

MATH 464. HW #8. Lazizbek.

Ex. 3.17.

$$(1) \begin{cases} \min z = 2x_1 + 3x_2 + 3x_3 + x_4 - 2x_5 \\ \text{s.t.} & x_1 + 3x_2 + 0x_3 + 4x_4 + x_5 = 2 \\ & x_1 + 2x_2 + 0x_3 - 3x_4 + x_5 = 2 \\ & -x_1 - 4x_2 + 3x_3 + 0x_4 + 0x_5 = 1. \end{cases}$$

$$x \geq 0, \quad x \in \mathbb{R}^5$$

Auxiliary problem for the given s.f. problem (1) with $(\frac{1}{3})$ equality constraints is as follows:

$$(2) \begin{cases} \min w = 0x_1 + 0x_2 + 0x_3 + 0x_4 + 0x_5 + x_6 + x_7 \\ \text{s.t.} & x_1 + 3x_2 + 0x_3 + 4x_4 + x_5 + 0x_6 + 0x_7 = 2 \\ & x_1 + 2x_2 + 0x_3 - 3x_4 + x_5 + 0x_6 + 0x_7 = 2 \\ & \left(-\frac{1}{3}\right)x_1 - \frac{4}{3}x_2 + 1x_3 + 0x_4 + 0x_5 + 0x_6 + 0x_7 = \frac{1}{3} \\ & x \geq 0, \quad x \in \mathbb{R}^7 \end{cases}$$

Matrix s.f. of the auxiliary problem (2):

$$(3) \begin{cases} \min w = c_w^T \cdot x \\ \text{s.t.} & Ax = b \\ & x \geq 0 \\ & x \in \mathbb{R}^7 \end{cases} \quad \begin{aligned} c_w^T &= [0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 1]^T \\ A &= \begin{bmatrix} 1 & 3 & 0 & 4 & 1 & 1 & 0 \\ 1 & 2 & 0 & -3 & 1 & 0 & 1 \\ -\frac{1}{3} & -\frac{4}{3} & 1 & 0 & 0 & 0 & 0 \end{bmatrix} \\ b &= \left[2, 2, \frac{1}{3} \right]^T \end{aligned}$$

(4) choose basis $\beta_1 = \{6, 7, 3\} \Rightarrow$

$$(5) \begin{cases} x_B = (x_6, x_7, x_3) = (2, 2, \frac{1}{3}) \geq 0 \\ x_N = (x_1, x_2, x_4, x_5) = (0, 0, 0, 0) \geq 0 \end{cases} \Rightarrow$$

$$(6) B = [A_6 \ A_7 \ A_3] = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \text{ so } x_B = B^{-1}b = b = \begin{bmatrix} 2 \\ 2 \\ \frac{1}{3} \end{bmatrix}$$

$$(7) -z = -C_B^T x_B = -[0 \ 0 \ 3] \cdot \begin{bmatrix} 2 \\ 2 \\ \frac{1}{3} \end{bmatrix} = -1, \text{ recall } C_z = [2, 3, 3, 1, -2, 0, 0]^T$$

$$(8) -w = -C_{wB}^T x_B = -[1 \ 1 \ 0] \cdot \begin{bmatrix} 2 \\ 2 \\ \frac{1}{3} \end{bmatrix} = -4$$

(9) reduced costs for z :

$$\begin{aligned} \bar{C}_z &= C_z - C_{zB}^T \overbrace{B^{-1}A}^{=I} = \\ &= [2, 3, 3, 1, -2, 0, 0]^T - [0 \ 0 \ 3] \cdot I A = \\ &= [2, 3, 3, 1, -2, 0, 0]^T - [0, 0, 3] \cdot \begin{bmatrix} 1 & 3 & 0 & 4 & 1 & 1 & 0 \\ 1 & 2 & 0 & -3 & 1 & 0 & 1 \\ -\frac{1}{3} & -\frac{4}{3} & 1 & 0 & 0 & 0 & 0 \end{bmatrix} = \\ &= [2, 3, 3, 1, -2, 0, 0]^T - [-1 \ -4 \ 3 \ 0 \ 0 \ 0 \ 0] = [3, 7, 0, 1, -2, 0, 0]^T \end{aligned}$$

