LPmatrix.wxmx

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	DMratio (rowNum)	
(%i4)	<pre>load(draw)\$ set_draw_defaults(line_width=2, grid = [2,2], point_type = filled_d head_type = 'nofilled, head_angle = 20, head_length = 0.5, background_color = light_gray, draw_realpart=false)\$ fpprintprec : 5\$ ratprint : false\$</pre>	circle,
(%i5)	load (simplex);	
(%05)	C:/maxima-5.43.2/share/maxima/5.43.2/share/simplex/simplex.mac	
(%i6)	load ("econ1.mac");	

(%06) c:/work5/econ1.mac

1 Preface

In Example 1 of LPmatrix.wxmx we introduce a matrix simplex method in which the "z-row" is the bottom row. In addition to the matrix functions Mdefine, Mdisplay, and Mratio used in LPsimplex.wxmx, we introduce and use the new functions Mbtableau() and Mbpivot (nEnter, nLeave), in which the extra "b" denotes use with the z-row on the bottom.

In the second section of LPmatrix.wxmx are a number of examples which use Mbtableau and Mbpivot (together with Mratio) for matrix simplex maximization and minimization using both the Big M method and Butenko's Dual Simplex Method (the latter starts with solving the "auxiliary problem" of minimizing the sum of the artificial variables in the problem).

The third section introduces the B/N Dual Simplex Method, in which all given constraints are converted to <= constraints, and then slack variables are added to the left-hand side of each constraint to convert to constraint equations. When exhibited in the form of a simplex tableau, if the optimality condition is satisfied AND one or more basic variables have negative values, then this dual simplex method is applicable.

The (regular) simplex method moves the initial feasible but nonoptimal solution to an optimal solution while maintaining feasibility through an iterative procedure. On the other hand, the B/N dual simplex method moves the initial optimal but infeasible solution to a feasible solution while maintaining optimality through an iterative procedure.

The basic variable with the most negative value becomes the departing variable (D.V.) - leaves the basis. Call the row in which this value appears the work row. If more than one candidate for D.V. exists, choose one.

Form ratios by dividing the z-row coefficients of the non-basic variables by the corresponding negative coefficients of the work row. The nonbasic variable with the smallest absolute ratio becomes the entering variable (E.V.) - enters the basis. Designate this element in the work row as the pivot element and the corresponding column the work column. If more than one candidate for E.V. exists, choose one. If no element in the work row is negative, the problem has no feasible solution.

Use elementary row operations to convert the pivot element to 1 and then to reduce all the other elements in the work column to zero.

Repeat until all the variables have reached nonnegative values, so the optimal solution is feasible.

This B/N Dual Simplex Method is formulated in terms of matrices, and continues to use our functions Mbtableau() and Mbpivot (nEnter, nLeave), as well as DMratio (rowNum) which makes it easy to identify the entering variable starting with the work row number. The new functions are in the updated Econ1.mac file.

Dec. 22, 2022

2 References

Sergiy Butenko, ISEN 620, A Survey of Optimization, 82 videos https://www.youtube.com/playlist?list=PLY9yf2-4yyeQTLkCVFnGedyERuCjKW7kI

(B/N) Richard Bronson & G. Naadimuthu, Operations Research, 2nd ed, Schaum's Outline Series, McGraw-Hill, 1982.

3 Example 1

This matrix simplex example was presented in LPsimplex.wxmx using the convention that the "z-row" appears on top, using the functions Mtableau and Mpivot. Here we use the convention that the "z-row" appears on the bottom, using the functions Mbtableau and Mbpivot. With either convention we can use Mratio, Mdefine, and Mdisplay.

3.1 Using Mdefine, Mdisplay, Mbtableau, Mratio, Mbpivot

We use a maximization example from Butenko's Video 42.

Maximize $4^{*}x1 + 3^{*}x2 + 5^{*}x3$, subject to $x1 + 2^{*}x2 + 2^{*}x3 \le 4$, $3^{*}x1 + 4^{*}x3 \le 6$, $2^{*}x1 + x2 + 4^{*}x3 \le 8$, with x1, x2, x3 >= 0.

We first use maximize_lp (objective, condL, posL) from the simplex package:

```
(%i7) maximize_lp (4^{*}x1 + 3^{*}x2 + 5^{*}x3, [x1 + 2^{*}x2 + 2^{*}x3 \le 4, 3^{*}x1 + 4^{*}x3 \le 6,
2*x1 + x2 + 4^{*}x3 \le 8], [x1,x2,x3]);
```

(%07) [11,[x3=0,x2=1,x1=2]]

We get a different type of output using maxlp.

(%i8) maxlp $(4^*x1 + 3^*x2 + 5^*x3, [x1 + 2^*x2 + 2^*x3 \le 4, 3^*x1 + 4^*x3 \le 6, 2^*x1 + x2 + 4^*x3 \le 8])$ for z = 5 x3 + 3 x2 + 4 x1, such that, $2 x3 + 2 x2 + x1 \le 4$, $4 x3 + 3 x1 \le 6$, $4 x3 + x2 + 2 x1 \le 8$, $z^* = 11$ with x1 = 2, x2 = 1, x3 = 0,

Adding slack variables x4, x5, x6, this example becomes (c^t stands for the transpose of c):

Maximize $z = 4*x1 + 3*x2 + 5*x3 + 0*x4 + 0*x5 + 0*x6 = c^t \cdot X$, subject to A · X = b, or x1 + 2*x2 + 2*x3 + x4 + 0*x5 + 0*x6 = 4, 3*x1 + 0*x2 + 4*x3 + 0*x4 + x5 + 0*x6 = 6, 2*x1 + x2 + 4*x3 + 0*x4 + 0*x5 + x6 = 8, with x1, x2, x3, x4, x5, x6 >= 0, equivalent to X >= 0.

3.1.1 Define Initial Global Entities: X, c, A, b, NV, BV

The matrix method functions use the following global matrices:

c, X, A, b, N, B, Xn, Xb, cN, cB, and the global lists NV, BV.

The first four (c, X, A, b) remain the same throughout the optimum solution process. c, X, and b are matrix column vectors. A is a matrix of coefficients of the left-hand sides of the constraint equations. b is a matrix column vector of the right-hand side constants from the constraint equations. X is a matrix column vector whose elements are the variable names.

The remaining eight global entities (NV, BV, N, B, Xn, Xb, cN, cB) are redefined in each simplex algorithm step (each pivot).

We seek to optimize $z = c^t \cdot X = transpose (c) \cdot X$, subject to: A $\cdot X = b$, with $X \ge 0$.

With N a matrix of the columns of A corresponding to the non-basic variables (in order left to right) and with B a matrix of the columns of A corresponding to the basic variables (in order left to right), the matrix equation A . X = b turns into the matrix equation N . Xn + B . Xb = b, which can then be solved for Xb (the basic variables) in terms of the non-basic variables Xn:

 $Xb = -invert(B) \cdot N \cdot Xn + invert(B) \cdot b$

NV is an ordered list of the numbers associated with the non-basic variables for a particular tableau, and BV is an ordered list of the numbers associated with the basic variables of the same particular tableau.

cB is a matrix row vector with elements taken from the row vector transpose(c) which correspond to the variables in Xb. Likewise cN is a matrix row vector with elements taken from the row vector transpose(c) which correspond to the variables in Xn. Then we have the identity

 $z = c^{t} \cdot X = transpose(c) \cdot X = cB \cdot Xb + cN \cdot Xn$.

Using $Xb = -invert(B) \cdot N \cdot Xn + invert(B) \cdot b$, we can express z entirely in terms of the variables in Xn to get:

z = cB. invert(B). b + (cN - cB. invert(B). N). Xn.

The first term is some number (a constant) and the second term is a function of the non-basic variables at a given step. Since all non-basic variables, at a given step, are equal to zero, the second term is zero, and the current value of z is the first term.

The association of variables names with variable numbers (defined by our definition of the matrix column vector X), is (for this problem):

 $[x1,x2,x3,x4,x5,x6] \iff [1,2,3,4,5,6].$

```
We convert

Maximize z = 4^*x1 + 3^*x2 + 5^*x3

subject to

x1 + 2^*x2 + 2^*x3 + x4 = 4,

3^*x1 + 4^*x3 + x5 = 6,

2^*x1 + x2 + 4^*x3 + x6 = 8,

with x1, x2, x3, x4, x5, x6 \ge 0,

into matrix form.

Our initial feasible solution is x1 = x2 = x3 = 0 (so x1, x2, x3 are nonbasic variables for the

initial tableau), and x4 = 4, x5 = 6, x6 = 8 (so x4, x5, x6 are basic variables for the initial

tableau).
```

(%i14) X : cvec ([x1,x2,x3,x4,x5,x6])\$
 c : cvec ([4,3,5,0,0,0])\$
 A : matrix ([1,2,2,1,0,0], [3,0,4,0,1,0], [2,1,4,0,0,1])\$
 b : cvec ([4,6,8])\$
 NV : [1,2,3]\$
 BV : [4,5,6]\$

The above definitions of X, c, A, and b will remain unaltered throughout the optimization process. The definitions of NV and BV will change, swapping a basic and non-basic variable to produce the next simplex step (pivot).

3.1.2 Mdefine()\$ to define global N, B, Xb, Xn, cB, cN

Mdefine() uses global X, c, A, b, NV, and BV to define global N, B, Xn, Xb, cN, and cB. Unless details is set to true (default is false), the results are not displayed.

Here is a look at the Maxima code for Mdefine():

```
(%i17) display2d:false$
fundef (Mdefine);
display2d:true$
```

```
(%o16) Mdefine():=(N:newM(A,NV),B:newM(A,BV),Xn:part(X,NV),Xb:part(X,BV),
cN:transpose(part(c,NV)),cB:transpose(part(c,BV)),
if details then Mdisplay())
```

Here we call Mdefine():

(%i18) Mdefine()\$

3.1.3 Mdisplay()\$ to see NV, BV, N, B, Xb, Xn, cB, cN values

Since Mdefine(), by default, does not display the current definitions of N, B, Xn, Xb, cN, and cB, you can use Mdisplay()\$ to have the current values of these global variables displayed, and as a bonus, a reminder of NV and BV.

We first show our definition of the matrix A.

```
(%i19) display (A)$
```

 $A = \begin{pmatrix} 1 & 2 & 2 & 1 & 0 & 0 \\ 3 & 0 & 4 & 0 & 1 & 0 \\ 2 & 1 & 4 & 0 & 0 & 1 \end{pmatrix}$



cN and cB are each matrix row vectors.

3.1.4 Mbtableau(); to see current Matrix tableau with z-row on bottom

Once the lists NV and BV have been defined, and the values of N, B, Xn, Xb, cN, and cB have been defined, the function Mbtableau() creates a "matrix tableau" of the current state of the optimization process, with the "z-row" on the bottom.

Mbtableau() creates a matrix display in which the basic variables and their coefficients are on the left-hand side, the non-basic variables and their coefficients come next, reading left to right, and then comes the "rhs" values as a column, and finally a column showing the symbols standing for the current basic variables.

cB is a matrix row vector with elements taken from the row vector transpose(c) which correspond to the variables in Xb. Likewise cN is a matrix row vector with elements taken from the row vector transpose(c) which correspond to the variables in Xn. Then we have the identity

 $z = transpose(c) \cdot X = cB \cdot Xb + cN \cdot Xn$.

Using $Xb = -invert(B) \cdot N \cdot Xn + invert(B) \cdot b$, we can express z entirely in terms of the variables in Xn to get:

z = cB. invert(B). b + (cN - cB. invert(B). N). Xn.

The first term is some number (a constant) and the second term is a function of the non-basic variables at a given step. Since all non-basic variables, at a given step, are equal to zero, the second term is zero, and the current value of z is the first term.

The pure matrix form was used in LPsimplex.wxmx. Putting the z-row on the bottom here, that pure form matrix display has the form:

transpose (Xb)	transpose(Xn)	"rhs"	"Basis"
ident (length(Xb)	invert(B) . N	invert(B) . b	Xb
makelist(0,j,1,length(Xb))	-(cN - cB . invert(B) . N)	cB . invert(B) . b	"z"

In this file, Mbtableu() is used which converts the above pure matrix display into a matrix of numbers and symbols which is easier to read when the number of basic variables is large (a large number of conditions).

