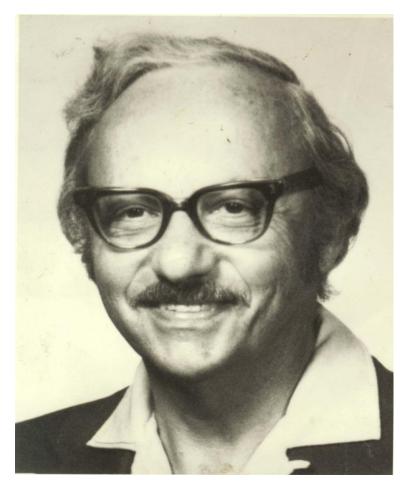
MDI210 : Linear Programming

Robert M. Gower





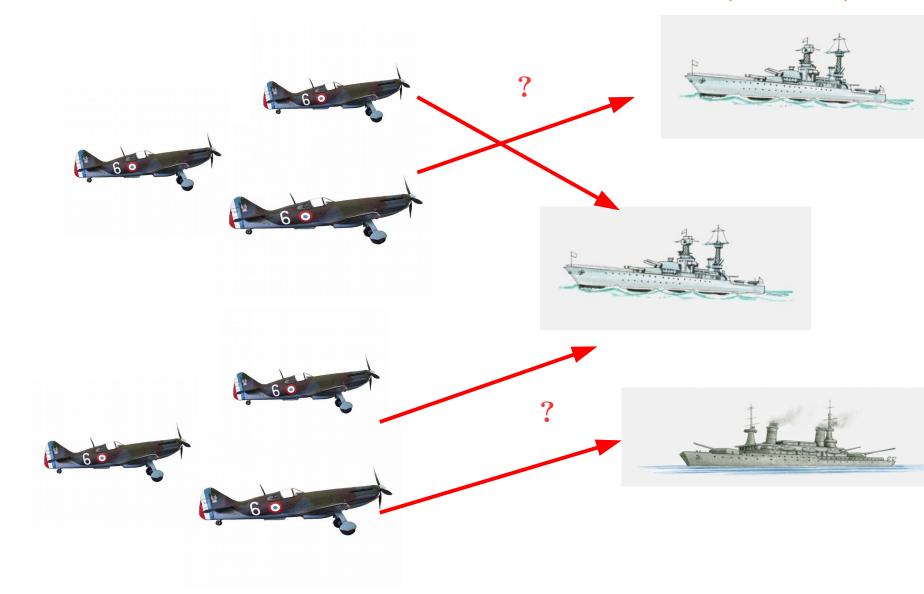
Linear Programming History (1939)



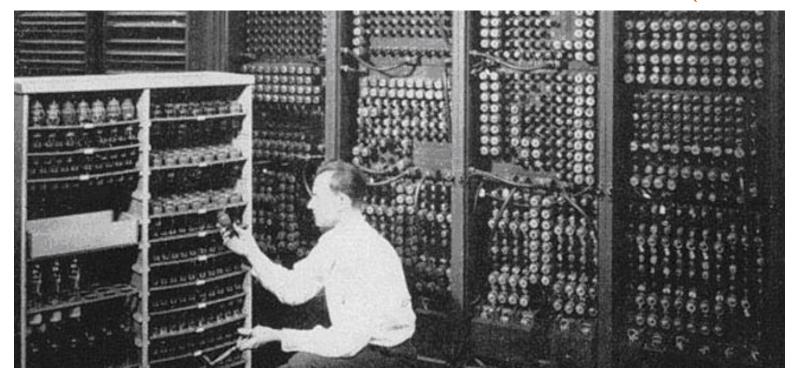


- 1947: George Dantzig, advising U.S. Air Force, invents Simplex.
- Assignment 70 people to 70 jobs (more possibilities than particles).

Linear Programming History (1941)



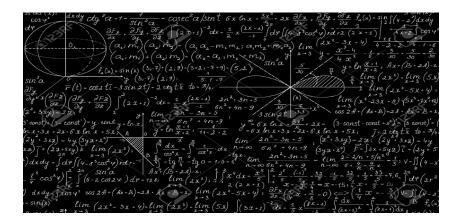
Army Builds Killing Machine (1949)



1949 SCOOP: Scientific Computation Of Optimal Programs

Mathematical Programming: Math used to figured out Flight and logistic programs/schedules

Dantzig the Urban Legend



Dantzig, George B. "On the Non-Existence of Tests of 'Student's' Hypothesis Having Power Functions Independent of Sigma." Annals of Mathematical Statistics. No. 11; 1940 (pp. 186-192).

Dantzig, George B. and Abraham Wald. "On the Fundamental Lemma of Neyman and Pearson." Annals of Mathematical Statistics. No. 22; 1951 (pp. 87-93).

