## **CHAPTER 11**

**11.1** First, compute the LU decomposition The matrix to be evaluated is

$$\begin{bmatrix} 10 & 2 & -1 \\ -3 & -6 & 2 \\ 1 & 1 & 5 \end{bmatrix}$$

Multiply the first row by  $f_{21} = -3/10 = -0.3$  and subtract the result from the second row to eliminate the  $a_{21}$  term. Then, multiply the first row by  $f_{31} = 1/10 = 0.1$  and subtract the result from the third row to eliminate the  $a_{31}$  term. The result is

$$\begin{bmatrix} 10 & 2 & -1 \\ 0 & -5.4 & 1.7 \\ 0 & 0.8 & 5.1 \end{bmatrix}$$

Multiply the second row by  $f_{32} = 0.8/(-5.4) = -0.148148$  and subtract the result from the third row to eliminate the  $a_{32}$  term.

$$\begin{bmatrix} 10 & 2 & -1 \\ 0 & -5.4 & 1.7 \\ 0 & 0 & 5.351852 \end{bmatrix}$$

Therefore, the LU decomposition is

$$[L]{U} = \begin{bmatrix} 1 & 0 & 0 \\ -0.3 & 1 & 0 \\ 0.1 & -0.148148 & 1 \end{bmatrix} \begin{bmatrix} 10 & 2 & -1 \\ 0 & -5.4 & 1.7 \\ 0 & 0 & 5.351852 \end{bmatrix}$$

The first column of the matrix inverse can be determined by performing the forward-substitution solution procedure with a unit vector (with 1 in the first row) as the right-hand-side vector. Thus, the lower-triangular system, can be set up as,

$$\begin{bmatrix} 1 & 0 & 0 \\ -0.3 & 1 & 0 \\ 0.1 & -0.148148 & 1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

and solved with forward substitution for  $\{d\}^T = \begin{bmatrix} 1 & 0.3 & -0.055556 \end{bmatrix}$ . This vector can then be used as the right-hand side of the upper triangular system,

$$\begin{bmatrix} 10 & 2 & -1 \\ 0 & -5.4 & 1.7 \\ 0 & 0 & 5.351852 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0.3 \\ -0.055556 \end{bmatrix}$$

which can be solved by back substitution for the first column of the matrix inverse,

$$[A]^{-1} = \begin{bmatrix} 0.110727 & 0 & 0 \\ -0.058824 & 0 & 0 \\ -0.010381 & 0 & 0 \end{bmatrix}$$

To determine the second column, Eq. (9.8) is formulated as

$$\begin{bmatrix} 1 & 0 & 0 \\ -0.3 & 1 & 0 \\ 0.1 & -0.148148 & 1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

This can be solved with forward substitution for  $\{d\}^T = \begin{bmatrix} 0 & 1 & 0.148148 \end{bmatrix}$ , and the results are used with [U] to determine  $\{x\}$  by back substitution to generate the second column of the matrix inverse,

$$[A]^{-1} = \begin{bmatrix} 0.110727 & 0.038062 & 0 \\ -0.058824 & -0.176471 & 0 \\ -0.010381 & 0.027682 & 0 \end{bmatrix}$$

Finally, the same procedures can be implemented with  $\{b\}^T = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}$  to solve for  $\{d\}^T = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}$ , and the results are used with [U] to determine  $\{x\}$  by back substitution to generate the third column of the matrix inverse,

$$[A]^{-1} = \begin{bmatrix} 0.110727 & 0.038062 & 0.00692 \\ -0.058824 & -0.176471 & 0.058824 \\ -0.010381 & 0.027682 & 0.186851 \end{bmatrix}$$

This result can be checked by multiplying it times the original matrix to give the identity matrix. The following MATLAB session can be used to implement this check,

```
>> A = [10 2 -1; -3 -6 2; 1 1 5];

>> AI = [0.110727 0.038062 0.00692;

-0.058824 -0.176471 0.058824;

-0.010381 0.027682 0.186851];

>> A*AI

ans =

1.0000 -0.0000 -0.0000

0.0000 1.0000 -0.0000

-0.0000 0.0000 1.0000
```