The first set of rows can be read off as a set of equations which must be satisfied in the current tableau. The last row is called the "z-row". The z-row coefficients of the basic variables Xb are all equal to zero, and the z-row coefficients of the non-basic variables Xn in the middle section provide possible further maximizing opportunities, provided at least one of the coefficients is negative.

In each step of the simplex algorithm, one non-basic variable and one basic variable trade roles, and appear in the opposite set of columns, maintaining the established order set by the constant vector X.

Here is the Step 0 tableau, the initial feasible solution tableau, with the basic variables (x4,x5,x6) appearing first on the left, and then the non-basic variables (x1,x2,x3) at this stage appearing in the middle section, with the condition coefficients appearing underneath in the next three rows (we have three conditions in this LP).

(%i21) Mbtableau();

	x4	x5	x6	Ι	x1	x2	х3	I	rhs	Basis
	1	0	0	Ι	1	2	2	Ι	4	х4
(%o21)	0	1	0	Ι	3	0	4	Ι	6	x5
	0	0	1	Ι	2	1	4	Ι	8	x6
	0	0	0	Ι	-4	-3	-5	I	0	z

The bottom row, the z-row, is meant to be read as:

 $z - 4^{*}x1 - 3^{*}x2 - 5^{*}x3 = 0$, (since x1 = x2 = x3 = 0).

The three constraint condition rows can be read off in a manner not too different from our previous work with tableau.

The first constraint row says:

 $x1 + 2^{*}x2 + 2^{*}x3 + x4 = 4$ (with x4 = 4 since x1 = x2 = x3 = 0), and so on.

If you put a dollar sign on the end instead of a semi-colon you do not see the output.

3.1.5 Mratio(colNum); to see ratios for the minimum ratio test

The coefficient of the current non-basic variable in the z-row which is most negative is normally used to define which of the non-basic variables will enter the Basis. Here column 3, the x3 column here, is used as the column number to use with Mratio(Ncol). The x3 variable will be the "pivot variable", and "x3 enters the Basis" with a positive value.

In more detail, the code for Mratio(colNum) has the line:

NC : col (invert (B) . N, colNum),

in which if colNum is 3, NC is the third column of the matrix invert(B) . N

```
(%i22) invert(B) . N;

(%o22) \begin{pmatrix} 1 & 2 & 2 \\ 3 & 0 & 4 \\ 2 & 1 & 4 \end{pmatrix}

(%i23) col (invert(B) . N, 3);

(%o23) \begin{pmatrix} 2 \\ 4 \\ 4 \end{pmatrix}
```

The individual elements of this column are then divided into the corresponding elements of invert(B). b:



Then an extra column with the elements of the current Xb are added on the right.



The "minimum ratio test" is won by the variable x5, which is the fifth variable in X (variable number 5).

3.1.6 Mbpivot (n_Enter, n_Leave); for one simplex step

In Mbpivot(n_Enter, n_Leave), n_Enter is the number of the variable entering the Basis and n_Leave is the number of the variable leaving the basis, based on the choice of the user. The simplex algorithm says make this choice based on the non-basic variable with the most negative coefficient in the z-row and the basic variable for which the ratio test produces the smallest value. The "variable number" depends on the order of the variable in the constant vector X.

Step 1 tableau with the third variable in X entering and the fifth variable in X leaving:

(%i27) Mbpivot (3, 5); x3 enters, x5 leaves Basis $\begin{bmatrix} x3 & x4 & x6 & | & x1 & x2 & x5 & | & rhs & Basis \\ 1 & 0 & 0 & | & \frac{3}{4} & 0 & \frac{1}{4} & | & \frac{3}{2} & x3 \\ 0 & 1 & 0 & | & -\frac{1}{2} & 2 & -\frac{1}{2} & | & 1 & x4 \\ 0 & 0 & 1 & | & -1 & 1 & -1 & | & 2 & x6 \\ 0 & 0 & 0 & | & -\frac{1}{4} & -3 & \frac{5}{4} & | & \frac{15}{2} & z \end{bmatrix}$

The first simplex pivot operation has caused z to increase from 0 to 15/2 = 7.5. The variable x3 has increased from 0 to 3/2 = 1.5.

The most negative coefficient among the non-basic z-row coefficients is in column 2 of the non-basic variable set, the x2 variable.

(%i28) Mratio(2); (%o28) $\begin{pmatrix} - & x3 \\ 0.5 & x4 \\ 2.0 & x6 \end{pmatrix}$

Since the most negative coefficient in the z-row corresponds to the second variable in the matrix vector X, and the minimum ratio test is won by the row corresponding to the fourth variable in X, we have n_Enter = 2, n_Leave = 4.

Step 2 tableau:

(%i29) Mbpivot (2, 4);

 The second simplex pivot operation has caused z to increase from 7.5 to 9.

At this stage, there is only one negative coefficient in the z-row, corresponding to the first column of the non-basic variables, and in that first column there is only one positive coefficient (3/4) corresponding to the variable x3, the third variable in X.

```
(%i30) Mratio(1);
```

 $(\%030) \begin{pmatrix} - & x2 \\ 2.0 & x3 \\ - & x6 \end{pmatrix}$

Step 3 tableau:

(%i31) Mbpivot (1,3);

In this matrix tableau, there are no negative coefficients in the z-row, z has reached the value 11, and the variables have the values:

x1 = 2, x2 = 1, x3 = 0, x4 = 0, x5 = 0, x6 = 3.

This is an optimum solution and agrees with the optimum found by maximize_lp or maxlp.

4 Example 2 : Matrix Maximization with Artificial Var. B/N Prob. 3.4

This is B/N Prob. 3.4. maximize $z = 2^*x1 + 3^*x2$, subject to $x1 + 2^*x2 \le 2$, $6^*x1 + 4^*x2 \ge 24$, with $x1,x2 \ge 0$. (%i32) maxlp (2*x1 + 3*x2, [x1 + 2*x2 <= 2, 6*x1 + 4*x2 >= 24]); (%o32) Problem not feasible!

4.1 Matrix Big M Method

Put this problem in standard form by introducing slack variable x3 in the first constraint and both a surplus variable x4 and an artificial variable x5 in the second constraint, to get maximize $z = 2^{*}x1 + 3^{*}x2 + 0^{*}x3 + 0^{*}x4 - M^{*}x5$ subject to $x1 + 2^{*}x2 + x3 = 2$, $6^{*}x1 + 4^{*}x2 - x4 + x5 = 24$, with x1,x2,x3,x4,x5 >= 0.

An initial feasible solution is x1 = x2 = x4 = 0, x3 = 2, x5 = 24. X^t = (x1,x2,x3,x4,x5), c^t = (2,3,0,0, - M),

```
(%i42) X : cvec ([x1,x2,x3,x4,x5])$
            c : cvec ([2,3,0,0,- M])$
            z = transpose (c) \cdot X;
            A : matrix ( [1, 2, 1, 0, 0], [6, 4, 0, -1, 1] )$
            b : cvec ([2, 24])$
            A \cdot X = b;
            NV : [1, 2, 4]$
            BV : [3, 5]$
            Mdefine()$
            Mbtableau();
(\%035) z=-M x5+3 x2+2 x1
                x3+2 x2+x1
             x5 - x4 + 4x2 + 6x1  (24)
             \begin{vmatrix} x^3 & x^5 & | & x^1 & x^2 & x^4 & | & rhs \\ 1 & 0 & | & 1 & 2 & 0 & | & 2 \\ 0 & 1 & | & 6 & 4 & -1 & | & 24 \\ 0 & 0 & | & -6 M - 2 & -4 M - 3 & M & | & -24 M \end{vmatrix} 
                                                                            Basis
                                                                             хЗ
(%042)
                                                                   24
                                                                             х5
                                                                              Ζ
```

```
(%i43) Mratio (1);
```

(%043) $\begin{pmatrix} 2.0 & x3 \\ 4.0 & x5 \end{pmatrix}$

(%i44)	Mbpivot (1, 3);										
	x1	ente	ers	, x3 le	eaves B	asis					
	x1	х5	Ι	x2	х3	x4	Ι	rhs	Basis		
(0/-4.4)	1	0	Ι	2	1	0	Ι	2	x1		
(%044)	0	1	Ι	-8	-6	-1	Ι	12	x5		
	0	0	Ι	8 <i>M</i> +1	6 <i>M</i> +2	М	Ι	4–12 <i>M</i>	Ζ		

There are no negative z-row coefficients of the nonbasic variables $x_{2,x_{3,x_{4}}}$, so optimality has been achieved but the artificial variable is still basic ($x_{5} = 12$). The original program has no solution since the constraint conditions and the nonnegativity conditions cannot be satisfied simultaneously. We have $x_{1} = 2$, $x_{2} = 0$, $x_{3} = 0$, $x_{4} = 0$, $x_{5} = 12$.

The first constraint condition reads

 $x1 + 2^{*}x2 + x3 = 2$, or since x2 = x3 = 0, x1 = 2.

The second constraint condition reads

 $x5 - 8^{*}x2 - 6^{*}x3 - x4 = 12$, or since x2 = x3 = x4 = 0, x5 = 12.

We need $x1 + 2^{*}x2 \le 2$, or $x1 \le 2$ (since x2 = 0), so the first condition is satisfied. We then need $6^{*}x1 + 4^{*}x2 \ge 24$, or since x2 = 0, we need $x1 \ge 24/6 = 4$. But x1 = 2 does not satisfy this second condition.

4.2 Matrix (Butenko) Dual Simplex Method

Step 1 (Phase I) is the Auxiliary problem of Butenko's dual simplex method: We minimize the sum of the artificial variables subject to the constraints, and with only one artificial variable, we minimize w = x5, or maximize z = -w = -x5.

Auxiliary Problem: maximize z = -x5, subject to $x1 + 2^{*}x2 + x3 = 2$, $6^{*}x1 + 4^{*}x2 - x4 + x5 = 24$, with x1,x2,x3,x4,x5 >= 0.

We don't need to redefine X, A, nor b.

```
(%i45) display (X,A,b)$

X = \begin{cases} x1 \\ x2 \\ x3 \\ x4 \\ x5 \end{cases}
A = \begin{pmatrix} 1 & 2 & 1 & 0 & 0 \\ 6 & 4 & 0 & -1 & 1 \end{pmatrix}
b = \begin{pmatrix} 2 \\ 24 \end{pmatrix}
```

The initial feasible solution has x1 = x2 = x4 = 0 (three non-basic variables to start) and x3 = 2, x5 = 24 (two basic variables).

```
(%i52) c: cvec ([0, 0, 0, 0, -1])$
           z = transpose (c) . X;
           A \cdot X = b;
           NV : [1, 2, 4]$
           BV:[3,5]$
            Mdefine()$
           Mbtableau();
(\%047) z = -x5
              x3+2 x2+x1
                                        2
(%048)
            x5-x4+4 x2+6 x1 24
             \begin{vmatrix} x3 & x5 & | & x1 & x2 & x4 & | & rhs \\ 1 & 0 & | & 1 & 2 & 0 & | & 2 \\ 0 & 1 & | & 6 & 4 & -1 & | & 24 \\ \end{vmatrix} 
                                                      Basis
                                                          хЗ
                                                          х5
             0
                                                -24
                           -6
                                                           Ζ
```

We need to drive z to a zero value, starting with z = -24.



(%054

(%i54) Mbpivot (1, 3);

	x1 enters, x3 leaves Basis										
	x1	х5	I	x2	хЗ	x4	I	rhs	Basis		
	1	0	I	2	1	0	I	2	x1		
)	0	1	I	-8	-6	- 1	I	12	x5		
	0	0	I	8	6	1	I	-12	z		

Since there are no negative coefficients of the current nonbasic variables $x_{2,x_{3,x_{4}}}$ in the bottom z-row, this is an optimum solution of the Auxiliary problem, and the maximum value of $z = -x_{5}$ is -12, the minimum value of x_{5} is 12 > 0. The original problem is not feasible.

5 Example 3: Matrix Minimization with Artificial Var.

minimize w = $80^{*}x1 + 60^{*}x2$ subject to $0.2^{*}x1 + 0.32^{*}x2 \le 0.25$, x1 + x2 = 1, $x1,x2 \ge 0$.

minimize_lp likes whole number fractions better than decimals.