Optimization and Numerical Analysis: Linear Programming

Robert Gower



September 19, 2019

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The Problem: Linear Programming

$$\max_{x} z \stackrel{\text{def}}{=} c^{ op} x$$

subject to $Ax \leq b,$
 $x \geq 0,$

where $c, x \in \mathbb{R}^n$, $A \in \mathbb{R}^{m \times n}$, and $b \in \mathbb{R}^m$. Equivalently

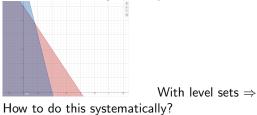
$$\max_{x} z \stackrel{\text{def}}{=} \sum_{j=1}^{n} c_{j} x_{j}$$

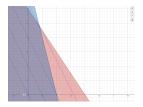
subject to $\sum_{j=1}^{n} a_{ij} x_{j} \le b_{i}$, for $i = 1, \dots, m$.
 $x \ge 0$.

The problem

$$\begin{array}{ll} \max & 4x_1 + 2x_2 \\ & 3x_1 + 2x_2 & \leq 600 \\ & 4x_1 + 1x_2 & \leq 400 \\ & x_1 \geq 0, x_2 \geq 0. \end{array}$$

We can solve this graphically:





The problem

$$\begin{array}{ll} \max & 4x_1 + 2x_2 \\ & 3x_1 + 2x_2 & \leq 600 \\ & 4x_1 + 1x_2 & \leq 400 \\ & x_1 \geq 0, x_2 \geq 0. \end{array}$$

Can be transformed into

where x_3 and x_4 are *slack variables*. This is known as the the *dictionary* format and is often written as:

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The dictionary format

<i>x</i> 3	=	600	_	$3x_1$	_	$2x_{2}$
<i>x</i> ₄	=	400	_	$4x_1$	—	<i>x</i> ₂
Ζ	=			$4x_1$	+	$2x_2$

admits obvious solution

$$(x_1^*, x_2^*, x_3^*, x_4^*) = (0, 0, 600, 400).$$

The objective z will improve if $x_1 > 0$. Increasing x_1 as much as possible

$$\begin{array}{rrrr} x_3 \geq 0 & \Rightarrow & 600-3x_1 \geq 0 & \Rightarrow & x_1 \leq 200, \\ x_4 \geq 0 & \Rightarrow & 400-4x_1 \geq 0 & \Rightarrow & x_1 \leq 100. \end{array}$$

Thus $x_1 \leq 100$ to guarantee $x_4 \geq 0$. This means x_4 will leave the basis and x_1 will enter the basis. Using row operations $z \leftarrow z + r_2$ and $r_1 \leftarrow r_1 - \frac{3}{4}r_2$ to isolate x_1 on row₂.

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From

Now we are at the vertex $(x_1^*, x_2^*) = (100, 0)$. Next we see that increasing x_2 increases the objective value but

$$\begin{array}{rrrr} x_3 \geq 0 & \Rightarrow & 300 - \frac{5}{4}x_2 \geq 0 & \Rightarrow & 240 \geq x_2, \\ x_1 \geq 0 & \Rightarrow & 100 - \frac{x_2}{4} \geq 0 & \Rightarrow & 400 \geq x_2. \end{array}$$

Increase x_2 upto 240 while respecting the positivity constraints of x_3 . Thus x_3 will *leave* the basis and x_2 will *enter* the basis. Performing a row elimination again via $z \leftarrow z + \frac{4}{5}r_1$ and $r_2 \leftarrow r_2 - \frac{1}{5}r_1$, we have that

Now $(x_1^*, x_2^*) = (40, 240)$. Increasing x_4 or x_3 will decrease z. THE END

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Theorem (Fundamental Theorem of Linear Programming)

Let
$$P = \{x \mid Ax = b, x \ge 0\}$$
 then either

- $P = \{ \emptyset \}$
- *P* ≠ {∅} and there exists a vertex *v* of *P* such that *v* ∈ arg min_{x∈P} c[⊤]x
- There exists x, d ∈ ℝⁿ such that x + td ∈ P for all t ≥ 0 and lim_{t→∞} c^T(x + td) = ∞.



Problem Notation

We will now formalize the definitions we introduced in the examples.

- There are n variables and m constraints
- The linear objective function $z = \sum_{j=1}^{n} c_j x_j$
- The m inequality constraints in standard form

$$\sum_{j=1}^n a_{ij} x_j \le b_i, \text{ for } i \in \{1, \dots, m\}.$$

- ▶ The *n* positivity constraints $x_j \ge 0$, for $j \in \{1, ..., n\}$.
- x_i^* denotes the value of *i*th variable.
- We call (x₁^{*},...,x_n^{*}) ∈ ℝⁿ a feasible solution if it satisfies the inequality and positivity constraints.

