

MATH464 - Lecture 16 (03/02/2023)

Today: * Example for revised simplex method
 * full tableau implementation

Problems from Hw6

S =

5. **Exercise 2.16** Consider the set $\{ \mathbf{x} \in \mathbb{R}^n \mid x_1 = \dots = x_{n-1} = 0, 0 \leq x_n \leq 1 \}$. Could this be the feasible set of a problem in standard form?

P in standard form is $\{ \bar{\mathbf{x}} \in \mathbb{R}^n \mid A\bar{\mathbf{x}} = \bar{b}, \bar{\mathbf{x}} \geq 0 \}$, with $\text{rank}(A) = m, m \leq n$.

The given set (let's call it S) is obviously not in standard form.
 The question is whether this set could describe a set that is in standard form.

You should assume $n \geq 2$.

Note that $\bar{\mathbf{x}} = \bar{0} \in S$. So, if $S = P$, then $\bar{0} \in P$ as well.

$\Rightarrow A(\bar{0}) = \bar{b} \Rightarrow \bar{b} = \bar{0}$ in the standard form.

with $\bar{b} = \bar{0}$, if $\bar{\mathbf{x}} \in P$, $\lambda \bar{\mathbf{x}} \in P \quad \forall \lambda \geq 0$

$$A(\lambda \bar{\mathbf{x}}) = \lambda A(\bar{\mathbf{x}}) = \lambda (\bar{b}) = \lambda \bar{0} = \bar{0}. \quad \& \quad \lambda \bar{\mathbf{x}} \geq \bar{0}.$$

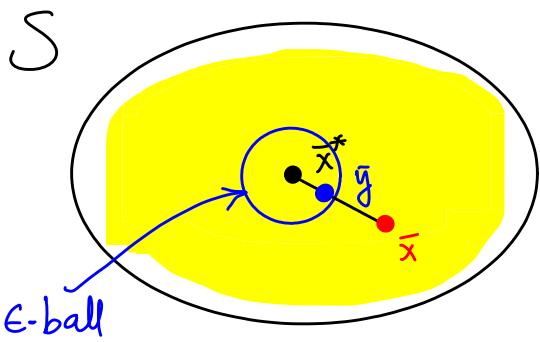
So $\lambda > 1$ is also good.

But $\lambda \bar{x}_n \neq 1$ for $\lambda > 1$.

So $\lambda \bar{\mathbf{x}} \notin S$ for $\lambda > 1$, giving an obstruction

for $S = P$.

6. Problem 3.6. Show that a local minimum \bar{x}^* of a convex function over a convex set S is the global minimum.



\bar{x}^* is a local optimum of $f \Rightarrow$

$\exists \epsilon > 0$, s.t.

$$f(\bar{x}^*) \leq f(\bar{x}) \quad \forall \bar{x} \text{ s.t. } \|\bar{x} - \bar{x}^*\| \leq \epsilon$$

Need to prove \bar{x}^* is global minimum, i.e., $f(\bar{x}^*) \leq f(\bar{x}) \quad \forall \bar{x} \in S$
(and not just inside the ϵ -ball around \bar{x}^*).

Proof by contradiction: Assume $\exists \bar{x}$ outside the ϵ -ball centered at \bar{x}^* with $f(\bar{x}) < f(\bar{x}^*)$, i.e., \bar{x}^* is not a global minimum.

Consider a point \bar{y} on the line segment connecting \bar{x}^* and \bar{x} that is within the ϵ -ball @ \bar{x}^* , i.e., $\|\bar{y} - \bar{x}^*\| < \epsilon$.

$$\Rightarrow f(\bar{x}^*) \leq f(\bar{y}) \quad (1)$$

Q. Why are guaranteed to find such a \bar{y} ?

\bar{y} is a point on the line segment joining \bar{x}^* and \bar{x}

$$\Rightarrow \bar{y} = \lambda \bar{x}^* + (1-\lambda) \bar{x} \text{ for some } \lambda \in (0, 1).$$

Q. Why do we choose λ strictly between 0 and 1?

$$\Rightarrow f(\bar{y}) = f(\lambda \bar{x}^* + (1-\lambda) \bar{x}) \leq \lambda f(\bar{x}^*) + (1-\lambda) \bar{x} \quad (2)$$

Combine (1) & (2) to get a contradiction...

why?

Hw7: AMPL problems. You **must** separate the model from data. A good rule to follow: do **not** include any actual numbers, i.e.; data in your model file.

