

ORF 522

Linear Programming and Convex Analysis

The Simplex Method, Tableaux and Dictionaries

Marco Cuturi

Reminder: Basic Feasible Solutions, Extreme points, Optima

Three fundamental theorems:

- Let \mathbf{x} be a **basic feasible solution (BFS)** to a LP with index set I and objective value z . If $\exists e, 1 \leq e \leq n, e \notin I$ such that $c_e - z_e > 0$ **and** at least one $y_{i,e} > 0$, then we can have a **better basic feasible** solution by replacing an index in I by e with a new objective $\hat{z} \geq z$, **strictly** if x_I is non-degenerate.
- Let \mathbf{x}^* be a **basic feasible solution (BFS)** to a LP with index set I and objective value z^* . If $c_i - z_i^* \leq 0$ for all $1 \leq i \leq n$ then \mathbf{x}^* is optimal.
- Let \mathbf{x} be a **basic feasible solution (BFS)** to a LP with index set I . If \exists an index $e \notin I$ such that $y_e \leq 0$ then the **feasible region** is **unbounded**. If moreover for e the reduced cost $c_e - z_e > 0$ then there exists a feasible solution with at most $m + 1$ nonzero variables and an **arbitrary large objective function**.

Today

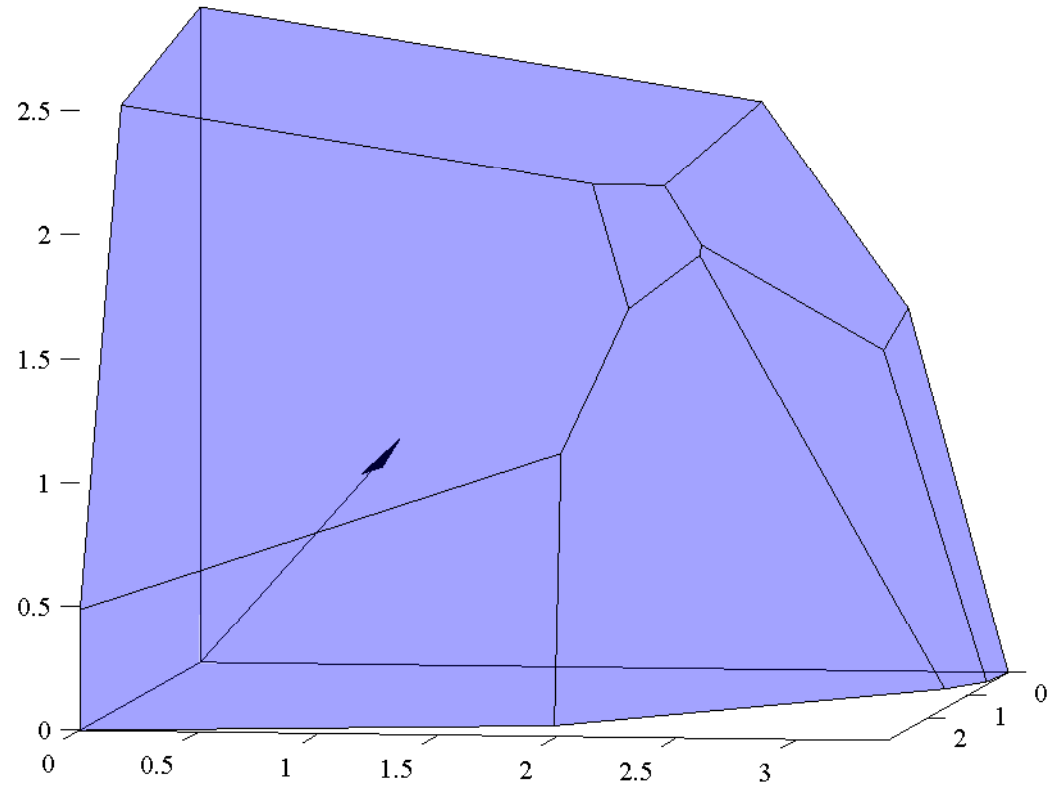
- Visualizing the simplex.
- Example of tableaux in canonical feasible form.