(%i55) rat(0.32); (%o55)/R/ 8/25

(%i56) minimize_lp ($80^{x}1 + 60^{x}2$, [$x1/5 + 8^{x}2/25 \le 1/4$, x1 + x2 = 1], [x1, x2]);

(%056)
$$\left[\frac{215}{3}, \left[x^2 = \frac{5}{12}, x^1 = \frac{7}{12}\right]\right]$$

(%i57) float(%);

(%057) [71.667, [x2=0.41667, x1=0.58333]]

(%i58) maxlp (- 80*x1 - 60*x2, [x1/5 + 8*x2/25 <= 1/4, x1 + x2 = 1]); for z = -60 x2 - 80 x1, such that, $\frac{8 x^2}{25} + \frac{x1}{5} \le \frac{1}{4}$, x2 + x1 = 1, $z^* = -\frac{215}{3}$ with x1 = 7/12, x2 = 5/12, (%o58) $\left[-\frac{215}{3}, [x1 = \frac{7}{12}, x2 = \frac{5}{12}] \right]$

5.1 Convert to a Maximization Problem in Standard Form

1. Convert to a maximization problem with z = -w: maximize z = -80*x1 - 60*x2subject to $0.2*x1 + 0.32*x2 \le 0.25$, x1 + x2 = 1, $x1,x2 \ge 0$.

2. Write in standard form. Use slack variable x3 in the first constraint condition. We need an artificial variable x4 in the second constraint in order to have a second non-zero basic variable (x4 = 1) in the initial feasible solution (x1 = 0, x2 = 0, x3 = 1/4, x4 = 1), since we have two constraint conditions.

5.2 Matrix Big M Method

maximize z = -80*x1 - 60*x2 - M*x4, subject to 0.2*x1 + 0.32*x2 + x3 = 0.25, x1 + x2 + x4 = 1, $x1,x2,x3,x4 \ge 0$.

To be able to compare with the output of minimize_lp, we will use: 0.2 = 2/10 = 1/5, 0.32 = 8/25, 0.25 = 1/4 in the constraint equations.

(%i59) rat(0.32); (%o59)/R/ 8/25

The initial feasible solution is x1 = x2 = 0, x3 = 1/4, x4 = 1.

(%i69) X : cvec ([x1,x2,x3,x4])\$ c : cvec ([-80,-60,0, - M])\$ A : matrix ([1/5,8/25,1,0], [1,1,0,1])\$ b : cvec ([1/4,1])\$ NV : [1,2]\$ BV: [3, 4]\$ z = transpose(c) . X; A . X = b; Mdefine()\$ Mbtableau(); $(\% \circ 66)$ z = -M x4 - 60 x2 - 80 x1 $(\%067) \begin{pmatrix} x3 + \frac{8 x^2}{25} + \frac{x1}{5} \\ x4 + x2 + x1 \end{pmatrix} = \begin{pmatrix} 1 \\ 4 \\ 1 \end{pmatrix}$ $(\%069) \begin{pmatrix} x3 & x4 & | & x1 & x2 & | & rhs & Basis \\ 1 & 0 & | & \frac{1}{5} & \frac{8}{25} & | & \frac{1}{4} & x3 \\ 0 & 1 & | & 1 & 1 & | & 1 & x4 \\ 0 & 0 & | & 80 - M & 60 - M & | & -M & z \end{pmatrix}$

In the Big M method, M is an arbitrarily large positive number.

The coefficient of x2, (60 - M), is more negative than the coefficient of x1, (80 - M), in the z-row. We then use Mratio (colNum) (here the second column of the Xn matrix) to perform the minimum ratio test for the second column (which happens to be the x2 column):

```
(%i70) Mratio(2);
```

(%070) $\begin{pmatrix} 0.78125 & x3 \\ 1.0 & x4 \end{pmatrix}$

We then call Mbpivot (n Enter, n Leave) with n Enter = 2 (variable x2, the second variable in X, and with n Leave = 3 (variable x3, the third variable in X):

(%i71) Mbpivot(2, 3); x2 enters, x3 leaves Basis $(\%071) \begin{pmatrix} x2 & enters, & x3 & reaves Basis \\ x2 & x4 & | & x1 & x3 & | & rhs & Basis \\ 1 & 0 & | & \frac{5}{8} & \frac{25}{8} & | & \frac{25}{32} & x2 \\ 0 & 1 & | & \frac{3}{8} & -\frac{25}{8} & | & \frac{7}{32} & x4 \\ 0 & 0 & | & \frac{85}{2} - \frac{3M}{8} & \frac{25M}{8} - \frac{375}{2} & | & -\frac{7M}{32} - \frac{375}{8} & z \\ \end{pmatrix}$ (%i72) Mratio(1); $(\%072) \begin{pmatrix} 1.25 & x2 \\ 0.58333 & x4 \end{pmatrix}$ (%i73) Mbpivot (1, 4); x1 enters, x4 leaves Basis $(\%073) \begin{pmatrix} x1 & x2 & | & x3 & x4 & | & rhs & Basis \\ x1 & x2 & | & x3 & x4 & | & rhs & Basis \\ 1 & 0 & | & -\frac{25}{3} & \frac{8}{3} & | & \frac{7}{12} & x1 \\ 0 & 1 & | & \frac{25}{3} & -\frac{5}{3} & | & \frac{5}{12} & x2 \\ 0 & 0 & | & \frac{500}{3} & M - \frac{340}{3} & | & -\frac{215}{3} & z \\ \end{pmatrix}$

The large positive parameter M has disappeared the current value of z and from the values of the decision variables x1 and x2. The values of both the slack variable x3 and the artificial variable x4 are zero. This is the optimum solution since in the limit that M is an arbitrarily large positive number, there is no negative nonbasic variable coefficient in the bottom z row.

Since $z^* = -215/3$, the minimum value of w = -z is $w^* = 215/3$ with $x1^* = 7/12$, $x2^* = 5/12$ in agreement with minimize_lp.

(%i74) float(%);

 $(\%074) \begin{pmatrix} x1 & x2 & | & x3 & x4 & | & rhs & Basis \\ 1.0 & 0.0 & | & -8.3333 & 2.6667 & | & 0.58333 & x1 \\ 0.0 & 1.0 & | & 8.3333 & -1.6667 & | & 0.41667 & x2 \\ 0.0 & 0.0 & | & 166.67 & M-113.33 & | & -71.667 & z \end{pmatrix}$

5.3 Matrix (Butenko) Dual Simplex Method



$$X = \begin{pmatrix} x1\\ x2\\ x3\\ x4 \end{pmatrix}$$
$$A = \begin{pmatrix} \frac{1}{5} & \frac{8}{25} & 1 & 0\\ 1 & 1 & 0 & 1 \end{pmatrix}$$
$$b = \begin{pmatrix} \frac{1}{4}\\ 1 \end{pmatrix}$$

Phase 1 of Butenko's dual simplex method tests feasibility of given problem by seeking to minimize w = x4 (the single artificial variable in the problem), or maximize z = -w = -x4 subject to the two given constraint equations.

```
(\%i84) \ c: cvec ([0,0,0,-1]) \ NV: [1,2] \ BV: [3,4] \ BV: [3,4] \ "maximize"; 
z = transpose(c) . X; 
"subject to "; 
A . X = b; 
Mdefine() \ Mbtableau(); 
(%o79) maximize 
(%o80) z = -x4 
(%o81) subject to 
(%o82) <math display="block"> \begin{pmatrix} x3 + \frac{8x^2}{25} + \frac{x1}{5} \\ x4 + x2 + x1 \end{pmatrix} = \begin{pmatrix} 1 \\ 4 \\ 1 \end{pmatrix} 
(%o84)  \begin{pmatrix} x3 x4 | x1 x2 | rhs Basis \\ 1 & 0 | \frac{1}{5} \frac{8}{25} | \frac{1}{4} x3 \\ 0 & 1 | 1 & 1 | 1 x4 \\ 0 & 0 | -1 & -1 | -1 z \end{pmatrix}
```

We arbitrarily choose the first column of the nonbasic variable coefficients, corresponding to variable x1 (#1 in X) entering the Basis.

```
(%i85) Mratio (1);

(%o85) \begin{pmatrix} 1.25 & x3 \\ 1.0 & x4 \end{pmatrix}

(%i86) Mbpivot (1, 4);

x1 enters, x4 leaves Basis

\begin{pmatrix} x1 & x3 & | & x2 & x4 & | & rhs & Basis \\ 1 & 0 & | & 1 & 1 & | & 1 & x1 \\ 0 & 1 & | & \frac{3}{25} & -\frac{1}{5} & | & \frac{1}{20} & x3 \\ 0 & 0 & | & 0 & 1 & | & 0 & z \end{pmatrix}
```

With no more negative coefficients of the nonbasic variables in the bottom z-row, we have the optimum tableau with x4 nonbasic and hence x4 = 0. Hence the original LP is feasible.

Phase II of the dual simplex method.

From the final matrix tableau of Phase I of the dual simplex method, we eliminate the artificial variable (x4) column, work now with the three remaining variables x1, x2, x3, and rewrite the z-row in the original maximization form

 $\max z = -80^{*}x1 - 60^{*}x2 + 0^{*}x3$,

and continue using the condition equations governing x1,x2,x3 as they appear in the final tableau of Phase I.

Reading off the condition rows of the final Phase I matrix tableau (ignoring the x4 column): x1 + x2 = 1,

x3 + (3/25)*x2 = 1/20,

to define the matrices A and b. Because the basic variables in the final Phase I matrix tableau are x1 and x3, we set BV = [1,3]. The variable x2 is a nonbasic variable in the final Phase I matrix tableau, so we set NV = [2].

(%i98) X : cvec([x1,x2,x3])\$ c: cvec ([-80, -60, 0])\$ A : matrix ([1, 1, 0], [0, 3/25, 1])\$ b : cvec ([1, 1/20])\$ NV : [2]\$ BV: [1, 3]\$ "maximize"; $z = transpose(c) \cdot X;$ "subject to"; $A \cdot X = b;$ Mdefine()\$ Mbtableau(); (%093) maximize (%094) z=-60 x2-80 x1 (%095) subject to x2+x1 $(\%096) \left(x3 + \frac{3 x2}{25} \right) = \left(\frac{1}{20} \right)$ $(\%098) \begin{pmatrix} x1 & x3 & | & x2 & | & rhs & Basis \\ 1 & 0 & | & 1 & | & 1 & x1 \\ 0 & 1 & | & \frac{3}{25} & | & \frac{1}{20} & x3 \\ 0 & 0 & | & -20 & | & -80 & z \end{pmatrix}$

Mratio (Ncol) calculates the ratios of the rhs elements to the elements in column Ncol of the nonbasic variable coefficients. Here there is only one column, so we use Mratio (1). The coefficients in this column are the coefficients of x2 (variable #2 in X), so x2 will enter the basis.

```
(%i99) Mratio (1);
```

 $(\%099) \begin{pmatrix} 1.0 & x1 \\ 0.41667 & x3 \end{pmatrix}$

The minimum ratio test is won by the second constraint row, corresponding to x3 (variable #3 in X), so x3 leaves the basis. We then call Mbpivot (Enter, Leave) = Mbpivot (2, 3).

(%i100) Mbpivot (2, 3);

	х2	ente	ers,	х3	lea	aves Ba	SIS
	x1	x2	I	х3	Ι	rhs	Basis
	1	0	.	- <u>25</u> 3	Ι	7 12	x1
100)	0	1	I	25 3	Ι	5 12	x2
	0	0		500 3	Ι	$-\frac{215}{3}$	z

No more x3 negative coefficient in the bottom z-row implies a maximum of z = -w: $z^* = -215/3$ with $x1^* = 7/12$, $x2^* = 5/12$. This implies a minimum of w with $w^* = 215/3$ at the same point.