An example for the Revised Simplex Method

Recall how we maintain and update \bar{B}^{-1} :

$$[\bar{B}^{-1} | -\bar{d}_{\bar{B}}] \xrightarrow{\text{EROs}} [(\bar{B}')^{-1} | \bar{e}_l]$$

Solve using the revised simplex method.

$$\left. \begin{array}{l} \min -x_1 - x_2 \\ \text{s.t. } 2x_1 + x_2 \leq 4 \\ 3x_1 + 5x_2 \leq 15 \\ x_1, x_2 \geq 0 \end{array} \right\} \quad \left. \begin{array}{l} \min -x_1 - x_2 \\ \text{s.t. } 2x_1 + x_2 + x_3 = 4 \\ 3x_1 + 5x_2 + x_4 = 15 \\ x_j \geq 0 \forall j \end{array} \right\} \quad \left. \begin{array}{l} \bar{C}^T = [1 \ 2 \ 3 \ 4] \\ A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 3 & 5 & 0 & 1 \end{bmatrix} \\ \bar{b} = \begin{bmatrix} 4 \\ 15 \end{bmatrix} \\ m=2, n=4 \end{array} \right\}$$

Iteration 1

$$B(1) = 3, B(2) = 4 \quad \mathcal{J} = \{3, 4\}, \mathcal{N} = \{1, 2\}. \quad B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, B^{-1} = B = I.$$

$$\bar{C}_B^T = [0 \ 0]. \quad \bar{C}_N^T = \bar{C}_N^T - \cancel{\bar{C}_B^T \bar{B}^T N} = [-1 \ -1] \quad j=1 \quad (x_1 \text{ enters}).$$

$$\bar{x}_B = \begin{bmatrix} x_3 \\ x_4 \end{bmatrix} = \bar{B}^{-1} \bar{b} = \begin{bmatrix} 4 \\ 15 \end{bmatrix}. \quad \bar{d}_B = -\bar{B}^{-1} A_1 = -\begin{bmatrix} 2 \\ 3 \end{bmatrix}.$$

$$\theta^* = \min \left\{ \frac{-4}{-2}, \frac{-15}{-3} \right\} = 2 \quad l=1 \quad B(1)=3 \text{ leaves the basis} \\ (x_3 \text{ leaves}).$$

$$\mathcal{J}' = \{1, 4\}. \quad \bar{x}' = \bar{x} + \theta^* \bar{d} = \begin{bmatrix} 0 \\ 0 \\ 4 \\ 15 \end{bmatrix} + 2 \begin{bmatrix} 1 \\ 0 \\ -2 \\ -3 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 0 \\ 9 \end{bmatrix}.$$

$$[\bar{B}^{-1} | -\bar{d}_{\bar{B}}] \xrightarrow{\text{EROs}} [(\bar{B}')^{-1} | \bar{e}_l] \quad \bar{e}_l = \bar{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ here.}$$

$$\left[\begin{array}{c|cc} 1 & 0 & 2 \\ 0 & 1 & 3 \end{array} \right] \xrightarrow[\substack{\text{then} \\ R_3 - 3R_1}]{R_1 \times \frac{1}{2}} \left[\begin{array}{cc|c} \frac{1}{2} & 0 & 1 \\ -\frac{3}{2} & 1 & 0 \end{array} \right] \Rightarrow \text{new } B^{-1} = \begin{bmatrix} \frac{1}{2} & 0 \\ -\frac{3}{2} & 1 \end{bmatrix}.$$

Iteration 2

$$\mathcal{J} = \{1, 4\}, N = \{2, 3\}. \quad \bar{C}_B^T = [-1 \ 0]$$

$$\begin{aligned}\bar{C}_N^{T'} &= \bar{C}_N^T - \bar{C}_B^T B^{-1} N = \begin{bmatrix} 2 & 3 \\ -1 & 0 \end{bmatrix} - \begin{bmatrix} -1 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} \frac{1}{2} & 0 \\ -\frac{3}{2} & 1 \end{bmatrix}}_{\begin{bmatrix} -\frac{1}{2} & 0 \\ -\frac{1}{2} & -\frac{1}{2} \end{bmatrix}} \begin{bmatrix} 1 & 1 \\ 5 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 2 & 3 \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix} \quad j=2 (\bar{x}_2 \text{ enters}). \quad \underbrace{\begin{bmatrix} -\frac{1}{2} & 0 \\ -\frac{1}{2} & -\frac{1}{2} \end{bmatrix}}_{\text{no tie to break here!}}\end{aligned}$$

$$\bar{d}_B = -B^{-1} A_2 = -\begin{bmatrix} \frac{1}{2} & 0 \\ -\frac{3}{2} & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 5 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} \\ -\frac{7}{2} \end{bmatrix} \quad \bar{x} = \begin{bmatrix} 2 \\ 0 \\ 0 \\ 9 \end{bmatrix}$$

$$\theta^* = \min \left\{ \frac{-2}{-\frac{1}{2}}, \frac{-9}{-\frac{7}{2}} \right\} = \frac{18}{7}, l=2 \text{ here. } B(2)=4 \text{ leaves.}$$

$$\mathcal{J}' = \{1, 2\} \text{ (new basis). } \bar{x}' = \bar{x} + \theta^* \bar{d} = \begin{bmatrix} 2 \\ 0 \\ 0 \\ 9 \end{bmatrix} + \frac{18}{7} \begin{bmatrix} -\frac{1}{2} \\ 1 \\ 0 \\ -\frac{7}{2} \end{bmatrix} = \begin{bmatrix} 5/7 \\ 18/7 \\ 0 \\ 0 \end{bmatrix}.$$

