

## Lecture 9: May 23

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# 1 One iteration of the Simplex algorithm a.k.a. a "pivot"

## 1.1 Pseudocode

1. Start with basis matrix  $A_{B(1)}, \dots, A_{B(m)} \rightarrow$  basic feasible solution  $\mathbf{x}$ .
2. Compute reduced costs  $\bar{c}_j = c_j - c_j^\top A_B^{-1} A_j$  for each nonbasic variable  $x_j$ .
  - (a) if all  $\bar{c}_j \geq 0$  then we are optimal. STOP.
  - (b) choose *some*  $j$  with  $\bar{c}_j < 0$
3. Compute  $u = A_B^{-1} A_j = -d_B^j$ . If  $u \leq 0$  then optimum =  $-\infty$  and we stop.
4. Choose index  $l$  such that  $u_l > 0$  and

$$\frac{x_{B(l)}}{u_l} = \theta^* = \min \left\{ \frac{x_{B(i)}}{u_i} \mid i \in [m] \text{ and } u_i > 0 \right\} \quad (1)$$

5. Form new basis by replacing  $A_{B(l)}$  with  $A_j$ .

## 1.2 Introduction

The computational most expensive operation used in a pivot is the generation of  $A_B^{-1}$  which has to be recomputed whenever the basis  $B$  changes. Computing the inverse of a matrix costs  $O(n^3)$  by applying brute-force, e.g. Gauss-Elimination.

We observe that only one column of  $B$  will change per pivot. Assume that  $A_B^{-1}$  is at our disposal from the last pivot and we want to compute  $A_{\bar{B}}^{-1}$  where  $\bar{B}$  denotes a new basis.

The idea we get is that  $A_{\bar{B}}^{-1}$  might look similar to  $A_B^{-1}$  if  $A_{\bar{B}}$  and  $A_B$  differ only by one column. This lecture introduces a method exploiting elementary row operations within a *tableau*-scheme to compute  $A_{\bar{B}}^{-1}$  from  $A_B^{-1}$  in  $O(n^2)$ .

### 1.2.1 Mathematical point of view

$$\begin{aligned}
 A_B^{-1}A_B &= I \\
 A_B^{-1}A_{B(i)} &= e_i \\
 A_B^{-1}A_j &= u
 \end{aligned}$$

$$A_B^{-1}A_{\bar{B}} = \begin{bmatrix} 1 & 0 & 0 & \cdots & u_1 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & u_2 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & u_3 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & u_l & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & u_{n-1} & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & u_n & 0 & 0 & 1 \end{bmatrix} \quad (2)$$

Our goal is to find a matrix  $Q$  such that  $QA_B^{-1}A_{\bar{B}} = I$ . In other words we search for elementary row operations transforming  $A_B^{-1}$  into  $A_{\bar{B}}^{-1}$  via  $Q$  ( $QA_B^{-1} = A_{\bar{B}}^{-1}$ ).

### 1.3 Elementary row operations

We will use elementary row operations to setup the matrix  $Q$ . A short recap, there are three different elementary row operations:

1. Switching: A row within the matrix can be switched with another row,
2. Multiplication: Components of a row-vector may be multiplied with a non-zero scalar  $\lambda \in \mathbb{R}$ ,
3. Addition: Multiples of a row  $j$  can be added to another row  $i$  where  $i \neq j$ .

Starting from equation 2 our first goal is to bring the entry  $(l, j) = u_l$  to 1. Multiplying  $i$ -th row by some  $\alpha \neq 0$  is equivalent to multiply with  $Q_1$  from left. For simplicity we assume  $l = j$  here.

$$Q_1 = \begin{pmatrix} 1 & 0 & 0 & \cdots & 0 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \alpha & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \quad (3)$$

Since  $Q_1$  is a square matrix with full rank, it is invertible. Another argumentation is that  $\det(Q_1) \neq 0$ .