6 B/N Prob. 3.5

Maximize z = -x5subject to five conditions: $3^{*}x1 - 2^{*}x2 - 4^{*}x3 + 6^{*}x4 - x5 \le 0$ $-4^{*}x1 + 2^{*}x2 - x3 - 8^{*}x4 - x5 \le 0$ $-3^{*}x^{2}-2^{*}x^{3}-x^{4}-x^{5} <= 0$. x1 + x2 + x3 + x4<= 1, -x1 - x2 - x3 - x4 <= -1, with $x1,x2,x3,x4 \ge 0$, and x5 unrestricted. (%i101) maximize lp (- x5, [3*x1 - 2*x2 - 4*x3 + 6*x4 - x5 <= 0, -4*x1 + 2*x2 - x3 - 8*x4 - x5 <= 0. $-3^{*}x^{2} - 2^{*}x^{3} - x^{4} - x^{5} <= 0$ x1 + x2 + x3 + x4 <= 1,-x1 - x2 - x3 - x4 <= -1], -x1 - x2 - x3 - x4 [x1,x2,x3,x4]); (%0101) $\left[\frac{29}{15}, \left[x5 = -\frac{29}{15}, x4 = \frac{11}{60}, x3 = \frac{7}{10}, x2 = \frac{7}{60}, x1 = 0\right]\right]$

(%i102) float(%);

(%0102) [1.9333, [x5=-1.9333, x4=0.18333, x3=0.7, x2=0.11667, x1=0.0]]

Since x5 is unrestricted, we set x5 = x6 - x7, where both x6 and x7 are nonnegative; then all variables left are nonnegative. We multiply the last constraint by -1, thereby forcing a positive right-hand side. Finally, we achieve standard form by adding slack variables x8 through x11, respectively, to the left-hand sides of the first four constraints, and subtracting surplus variable x12 and adding artificial variable x13 to the left-hand side of the last constraint. We then have the standard form

Maximize z = -x6 + x7, subject to $3^{*}x1 - 2^{*}x2 - 4^{*}x3 + 6^{*}x4 - x6 + x7 + x8 = 0$, $-4^{*}x1 + 2^{*}x2 - x3 - 8^{*}x4 - x6 + x7 + x9 = 0$, $-3^{*}x2 - 2^{*}x3 - x4 - x6 + x7 + x10 = 0$, x1 + x2 + x3 + x4 + x11 = 1, x1 + x2 + x3 + x4 - x12 + x13 = 1, with x1, x2, x3, x4, x6, x7, x8, x9, x10, x11, x12, x13 >= 0, and x13 is an artificial variable.

An initial feasible solution is

x1,x2,x3,x4,x6,x7,x12 = 0, with five basis variables associated with the five constraints: x8 = 0, x9 = 0, x10 = 0, x11 = 1,x13 = 1.

6.1 Nonmatrix Big M "two phase" tableau pivot1 method

B/N's "two phase" Big M procedure:

The bottom "z-row" is split into two rows, with the second row being the coefficients of M. The simplex method is applied to that second "z-row", until this bottom row contains no negative elements. Then the simplex method is applied to those elements in the next-to-last row that are positioned over zeros in the last row

Whenever an artificial variable ceases to be basic, it is deleted from the top row of the tableau, as is the entire column under it. (This modification simplifies hand calculations but is not implemented in many computer programs.) The last row can be deleted from the tableau whenever it contains all zeros.

If nonzero artificial variables are present in the final basic set, then the program has no solution. Zero-valued artificial variables may appear as basic variables in the final solution when one or more of the original constraint equations is redundant.

With Xs^At = (x1,x2,x3,x4,x6,x7,x8,x9,x10,x11,x12,x13), x6 is the fifth element of Xs, and x13 is the twelfth element of Xs. Using the Big M method, we have the objective: Maximize $z = -x6 + x7 - M^*x13 = Cs^At$. Xs, subject to the conditions above, As . Xs = E. We have Cs^At = (0,0,0,0,-1,1,0,0,0,0,0,-M), E^At = (0,0,0,1,1), Xso^At = (x8,x9,x10,x11,x13), and Cso^At = (0,0,0,0,-M). [see below for definition of Xso and Cso]. Given the Step 0 LP: maximize $z = Cs^{t}$. Xs, such that As . Xs = E, with Xs >= 0,

Xso is the known initial feasible solution Basis vector of symbols.

Xso is defined using the Basis variable order in the constraint equations, and not necessarily with the order in Xs.

Xso has the same number of variables as the rhs vector E.

Cso is the vector of objective coefficients, taken from Cs, associated with the initial basis vector Xso, and in the same order as Xso.

With these conventions, the step 0 maximization tableau is, using these matrices:

	Xs^t		rhs	Basis
	As		E	Xso
(Cso^t . As - Cs^t		Cso⁄	`t.E∣z

With $X^t = (x_{1,x_{2,x_{3,x_{4,x_{6,x_{7,x_{8,x_{9,x_{10,x_{11,x_{12,x_{13}}}}}}, x_6} is the fifth element of X, and x_{13} is the twelfth element of X. Using the Big M method, we have the objective:$ $Maximize <math>z = -x_6 + x_7 - M^*x_{13}$, subject to the conditions above.

```
(%i113) Xs : cvec ([x1,x2,x3,x4,x6,x7,x8,x9,x10,x11,x12,x13])$
            Cs : cvec ([0,0,0,0,-1,1,0,0,0,0,0, - M])$
            " maximize z " = transpose (Cs) . Xs;
            Xso : cvec([x8,x9,x10,x11,x13])$
            Cso: cvec ([0,0,0,0,-M])$
            As : matrix ([3,-2,-4,6,-1,1,1,0,0,0,0,0],
                            [-4,2,-1,-8,-1,1,0,1,0,0,0,0],
                            [0, -3, -2, -1, -1, 1, 0, 0, 1, 0, 0, 0],
                            [1,1,1,1,0,0,0,0,0,1,0,0],
                            [1,1,1,1,0,0,0,0,0,0,-1,1])
            E : cvec ([0,0,0,1,1])$
            "subject to";
            As Xs = E;
            "Cso<sup>t</sup>. As - Cs<sup>t</sup>" = transpose(Cso). As - transpose(Cs);
            "Cso<sup>t</sup>. E = transpose(Cso). E;
(%0105) maximize z = x7 - x6 - M x13
(%o110) subject to
             x8+x7-x6+6 x4-4 x3-2 x2+3 x1
 (\%0111) \begin{bmatrix} x8+x7-x6+6x4-4x3-2x2+3x7\\x9+x7-x6-8x4-x3+2x2-4x1\\x7-x6-x4-2x3-3x2+x10\\x4+x3+x2+x11+x1\\x4+x3+x2+x13-x12+x1 \end{bmatrix} = \begin{bmatrix} 0\\0\\1\\1\\1 \end{bmatrix} 
 (\%0112) Cso^{t} \cdot As - Cs^{t} = \begin{pmatrix} -M & -M & -M & -M & 1 & -1 & 0 & 0 & 0 & M & 0 \end{pmatrix} 
(\% 0113) Cso<sup>t</sup>. E = -M
```

The last "z-row" is split into two rows (r6 and r7) according to B/N's "two phase method", page 33, see examples in LPduality.wxmx. With two "z-rows" on the bottom, we can use b2ratio(RL, ncol) to see the ratios needed in the minimum ratio test.

Tableau 1

```
(%i124) vL : [x1,x2,x3,x4,x6,x7,x8,x9,x10,x11,x12,x13,rhs,Basis]$
          bL : [x8,x9,x10,x11,x13,z1,z2]$
         r1 : endcons (E[1,1], As[1])$
         r2 : endcons (E[2,1], As[2])$
          r3 : endcons (E[3,1], As[3])$
          r4 : endcons (E[4,1], As[4])$
         r5 : endcons (E[5,1], As[5])$
          r6 : [0,0,0,0,1,-1,0,0,0,0,0,0,0]$
         r7 : [-1,-1,-1,-1,0,0,0,0,0,0,1,0,-1]$
          RL : [r1,r2,r3,r4,r5,r6,r7]$
          tableau(RL);
           x1 x2 x3 x4 x6 x7 x8 x9 x10 x11 x12 x13 rhs Basis
           3 -2 -4 6 -1 1 1
                                              0
                                                    0
                                                          0
                                                                0
                                                                      0
                                                                           0
                                                                                 x8
           -4 2 -1 -8 -1 1 0
                                             1
                                                    0
                                                          0
                                                                0
                                                                      0
                                                                           0
                                                                                 х9
           0 -3 -2 -1 -1 1 0 0 1 0 0 0
1 1 1 1 0 0 0 0 0 1 0 0 1
                                                                                 x10
                                                                                x11

      1
      1
      1
      0
      0
      0
      0
      0
      -1
      1
      1

      0
      0
      0
      0
      1
      -1
      0
      0
      0
      0
      0
      0
      0
      0

                                                                                x13
                                                                                 z1
                                0
                                          0
                                                          0
                                                                      0
                -1 -1 -1
                                     0
                                               0
                                                    0
                                                               1
                                                                                 z2
                                                                          -1
```

Choose the x1 column as the pivot column so x1 enters the Basis. With two "z-rows", we need to use b2ratio instead of bratio.

(%i125) b2ratio (RL, 1);

	0.0	x8	
	-	х9	
(%o125)	-	x10	
	1.0	x11	
	1.0	x13	

٢

The minimum ratio test is won by row 1, the x8 row, so x8 leaves the Basis.

Tableau 2 using pivot1 (RL, [row, col]):

(%i126) RL : pivot1 (RL, [1, 1])\$

pivo	pivot row = 1 pivot col = 1 value = 3												
x1	x1 enters Basis, x8 leaves Basis												
x1	x2	х3	х4	x6	х7	x8	x9	x10	x11	x12	x13	rhs	Basis
1	$-\frac{2}{3}$	$-\frac{4}{3}$	2	$-\frac{1}{3}$	$\frac{1}{3}$	<u>1</u> 3	0	0	0	0	0	0	x1
0	$-\frac{2}{3}$	$-\frac{19}{3}$	0	$-\frac{7}{3}$	$\frac{7}{3}$	$\frac{4}{3}$	1	0	0	0	0	0	x9
0	-3	-2	-1	-1	1	0	0	1	0	0	0	0	x10
0	5 3	$\frac{7}{3}$	-1	<u>1</u> 3	$-\frac{1}{3}$	$-\frac{1}{3}$	0	0	1	0	0	1	x11
0	5 3	7 3	-1	1 3	$-\frac{1}{3}$	$-\frac{1}{3}$	0	0	0	-1	1	1	x13
0	0	0	0	1	-1	0	0	0	0	0	0	0	z1
0	$-\frac{5}{3}$	$-\frac{7}{3}$	1	$-\frac{1}{3}$	$\frac{1}{3}$	<u>1</u> 3	0	0	0	1	0	-1	z2

(%i127) b2ratio (RL, 3);

(%0127) - x1 0.42857 x11 0.42857 x13

Choose row 4, corresponding to x11 leaving the Basis, as the pivot row.

Tableau 3 using pivot1 (RL, [row, col]):

(%i128) RL : pivot1 (RL, [4, 3])\$

pivot row = 4 pivot col = 3 value = $\frac{7}{3}$ x3 enters Basis, x11 leaves Basis x6 x7 x8 x9 x10 x11 x12 x13 rhs Basis x1 x2 х3 x4 $\frac{2}{7} \quad 0 \quad \frac{10}{7} \quad -\frac{1}{7} \quad \frac{1}{7} \quad \frac{1}{7} \quad 0 \quad 0 \quad \frac{4}{7}$ 0 0 1 х1 $0 \quad \frac{27}{7} \quad 0 \quad -\frac{19}{7} \quad -\frac{10}{7} \quad \frac{10}{7} \quad \frac{3}{7} \quad 1 \quad 0 \quad \frac{19}{7} \quad 0 \quad 0$ 19 7 <u>6</u> 7 х9 $0 \quad -\frac{11}{7} \quad 0 \quad -\frac{13}{7} \quad -\frac{5}{7} \quad \frac{5}{7} \quad -\frac{2}{7} \quad 0 \quad 1 \quad \frac{6}{7} \quad 0 \quad 0$ x10 0 0 0 0 0 0 0 0 0 0 1 1 0 z2

With no more negative coefficients in the bottom z-row, we look at coefficients in the upper z-row which are over the zero. The sixth column, the x7 column, is chosen as the pivot column. We still need to use b2ratio since we have two z-rows.