Dictionary Notation

- ▶ The slack variables $(x_{n+1}, ..., x_{n+m}) \in \mathbb{R}^m$ (variables d'écart)
- ► The initial dictionary

$$x_{n+1} = b_1 - \sum_{j=1}^n a_{1j} x_j$$

$$\vdots$$

$$x_{n+i} = b_i - \sum_{j=1}^n a_{ij} x_j$$

$$\vdots$$

$$x_{n+m} = b_m - \sum_{j=1}^n a_{mj} x_j$$

$$z = \sum_{j=1}^n c_j x_j$$

- Valid dictionary if m of the variables (x₁,..., x_{n+m}) can be expressed as function of the remaining n variables.
- The *m* variables on the left-hand side are the basic variable (variable de base). The *n* variables on the right-hand side are the non-basic (variable hors-base).

Dictionary Notation

After row elimination operations we have a new basis.

- ▶ Basic variable set $I \subset \{1, ..., n + m\}$ and non-basic set $J = \{1, ..., n + m\} \setminus I$ with |I| = m and |J| = n
- Current objective value $z^* = \sum_{j=1}^n c_j x_j^*$.

For each basis set I there is a corresponding dictionary

$$\begin{array}{rcl} x_i &=& b_i' - \sum_{j \in J} a_{ij}' x_j, \ \text{for } i \in I \\ z &=& z^* + \sum_{j \in J} c_j' x_j, \end{array}$$

where $a'_{ij}, b'_i, z^* \in \mathbb{R}$ are coefficients resulting from the row operations. For this to a feasible dictionary we require that $b'_i \geq 0$.

▶ A basic solution: $x_i^* = b_i'$ for $i \in I$ and $x_j^* = 0$ for $j \in J$.

If j₀ ∈ J with c'_{j0} > 0 then increasing x_{j0} will improve the objective since

$$z = z^* + \sum_{j \in J} c'_j x_j$$

If j₀ ∈ J with c'_{j0} > 0 then increasing x_{j0} will improve the objective since

$$z=z^*+\sum_{j\in J}c'_jx_j.$$

• How much can we increase x_{i_0} ? Until there is a $x_i = 0$ since

$$x_i^* = b_i' - a_{ij_0}' x_{j_0}^* \ge 0$$

If j₀ ∈ J with c'_{j0} > 0 then increasing x_{j0} will improve the objective since

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• How much can we increase x_{i_0} ? Until there is a $x_i = 0$ since

$$x_i^* = b_i' - a_{ij_0}' x_{j_0}^* \ge 0 \quad \Rightarrow \quad a_{ij_0}' x_{j_0}^* \le b_i', \quad \forall i \in I.$$

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▶ If $a'_{ij_0} \leq 0$, then increasing $x^*_{j_0}$ will increase x^*_i

If j₀ ∈ J with c'_{j0} > 0 then increasing x_{j0} will improve the objective since

$$z=z^*+\sum_{j\in J}c'_jx_j.$$

• How much can we increase x_{j_0} ? Until there is a $x_i = 0$ since

$$x_i^* = b_i' - a_{ij_0}' x_{j_0}^* \ge 0 \quad \Rightarrow \quad a_{ij_0}' x_{j_0}^* \le b_i', \quad \forall i \in I.$$

▶ If $a'_{ij_0} \leq 0$, then increasing $x^*_{j_0}$ will increase x^*_i ▶ If $a'_{ij} > 0$, then $x^*_{j_0} \leq b'_i / a'_{ij_0}$

If j₀ ∈ J with c'_{j0} > 0 then increasing x_{j0} will improve the objective since

$$z = z^* + \sum_{j \in J} c'_j x_j.$$

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$$x_i^* = b_i' - a_{ij_0}' x_{j_0}^* \ge 0 \quad \Rightarrow \quad a_{ij_0}' x_{j_0}^* \le b_i', \quad \forall i \in I.$$

ln this case, which $x_i^* = 0$ (which *i* leaves the basis?)

A Step of the Simplex Method Input: $I = \{n + 1, \dots, n + m\}, J = \{1, \dots, n\}, a'_{ii} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

Input: $I = \{n + 1, ..., n + m\}, J = \{1, ..., n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

if $c'_i \leq 0$ for all $i \in J$ then STOP; # Optimal point found.

Input: $I = \{n + 1, \dots, n + m\}, J = \{1, \dots, n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

if $c'_i \leq 0$ for all $i \in J$ then

STOP; # Optimal point found.

Choose a variable j_0 to enter the basis from the set $j_0 \in \{j \in J : c'_j > 0\}$. if $a'_{ij_0} \leq 0$ for all $i \in J$ then

STOP; # The problem is unbounded.