**11.2** The system can be written in matrix form as

$$[A] = \begin{bmatrix} -8 & 1 & -2 \\ 2 & -6 & -1 \\ -3 & -1 & 7 \end{bmatrix} \qquad \{b\} = \begin{cases} -38 \\ -34 \\ -20 \end{cases}$$

Forward eliminate

$$f_{21} = 2/(-8) = -0.25$$

$$[A] = \begin{bmatrix} -8 & 1 & -2 \\ 0 & -5.75 & -1.5 \\ 0 & -1.375 & 7.75 \end{bmatrix}$$

Forward eliminate

$$f_{32} = -1.375/(-5.75) = 0.23913$$

$$[A] = \begin{bmatrix} -8 & 1 & -2 \\ 0 & -5.75 & -1.5 \\ 0 & 0 & 8.108696 \end{bmatrix}$$

Therefore, the LU decomposition is

$$[L]{U} = \begin{bmatrix} 1 & 0 & 0 \\ -0.25 & 1 & 0 \\ 0.375 & 0.23913 & 1 \end{bmatrix} \begin{bmatrix} -8 & 1 & -2 \\ 0 & -5.75 & -1.5 \\ 0 & 0 & 8.108696 \end{bmatrix}$$

The first column of the matrix inverse can be determined by performing the forward-substitution solution procedure with a unit vector (with 1 in the first row) as the right-hand-side vector. Thus, the lower-triangular system, can be set up as,

$$\begin{bmatrix} 1 & 0 & 0 \\ -0.25 & 1 & 0 \\ 0.375 & 0.23913 & 1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

and solved with forward substitution for  $\{d\}^T = \begin{bmatrix} 1 & 0.25 & -0.434783 \end{bmatrix}$ . This vector can then be used as the right-hand side of the upper triangular system,

$$\begin{bmatrix} -8 & 1 & -2 \\ 0 & -5.75 & -1.5 \\ 0 & 0 & 8.108696 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0.25 \\ -0.434783 \end{bmatrix}$$

which can be solved by back substitution for the first column of the matrix inverse,

$$[A]^{-1} = \begin{bmatrix} -0.115282 & 0 & 0 \\ -0.029491 & 0 & 0 \\ -0.053619 & 0 & 0 \end{bmatrix}$$

To determine the second column, Eq. (9.8) is formulated as

$$\begin{bmatrix} 1 & 0 & 0 \\ -0.25 & 1 & 0 \\ 0.375 & 0.23913 & 1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

This can be solved with forward substitution for  $\{d\}^T = \begin{bmatrix} 0 & 1 & -0.23913 \end{bmatrix}$ , and the results are used with [U] to determine  $\{x\}$  by back substitution to generate the second column of the matrix inverse,

$$[A]^{-1} = \begin{bmatrix} -0.115282 & -0.013405 & 0\\ -0.029491 & -0.16622 & 0\\ -0.053619 & -0.029491 & 0 \end{bmatrix}$$

Finally, the same procedures can be implemented with  $\{b\}^T = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}$  to solve for  $\{d\}^T = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}$ , and the results are used with [U] to determine  $\{x\}$  by back substitution to generate the third column of the matrix inverse.

$$[A]^{-1} = \begin{bmatrix} -0.115282 & -0.013405 & -0.034853 \\ -0.029491 & -0.16622 & -0.032172 \\ -0.053619 & -0.029491 & 0.123324 \end{bmatrix}$$

**11.3** The following solution is generated with MATLAB.

```
(a)
>> A = [15 -3 -1; -3 18 -6; -4 -1 12];
>> format long
>> AI = inv(A)
AI =
   0.07253886010363 \qquad 0.01278065630397 \qquad 0.01243523316062
   0.02072538860104 0.06079447322971 0.03212435233161
   0.02590673575130 0.00932642487047 0.09015544041451
(b)
>> b = [3800 1200 2350]';
>> format short
>> c = AI*b
C =
 320.2073
  227.2021
  321.5026
```