Reflecting on the Algorithm

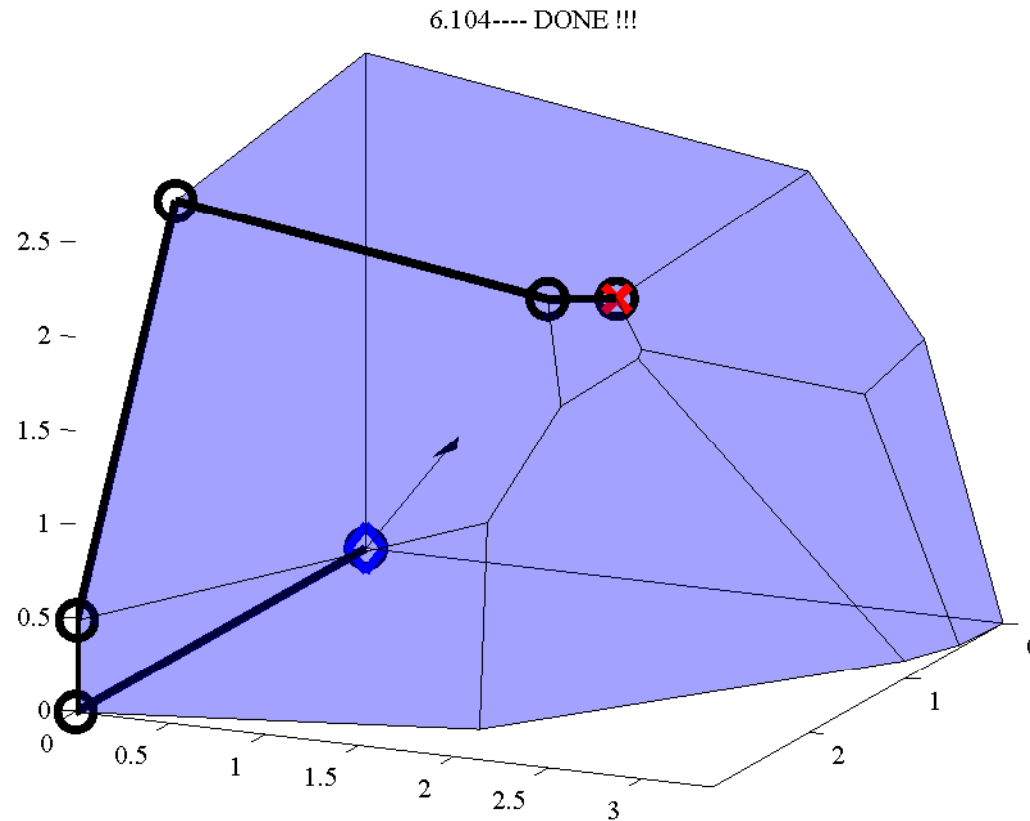
So far, what is the simplex?

- The simplex is a family of algorithms which do the following:
 1. Obtains an **initial** Basic feasible solution. **more on that later.**
 2. iterates: move from one BFS \mathbf{I} to a **better** BFS \mathbf{I}' :
 - check reduced cost coefficients $c_j - c_{\mathbf{I}}^T B_{\mathbf{I}}^{-1} \mathbf{a}_j$, $j \in \mathbf{O}$. **if all negative \mathbf{I} is optimal, OVER.**
 - otherwise, pick **one** index e for which it is positive. this will enter \mathbf{I} .
 - Check coordinates of $\mathbf{y}_e = B_{\mathbf{I}}^{-1} \mathbf{a}_e$. **if all ≤ 0 then optimum is unbounded, OVER.**
 - otherwise, take the index r such that it achieves the minimum in $\{\frac{x_{i_j}}{y_{j,e}} | y_{j,e} > 0, 1 \leq j \leq m\}$, this will ensure feasibility. The r th index of the base \mathbf{I} is $i_r \leq n$.
 - $\mathbf{I}' = \{\mathbf{I} \setminus i_r\} \cup e$.
 - We have improved on the objective. If $x_{\mathbf{I}}$ was **not** degenerate, we have **strictly** improved.
 - $\mathbf{I} \leftarrow \mathbf{I}'$
- The loop is on a finite set of extreme points. it either exits early (unbounded), exits giving an answer (optimum \mathbf{I}^* and corresponding solution \mathbf{x}^*) or loops indefinitely (degeneracy).