Input: $I = \{n + 1, \dots, n + m\}, J = \{1, \dots, n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

if $c'_i \leq 0$ for all $i \in J$ then STOP; # Optimal point found. Choose a variable j_0 to enter the basis from the set $j_0 \in \{j \in J : c'_j > 0\}$. if $a'_{ij_0} \leq 0$ for all $i \in J$ then STOP; # The problem is unbounded.

Choose a variable i_0 to **leave the basis** from the set $i_0 \in \arg\min_{i \in I, a'_{iin} > 0} \left\{ \frac{b'_i}{a'_{iin}} \right\}$.

Input: $I = \{n + 1, ..., n + m\}, J = \{1, ..., n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

 $\begin{array}{ll} \text{if } c_i' \leq 0 \text{ for all } i \in J \text{ then} \\ \textbf{STOP}; & \# \text{ Optimal point found.} \\ \text{Choose a variable } j_0 \text{ to enter the basis from the set } j_0 \in \{j \in J : c_j' > 0\}. \\ \text{if } a_{ij_0}' \leq 0 \text{ for all } i \in J \text{ then} \\ \textbf{STOP}; & \# \text{ The problem is unbounded.} \\ \text{Choose a variable } i_0 \text{ to leave the basis from the set } i_0 \in \arg\min_{i \in I, a_{ij_0}' > 0} \left\{ \frac{b_i'}{a_{ij_0}'} \right\}. \\ I \leftarrow (I \setminus \{i_0\}) \quad \text{and} \quad J \leftarrow J \cup \{i_0\} \quad \triangleright \text{ Move } i_0 \text{ from basic to non-basic for } i \in I \text{ do} \\ a_{i:}' \leftarrow a_{i:}' - \frac{a_{ij_0}'}{a_{i_0j_0}'} a_{i_0}': \quad \triangleright \text{ Row elimination on pivot } (i_0, j_0). \end{array}$

 $a'_{i:} \leftarrow a'_{i:} - rac{a'_{ij_0}}{a'_{i_0j_0}}a'_{i_0:}$ $a'_{i_0:} \leftarrow rac{1}{a'_{i_0j_0}}a'_{i_0:}$ and $a'_{i_0j_0} \leftarrow rac{1}{a'_{i_0j_0}}$

Input: $I = \{n + 1, ..., n + m\}, J = \{1, ..., n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

if $c'_i \leq 0$ for all $i \in J$ then STOP; # Optimal point found. Choose a variable j_0 to enter the basis from the set $j_0 \in \{j \in J : c'_j > 0\}$. if $a'_{ij_0} \leq 0$ for all $i \in J$ then STOP; # The problem is unbounded.

Choose a variable i_0 to **leave the basis** from the set $i_0 \in \arg \min_{i \in I, a'_{ij_0} > 0} \left\{ \frac{b'_i}{a'_{ij_0}} \right\}$. $I \leftarrow (I \setminus \{i_0\})$ and $J \leftarrow J \cup \{i_0\}$ \triangleright Move i_0 from basic to non-basic for $i \in I$ do

▷ Row elimination on pivot (i_0, j_0) .

 \triangleright Normalize the coefficient of $a'_{i_0 j_0}$

Input: $I = \{n + 1, ..., n + m\}, J = \{1, ..., n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

 $\begin{array}{ll} \text{if } c_i' \leq 0 \text{ for all } i \in J \text{ then} \\ & \text{STOP}; \quad \# \text{ Optimal point found.} \\ \text{Choose a variable } j_0 \text{ to enter the basis from the set } j_0 \in \{j \in J : c_j' > 0\}. \\ & \text{if } a_{ij_0}' \leq 0 \text{ for all } i \in J \text{ then} \\ & \text{STOP}; \qquad \# \text{ The problem is unbounded.} \end{array}$

Choose a variable i_0 to **leave the basis** from the set $i_0 \in \arg \min_{i \in I, a'_{ij_0} > 0} \left\{ \frac{b'_i}{a'_{ij_0}} \right\}$. $I \leftarrow (I \setminus \{i_0\})$ and $J \leftarrow J \cup \{i_0\}$ \triangleright Move i_0 from basic to non-basic

for $i \in I$ do $a'_{i:} \leftarrow a'_{i:} - \frac{a'_{i_{0}}}{a'_{i_{0}j_{0}}}a'_{i_{0}}$ $a'_{i_{0}:} \leftarrow \frac{1}{a'_{i_{0}j_{0}}}a'_{i_{0}:}$ and $a'_{i_{0}j_{0}} \leftarrow \frac{1}{a'_{i_{0}j_{0}}}$ $c' \leftarrow c' - \frac{c'_{j_{0}}}{a'_{i_{0}i_{0}}}a'_{i_{0}:}$

▷ Row elimination on pivot (i_0, j_0) .

