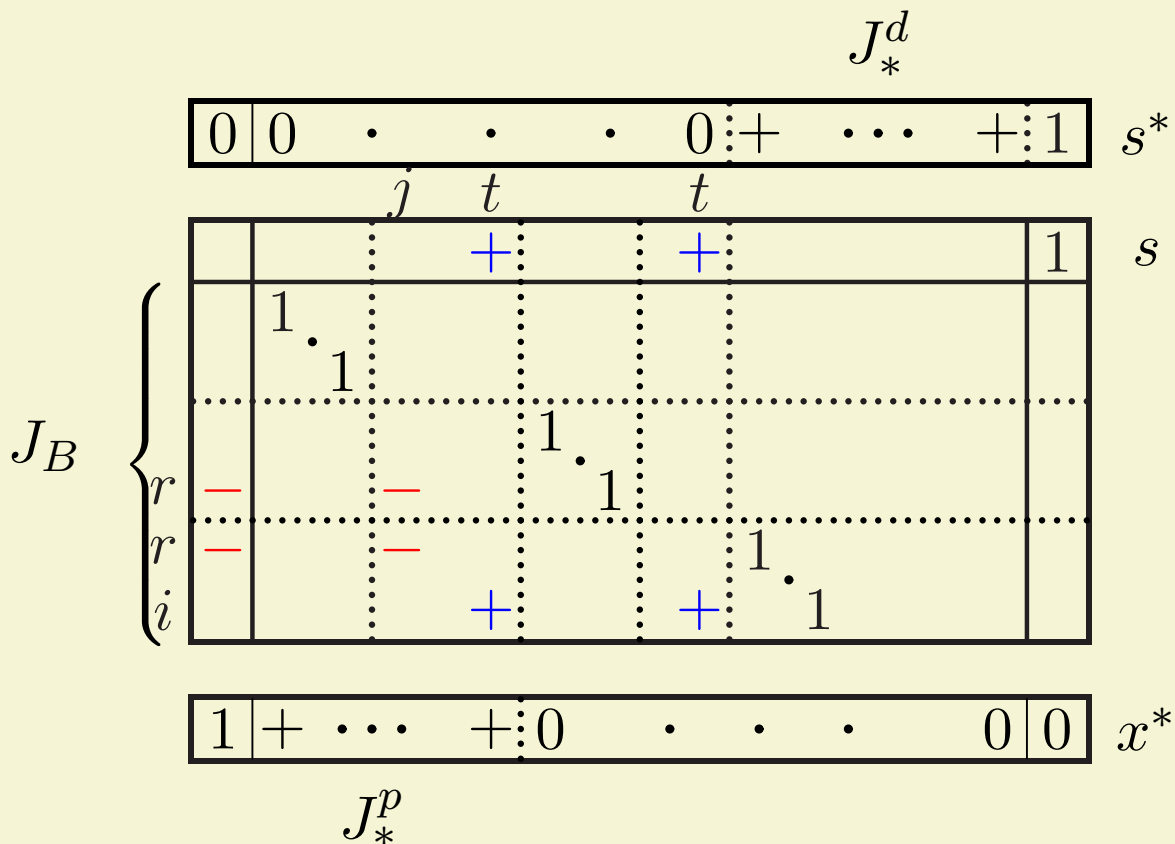
Primal infeasible tableau

						1
						0
						.
						.
\ominus	\oplus	.	.	.	\oplus	.
						.
						0

Dual infeasible tableau

						1
						0
						.
						.
						.
						0

Either there is an $r \in J_B \cap \bar{J}_*^p$ with $x_r < 0$ and an $j \in J_*^p$ such that pivot (r, j) is admissible; or there is a $t \in \bar{J}_*^d$ with $s_t < 0$ and an $i \in J_B \cap \bar{J}_*^d$ such that pivot (i, t) is admissible.



Do pivot at (r, j) or at (i, t) ;

Update the data structure.

Restart the algorithm. Here neither x^* nor (y^*, s^*) change, only the basis.

Here $x_i \geq 0$ for all $i \in \bar{J}_B^p$, and

$s_j \geq 0$ for all $j \in \bar{J}_*^d$

but there is either an $i \in J_*^p$ such that $x_i < 0$,
or there is a $j \in J_*^d$ such that $s_j < 0$.

We eliminate either a coordinate of x^* or s^* .

		J_*^d											
		0	0	·	·	·	0	+	·	·	+	1	s^*
				⊖	·	⊖		⊖	⊖		+	1	s
{	J_B	-	1	·	1								
	⊕					1	·						
	·						1						
	·								1	·			
	⊕									1	·	1	
		1	+	·	·	+	0	·	·	·	0	0	x^*
		J_*^p											

(This structure is justified on the next sheet.)

Keeping optimality, we eliminate a coordinate of x^* or s^* .

1. Because no desired admissible pivot exists, so we have
 - $-s_i \leq 0$ for $i \in \bar{J}_B \cap J_*^p$;
 - $x_i \geq 0$ for $i \in J_B \cap J_*^d$.
2. Although the current basis might be both primal and dual infeasible, **the current objective value equals to the optimal one.**
Proof: Else, by taking convex combinations either better primal or better dual solution could be obtained.
3. Now **it follows that $x_i = 0$ for $i \in J_B \cap J_*^d$;** else, by taking convex combinations of the current and the optimal x would give an optimal x not complementaire with the optimal s^* .
4. Similarly, **it follows that $s_i = 0$ for $i \in \bar{J}_B \cap J_*^p$;** else, by taking convex combinations of the current and the optimal s would give an optimal s not complementaire with the optimal x^* .

In the case $s_j < 0$:

$$\text{Let } \lambda := \min \left\{ \frac{s_i^*}{s_i^* - s_i} \mid s_i < 0 \right\} = \frac{s_t^*}{s_t^* - s_t}.$$

Then we have $A^T y^* + s^* = c$ and $s^* \geq 0$ with $(y^*, s^*) := \lambda(y, s) + (1 - \lambda)(y^*, s^*)$ with $s_t^* = 0$.

Restart with the new updated data.

J_*^d has decreased!

One can operate analogously if an $x_i < 0$.

In the case $x_i < 0$:

$$\text{Let } \lambda := \min \left\{ \frac{x_i^*}{x_i^* - x_i} \mid x_i < 0 \right\} = \frac{x_r^*}{x_r^* - x_r}.$$

Then we have $Ax^* = b$ and $x^* \geq 0$ with $x^* := \lambda x + (1 - \lambda)x^*$ where $x_r^* = 0$.

Restart with the new updated data.

J_*^p has decreased!

Theorem 3 *Let us consider the LO problem*

$$\min \{ c^T x \mid Ax = b, \quad x \geq 0 \},$$

and its dual. Let a pair of optimal solutions x^ ; (y^*, s^*) and any basis B be given. Then there is*

- *a path of basis solutions initiated at B ,*
- *leading to an optimal basis and*
- *connected by admissible pivots.*

The length of the path is at most n .

Proof: The algorithm presented above produces such an admissible pivot path.

At the step **Reduce the basis:** either $|J_B \cap \bar{J}_*^d|$ reduces by one, what might happen at, most $|J_B \cap J_*^d|$ times;

or $|J_B \cap J_*^p|$ increases by one, what might happen at most $|J_*^p|$ times.

Because $|J_B \cap J_*^d| + |J_*^p| \leq |J_*^d| + |J_*^p| \leq n$ we need at most n pivots to find an optimal basis. ●

Note: The information that an optimal solution is known was heavily used. **Basis identification in IPMs.**

The algorithm does not give a strongly polynomial algorithm to solve the LO problem.