(%i129) b2ratio (RL, 6);

(%0129) (%012) (%012)

Tableau 4

(%i130) RL : pivot1 (RL, [3, 6])\$

(%i131) b2ratio (RL, 4);

(%0131) (%0

Tableau 5

(%i132) RL : pivot1 (RL, [1, 4])\$

(%i133) b2ratio (RL, 2);

(%0133) 0.66667 x4 0.11667 x9 - x7 1.1667 x3 - x13

Tableau 6

(%i134) RL : pivot1 (RL, [2, 2])\$

pivot row = 2 pivot col = 2 value = $\frac{20}{3}$ x2 enters Basis, x9 leaves Basis x9 x10 x11 x12 x13 rhs Basis x2 x3 x4 x6 x7 x8 x1 $\frac{1}{15}$ $-\frac{1}{20}$ $-\frac{1}{60}$ 11 60 1 0 0 0 х4 1 12 0 0 0 0 0 х2 0 0 -1 1 0 0 х7 7 10 х3 0 x13 29 15 z1 0 0 0 0 0 0 1 1 0 0 z2

This is the optimum tableau. $z^* = 29/15 = 1.93$, x1 = 0, x2 = 7/60 = 0.117, x3 = 7/10 = 0.7, x4 = 11/60 = 0.183, x5 = x6 - x7 = -x7 = -29/15 = -1.93, which is the same maximum as found by maxima_lp.

Note that the artificial variable x13, although still a basic variable, is equal to 0. In contrast, if nonzero artificial variables are present in the final basic set, then the LP has no solution.

6.2 Matrix Big M method

With $X^t = (x_{1,x_{2,x_{3,x_{4,x_{6,x_{7,x_{8,x_{9,x_{10,x_{11,x_{12,x_{13}}}}}}, x_6} is the fifth element of X, and so on, finally x_{13} is the twelfth element of X. Using the Big M method, we have the objective:$ $Maximize <math>z = -x_6 + x_7 - M^*x_{13} = c^t \cdot X$, subject to the conditions A $\cdot X = b$, with X >= 0.

An initial feasible solution is

x1,x2,x3,x4,x6,x7,x12 = 0, with five basis variables associated with the five constraints: x8 = 0, x9 = 0, x10 = 0, x11 = 1,x13 = 1. Note the definitions of NV and BV take into account the missing x5 variable in X.

```
(%i144) X : cvec ([x1,x2,x3,x4,x6,x7,x8,x9,x10,x11,x12,x13])$
       c : cvec ([0,0,0,0,-1,1,0,0,0,0,0, - M])$
       A : matrix ( [3,-2,-4,6,-1,1,1,0,0,0,0,0,0,],
                 [-4,2,-1,-8,-1,1,0,1,0,0,0,0],
                  [0, -3, -2, -1, -1, 1, 0, 0, 1, 0, 0, 0],
                  [1,1,1,1,0,0,0,0,0,1,0,0],
                  [1,1,1,1,0,0,0,0,0,0,-1,1])$
       b : cvec ([0,0,0,1,1])$
       NV : [1,2,3,4,5,6,11]$
       BV : [7,8,9,10,12]$
       z = transpose(c) . X;
       A \cdot X = b:
       Mdefine()$
       Mbtableau();
(\%0141) z = x7 - x6 - M x13
       x8+x7-x6+6 x4-4 x3-2 x2+3 x1
                                       0
        x9+x7-x6-8 x4-x3+2 x2-4 x1
                                       0
                                     = 0
(%0142)
         x7-x6-x4-2 x3-3 x2+x10
             x4+x3+x2+x11+x1
                                       1
                                       1
           x4 + x3 + x2 + x13 - x12 + x1
        x8 x9 x10 x11 x13 | x1
                                       х3
                                           x4 x6 x7 x12 | rhs Basis
                                   x2
           0
                0
                    0
                         0 | 3
                                   -2 -4
                                            6 -1 1
        1
                                                        0 | 0
                                                                    х8
        0
           1
                0
                    0 0 | -4
                                   2
                                       -1
                                           -8 -1 1 0 | 0
                                                                    х9
        0 0 1 0 0 0 0 0
0 0 0 1 0 1 1
                                   -3 -2 -1 -1 1 0 | 0
                                                                   x10
                                   1 1 1 0 0 0 1
                                                                   x11
                    0 1 | 1
                                        1
                                            1 0 0 -1 |
        0
            0
                0
                                    1
                                                             1
                                                                   x13
        0
                0
                    0
                         0
                            | -M -M -M -M
                                               1
                                                        М
            0
                                                              -M
                                                                    Ζ
```

Taking into account that in the Big M method, M is an arbitrarily large positive number, we choose x1, ncol = 1 (which looks only at the non-basic variables in the center section) to enter the Basis, and use Mratio (ncol) to show results of the minimum ratio test.

The second block of coefficients are the coefficients of the current non-basic variables in Xn.

(%i145) transpose (Xn); (%o145) (*x1 x2 x3 x4 x6 x7 x12*)

and the current matrix N holds these coefficients:

(%i146)N;

	3	-2	-4	6	-1	1	0	
	-4	2	-1	-8	-1	1	0	
(%o146)	0	-3	-2	-1	-1	1	0	
	1	1	1	1	0	0	0	
	1	1	1	1	0	0	-1	

so the column number (1) we are using refers to the first column of the matrix N. This is because B is the identity matrix (1's on the diagonal, 0's elsewhere), and its inverse is also the identity matrix, so invert(B). N = N. In the code for Mratio, the column chosen is that of the matrix invert(B). N.

(%i148)B;

We have chosen variable x1 (element 1 of X) to enter the Basis.

Choose x11 to leave the Basis (x11 = element 10 of X), then Mbpivot (nEnter, nLeave) is Mbpivot (1, 10).

(%i150) Mbpivot (1, 10);

x1 enters, x11 leaves Basis

(%0150)

x1	x8	x9	x10	x13	I	x2	х3	х4	x6	х7	x11	x12	Ι	rhs	Basis
1	0	0	0	0	I	1	1	1	0	0	1	0	I	1	x1
0	1	0	0	0	I	-5	-7	3	- 1	1	-3	0	I	-3	x8
0	0	1	0	0	I	6	3	-4	- 1	1	4	0	I	4	x9
0	0	0	1	0	I	-3	-2	-1	-1	1	0	0	Ι	0	x10
0	0	0	0	1	I	0	0	0	0	0	-1	-1	Ι	0	x13
0	0	0	0	0	Ι	0	0	0	1	-1	М	М	Ι	0	z

x7 (ncol = 5 of the new N) has the most negative coefficient in the bottom z-row. x7 (element 6 in X) will enter the Basis.

(%i151) Mratio(5);

	-	x1
	-3.0	x8
(%o151)	4.0	х9
	-	x10
	-	x13

The x8 row wins the minimum ratio test; x8 (element 7 in X) leaves the Basis, so Mbpivot (nEnter, nLeave) is Mbpivot (6, 7).

(%i152) Mbpivot (6, 7);

x7 enters, x8 leaves Basis

(%0152)

x1	х7	х9	x10	x13	Ι	x2	х3	х4	x6	x8	x11	x12	Ι	rhs	Basis
1	0	0	0	0	I	1	1	1	0	0	1	0	I	1	x1
0	1	0	0	0	Ι	-5	-7	3	-1	1	-3	0	Ι	-3	х7
0	0	1	0	0	Ι	11	10	-7	0	-1	7	0	Ι	7	x9
0	0	0	1	0	Ι	2	5	-4	0	-1	3	0	Ι	3	x10
0	0	0	0	1	Ι	0	0	0	0	0	-1	-1	Ι	0	x13
0	0	0	0	0	Ι	-5	-7	3	0	1	<i>M</i> -3	М	Ι	-3	z

x3 (ncol = 2, element 3 in X) has the most negative coefficient in the z-row. x3 is the third element in X.

(%i153)	Mrat	tio <mark>(2)</mark> ;	
	1.0	x1	
	-	x7	
(%o153)	0.7	x9	
	0.6	x10	
	-	x13	

The x10 row wins the minimum ratio test. x10 (element 9 in X) leaves the Basis:

(%i154) Mbpivot (3, 9);

x3 enters, x10 leaves Basis

(%0154)

x1	хЗ	х7	x9	x13	Ι	x2	x4	x6	x8	x10	x11	x12	I	rhs	Basis
1	0	0	0	0		<u>3</u> 5	<u>9</u> 5	0	<u>1</u> 5	- <u>1</u> 5	2 5	0	I	2 5	x1
0	1	0	0	0		2 5	$-\frac{4}{5}$	0	- <u>1</u> 5	<u>1</u> 5	<u>3</u> 5	0	I	3 5	x3
0	0	1	0	0		- <u>11</u> 5	- <u>13</u> 5	-1	$-\frac{2}{5}$	7 5	<u>6</u> 5	0	I	6 5	х7
0	0	0	1	0	I	7	1	0	1	-2	1	0	I	1	x9
0	0	0	0	1	Ι	0	0	0	0	0	-1	-1	I	0	x13
0	0	0	0	0	I	- <u>11</u> 5	- <u>13</u> 5	0	$-\frac{2}{5}$	7 5	$M + \frac{6}{5}$	М	I	6 5	z

(%i155) Mratio (2);

(,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	. · · · · ·	, ,
	0.22222	x1
	-	х3
(%o155)	-	x7
	1.0	x9
	_	x13

(%i156) Mbpivot (4, 1);

x4 enters, x1 leaves Basis

(%0156)

х3	x4	х7	x9	x13	Ι	x1	x2	x6	x8	x10	x11	x12	Ι	rhs	Basis
1	0	0	0	0	Ι	4 9	2 3	0	$-\frac{1}{9}$	<u>1</u> 9	7 9	0	I	7 9	х3
0	1	0	0	0	I	5 9	<u>1</u> 3	0	<u>1</u> 9	$-\frac{1}{9}$	<u>2</u> 9	0		2 9	x4
0	0	1	0	0	I	<u>13</u> 9	$-\frac{4}{3}$	-1	$-\frac{1}{9}$	<u>10</u> 9	<u>16</u> 9	0	I	<u>16</u> 9	x7
0	0	0	1	0	I	- <u>5</u> 9	<u>20</u> 3	0	<u>8</u> 9	- <u>17</u> 9	7 9	0	I	7 9	x9
0	0	0	0	1		0	0	0	0	0	-1	-1	I	0	x13
0	0	0	0	0	I	<u>13</u> 9	$-\frac{4}{3}$	0	- <u>1</u> 9	<u>10</u> 9	$M + \frac{16}{9}$	М	I	<u>16</u> 9	z

(%i157) Mratio (2);

1.1667	х3	
0.66667	х4	
-	х7	
0.11667	х9	
-	x13	
	1.1667 0.66667 - 0.11667 -	1.1667 x3 0.666667 x4 - x7 0.11667 x9 - x13

(%i158) Mbpivot (2, 8);

x2 enters, x9 leaves Basis

(%0158)

x2	х3	x4	х7	x13	Ι	x1	x6	x8	x9	x10	x11	x12	I	rhs	Basis
1	0	0	0	0	I	- <u>1</u> 12	0	2 15	3 20	$-\frac{17}{60}$	7 60	0	Ι	7 60	x2
0	1	0	0	0	I	<u>1</u> 2	0	$-\frac{1}{5}$	- <u>1</u> 10	3 10	7 10	0	I	7 10	х3
0	0	1	0	0	I	7 12	0	1 15	$-\frac{1}{20}$	- <u>1</u> 60	<u>11</u> 60	0	I	<u>11</u> 60	x4
0	0	0	1	0	I	$\frac{4}{3}$	-1	1 15	<u>1</u> 5	11 15	29 15	0	I	29 15	x7
0	0	0	0	1	Ι	0	0	0	0	0	-1	-1	Ι	0	x13
0	0	0	0	0	I	<u>4</u> 3	0	1 15	<u>1</u> 5	11 15	$M + \frac{29}{15}$	М	I	29 15	z

This is the optimum tableau. $z^* = 29/15 = 1.93$, x1 = 0, x2 = 7/60 = 0.117, x3 = 7/10 = 0.7, x4 = 11/60 = 0.183, x5 = x6 - x7 = -x7 = -29/15 = -1.93, which is the same maximum as found by maxima_lp.