 \triangleright Normalize the coefficient of $a'_{i_0 i_0}$

▷ Update the cost coefficients.

Input: $I = \{n + 1, \dots, n + m\}, J = \{1, \dots, n\}, a'_{ij} \in \mathbb{R}, b'_i \ge 0, c'_i \in \mathbb{R}.$

 $\begin{array}{ll} \text{if } c_i' \leq 0 \text{ for all } i \in J \text{ then} \\ & \text{STOP}; \quad \# \text{ Optimal point found.} \\ \text{Choose a variable } j_0 \text{ to enter the basis from the set } j_0 \in \{j \in J : c_j' > 0\}. \\ & \text{if } a_{ij_0}' \leq 0 \text{ for all } i \in J \text{ then} \\ & \text{STOP}; \qquad \# \text{ The problem is unbounded.} \end{array}$

Choose a variable i_0 to **leave the basis** from the set $i_0 \in \arg \min_{i \in I, a'_{ij_0} > 0} \left\{ \frac{b'_i}{a'_{ij_0}} \right\}$. $I \leftarrow (I \setminus \{i_0\})$ and $J \leftarrow J \cup \{i_0\}$ \triangleright Move i_0 from basic to non-basic

 $I \leftarrow (I \setminus \{i_0\})$ and $J \leftarrow J \cup \{i_0\}$ for $i \in I$ do

$$a_{i:}^{\prime} \leftarrow a_{i:}^{\prime} - rac{a_{ij_0}^{\prime}}{a_{i_0i_0}^{\prime}}a_{i_0:}^{\prime}$$

$$\begin{aligned} a_{i_0:}' \leftarrow \frac{1}{a_{i_0j_0}'} a_{i_0:}' & \text{and} & a_{i_0j_0}' \leftarrow \frac{1}{a_{i_0j_0}'} \\ c' \leftarrow c' - \frac{C_{j_0}'}{a_{i_0j_0}'} a_{i_0:}' \\ I \leftarrow I \cup \{j_0\} & \text{and} & J \leftarrow (J \setminus \{j_0\}) \end{aligned}$$

 \triangleright Row elimination on pivot (i_0, j_0) .

 \triangleright Normalize the coefficient of $a'_{i_0 j_0}$

▷ Update the cost coefficients.

 \triangleright Move j_0 from non-basic to basic

How to choose who enters the basis?

$$j_0 \in \{j \in J : c'_j > 0\}$$

- **(**) The mad hatter rule: Choose the first one you see costs: O(1)
- 2 Dantzig's 1st rule: $j_0 = \arg \max_{\substack{i \in J}} c_i \quad \text{cost: O(n)}$
- **③** Dantzig's 2nd rule: Choose j_0 that maximizes the increase in z.

$$j_0 = \arg \max_{j \in J} \left\{ c_j \min_{i \in I, a_{ij} > 0} \left\{ \frac{b_i}{a_{ij}} \right\} \right\} \quad costs : O(nm)$$

Effective, but computationally expensive.

4 Bland's rule: Choose the smallest indices j_0 and i_0 . That is, choose

$$j_0 = \arg\min\{j \in J : c_j > 0\} \quad costs : O(n)$$
$$i_0 = \min\left\{\arg\min_{i \in I, a_{ij_0} > 0}\left\{\frac{b_i}{a_{ij_0}}\right\}\right\}.$$

Degeneracy

Consider the problem

$$\begin{array}{ll} \max & 2x_1 - x_2 + 8x_3 \\ & 2x_3 & \leq 1 \\ & 2x_1 - 4x_2 + 6x_3 & \leq 3 \\ & -x_1 + 3x_2 + 4x_3 & \leq 2 \\ & x_1, x_2, x_3 \geq 0. \end{array}$$

Adding slack variables we have that

$$x_4 = 1 + 0 + 0 - 2x_3$$

$$x_5 = 3 - 2x_1 + 4x_2 - 6x_3$$

$$x_6 = 2 + x_1 - 3x_2 - 4x_3$$

$$z = 0 + 2x_1 - x_2 + 8x_3$$

Degeneracy

If any of the basic variables are zero, then we say that the solution is degenerate.

Consider the initial dictionary:

If x_3 enters then who leaves?

Degeneracy

If any of the basic variables are zero, then we say that the solution is degenerate.