A Matlab Demo With Polyhedrons Containing the Origin



A Matlab Demo With Polyhedrons Containing the Origin



now with the real matlab demo...

Tableaux with Canonical Feasible Form

WHY tableaux ?

- Last time: an example where we move from a base \mathbf{I} to a new base \mathbf{I}' , compute $B_{\mathbf{I}'}^{-1}$, do the multiplications etc.. and reach the optimum. This is the simplex.
- **Double issue:**
 - **Computational 1:** inverting matrices costs time & money. One column is different between $B_{\mathbf{I}}$ and $B_{\mathbf{I}'}$, can we do better than inverting everything again?
 - **Computational 2:** multiplying matrices costs time & money. $B_{\mathbf{I}}^{-1}A$ and $B_{\mathbf{I}'}^{-1}A$ are related.

WHY tableaux ?

- **Down to what we really need at each iteration:**

- reduced cost coefficients vector $(c_i - z_i)$ of \mathbf{R}^n to pick an index e and check optimality,
- All column vectors of A in the base \mathbf{I} , that is Y , to check boundedness and choose r , namely all coordinates of $\mathbf{y}_e = B_{\mathbf{I}}^{-1}a_e$ in particular.
- The current basic solution vector, $B_{\mathbf{I}}^{-1}\mathbf{b}$ both to choose r and on exit.
- Having also the objective $c_{\mathbf{I}}^T B_{\mathbf{I}}^{-1}\mathbf{b}$ would help.

- Summing up, we need something that keeps track of

\ddots	\dots	\ddots	\vdots
\vdots	$B_{\mathbf{I}}^{-1}A$	\vdots	$B_{\mathbf{I}}^{-1}\mathbf{b}$
\ddots	\dots	\ddots	\vdots
\dots	$(\mathbf{c} - \mathbf{z})'$	\dots	$c_{\mathbf{I}}^T B_{\mathbf{I}}^{-1}\mathbf{b}$

Canonical Feasible Form: We know an initial BFS to corresponding Standard Form

- let's **standardize** a **feasible** (*i.e.* $\mathbf{b} \geq 0$) canonical form:

$$\begin{array}{ll} \text{maximize} & \alpha^T \mathbf{u} \\ \text{subject to} & \begin{cases} M\mathbf{u} \leq \mathbf{b} \\ \mathbf{u} \geq 0 \end{cases} \end{array}$$

- We assume that $\mathbf{u}, \alpha \in \mathbf{R}^d$ for a d dimensional objective and $M \in \mathbf{R}^{m \times d}$ and $\mathbf{b} \in \mathbf{R}^m$ for m constraints.

Canonical Feasible Form: We know an initial BFS to corresponding Standard Form

- Slack variables x_{d+1}, \dots, x_{d+m} can be added so that $[A, I_m] \begin{bmatrix} \mathbf{u} \\ x_{d+1} \\ \vdots \\ x_{d+m} \end{bmatrix} = \mathbf{b}$ and the problem is now with $\mathbf{c} = [\alpha, \underbrace{0, \dots, 0}_m] \in \mathbf{R}^{d+m}$

$$\begin{aligned} &\text{maximize} && x_0 = \mathbf{c}^T \mathbf{x} \\ &\text{subject to} && \begin{cases} [M, I_m] \mathbf{x} = \mathbf{b} \\ \mathbf{x} \geq 0 \end{cases} \end{aligned}$$

- $\mathbf{x}, \mathbf{c} \in \mathbf{R}^{m+d}$, $\mathbf{c} = \begin{bmatrix} \alpha \\ \mathbf{0} \end{bmatrix}$, $A = [M, I_m] \in \mathbf{R}^{m \times (m+d)}$ and same $\mathbf{b} \in \mathbf{R}^m$.
- The dimensionality of the problem is now $n = d + m$.