Note that the artificial variable x13, although still a basic variable, is equal to 0. In contrast, if nonzero artificial variables are present in the final basic set, then the LP has no solution.

7 B/N Prob. 3.9 using the Big M Method

Minimize w = $2^{x}1 + x^{2} + 3^{x}3$, subject to x1 - $2^{x}x^{2} + x^{3} \ge 4$, $2^{x}x^{1} + x^{2} + x^{3} \le 8$, x1 - $x^{3} \ge 0$, with x1,x2,x3 >= 0.

(%i159) minimize_lp (2*x1 + x2 + 3*x3, [x1 - $2*x2 + x3 \ge 4$, 2*x1 + x2 + x3 <= 8, x1 - x3 ≥ 0], [x1,x2,x3]);

$$(\%0159)$$
 [8,[x3=0,x2=0,x1=4]]

(%i160) maxlp (- $2^{x}x1 - x2 - 3^{x}x3$, [x1 - $2^{x}x2 + x3 \ge 4$, $2^{x}x1 + x2 + x3 \le 8$, x1 - x3 ≥ 0]); for z = -3 x3 - x2 - 2 x1, such that, $x3 - 2 x2 + x1 \ge 4$, $x3 + x2 + 2 x1 \le 8$, $x1 - x3 \ge 0$, $z^{*} = -8$ with x1 = 4, x2 = 0, x3 = 0, (%o160) [-8,[x1=4,x2=0,x3=0]]

Subtract suplus variable x4 and add artificial variable x5 to left-hand side of the first condition. Add slack variable x6 to left-hand side of the the second condition, Subtract surplus variable x7 from left-hand side of cond. 3.

Maximize $z = -w = -2^{*}x1 - x2 - 3^{*}x3 + 0^{*}x4 - M^{*}x5 + 0^{*}x6 + 0^{*}x7 = c^{*}t \cdot X$ subject to A . X = b, X >=0, or x1 - 2^{*}x2 + x3 - x4 + x5 = 4, 2^{*}x1 + x2 + x3 + x6 = 8, x1 - x3 - x7 = 0, with x1,x2,x3,x4,x5,x6,x7 >= 0. Initial feasible solution is then x1 = x2 = x3 = x4 = x7 = 0, x5 = 4, x6 = 8, in a solution that does not add an artificial variable x8 to the lhs of cond. 3. The basic variables in the initial tableau are x5,x6,x7, with the last basic variable (x7) equal to 0.

```
(%i170) X : cvec ([x1,x2,x3,x4,x5,x6,x7])$
        c : cvec ([-2,-1,-3, 0, -M, 0, 0])$
        A : matrix ( [1, -2,1,-1,1,0,0 ], [2,1,1,0,0,1,0], [1,0,-1,0,0,0,-1])$
        b : cvec ([4,8,0])$
        NV : [1,2, 3, 4]$
        BV : [5, 6, 7]$
        z = transpose(c) \cdot X;
        A \cdot X = b;
        Mdefine()$
        Mbtableau();
(\%0167) z = -M x5 - 3 x3 - x2 - 2 x1
        x5-x4+x3-2 x2+x1
(\%0168) \qquad x6 + x3 + x2 + 2 x1 = 8
                         Jo
            -x7-x3+x1
        x5 x6 x7 | x1 x2 x3 x4 |
                                                      Basis
                                                rhs
(%o170) 1 0 0 1 1 -2 1
0 1 0 2 1 1
0 0 1 1 -1 0 1
                                         -1 |
                                                 4
                                                       х5
                                   1 0 |
                                                       x6
                                                 8
                                          0 |
                                                 0
                                                       х7
             0 0 | 2-M 2M+1 3-M M | -4M
         0
                                                       Ζ
```

With M a positive number, 2 - M is more negative than 3 - M, so choose x1 to enter the basis.

(%i171) Mratio(1); (%o171) $\begin{pmatrix} 4.0 & x5 \\ 4.0 & x6 \\ - & x7 \end{pmatrix}$

Choose x5 (the single artificial variable) to leave the basis and use Mbpivot(Enter, Leave).

(%i172) Mbpivot (1,5);

x1 enters, x5 leaves Basis

	x1	x6	х7	Ι	x2	х3	х4	х5	Ι	rhs	Basis
	1	0	0	I	-2	1	-1	1	I	4	x1
60172)	0	1	0	Ι	5	-1	2	-2	Ι	0	x6
	0	0	1	I	-2	2	-1	1	I	4	х7
	0	0	0	I	5	1	2	<i>M</i> -2	I	-8	z

The artificial variable has left the basis with value 0 (down from x5 = 4) and x1 has entered the basis with value x1 = x4. No more negative non-basic variable coefficients in the z-row, optimum soln for max z = - w is z* = -8 (w* = 8) with x1 = 4,x2 = x3 = 0, in agreement with the solution found by minimize_lp with w.

8 B/N Prob. 3.11 using the Big M Method

Minimize w = 4*x1 + 3*x2 + 2*x3 + 5*x4, subject to $x1 + 2*x2 + 3*x3 + x4 \ge 5$, $2*x1 - x2 + 5*x3 - x4 \ge 1$, $2*x1 + x2 + x3 + 3*x4 \ge 10$, with $x1,x2,x3,x4 \ge 0$.

(%i173) minimize_lp (4*x1 + 3*x2 + 2*x3 + 5*x4, [x1 + 2*x2 + 3*x3 + x4 >= 5, 2*x1 - x2 + 5*x3 - x4 >= 1, 2*x1 + x2 + x3 + 3*x4 >= 10], [x1,x2,x3,x4]);

(%0173)
$$\left[\frac{271}{16}, \left[x4 = \frac{49}{16}, x3 = \frac{13}{16}, x2 = 0, x1 = 0\right]\right]$$

(%i174) float(%); (%o174) [16.938, [x4=3.0625, x3=0.8125, x2=0.0, x1=0.0]] Convert to a maximization problem, subtract surplus variable x5 and add artificial variable x6 to the left-hand side of the first constraint to convert to an equality constraints, and proceed to so similar on the second and third constraints.

```
Maximize z = - w = - 4*x1 - 3*x2 - 2*x3 - 5*x4 + 0*x5 - M*x6 + 0*x7 - M*x8 + 0*x9 - M*x10
                             = c^t . X.
         subject to A \cdot X = b, X \ge 0, or
         x1 + 2^{*}x2 + 3^{*}x3 + x4 - x5 + x6 = 5,
       2^{x}1 - x^{2} + 5^{x}3 - x^{4} - x^{7} + x^{8} = 1,
       2^{*}x1 + x2 + x3 + 3^{*}x4 - x9 + x10 = 10,
         with x1,x2,x3,x4,x5,x6,x7,x8,x9,x10 >= 0.
(%j184) X : cvec ([x1,x2,x3,x4,x5,x6,x7,x8,x9,x10])$
        c : cvec ([-4,-3,-2, -5, 0, -M, 0, -M, 0, -M ])$
        A : matrix ([1, 2, 3, 1, -1, 1, 0, 0, 0, 0],
                   [2, -1, 5, -1, 0, 0, -1, 1, 0, 0],
                   [2, 1, 1, 3, 0, 0, 0, 0, -1, 1])$
        b : cvec ([5, 1, 10])$
        NV : [1,2, 3, 4, 5, 7, 9]$
        BV : [6, 8, 10 ]$
        z = transpose(c) \cdot X;
        A . X = b;
        Mdefine()$
        Mbtableau();
(\%0181) z=-M x8-M x6-5 x4-2 x3-3 x2-M x10-4 x1
         x6-x5+x4+3 x3+2 x2+x1 5
         x8 - x7 - x4 + 5x3 - x2 + 2x1 = 1
         -x9+3 x4+x3+x2+x10+2 x1
                                                                            Basis
        x6 x8 x10 |
                         х1
                                 x2
                                        х3
                                                x4
                                                      x5 x7 x9 |
                                                                      rhs
                 0
                         1
                                 2
                                         3
                                                      -1 0
                                                               0 |
                                                                             x6
             0
                                                1
                                                                       5
           1 0 | 2
                                -1
         0
                                         5
                                                -1
                                                      0 -1 0 |
                                                                       1
                                                                             х8
                         2
                                 1
                                                3
         0
             0
                 1
                                         1
                                                       0
                                                           0
                                                             -1 |
                                                                       10
                                                                             x10
                       4-5M 3-2M 2-9M 5-3M M M
         0
                                                                    -16 M
                                                             M |
                                                                              Ζ
```

The x3 column has the most negative coefficient in the z-row, ncol = 3, use Mratio(ncol):

(%i185) Mratio (3);

	1.6667	x6
(%0185)	0.2	x8
	10.0	x10

The x8 row wins the minimum ratio test, variable x8 leaves the basis. Use Mbpivot (Enter, Leave):

(%i186) Mbpivot (3, 8);

x3 enters, x8 leaves Basis

(%0186)

x3	x6	x10	Ι	x1	x2	x4	x5	х7	x8	x9	Ι	rhs	Basis
1	0	0	I	2 5	$-\frac{1}{5}$	$-\frac{1}{5}$	0	$-\frac{1}{5}$	<u>1</u> 5	0	I	<u>1</u> 5	х3
0	1	0		$-\frac{1}{5}$	<u>13</u> 5	<u>8</u> 5	-1	$\frac{3}{5}$	$-\frac{3}{5}$	0		<u>22</u> 5	x6
0	0	1	Ι	<u>8</u> 5	<u>6</u> 5	<u>16</u> 5	0	<u>1</u> 5	$-\frac{1}{5}$	-1	Ι	<u>49</u> 5	x10
0	0	0	Ι	$\frac{16}{5} - \frac{7 M}{5}$	$\frac{17}{5} - \frac{19 M}{5}$	$\frac{27}{5} - \frac{24}{5} \frac{M}{5}$	М	$\frac{2}{5} - \frac{4 M}{5}$	$\frac{9 M}{5} - \frac{2}{5}$	М	Ι	$-\frac{71 M}{5}-\frac{2}{5}$	z

The x4 variable (ncol = 3) has the most negative coefficient in the z-row.

(%i187) Mratio (3);

	-	xЗ
(%o187)	2.75	x6
	3.0625	x10

(%i188) Mbpivot (4, 6);

x4 enters, x6 leaves Basis

```
(%0188)
```

хЗ	8 x4	x10	Ι	x1	x2	x5	x6	х7	x8	х9	Ι	rhs	Basis
1	0	0	I	<u>3</u> 8	<u>1</u> 8	$-\frac{1}{8}$	<u>1</u> 8	$-\frac{1}{8}$	<u>1</u> 8	0	I	<u>3</u> 4	xЗ
0	1	0	Ι	$-\frac{1}{8}$	<u>13</u> 8	$-\frac{5}{8}$	<u>5</u> 8	$\frac{3}{8}$	$-\frac{3}{8}$	0	I	<u>11</u> 4	x4
0	0	1	Ι	2	-4	2	-2	-1	1	-1	Ι	1	x10
0	0	0	I	$\frac{31}{8}$ - 2 M	$4 M - \frac{43}{8}$	$\frac{27}{8}$ - 2 M	$3 M - \frac{27}{8}$	$M - \frac{13}{8}$	<u>13</u> 8	М	I	$-M-\frac{61}{4}$	Z

The x5 variable (ncol = 3) has the most negative coefficient in the z-row.

(%i189) Mratio (3); (%o189) $\begin{pmatrix} - & x3 \\ - & x4 \\ 0.5 & x10 \end{pmatrix}$

(%i190) Mbpivot (5, 10);

x5 enters, x10 leaves Basis

(%o190)

х3	х4	х5	I	x1	x2	x6	х7	x8	x9	x10	Ι	rhs	Basis
1	0	0	I	<u>1</u> 2	$-\frac{1}{8}$	0	- <u>3</u> 16	3 16	- <u>1</u> 16	1 16	I	<u>13</u> 16	х3
0	1	0	I	<u>1</u> 2	<u>3</u> 8	0	1 16	- <u>1</u> 16	- <u>5</u> 16	5 16		49 16	x4
0	0	1	I	1	-2	-1	$-\frac{1}{2}$	<u>1</u> 2	$-\frac{1}{2}$	<u>1</u> 2	I	<u>1</u> 2	x5
0	0	0	I	<u>1</u> 2	<u>11</u> 8	М	1 16	$M - \frac{1}{16}$	27 16	$M - \frac{27}{16}$		- <u>271</u> 16	z

In the limit of large positive M, no more negative non-basic variable coefficients in the z-row, optimum solution for w = - z is w* = 271/16 with x1 = x2 = 0, x3 = 13/16, x4 = 49/16, no artificial variables left in the basis, in agreement with solution found by minimize_lp.