Consider the initial dictionary:

If x_3 enters then who leaves? Both x_5 and x_6 are set to zero, so either one. Choosing x_4 and pivoting on row 1 and column 4 we have.

$$x_{3} = 0.5 + 0 + 0 - 0.5x_{4}$$

$$x_{5} = 0 - 2x_{1} + 4x_{2} + 3x_{4}$$

$$x_{6} = 0 + x_{1} - 3x_{2} + 2x_{4}$$

$$z = 4 + 2x_{1} - x_{2} - 4x_{4}$$

Only x_1 can enter the basis, but it doesn't increase in value :(Full example in lecture notes.

Bland's rule for degeneracy

Bland's rule

Choose the smallest indices j_0 and i_0 . That is, choose

$$j_0=rgmin\{j\in J\,:\,c_j>0\}.$$

$$i_0 = \min\left\{\arg\min_{i\in I, a_{ij_0}>0}\left\{\frac{b_i}{a_{ij_0}}\right\}\right\}.$$

Definition

A dictionary is degenerate if there are basic variables equal to zero.

Theorem

If Bland's rule is used on all degenerate dictionaries, then the simplex algorithm will not cycle.

Finding an initial feasible dictionary

The point $(x_1^*, x_2^*, x_3^*) = (0, 0, 0)$ is not feasible.

Finding an initial feasible dictionary

r

The point $(x_1^*, x_2^*, x_3^*) = (0, 0, 0)$ is not feasible. Setup an auxiliary problem

For x_0 big enough, it will be feasible. Setup initial dictionary

Pivot on "most infeasible" variable in the basis with the most negative value. Thus x_5 leaves the basis and x_0 enters the basis. Pivoting on row 2 and column 5:

Pivot on "most infeasible" variable in the basis with the most negative value. Thus x_5 leaves the basis and x_0 enters the basis. Pivoting on row 2 and column 5:

$$r_1 \leftarrow r_1 - r_2.$$

$$r_3 \leftarrow r_3 - r_2.$$

$$w \leftarrow w + r_2.$$

Pivot on "most infeasible" variable in the basis with the most negative value. Thus x_5 leaves the basis and x_0 enters the basis. Pivoting on row 2 and column 5:

$$r_{1} \leftarrow r_{1} - r_{2}.$$

$$r_{3} \leftarrow r_{3} - r_{2}.$$

$$w \leftarrow w + r_{2}.$$

$$x_{4} = 9 \quad +0 \quad -2x_{2} \quad -x_{3} \quad +x_{5}$$

$$x_{0} = 5 \quad 2x_{1} \quad -3x_{2} \quad +x_{3} \quad +x_{5}$$

$$x_{6} = 4 \quad +3x_{1} \quad -4x_{2} \quad +3x_{3} \quad +x_{5}$$

$$w = -5 \quad -2x_{1} \quad +3x_{2} \quad -x_{3} \quad -x_{5}.$$

Now x_2 enters and who leaves?

Pivot on "most infeasible" variable in the basis with the most negative value. Thus x_5 leaves the basis and x_0 enters the basis. Pivoting on row 2 and column 5:

$$r_{1} \leftarrow r_{1} - r_{2}.$$

$$r_{3} \leftarrow r_{3} - r_{2}.$$

$$w \leftarrow w + r_{2}.$$

$$x_{4} = 9 \quad +0 \quad -2x_{2} \quad -x_{3} \quad +x_{5}$$

$$x_{0} = 5 \quad 2x_{1} \quad -3x_{2} \quad +x_{3} \quad +x_{5}$$

$$x_{6} = 4 \quad +3x_{1} \quad -4x_{2} \quad +3x_{3} \quad +x_{5}$$

$$w = -5 \quad -2x_{1} \quad +3x_{2} \quad -x_{3} \quad -x_{5}.$$

Now x_2 enters and who leaves? x_6 leaves the basis

└─Simplex Algorithm └─Finding an initial feasible dictionary

Now x_2 enters and who leaves? x_6 leaves the basis. After pivoting

Simplex Algorithm

Finding an initial feasible dictionary

Now x_2 enters and who leaves? x_6 leaves the basis. After pivoting

Who enters the basis now?

Simplex Algorithm

Finding an initial feasible dictionary

Now x_2 enters and who leaves? x_6 leaves the basis. After pivoting

Who enters the basis now? x_3 Who leaves the basis?