Simplex Method: Tableau

Let us represent this by an (annotated) tableau:

	O						I						b
	x_1	x_2	\cdots	x_e	\cdots	x_d	x_{d+1}	x_{d+2}	\cdots	x_{d+r}	\cdots	x_{d+m}	
x_{d+1}	m_{11}	m_{12}	\cdots	m_{1e}	\cdots	m_{1d}	1	0	\cdots	0	\cdots	0	b_1
x_{d+2}	m_{21}	m_{22}	\cdots	m_{2e}	\cdots	m_{2d}	0	1	\cdots	0	\cdots	0	b_2
\vdots	\vdots	\vdots	\ddots	\vdots	\ddots	\vdots	\vdots	\vdots	\ddots	\vdots	\ddots	\vdots	\vdots
x_{d+r}	m_{r1}	m_{r2}	\cdots	m_{re}	\cdots	m_{rd}	0	0	\cdots	1	\cdots	0	b_r
\vdots	\vdots	\vdots	\ddots	\vdots	\ddots	\vdots	\vdots	\vdots	\ddots	\vdots	\ddots	\vdots	\vdots
x_{d+m}	m_{m1}	m_{m2}	\cdots	m_{me}	\cdots	m_{md}	0	0	\cdots	0	\cdots	1	b_m
x_0	c_1	c_2	\cdots	c_e	\cdots	c_d	0	0	\cdots	0	\cdots	0	0

- Since $\mathbf{b} \geq 0$, take an original BFS as $\left[\underbrace{0, \dots, 0}_d, b_1, b_2, \dots, b_m \right]^T$
- Why:
 - **basic**: $I = \{d + 1, \dots, d + m\}$
 - **feasible**: $[0, \dots, 0, b_1, b_2, \dots, b_m]^T \geq 0$.

Simplex Method: Tableau

- the structure of the tableau so far,

A	\mathbf{b}
\mathbf{c}^T	0

- The index set \mathbf{I} so far $\{d + 1, d + 2, \dots, d + m\}$.
- $B_{\mathbf{I}} = I_m$, $B_{\mathbf{I}}^{-1}\mathbf{b} = \mathbf{b}$, $B_{\mathbf{I}}^{-1}A = A$ etc..
- The lower-right coincides with the objective so far... 0
- \mathbf{c} is actually equal to $(\mathbf{c} - \mathbf{z}_{\mathbf{I}})$ when \mathbf{I} only describes slack variables.

Simplex Method without non-negativity and objectives...

- Remember: a basis \mathbf{I} gives a sparse solution $\mathbf{x}_{\mathbf{I}}$.
- **there's one basis \mathbf{I}^*** which is the good one.
- The solution is \mathbf{x} such that $\mathbf{x}_{\mathbf{I}^*}^* = B_{\mathbf{I}^*}^{-1}\mathbf{b}$ and the rest is zero.
- We can start with the **slack variables** as a basis in canonical **feasible** form.
- Under this form, the first matrix basis is $B_{\mathbf{I}} = I_m$ the identity matrix.
- We will **move** from one basis to the other. We've proved this is possible.
- In doing so, we also have to recast the cost.
- Let's check how it looks in practice, without looking at feasibility and objective related concepts.

...the Gauss pivot...

- Consider now taking a variable out of \mathbf{I} to replace it by a variable in \mathbf{O} .
- The r th index of \mathbf{I} , i_r leaves the basis, e initially in \mathbf{O} is removed.

...the Gauss pivot...