9 B/N Dual Simplex Method, DMratio (rowNum)

On p. 34, B/N list the elements of their Dual Simplex Method. We quote, with some editing:

```
THE DUAL SIMPLEX METHOD
```

The (regular) simplex method moves the initial feasible but nonoptimal solution to an optimal solution while maintaining feasibility through an iterative procedure. On the other hand, the dual simplex method moves the initial optimal but infeasible solution to a feasible solution while maintaining optimality through an iterative procedure.

Iterative procedure of the B/N Dual Simplex Method for maximization.

STEP 1: Rewrite the linear programming problem by expressing all the constraints in \leq form and transforming them into equations through slack variables. If a particular given constraint has the form (for example)

a1*x1 + a2*x2 = d, replace that constraint by the two conditions a1*x1 + a2*x2 <= d, a1*x1 + a2*x2 >= d.

Then continue to replace >= conditions by <= conditions, getting

a1*x1 + a2*x2 <= d, - a1*x1 - a2*x2 <= - d.

STEP 2: Exhibit the above problem in the form of a simplex tableau. IF the optimality condition is satisfied AND one or more basic variables have negative values, the dual simplex method is applicable.

STEP 3: Feasibility Condition: The basic variable with the most negative value becomes the departing variable (D.V.) - leaves the basis. Call the row in which this value appears the work row. If more than one candidate for D.V. exists, choose one.

STEP 4: Optimality Condition: Form ratios by dividing the z-row coefficients of the non-basic variables by the corresponding negative coefficients of the work row. The nonbasic variable with the smallest absolute ratio becomes the entering variable (E.V.) - enters the basis. Designate this element in the work row as the pivot element and the corresponding column the work column. If more than one candidate for E.V. exists, choose one. If no element in the work row is negative, the problem has no feasible solution.

STEP 5: Use elementary row operations to convert the pivot element to 1 and then to reduce all the other elements in the work column to zero.

STEP 6: Repeat steps 3 through 5 until all the variables have reached nonnegative values, so the optimal solution is feasible.

9.1 B/N Prob. 3.9 using B/N Dual Simplex Method

We worked B/N Prob. 3.9 above using the Big M Method.

```
Minimize w = 2^{x}1 + x^{2} + 3^{x}3,
subject to
x1 - 2^{x}x^{2} + x^{3} \ge 4,
2^{x}x^{1} + x^{2} + x^{3} \le 8,
x1 - x^{3} \ge 0,
with x1,x2,x3 >= 0.
```

```
(%i191) minimize_lp (2^{x}x1 + x2 + 3^{x}x3, [x1 - 2^{x}x2 + x3 \ge 4, 2^{x}x1 + x2 + x3 \le 8,
 x1 - x3 \ge 0], [x1,x2,x3]);
```

$$(\%0191)$$
 [8,[x3=0,x2=0,x1=4]]

Step 1a: Convert to a maximization problem, and convert >= inequality conditions to <= conditions

```
Maximize z = -w = -2^{*}x1 - x2 - 3^{*}x3,
subject to
-x1 + 2^{*}x2 - x3 <= -4,
2^{*}x1 + x2 + x3 <= 8,
-x1 + x3 <= 0,
with x1,x2,x3 >= 0.
```

Step1b:

Add slack variables x4,x5,x6 to the left-hand sides of the three constraint conditions (respectively) to arrive at constraint equations.

Maximize $z = -w = -2^{*}x1 - x2 - 3^{*}x3 + 0^{*}x4 + 0^{*}x5 + 0^{*}x6 = c^{t} \cdot X$, subject to A . X = b, X >= 0, or: $-x1 + 2^{*}x2 - x3 + x4 = -4$, $2^{*}x1 + x2 + x3 + x5 = 8$, -x1 + x3 + x6 = 0, with x1,x2,x3,x4,x5,x6 >= 0.

Take x4,x5,x6 as the basic variables, x1,x2,x3 as the non-basic variables in the initial tableau.

(%i201)	X :	cve	c <mark>(</mark>)	(1,)	x2,x	3,x4	,x5,	x6])\$					
	c : c	cvec	; <mark>([-</mark> 2	2,-	1,-3	, 0, (0,0])\$	5					
	A :	mat	rix ([-	1, +2	2, -1	,1, (), (0],[2,1,1,	0,1, 0], [-1, 0,1, 0, 0, 1])\$			
	b:(cved) ([-	4,	8, 0)])\$								
	NV	:[1,	2,	3]	\$									
	ΒV	: [4,	5,	6]	\$									
	z =	tran	spc	se	e(c)	X ;								
	Α.	X =	b;											
	Mde	efine	∋() \$											
Mbtableau();														
(%o198)	z=-	-3 x	3-	x2	-22	x1								
	(x4 -	-x3+	+2 x	2 –	x1	[-4								
(%0199)	x5 -	+x3+	+x2·	+2	x1 =	= 8								
	ļ	xo+.	x3-	XI	J	ĺ	J			,				
	x4	х5	x6	Ι	x1	x2	х3	Ι	rhs	Basis				
	1	0	0	Ι	- 1	2	-1	Ι	-4	x4				
(%o201)	0	1	0	Ι	2	1	1	Ι	8	x5				
	0	0	1	Ι	-1	0	1	Ι	0	x6				
	0	0	0	Ι	2	1	3	Ι	0	z				

The B/N dual simplex method is applicable since the optimality condition is satisfied (no z-row negative coefficients of the non-basic variables) AND one or more of the basic variables have negative values.

The "work row" is row 1, x4 is the "departing value" (D.V.) - leaves the basis, since only basic variable x4 has a negative value. To determing the "entering variable" (E.V.) which enters the basis, we first check that there is at least one negative coefficient in the work row coefficients of the non-basic variables x1,x2,x3. (If there are none, the LP is NOT feasible). Here there are two negative coefficients corresponding to x1 and x3. We can make a table.

x1 x2 x3 2 1 3 z-r

2 1 3 | z -row -1 2 -1 | x4 - work row

-2 -- -3 | z-row/work-row

2 -- 3 | abs(zr/wr)

The column with the minimum absolute ratio is the x1 column, so x1 is the E.V., the "entering variable".

We can use our function DMratio (rowNum) in which rowNum is the number of the work row (top row = 1, next lower row = 2, etc.) and a two rowed matrix is returned with the non-basic variable names as the first row and the absolute values of the needed ratios as the second row.

```
(%i202) DMratio (1);
```

 $(\%0202) \begin{pmatrix} x1 & x2 & x3 \\ 2.0 & - & 3.0 \end{pmatrix}$

We then call Mbpivot (nEnter, nLeave), in which nEnter refers to the number of the entering variable according to its place in the vector X, and likewise nLeave refers to the number of the leaving variable according to its place in the vector X.

(%i203) Mbpivot (1, 4);

```
x1 enters, x4 leaves Basis
```

	x1	х5	x6	I	x2	х3	х4	I	rhs	Basis
	1	0	0	I	-2	1	-1	Ι	4	x1
(%o203)	0	1	0	Ι	5	-1	2	Ι	0	x5
	0	0	1	I	-2	2	-1	Ι	4	x6
	0	0	0	I	5	1	2	Ι	-8	z

Since all the variables have reached nonnegative values, the above optimal solution is feasible. $z^* = -w^* = -8$ with x1 = 4, x2 = x3 = 0 in agreement with our previous optimum solutions for this problem.

9.2 B/N Prob. 3.10 using B/N Dual Simplex Method

```
Use the Dual simplex method to solve:
maximize z = -2^*x1 - 3^*x2,
subject to
x1 + x2 \ge 2,
2^*x1 + x2 \le 10,
x1 + x2 \le 8,
with x1,x2 \ge 0,
```

 $(\%i204) \max \left[(-2^*x1 - 3^*x2, [x1 + x2 >= 2, 2^*x1 + x2 <= 10, x1 + x2 <= 8] \right];$ for z = -3 x2 - 2 x1, such that, $x2 + x1 \ge 2$, $x2 + 2 x1 \le 10$, $x2 + x1 \le 8$, $z^* = -4$ with x1 = 2, x2 = 0, (%o204) [-4, [x1 = 2, x2 = 0]]

Convert first constraint to \leq form and add slack variables x3,x4,x5 to get equations.

```
maximize z = -2^{*}x1 - 3^{*}x2 + 0^{*}x3 + 0^{*}x4 + 0^{*}x5,
              subject to
              -x1 - x2 + x3 = -2,
             2^{*}x1 + x2 + x4 = 10.
              x1 + x2 + x5 = 8,
           with x_{1,x_{2,x_{3,x_{4,x_{5}}}} = 0.
(%i214) X : cvec ([x1,x2,x3,x4,x5])$
            c: cvec ([-2,-3, 0, 0, 0])$
            A : matrix ( [-1, -1, 1,0, 0 ], [ 2,1,0, 1,0 ], [1,1,0, 0, 1])$
            b : cvec ([-2, 10, 8])$
            NV : [1, 2]$
            BV : [3, 4, 5]$
            z = transpose(c) \cdot X;
            A \cdot X = b;
            Mdefine()$
            Mbtableau();
(\% 0211) z = -3 x 2 - 2 x 1
\binom{\%02111}{2} \begin{pmatrix} x3 - x2 - x1 \\ x4 + x2 + 2 x1 \\ x5 + x2 + x1 \end{pmatrix} = \begin{pmatrix} -2 \\ 10 \\ 8 \end{pmatrix}
```

 $(\%o214) \begin{bmatrix} x & 3 & x & x & x & 5 & | & x & 1 & x & 2 & | & rhs & Basis \\ 1 & 0 & 0 & | & -1 & -1 & | & -2 & x & 3 \\ 0 & 1 & 0 & | & 2 & 1 & | & 10 & x & 4 \\ 0 & 0 & 1 & | & 1 & 1 & | & 8 & x & 5 \\ 0 & 0 & 0 & | & 2 & 3 & | & 0 & z \end{bmatrix}$

(%i215) DMratio (1);

(%0215) $\begin{pmatrix} x1 & x2 \\ 2.0 & 3.0 \end{pmatrix}$

	x1	ente	ərs,	Х	x3 leaves Basis						
	x1	x4	x5	I	x2	хЗ	I	rhs	Basis		
	1	0	0	Ι	1	- 1	I	2	x1		
(%o216)	0	1	0	Ι	-1	2	I	6	x4		
	0	0	1	Ι	0	1	I	6	x5		
	0	0	0	Ι	1	2	Ι	-4	z		

Since all the variables have reached nonnegative values, the above optimal solution is feasible, with $z^* = -4$, x1 = 2, x2 = 0.

9.3 B/N Prob. 3.11 using B/N Dual Simplex Method

Minimize w = 4*x1 + 3*x2 + 2*x3 + 5*x4, subject to $x1 + 2*x2 + 3*x3 + x4 \ge 5$, $2*x1 - x2 + 5*x3 - x4 \ge 1$, $2*x1 + x2 + x3 + 3*x4 \ge 10$, with $x1,x2,x3,x4 \ge 0$.

Convert to a maximization problem and convert >='s to <='s. Then add slack variables x5,x6,x7 to left-hand sides.

Maximize $z = -w = -4^{*}x1 - 3^{*}x2 - 2^{*}x3 - 5^{*}x4 + 0^{*}x5 + 0^{*}x6 + 0^{*}x7 = c^{t} \cdot X$, subject to A . X = b, X >= 0, or $-x1 - 2^{*}x2 - 3^{*}x3 - x4 + x5 = -5$, $-2^{*}x1 + x2 - 5^{*}x3 + x4 + x6 = -1$, $-2^{*}x1 - x2 - x3 - 3^{*}x4 + x7 = -10$, with x1,x2,x3,x4,x5,x6,x7 >= 0.