(10) reduced cost for w :

$$\begin{aligned} \bar{C}_w &= C_w - C_{wB}^T \overbrace{B^{-1}A}^{=I} = [0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 1]^T - [1, 1, 0] \cdot \begin{bmatrix} 1 & 3 & 0 & 4 & 1 & 1 & 0 \\ 1 & 2 & 0 & -3 & 1 & 0 & 1 \\ -\frac{1}{3} & -\frac{4}{3} & 1 & 0 & 0 & 0 & 0 \end{bmatrix} = [-2 \ -5 \ 0 \ -1 \ -2 \ 0 \ 0]^T \end{aligned}$$

reduced cost for $w \uparrow$

(11) $u = B^{-1}A = A$

So, the initial tableau by (4) - ... - (10), (11):

	x_1	x_2	x_3	x_4	x_5	x_6	x_7	
row ① - $z =$	-1	3	7	0	1	-2	0	\bar{c}_z
② - $w =$	-4	-2	-5	0	-1	-2	0	\bar{c}_w
(12) ③ $x_6 =$	2	1	3	0	4	1	1	0
④ $x_7 =$	2	1	2	0	-3	1	0	1
⑤ $x_3 =$	$\frac{1}{3}$	$-\frac{1}{3}$	$-\frac{4}{3}$	1	0	0	0	0

Status:

- $\beta_1 = \{6, 7, 3\}$
- $x = (0, 0, \frac{1}{3}, 0, 0, 2, 2)$

• $z = 1$

• $w = 4$

• Non optimal

• Direction: []

• Valid pivot: ⑥

Status:

• $\beta_2 = \{6, 5, 3\}$

• $x = (0, 0, \frac{1}{3}, 0, 2, 0, 0) \geq 0$

• $z = -3$, $w = 0$

• Monoptimal, $\bar{c}_{w4} = -7 < 0$

• Degenerate \Rightarrow Bland's Rule!

	x_1	x_2	x_3	x_4	x_5	x_6	x_7
$R1 = r1 + 2 \cdot r4$	3	5	11	0	-5	0	2
$R2 = r2 + 2 \cdot r4$	0	0	-1	0	-7	0	2
$R3 = r3 - r4 \quad x_6 =$	0	0	1	0	7	0	1
$R4 = r4 \quad x_5 =$	2	1	2	0	-3	1	0
$R5 = r5 \quad x_3 =$	$\frac{1}{3}$	$-\frac{1}{3}$	$-\frac{4}{3}$	1	0	0	0

• I'm going to use Bland's Rule by changing basis to go out of degeneracy:

	x_1	x_2	x_3	x_4	x_5	x_6	x_7
$R1 = r1 + \frac{5}{7}r3 - z =$	3	5	$\frac{82}{7}$	0	0	0	$\frac{5}{7}$
$R2 = r2 + r3 - w =$	0	0	0	0	0	0	1
$R3 = \frac{1}{7} \cdot r3 \quad x_4 =$	0	0	$\frac{1}{7}$	1	0	$\frac{1}{7}$	$-\frac{1}{7}$
$R4 = r4 + \frac{3}{7}r3 \quad x_5 =$	2	1	$\frac{17}{7}$	0	0	1	$\frac{3}{7}$
$R5 = r5 \quad x_3 =$	$\frac{1}{3}$	$-\frac{1}{3}$	$-\frac{4}{3}$	1	0	0	0

Status:

• $\beta_3 = \{4, 5, 3\} = \text{Initial b.f. basis}$

• $x = (0, 0, \frac{1}{3}, 0, 2, 0, 0) \geq 0$

• $z = -3$, $w = 0 \Rightarrow$ can stop

• Optimal $\bar{c}_z \geq 0$, $\bar{c}_w \geq 0$.