- Two ways of looking at the same operation:

- Through elementary row/column operations transfer a vector $\begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ where the 1 is in position r to get a similar basis vector in the e th column of A .

$$\begin{array}{c} \vdots \\ r \\ \vdots \\ i \\ \vdots \end{array} \begin{bmatrix} \cdots & i_r & \cdots & e & \cdots \\ \ddots & 0 & \ddots & \vdots & \ddots \\ \cdots & 1 & \cdots & a_{re} & \cdots \\ \ddots & 0 & \ddots & \vdots & \ddots \\ \cdots & 0 & \cdots & a_{ie} & \cdots \\ \ddots & 0 & \ddots & \vdots & \ddots \end{bmatrix} \Rightarrow \begin{array}{c} \vdots \\ r \\ \vdots \\ i \\ \vdots \end{array} \begin{bmatrix} \cdots & i_r & \cdots & e & \cdots \\ \ddots & \vdots & \ddots & 0 & \ddots \\ \cdots & a_{r,i_r} & \cdots & \mathbf{1} & \cdots \\ \ddots & \vdots & \ddots & 0 & \ddots \\ \cdots & a_{i,i_r} & \cdots & 0 & \cdots \\ \ddots & \vdots & \ddots & 0 & \ddots \end{bmatrix}$$

...the Gauss pivot...

- Consider the equalities $A\mathbf{x} = \mathbf{b}$ written in row form,

$$\mathbf{u}_i^T \mathbf{x} = b_i$$

where the \mathbf{u} 's are the rows of A .

- Putting variable x_e in the basis is equivalent to isolating x_e so that is present in **all but one** of the m equations, with coefficient 1. On the other hand we let x_{i_r} enter all equations again, that is

$$x_e = \tilde{b}_i - \sum_{i=1, i \neq e}^n \rho_i x_i$$

and x_e does not appear elsewhere.

...the Gauss pivot

- This is achieved through a pivot in the tableau.
- Once the r th element of basis \mathbf{I} , namely column $i_r \leq n$, and $e \leq n$ are agreed upon, the rules to update the tableau are:
 - (a) in pivot row $a_{rj} \leftarrow a_{rj}/a_{re}$.
 - (b) in pivot column $a_{re} \leftarrow 1, a_{ie} = 0$ for $i = 1, \dots, m, i \neq r$: the e th column becomes a matrix of zeros and a one.
 - (c) for all other elements $a_{ij} \leftarrow a_{ij} - \frac{a_{rj}a_{ie}}{a_{re}}$

The Gauss pivot

- Graphically,

$$\begin{array}{c} \vdots \\ i \\ \vdots \\ r \\ \vdots \end{array} \begin{array}{c} \dots \\ j \\ \dots \\ e \\ \dots \end{array} \begin{array}{c} \dots \\ \vdots \\ \dots \\ \vdots \\ \dots \end{array} \begin{array}{c} \dots \\ a_{ij} \\ \dots \\ a_{re} \\ \dots \end{array} \begin{array}{c} \dots \\ a_{ie} \\ \dots \\ a_{re} \\ \dots \end{array} \Rightarrow \begin{array}{c} \vdots \\ i \\ \vdots \\ r \\ \vdots \end{array} \begin{array}{c} \dots \\ j \\ \dots \\ e \\ \dots \end{array} \begin{array}{c} \dots \\ \vdots \\ \dots \\ \vdots \\ \dots \end{array} \begin{array}{c} \dots \\ a_{ij} - \frac{a_{rj}a_{ie}}{a_{re}} \\ \dots \\ a_{rj}/a_{re} \\ \dots \end{array} \begin{array}{c} \dots \\ 0 \\ \dots \\ 1 \\ \dots \end{array} \begin{array}{c} \dots \\ \vdots \\ \dots \\ \vdots \\ \dots \end{array}$$

- Look at how the column e is now a column of 0 and 1's. This makes sense

since $B_I^{-1} \mathbf{a}_e = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ with 1 in e th position means \mathbf{a}_e is in the basis.