The optimality condition is satisfied and one or more basic variables are negative, so the dual simplex method is applicable.

The most negative basic variable is x7 = -10 so the work row is row 3 which contains four negative coefficients.

```
(%i227) DMratio (3);
```

 $(\%0227) \begin{pmatrix} x1 & x2 & x3 & x4 \\ 2.0 & 3.0 & 2.0 & 1.6667 \end{pmatrix}$

The minimum absolute value of the needed ratios is in column 4, corresponding to x4 so x4 is the entering variable and x7 is the leaving variable.

(%i228) Mbpivot (4, 7), numer;

x4 enters, x7 leaves Basis

x4	x5	x6	I	x1	x2	х3	x7	I	rhs	Basis
1	0	0	I	0.66667	0.33333	0.33333	-0.33333	Ι	3.3333	x4
0	1	0	I	-0.33333	-1.6667	-2.6667	-0.33333	Ι	-1.6667	x5
0	0	1	I	-2.6667	0.66667	-5.3333	0.33333	Ι	-4.3333	x6
0	0	0	I	0.66667	1.3333	0.33333	1.6667	I	-16.667	z

The most negative basic variable is x6 in row 3 which becomes the work row, x6 leaves the basis. The work row has two negative non-basic variable coefficients.

```
(%i229) DMratio (3);
```

 $(\%0229) \begin{pmatrix} x1 & x2 & x3 & x7 \\ 0.25 & - & 0.0625 & - \end{pmatrix}$

The non-basic variable with the smallest absolute ratio is x3 with becomes the entering variable.

(%i230) Mbpivot (3, 6);

^{%1230})	IVID	pivo	τ (3,	6);								
	х3	ente	ers,	Х	x6 leaves Basis								
	х3	x4	х5	I	x1	x2	x6	х7	Ι	rhs	Basis		
	1	0	0	I	<u>1</u> 2	$-\frac{1}{8}$	- <u>3</u> 16	<u>- 1</u> 16	I	<u>13</u> 16	х3		
%o230)	0	1	0	I	<u>1</u> 2	<u>3</u> 8	1 16	- <u>5</u> 16	I	49 16	x4		
	0	0	1	I	1	-2	$-\frac{1}{2}$	$-\frac{1}{2}$	I	<u>1</u> 2	x5		
	0	0	0	I	<u>1</u> 2	<u>11</u> 8	1 16	27 16	I	- <u>271</u> 16	z		

(%i231) float(%);

	х3	х4	х5	Ι	x1	x2	x6	х7	I	rhs	Basis
	1.0	0.0	0.0	I	0.5	-0.125	-0.1875	-0.0625	I	0.8125	х3
(%o231)	0.0	1.0	0.0	I	0.5	0.375	0.0625	-0.3125	Ι	3.0625	x4
	0.0	0.0	1.0	I	1.0	-2.0	-0.5	-0.5	I	0.5	x5
	0.0	0.0	0.0	Ι	0.5	1.375	0.0625	1.6875	Ι	- 16.938	z

Since all the variables have reached nonnegative values, the above optimal solution is feasible, and we have the same solution as found earlier.

The B/N dual simplex method requires 2 pivots as compared with 3 pivots using the Big M method above.

9.4 B/N Prob. 3.13 using B/N Dual Simplex Method

Use the dual simplex method to solve minimize $w = 2^{*}x1 + x2$, subject to x1 + x2 = 4, $2^{*}x1 - x2 \ge 3$, with $x1,x2 \ge 0$.

The above problem is rewritten as follows:

minimize w = $2^{*}x1 + x2$, subject to x1 + x2 <= 4, x1 + x2 >= 4, $2^{*}x1 - x2 >= 3$, with x1,x2 >= 0.

We then convert to a maximization problem and convert >= conditions to <= conditions. The above problem is rewritten as follows:

maximize $z = -w = -2^{*}x1 - x2$, subject to x1 + x2 <= 4, -x1 - x2 <= -4, $-2^{*}x1 + x2 <= -3$, with x1,x2 >= 0.

Convert inequality conditions to equality conditions using slack variables x3,x4,x5. maximize $z = -w = -2^{*}x1 - x2 + 0^{*}x3 + 0^{*}x4 + 0^{*}x5 = c^{*}t$. X, subject to A . X = b, X >= 0, or x1 + x2 + x3 = 4, -x1 - x2 + x4 = -4, $-2^{*}x1 + x2 + x5 = -3$,

with x1,x2,x3,x4,x5 >= 0.

(%i241) X : cvec ([x1,x2,x3,x4,x5])\$ c: cvec ([-2,-1,0, 0, 0])\$ A : matrix ([1, 1, 1, 0, 0], [-1,-1,0, 1,0], [-2, 1,0, 0, 1])\$ b : cvec ([4, -4, -3])\$ NV : [1, 2]\$ BV: [3,4,5]\$ $z = transpose(c) \cdot X;$ A . X = b; Mdefine()\$ Mbtableau(); (% 0238) z = -x2 - 2x1x3+x2+x1 (%0239) | x4-x2-x1 | = | -4x5+x2-2 x1 x3 x4 x5 | x1 x2 | rhs Basis 1 0 0 | 1 1 | 4 х3 $(\%0241) \begin{vmatrix} 0 & 1 & 0 & | & -1 & -1 & | & -4 & x4 \\ 0 & 0 & 1 & | & -2 & 1 & | & -3 & x5 \end{vmatrix}$ 0 0 | 2 1 | 0 0 z

The optimality condition is satisfied and at least one of the basic variables is negative, so the dual simplex method is applicable. The basic variable with the most negative value becomes the departing variable (D.V.) - leaves the basis; that is x4 in row 2.

```
(%i242) DMratio (2);
```

```
(\%0242) \begin{pmatrix} x1 & x2 \\ 2.0 & 1.0 \end{pmatrix}
```

The nonbasic variable with the smallest absolute ratio becomes the entering variable (E.V.) - enters the basis; that is x2. Use Mbpivot (nEnter, nLeave).

```
(%i243) Mbpivot (2, 4);
```

The optimality condition is satisfied and one basic variable is negative (x5 in row 3).

(%i244) DMratio (3);
(%o244)
$$\begin{pmatrix} x1 & x4 \\ 0.33333 & - \end{pmatrix}$$

The non-basic variable with the smallest absolute ratio (x1) is the entering variable.

(%i245) Mbpivot (1, 5);

	x1	ente	ers,	Х	5 lea				
	x1	x2	х3	Ι	x4	x5	Ι	rhs	Basis
	1	0	0	I	$-\frac{1}{3}$	$-\frac{1}{3}$	I	$\frac{7}{3}$	x1
(%o245)	0	1	0	I	$-\frac{2}{3}$	$\frac{1}{3}$	I	5 3	x2
	0	0	1	Ι	1	0	Ι	0	х3
	0	0	0	I	$\frac{4}{3}$	$\frac{1}{3}$	I	$-\frac{19}{3}$	z

(%i246) float(%);

	x1	x2	хЗ	Ι	x4	x5	I	rhs	Basis
	1.0	0.0	0.0	I	-0.33333	-0.33333	Ι	2.3333	x1
(%o246)	0.0	1.0	0.0	I	-0.66667	0.33333	Ι	1.6667	x2
	0.0	0.0	1.0	I	1.0	0.0	I	0.0	х3
	0.0	0.0	0.0	I	1.3333	0.33333	Ι	-6.3333	z

Since all the variables have reached nonnegative values, the above optimal solution is feasible. $w^* = -z^* = 19/3$ with x1 = 7/3 = 2.33, x2 = 5/3 = 1.67.

9.5 B/N Prob. 3.14 using B/N Dual Simplex Method

minimize w = $6^{*}x1 + 3^{*}x2 + 4^{*}x3$, subject to x1 + $6^{*}x2 + x3 = 10$, $2^{*}x1 + 3^{*}x2 + x3 = 15$, with x1,x2,x3 >= 0.

```
maximize z = -w = -6^{*}x1 - 3^{*}x2 - 4^{*}x3,
subject to
x1 + 6^{*}x2 + x3 \le 10,
x1 + 6^{*}x2 + x3 \ge 10,
 2^{*}x1 + 3^{*}x2 + x3 \le 15
 2^{x}1 + 3^{x}2 + x^{3} \ge 15,
with x_{1,x_{2,x_{3}}} \ge 0.
maximize z = -w = -6^{*}x1 - 3^{*}x2 - 4^{*}x3,
subject to
x1 + 6^{*}x2 + x3 \le 10,
-x1 - 6^{*}x2 - x3 \le -10
 2^{*}x1 + 3^{*}x2 + x3 \le 15,
-2^{*}x1 - 3^{*}x2 - x3 <= -15
with x_{1,x_{2,x_{3}}} \ge 0.
maximize z = -w = -6^{*}x1 - 3^{*}x2 - 4^{*}x3 + 0^{*}x4 + 0^{*}x5 + 0^{*}x6 + 0^{*}x7 = c^{t}. X,
subject to A . X = b, X \ge 0, or
x1 + 6^{*}x2 + x3 + x4 = 10,
-x1 - 6^{*}x2 - x3 + x5 = -10,
 2^{*}x1 + 3^{*}x2 + x3 + x6 = 15,
-2^{*}x1 - 3^{*}x2 - x3 + x7 = -15,
```

with $x1, x2, x3, x4, x5, x6, x7 \ge 0$.

0

0

0 0

0 0

1 | 0

1

0

0

3

4

1 | 0

1

| -15

х6

Ζ

(%i256) X : cvec ([x1,x2,x3,x4,x5,x6,x7])\$ c: cvec ([-6,-3,-4, 0, 0, 0, 0])\$ A : matrix ([1, 6, 1, 1, 0, 0, 0], [-1, -6, -1, 0, 1, 0, 0], [2, 3, 1, 0, 0, 1, 0], [-2, -3, -1, 0, 0, 0, 1])\$ b : cvec ([10, -10, 15, -15])\$ NV : [1, 2, 3]\$ BV : [4, 5, 6, 7]\$ z = transpose(c) . X; $A \cdot X = b;$ Mdefine()\$ Mbtableau(); (% 0253) z = -4 x3 - 3 x2 - 6 x1x4+x3+6 x2+x1 10 x5-x3-6 x2-x1 -10 (%o254) x6+x3+3 x2+2 x1 15 -15 x7-x3-3 x2-2 x1 x4 x5 x6 x7 | x1 x2 x3 | rhs Basis 0 0 0 | 1 1 6 1 10 х4 1 0 0 | -1 -6 -1 | -10 0 1 0 | 2 3 1 | 15 0 х5 0 x6 0 0 1 | -2 -3 -1 | 0 -15 х7 0 0 0 0 | 6 3 4 0 Ζ (%i257) DMratio(4); x1 x2 x3 3.0 1.0 4.0 (%i258) Mbpivot (2, 7); x2 enters, x7 leaves Basis x2 x4 x5 x6 | x1 x3 x7 | rhs Basis 2 3 1 3 $-\frac{1}{3}$ 0 0 | 5 x2 1 0 1 0 0 | -3 -1 0 2 | -20 x4 0 1 0 | 3 1 -2 | 0 20 х5

(%i259) DMratio (2);

(%i260) Mbpivot (1, 4);

x1 enters, x4 leaves Basis

			-,								
	x1	x2	x5	x6	I	хЗ	x4	х7	I	rhs	Basis
	1	0	0	0	I	$\frac{1}{3}$	$-\frac{1}{3}$	$-\frac{2}{3}$	I	<u>20</u> <u>3</u>	x1
(%o260)	0	1	0	0	I	<u>1</u> 9	2 9	<u>1</u> 9		<u>5</u> 9	x2
	0	0	1	0	I	0	1	0	I	0	x5
	0	0	0	1	Ι	0	0	1	I	0	x6
	0	0	0	0	I	5 3	4/3	<u>11</u> 3	I	_ <u>125</u> 3	z

Since all the variables have reached nonnegative values, the above optimal solution is feasible. $w^* = -z^* = 125/3$ with x1 = 20/3, x2 = 5/9, x3 = 0.