To eliminate the non-diagonal entries of  $u$  a multiple of the  $l$ -th row can be subtracted from the  $i$ -th row. Written in terms of a matrix operation multiplied from left:

$$Q_2 = \begin{pmatrix} 1 & 0 & \cdots & \cdots & \cdots & 0 \\ 0 & 1 & 0 & \cdots & \beta & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 1 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \begin{matrix} j \\ i \end{matrix} \quad (4)$$

In summary, elementary row operations will help us to turn  $A_B^{-1}A_{\bar{B}}$  into  $I$ .

- For each  $i \neq l$  add  $l$ -th row  $-\frac{u_i}{u_l}$  times to the  $i$ -th row.
- Multiply  $l$ -th row by  $\frac{1}{u_l}$

In matrix notation we need a series of matrices  $Q_i$  such that:

$$\underbrace{Q_m \cdots Q_2 Q_1}_Q A_B^{-1} A_{\bar{B}} = I \implies Q A_B^{-1} = A_{\bar{B}}^{-1} \quad (5)$$

Implementation of the elementary row operations to accelerate the Simplex method is often done in a tableau-scheme.

### 1.4 Simplex: A full tableau implementation

Idea: Every iteration of the Simplex algorithm maintains a tableau (matrix) which stores important properties of the current Simplex iteration.

	Negated objective value of $\mathbf{x}$	Reduced costs		
	$-\mathbf{c}_B^T \mathbf{x}_B$	$\bar{c}_1$	$\cdots$	$\bar{c}_n$
$x_B$	$\left\{ \begin{array}{c} x_{B(1)} \\ \vdots \\ x_{B(m)} \end{array} \right.$	$A_B^{-1} A_1$	$\cdots$	$A_B^{-1} A_n$
		$\underbrace{\hspace{2cm}}_{u_1}$		$\underbrace{\hspace{2cm}}_{u_n}$



Minimal index  $l = 2 \implies A_1$  enters the basis and  $A_5$  leaves the basis. New basis:  $B(1) = 4$ ,

$$\begin{array}{c|cccccc} 100 & 0 & -7 & -2 & 0 & 5 & 0 \\ \hline 10 & 0 & 0.5 & 1 & 1 & -0.5 & 0 \\ 10 & 1 & 0.5 & 1 & 0 & 0.5 & 0 \\ 0 & 0 & 1 & -1 & 0 & -1 & 1 \end{array}$$

Table 3: Tableau after 1st iteration.

$B(2) = 1, B(3) = 6$ .

## 1.6 Lemma 44

The elementary row operations lead to tableau: where  $\bar{B}$  is obtained from adding  $j$  to  $B$

$$\begin{array}{c|c} -c_B^\top A_{\bar{B}}^{-1} b & c^\top - c_B^\top A_{\bar{B}}^{-1} A \\ \hline A_{\bar{B}}^{-1} b & A_{\bar{B}}^{-1} A \end{array}$$

Table 4: Tableau after performing elementary row operations.

and removing  $B(l)$  from  $B$ .

### 1.6.1 Proof

Consider entries of  $A_{\bar{B}}^{-1}$  and  $A_{\bar{B}}^{-1} A$ :

Elementary row operations are equivalent to left-multiplying with a matrix  $Q$  such that  $QA_{\bar{B}}^{-1} = A_{\bar{B}}^{-1}$ .

0-th row: We started with  $[\underbrace{0}_{\text{First entry}} \mid \underbrace{c^\top}_{\text{Second entry}}] - \underbrace{g^\top [b \mid A]}_{\text{Linear combination of rows of } A}$ , with  $g^\top = c_B^\top A_{\bar{B}}^{-1}$ .

After the  $i$ -th iteration:  $[0 \mid c^\top] - p^\top [b \mid A] \implies c_j - p^\top A_j = 0$  where  $j = \bar{B}(l)$ . Consider columns of old basis  $\rightarrow$  identity matrix.  $l$ -th row was zero at column  $j \rightarrow$  new reduced cost stays 0. Let  $i \neq l : \bar{c}_{B(i)} = 0$  and entry stays 0 after update.  $c_B^\top - p^\top A_{\bar{B}} = 0 \implies p^\top = c_B^\top A_{\bar{B}}^{-1} \implies 0$   $\square$