Linear system and pivoting

- Consider the linear system

$$\begin{cases} x_1 + x_2 - x_3 + x_4 & = 5 \\ 2x_1 - 3x_2 + x_3 + x_5 & = 3 \\ -x_1 + 2x_2 - x_3 + x_6 & = 1 \end{cases}$$

- The corresponding tableau

$$\begin{array}{ccccccc} \mathbf{a}_1 & \mathbf{a}_2 & \mathbf{a}_3 & \mathbf{a}_4 & \mathbf{a}_5 & \mathbf{a}_6 & \mathbf{b} \\ \left[\begin{array}{ccccccc} 1 & 1 & -1 & 1 & 0 & 0 & 5 \\ 2 & -3 & 1 & 0 & 1 & 0 & 3 \\ -1 & 2 & -1 & 0 & 0 & 1 & 1 \end{array} \right] \end{array}$$

Simplex Method: Swapping Indexes

- in the corresponding tableau,

$$\begin{array}{c}
 \mathbf{a}_4 \\
 \mathbf{a}_5 \\
 \mathbf{a}_6
 \end{array}
 \begin{bmatrix}
 \mathbf{a}_1 & \mathbf{a}_2 & \mathbf{a}_3 & \mathbf{a}_4 & \mathbf{a}_5 & \mathbf{a}_6 & \mathbf{b} \\
 1 & 1 & -1 & \mathbf{1} & 0 & 0 & 5 \\
 2 & -3 & 1 & 0 & \mathbf{1} & 0 & 3 \\
 -1 & 2 & -1 & 0 & 0 & \mathbf{1} & 1
 \end{bmatrix}$$

notice the structure:

$$\begin{array}{ccccccc}
 \dots & \dots & \dots & \dots & \dots & \dots & \vdots \\
 \vdots & M & \vdots & \vdots & I_3 & \vdots & \mathbf{b} \\
 \dots & \dots & \dots & \dots & \dots & \dots & \vdots
 \end{array}$$

- And the fact that by taking the obvious basis $\mathbf{I} = \{4, 5, 6\}$ we have $B_{\mathbf{I}} = I_3$ and $B_{\mathbf{I}}^{-1} = I_3$

Simplex Method: Let's pivot

- Let's pivot arbitrarily. We put **1** in the base and remove **4**.

$$\begin{array}{c}
 \mathbf{a}_4 \\
 \mathbf{a}_5 \\
 \mathbf{a}_6
 \end{array}
 \begin{array}{c}
 \mathbf{x}_1 \\
 \mathbf{a}_2 \\
 \mathbf{a}_3 \\
 \mathbf{a}_4 \\
 \mathbf{a}_5 \\
 \mathbf{a}_6 \\
 \mathbf{b}
 \end{array}
 \left[\begin{array}{ccccccc}
 \mathbf{1} & 1 & -1 & 1 & 0 & 0 & 5 \\
 2 & -3 & 1 & 0 & 1 & 0 & 3 \\
 -1 & 2 & -1 & 0 & 0 & 1 & 1
 \end{array} \right]$$

which yields

$$\begin{array}{c}
 \mathbf{a}_1 \\
 \mathbf{a}_5 \\
 \mathbf{a}_6
 \end{array}
 \begin{array}{c}
 \mathbf{a}_1 \\
 \mathbf{a}_2 \\
 \mathbf{a}_3 \\
 \mathbf{a}_4 \\
 \mathbf{a}_5 \\
 \mathbf{a}_6 \\
 \mathbf{b}
 \end{array}
 \left[\begin{array}{ccccccc}
 \mathbf{1} & 1 & -1 & 1 & 0 & 0 & 5 \\
 \mathbf{0} & -5 & 3 & -2 & 1 & 0 & -7 \\
 \mathbf{0} & 3 & -2 & 1 & 0 & 1 & 6
 \end{array} \right]$$

- $\mathbf{I} = \{1, 5, 6\}$, that is $B_{\mathbf{I}} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$. The basic solution is such that $\mathbf{x}_{\mathbf{I}} = B_{\mathbf{I}}^{-1}\mathbf{b}$
- Note that all coordinates of $\mathbf{a}_1, \dots, \mathbf{a}_6, \mathbf{b}$ in the table are given with respect to $\mathbf{a}_1, \mathbf{a}_5, \mathbf{a}_6$. In particular the last column corresponds to $B_{\mathbf{I}}^{-1}\mathbf{b}$...not feasible here BTW.