Simplex Algorithm

Finding an initial feasible dictionary

Now x_2 enters and who leaves? x_6 leaves the basis. After pivoting

Who enters the basis now? x_3 Who leaves the basis?

$$\begin{array}{rcl} x_0 \geq 0 & \Rightarrow & 2-1.25x_3 \geq 0 & \Rightarrow & x_3 \geq 2/1.25 = 1.6\\ x_4 \geq 0 & \Rightarrow & 7-2.5x_3 \geq 0 & \Rightarrow & x_3 \geq 7/2.5 = 2.8\\ x_0 \text{ leaves the basis!} \end{array}$$

Finding an initial feasible dictionary

Pivoting on row 2 and column 3:

$$r_1 \leftarrow r_1 + \frac{0.75}{1.25}r_2 = r_1 + 0.6r_2.$$

 $r_3 \leftarrow r_3 - 2r_2.$
 $w \leftarrow w + r_2.$

$$\begin{array}{c} x_2 = 2.2 + 0.6x_1 + 0.4x_5 + 0.2x_6 - 0.6x_0 \\ x_3 = 1.6 - 0.2x_1 + 0.2x_5 + 0.6x_6 - 0.8x_0 \\ x_4 = 3 - x_1 - x_6 + 2x_0 \\ \hline w = \end{array}$$

Feasible basis without x_0 !

Finding an initial feasible dictionary

Pivoting on row 2 and column 3:

$$r_1 \leftarrow r_1 + \frac{0.75}{1.25}r_2 = r_1 + 0.6r_2.$$

 $r_3 \leftarrow r_3 - 2r_2.$
 $w \leftarrow w + r_2.$

$$\begin{array}{c} x_2 = 2.2 + 0.6x_1 + 0.4x_5 + 0.2x_6 - 0.6x_0 \\ x_3 = 1.6 - 0.2x_1 + 0.2x_5 + 0.6x_6 - 0.8x_0 \\ x_4 = 3 - x_1 - x_6 + 2x_0 \\ \hline w = \end{array}$$

Feasible basis without x_0 ! Remove column with x_0 and replace w with z.

 $z = x_1 - x_2 + x_3$

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= $x_1 - (2.2 + 0.6x_1 + 0.4x_5 + 0.2x_6) + (1.6 - 0.2x_1 + 0.2x_5 + 0.6x_6)$
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So the initial basis is

Now apply the simplex again!

Upper Bounds Using Duality

The LP in standard form

$$\max_{x} z \stackrel{\text{def}}{=} c^{\top} x$$

subject to $Ax \leq b$,
 $x \geq 0$, (LP)

We want to find $w \in \mathbb{R}$ so that $z = c^{\top}x \leq w$ for all $x \in \mathbb{R}^n$. Combine rows of constraints?

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We want to find $w \in \mathbb{R}$ so that $z = c^{\top}x \le w$ for all $x \in \mathbb{R}^n$. Combine rows of constraints? Look for $y \ge 0 \in \mathbb{R}^m$ so that $y^{\top}A \ge c^{\top}$ so that

$$c^{\top}x \leq (y^{\top}A)x \leq y^{\top}b =: w.$$

Can we make this upper bound as tight as possible? Yes, by minimizing $y^{\top}b$. That is, we need to the *dual* linear program.

Dual definition

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subject to $Ax \leq b$,
 $x \geq 0$, (P) Primal (1)

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Lemma (Weak Duality) If $x \in \mathbb{R}^n$ is a feasible point for (1) and $y \in \mathbb{R}^m$ is a feasible point for (2) then $c^\top x < y^\top Ax < y^\top b.$ (3)

Lemma (Weak Duality)

If $x \in \mathbb{R}^n$ is a feasible point for (1) and $y \in \mathbb{R}^m$ is a feasible point for (2) then

$$c^{\top}x \le y^{\top}Ax \le y^{\top}b.$$
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Strong Duality

Theorem (Strong Duality)

If (1) or (2) is feasible, then $z^* = w^*$. Moreover, if c^* is the cost vector of the optimal dictionary of the primal problem (1), that is, if

$$z = z^* + \sum_{i=1}^{n+m} c_i^* x_i,$$
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then $y_i^* = -c_{n+i}^*$ for i = 1, ..., m.

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Proof: First $c_i^* \leq 0$ for i = 1, ..., m + n because dictionary is optimal. Consequently $y_i^* = -c_{n+i}^* \geq 0$ for i = 1, ..., m.

Strong duality: Proof I

By the definition of the slack variables we have that

$$x_{n+i} = b_i - \sum_{j=1}^n a_{ij} x_j, \quad \text{for } i = 1, \dots, m.$$
 (6)

Consequently, setting $y_i^* = -c_{n+i}^*$, we have that

2

$$z \stackrel{(5)}{=} z^{*} + \sum_{j=1}^{n} c_{j}^{*} x_{j} + \sum_{i=n+1}^{n+m} c_{i}^{*} x_{i}$$

$$\stackrel{(6)}{=} z^{*} + \sum_{j=1}^{n} c_{j}^{*} x_{j} - \sum_{i=1}^{m} y_{i}^{*} (b_{i} - \sum_{j=1}^{n} a_{ij} x_{j})$$

$$= z^{*} - \sum_{i=1}^{m} y_{i}^{*} b_{i} + \sum_{j=1}^{n} \left(c_{j}^{*} + \sum_{i=1}^{m} y_{i}^{*} a_{ij} \right) x_{j}$$

$$\stackrel{\text{def of } z}{=} \sum_{j=1}^{n} c_{j} x_{j}, \quad \forall x_{1}, \dots, x_{n}.$$
(7)

Last line followed by definition $z = \sum_{j=1}^{n} c_j x_j$. Since the above holds for all $x \in \mathbb{R}^n$, we can match the coefficients.