Update B^{-1} :

$$\left[\begin{array}{cc|c} \frac{1}{2} & 0 & \frac{1}{2} \\ -\frac{3}{2} & 1 & \frac{7}{2} \end{array} \right] \xrightarrow[\substack{\text{then} \\ R_1 - \frac{1}{2}R_2}]{R_2 \times \frac{2}{7}} \left[\begin{array}{cc|c} \frac{5}{7} & -\frac{1}{7} & 0 \\ -\frac{3}{7} & \frac{2}{7} & 1 \end{array} \right] \quad \text{New } B^{-1} = \begin{bmatrix} \frac{5}{7} & -\frac{1}{7} \\ -\frac{3}{7} & \frac{2}{7} \end{bmatrix}.$$

Iteration 3

$$\mathcal{J} = \{1, 2\}, N = \{3, 4\}. \quad \bar{C}_B^T = \begin{bmatrix} 1 & 2 \\ -1 & -1 \end{bmatrix} \quad \bar{C}_N^T = \begin{bmatrix} 3 & 4 \\ 0 & 0 \end{bmatrix}$$

$$\begin{aligned}\bar{C}_N^{T'} &= \bar{C}_N^T - \bar{C}_B^T B^{-1} N = \begin{bmatrix} 0 & 0 \end{bmatrix} - \begin{bmatrix} -1 & -1 \end{bmatrix} \underbrace{\begin{bmatrix} \frac{5}{7} & -\frac{1}{7} \\ -\frac{3}{7} & \frac{2}{7} \end{bmatrix}}_{\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} \frac{2}{7} & \frac{1}{7} \end{bmatrix} \geq 0\end{aligned}$$

$$\bar{x} = \begin{bmatrix} 5/7 \\ 18/7 \\ 0 \\ 0 \end{bmatrix} \text{ is an optimal bfs, with the optimal objective function value } Z^* = \bar{C}^T \bar{x} = -\frac{23}{7}.$$

Full-Tableau Implementation (LP in standard form)

For large LPs, the revised simplex method is the efficient way to solve them. But we will also introduce a compact way of representing the numbers, and implementing all the computations - the tableau simplex method.

We maintain and update $[\bar{x}_B | \bar{B}'A]$ or $\bar{B}^{-1}[\bar{b} | A]$. This is an $m \times (n+1)$ matrix, or table, called the **simplex tableau**.

$$[\bar{B}'\bar{b} | \bar{B}'A_1 \ \bar{B}'A_2 \dots \bar{B}'A_n]$$

Nothing new here!

$$\bar{B}'(A\bar{x} = \bar{b}) \Rightarrow$$

$$(\bar{B}'A)\bar{x} = \bar{B}'\bar{b}$$

0^{th} column: values of the basic variables

i^{th} column: has $\bar{B}'A_i$, for $i=1, \dots, n$.

If x_j is entering, the column having $\bar{B}'A_j = -\bar{d}_B$ is the **pivot column**.

If $x_{B(l)}$ leaves the basis ($1 \leq l \leq n$), then the l^{th} row is the **pivot row**. The (l, j) -element is the **pivot element** or just the pivot.

We include the costs \bar{c}^T and the objective function $z = \bar{c}^T \bar{x}$ at the top as the zero-th row.

\bar{c}' (reduced cost vector)

	0	1	2	...	j	...	n
0	$-\bar{c}_B^T \bar{B}'\bar{b}$	$\bar{c}^T - \bar{c}_B^T \bar{B}'A$					
1	$\bar{B}'\bar{b}$	$\bar{B}'A$					
2							
⋮							
m							

=

	0	1	2	...	j	...	n
0	$-\bar{c}_B^T \bar{B}'\bar{b}$	c'_1	c'_2	...	c'_j	...	c'_n
1	$x_{B(1)}$						
2	$x_{B(2)}$	$\bar{B}'A_1$	$\bar{B}'A_2$...	$\bar{B}'A_j$...	$\bar{B}'A_n$
⋮							
m	$x_{B(m)}$						

$\rightarrow -z = -\bar{c}_B^T \bar{x}_B$