Simplex Method: again...

- Let's pivot arbitrarily again, this time inserting **2** and removing the **second** variable of the basis, **5**.

$$\begin{array}{c}
 \mathbf{a}_1 \\
 \mathbf{a}_5 \\
 \mathbf{a}_6
 \end{array}
 \begin{array}{c}
 \mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3 \quad \mathbf{a}_4 \quad \mathbf{a}_5 \quad \mathbf{a}_6 \quad \mathbf{b} \\
 \left[\begin{array}{ccccccc}
 1 & 1 & -1 & 1 & 0 & 0 & 5 \\
 0 & -5 & 3 & -2 & 1 & 0 & -7 \\
 0 & 3 & -2 & 1 & 0 & 1 & 6
 \end{array} \right]
 \end{array}$$

$$\begin{array}{c}
 \mathbf{a}_1 \\
 \mathbf{a}_2 \\
 \mathbf{a}_6
 \end{array}
 \begin{array}{c}
 \mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3 \quad \mathbf{a}_4 \quad \mathbf{a}_5 \quad \mathbf{a}_6 \quad \mathbf{b} \\
 \left[\begin{array}{ccccccc}
 1 & 0 & -\frac{2}{5} & \frac{3}{5} & \frac{1}{5} & 0 & \frac{18}{5} \\
 0 & 1 & -\frac{3}{5} & \frac{2}{5} & -\frac{1}{5} & 0 & \frac{7}{5} \\
 0 & 0 & -\frac{1}{5} & -\frac{1}{5} & \frac{3}{5} & 1 & \frac{9}{5}
 \end{array} \right]
 \end{array}$$

- Notice how one can keep track of who is in the basis by checking where 0/1's columns are.
- The solution is now feasible... pure luck.

Simplex Method: and again...

- once again, pivot inserting **3** and removing the **third** variable of the basis, **6**.

$$\begin{array}{c}
 \mathbf{a}_1 \\
 \mathbf{a}_2 \\
 \mathbf{a}_6
 \end{array}
 \begin{array}{ccccccc}
 \mathbf{a}_1 & \mathbf{a}_2 & \mathbf{a}_3 & \mathbf{a}_4 & \mathbf{a}_5 & \mathbf{a}_6 & \mathbf{b} \\
 \left[\begin{array}{ccccccc}
 1 & 0 & -\frac{2}{5} & \frac{3}{5} & \frac{1}{5} & 0 & \frac{18}{5} \\
 0 & 1 & -\frac{3}{5} & \frac{2}{5} & -\frac{1}{5} & 0 & \frac{7}{5} \\
 0 & 0 & -\frac{1}{5} & -\frac{1}{5} & \frac{3}{5} & 1 & \frac{9}{5}
 \end{array} \right]
 \end{array}$$

$$\begin{array}{c}
 \mathbf{a}_1 \\
 \mathbf{a}_2 \\
 \mathbf{a}_3
 \end{array}
 \begin{array}{ccccccc}
 \mathbf{a}_1 & \mathbf{a}_2 & \mathbf{a}_3 & \mathbf{a}_4 & \mathbf{a}_5 & \mathbf{a}_6 & \mathbf{b} \\
 \left[\begin{array}{ccccccc}
 1 & 0 & \mathbf{0} & 1 & -1 & -2 & 0 \\
 0 & 1 & \mathbf{0} & 1 & -2 & -3 & -4 \\
 0 & 0 & \mathbf{1} & 1 & -3 & -5 & -9
 \end{array} \right]
 \end{array}$$

- horrible. moving randomly we have a now non-feasible degenerate basic solution.
- yet we knew that pivoting randomly based only on $y_{r,e} \neq 0$ would lead us nowhere.