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Strong duality: Proof II

$$z^* - \sum_{i=1}^m y_i^* b_i + \sum_{j=1}^n \left(c_j^* + \sum_{i=1}^m y_i^* a_{ij} \right) x_j = \sum_{j=1}^n c_j x_j.$$

Matching coefficients on x_i 's we have

$$z^* = \sum_{i=1}^m y_i^* b_i$$
 (8)

$$c_j = c_j^* + \sum_{i=1} y_i^* a_{ij}, \quad \text{for } j = 1, \dots, n.$$
 (9)

Since $c_j^* \leq 0$ for $j=1,\ldots,n,$ the above is equivalent to

$$z^* = \sum_{i=1}^m y_i^* b_i$$
 (10)

$$\sum_{i=1}^{m} y_i^* a_{ij} \leq c_j, \quad \text{for } j = 1, \dots, n.$$
 (11)

(11) $\Rightarrow y_i^*$ is feasible for (2). (10) $\Rightarrow z^* = \sum_{i=1}^m y_i^* b_i = w$, consequently by weak duality the y_i^* 's are dual optimal. \Box

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Re-writing the above in element form we have that

$$\sum_{j=1}^{n} (c_j - \sum_{i=1}^{m} a_{ij} y_i^*) x_j^* = 0 = \sum_{i=1}^{m} y_i^* (b_i - \sum_{j=1}^{n} a_{ij} x_j^*).$$

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Sum over positive numbers equal zero thus

$$y_i^* (b_i - \sum_{j=1}^n a_{ij} x_j^*) = 0, \quad \forall i = 1, \dots, m.$$

$$x_j^* (c_j - \sum_{i=1}^m a_{ij} y_i^*) = 0, \quad \forall j = 1, \dots, n.$$

/ 33

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This gives the following rule for computing y^* .

$$\sum_{i=1}^{n} a_{ij} y_i^* = c_j, \quad \forall j \in \{1, \dots, n\}, \quad x_j^* > 0.$$
$$y_i^* = 0, \quad \forall i \in \{1, \dots, m\}, \quad b_i > \sum_{j=1}^{n} a_{ij} x_j^*.$$

Question: If x^* is non-degenerate, how many $x_i^* > 0$?

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$$y_{i}^{*} = 0, \quad \forall i \in \{1, \dots, m\}, \ x_{n+i}^{*} > 0.$$

Finally

$$\sum_{i=1}^{n} a_{ij} y_{i}^{*} = c_{j} \quad \Rightarrow A_{J}^{\top} y^{*} = c_{J} \qquad (J \text{ indices of Basic variables})$$

Exercise on calculating dual variables

If
$$x_1^* = 3, x_2^* = 5$$

Then $y_1^* = \frac{3}{4}, y_2^* = 0, y_3^* = \frac{1}{4}$

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$$5x_1^* + 3x_2^* = 5 * 3 + 3 * 5 = 30 \quad \Rightarrow \quad y_1^* \neq 0$$
$$2x_1^* + 3x_2^* = 2 * 3 + 3 * 5 = 21 < 24 \quad \Rightarrow \quad y_2^* = 0$$

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$$x_1^* + 3x_2^* = 3 + 3 * 5 = 18 \implies y_3^* \neq 0.$$

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$$\begin{aligned} 5x_1^* + 3x_2^* &= 5 * 3 + 3 * 5 = 30 \quad \Rightarrow \quad y_1^* \neq 0 \\ 2x_1^* + 3x_2^* &= 2 * 3 + 3 * 5 = 21 < 24 \quad \Rightarrow \quad y_2^* = 0 \\ x_1^* + 3x_2^* &= 3 + 3 * 5 = 18 \quad \Rightarrow \quad y_3^* \neq 0. \end{aligned}$$

Setup linear system $\sum_{i=1} a_{ij}y_i^* &= c_j, \forall j \text{ with } x_j^* > 0: \end{aligned}$

$$\begin{array}{c|cccc} & 4 & 3 \\ y_1 & 5x_1 & +3x_2 \\ y_2 & 2x_1 & +3x_2 \\ y_3 & x_1 & +3x_2 \end{array}$$

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Test for complementarity:

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1 4

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G., R & P Richtárik, Randomized Iterative Methods for Linear Systems arXiv:1506.03296