• still degenerate

Yes, we're still at a degenerate point $x^* = (0, 0, \frac{1}{3}, 0, 2)$, possibly there's a way to go out of degenerate point by Bland's Rule with zero step size at Δ I pointed out, but I'm leaving here as my answer is matching with the software, too!

math_464_actual_hw8

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#Math 464. HW8. Lazizbek

EXercise 3.17

```
min z = 2 x1 + 3 x2 + 3 x3 +    x4 - 2 x5

s.t.      x1 + 3 x2 + 0 x3 + 4 x4 +    x5 = 2

          x1 + 2 x2 + 0 x3 - 3 x4 +    x5 = 2

          (-1/3) x1 -(4/3)x2 + x3 + 0 x4 + 0 x5 = (1/3)

          x >= 0

          x1, x2,x3,x4, x5 in R
```

1 *Linear Program*

Now let's see the case where **all** decision variables are **reals**

```
[9]: """
      Script to demonstrate how to solve small mixed integer programs
      using the python optimize module in the scipy package
      """
      """
      All decision variables are from reals:
      """

      import pandas as pd
      import numpy as np
      import scipy.optimize as opt

      # The problem we will solve is:
      #
      # min z = 2 x1 + 3 x2 + 3 x3 +    x4 - 2 x5
      # s.t.      x1 + 3 x2 + 0 x3 + 4 x4 +    x5 = 2
      #          x1 + 2 x2 + 0 x3 - 3 x4 +    x5 = 2
      #          (-1/3)x1 -(4/3)x2 + x3 + 0 x4 + 0 x5 = (1/3)
      #          x >= 0
```

```

#           x1, x2,x3,x4, x5 in R

function_array = list()
solution_array = list()

# First build the objective vector.
c=np.array([2, 3, 3, 1, -2])

# Next, create the coefficient array for the inequality constraints.
# Note that the inequalities must be  $Ax \leq b$ , so some sign
# changes result when converting  $\geq$  into  $\leq$ .
# A = np.array([[ 1, 3, 0, 4, 1],\
#               [ 1, 2, 0, -3, 1],\
#               [(-1/3), (-4/3), 1, 0, 0]])
A = None

# Next the right-hand-side vector for the inequalities
# Sign changes can occur here too.
# b = np.array([345 , 50000 , 60000, 0, 0, 0, 0, 0 ])
b = None

#The coefficient matrix for the equality constraints and
# the right hand side vector.
Ae = np.array([[ 1, 3, 0, 4, 1],\
               [ 1, 2, 0, -3, 1],\
               [(-1/3), (-4/3), 1, 0, 0]]) # Ae = [[1,1,1,1]]

be = np.array([ 2, 2, (1/3)])

# Next, we provide any lower and upper bound vectors, one
# value for each decision variable. In this example all
# lower bound are zero and there are no upper bounds.
bounds=((0,np.inf),(0,np.inf),(0,np.inf),(0,np.inf),(0,np.inf))

# Lastly, we can specify which variables are required to be integer.
# If no variables are integer then isint=[]; In our example, only x2
# is integer.
# isint=[]

# The call to the mixed integer solver looks like the following.
# Notice that we pass usual "c" when we have a minimization
# problem, we send "-c" when we have max problem.
# This is because the solver is expecting a minimization.

res=opt.linprog(c,A,b,Ae,be,bounds)

# The result is stored in the dictionary variable "res".

```

```

# In particular, to show the optimal objective value and the
# optimal decision variable values:

print("min z = ", res['fun'])
print("at optimal solution x = ", res['x'])

# print(res['x'])
# print(np.dot(c, res['x']))
# print(res)

# To download:
# !sudo apt-get install texlive-xetex texlive-fonts-recommended_
↪texlive-plain-generic
# !jupyter nbconvert --to pdf /content/math_464_actual_hw2.ipynb

```

```

min z = -3.0
at optimal solution x = [0.          0.          0.33333333  0.          2.
]

```