Adding the reduced costs

- What happens when we also pivot the last line?
- Remember the last line is equal to $\mathbf{v} \stackrel{\text{def}}{=} (\mathbf{c} - \mathbf{z})'$ in the beginning.
- Remember also that
 - (a) in pivot row $a_{rj} \leftarrow a_{rj}/a_{re}$.
 - (b) in pivot column $a_{re} \leftarrow 1, a_{ie} = 0$ for $i = 1, \dots, m, i \neq r$: the e th column becomes a matrix of zeros and a one.
 - (c) for all other elements $a_{ij} \leftarrow a_{ij} - \frac{a_{rj}a_{ie}}{a_{re}}$
- Here, (a) does not apply, we cannot be in the pivot row.
- we have
 - in pivot column $v_e = 0$: makes sense, reduced cost is zero for basis elements.
 - for all other elements $v_j \leftarrow v_j - \frac{a_{rj}v_e}{a_{re}}$.

Adding the reduced costs

- Recapitulating, at each iteration of the pivot the matrix is exactly

...	⋮
⋮	$B_I^{-1}M$	⋮	⋮	B_I^{-1}	⋮	$B_I^{-1}\mathbf{b}$
...	⋮
...	$(\mathbf{c} - \mathbf{z})$...	$-x_0$

- The pivot is thus applied on the $m + 1 \times n + 1$ tableau.
- The tableau contains **everything we need**, reduced costs, (minus)objective, the coordinates of $B_I^{-1}\mathbf{b}$ and $B_I^{-1}A$

Tableaux with Arbitrary Initial BFS

Working around to go back to previous situation

- Suppose that we are given an arbitrary BFS \mathbf{I} for the problem

$$\begin{array}{ll} \text{maximize} & x_0 = \mathbf{c}^T \mathbf{x} \\ \text{subject to} & \begin{cases} A\mathbf{x} = \mathbf{b} \\ \mathbf{x} \geq 0 \end{cases} \end{array}$$

- We try to go back to the previous situation.

Working around to go back to previous situation

- Perform a permutation of columns such that columns in positions i_1, \dots, i_m become in last positions $n - m + 1, \dots, n$.
- A is now $[N, B]$ (N for non-basic part) and the system can be written as

$$\begin{cases} N\mathbf{x}_N + B\mathbf{x}_B = \mathbf{b} \\ \mathbf{c}_N^T \mathbf{x}_N + \mathbf{c}_B^T \mathbf{x}_B = x_0 \end{cases}$$

- Multiplying the first line by B^{-1} ,

$$B^{-1}N\mathbf{x}_N + \mathbf{x}_B = B^{-1}\mathbf{b} \text{ thus } \mathbf{x}_B = B^{-1}\mathbf{b} - B^{-1}N\mathbf{x}_N$$

which when used for objective x_0 yields

$$x_0 = (\mathbf{c}_N - \mathbf{c}_B^T B^{-1}N)^T \mathbf{x}_N + \mathbf{c}_B^T B^{-1}\mathbf{b}$$

Working around to go back to previous situation

- We can now use the same tableau:

...	⋮
⋮	$B^{-1}N$	⋮	⋮	I	⋮	\mathbf{b}
...	⋮
...	$\mathbf{c}_N^T - \mathbf{z}_N^T$	$\mathbf{0}$...	$-x_0$

- And can apply the simplex as defined for canonical feasible forms.

Short Comment on Dictionaries

- A dictionary is a comparable compact form

$$\begin{array}{rcll}
 \text{maximize} & \zeta & = & 4x_2 + 3x_3 \\
 \text{subject to} & w_1 & = & 5 - 2x_1 - 3x_2 - x_3 \\
 & w_2 & = & 11 - 4x_1 - x_2 - 2x_3 \\
 & w_3 & = & 8 - 3x_1 - 4x_2 - 2x_3 \\
 & & & x_1, x_2, x_3, w_1, w_2, w_3 \geq 0.
 \end{array}$$

where basic variables are kept on the top and non-basic are kept on the left.

- We save space (1/0 columns) but need to **keep track of variable names**.
- The constants on the left correspond to the last column in tableaux.
- The first line stands for reduced cost coefficients of **nonbasic variables**.
- The lower-right corresponds to minus the $B_I^{-1}A$ matrix for indices in \mathbf{O} .
- Equivalent to Tableaux, rather used for educational purposes.