(c) The impact of a load to reactor 3 on the concentration of reactor 1 is specified by the element  $a_{13}^{-1} = 0.0124352$ . Therefore, the increase in the mass input to reactor 3 needed to induce a 10 g/m<sup>3</sup> rise in the concentration of reactor 1 can be computed as

$$\Delta b_3 = \frac{10}{0.0124352} = 804.1667 \frac{g}{d}$$

(d) The decrease in the concentration of the third reactor will be

$$\Delta c_3 = 0.0259067(500) + 0.009326(250) = 12.9534 + 2.3316 = 15.285 \frac{\mathrm{g}}{\mathrm{m}^3}$$

11.4 The mass balances can be written and the result written in matrix form as

$$\begin{bmatrix} 6 & 0 & -1 & 0 & 0 \\ -3 & 3 & 0 & 0 & 0 \\ 0 & -1 & 9 & 0 & 0 \\ 0 & -1 & -8 & 11 & -2 \\ -3 & -1 & 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \\ c_5 \end{bmatrix} = \begin{bmatrix} Q_{01}c_{01} \\ 0 \\ Q_{03}c_{03} \\ 0 \\ 0 \end{bmatrix}$$

MATLAB can then be used to determine the matrix inverse

The concentration in reactor 5 can be computed using the elements of the matrix inverse as in,

$$c_5 = a_{51}^{-1}Q_{01}c_{01} + a_{53}^{-1}Q_{03}c_{03} = 0.1698(5)20 + 0.0189(8)50 = 16.981 + 7.547 = 24.528$$

# 11.5 The problem can be written in matrix form as

$$\begin{bmatrix} 0.866 & 0 & -0.5 & 0 & 0 & 0 \\ 0.5 & 0 & 0.866 & 0 & 0 & 0 \\ -0.866 & -1 & 0 & -1 & 0 & 0 \\ -0.5 & 0 & 0 & 0 & -1 & 0 \\ 0 & 1 & 0.5 & 0 & 0 & 0 \\ 0 & 0 & -0.866 & 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \\ F_3 \\ H_2 \\ V_2 \\ V_3 \end{bmatrix} = \begin{bmatrix} F_{1,h} \\ F_{1,v} \\ F_{2,h} \\ F_{2,v} \\ F_{3,h} \\ F_{3,v} \end{bmatrix}$$

MATLAB can then be used to solve for the matrix inverse,

The forces in the members resulting from the two forces can be computed using the elements of the matrix inverse as in,

$$F_1 = a_{12}^{-1} F_{1,v} + a_{15}^{-1} F_{3,h} = 0.5(-2000) + 0(-500) = -1000 + 0 = -1000$$

$$F_2 = a_{22}^{-1} F_{1,v} + a_{25}^{-1} F_{3,h} = -0.433(-2000) + 1(-500) = 866 - 500 = 366$$

$$F_3 = a_{32}^{-1} F_{1,\nu} + a_{35}^{-1} F_{3,h} = 0.866(-2000) + 0(-500) = -1732 + 0 = -1732$$

**11.6** The matrix can be scaled by dividing each row by the element with the largest absolute value

MATLAB can then be used to determine each of the norms.

```
>> norm(A,'fro')
ans =
          1.9920
>> norm(A,1)
ans =
          2.8000
>> norm(A,inf)
ans =
          2
```

# **11.7** Prob. 11.2:

```
>> A = [-8 1 -2;2 -6 -1;-3 -1 7];
>> norm(A,'fro')

ans =
    13
>> norm(A,inf)

ans =
    11
```

#### Prob. 11.3:

## 11.8 (a) Spectral norm

```
>> A = [1 4 9 16;4 9 16 25;9 16 25 36;16 25 36 49];
>> cond(A)

ans =
  8.8963e+016
```

### (b) Row-sum norm

```
>> cond(A,inf)
```

#### 11.9 (a) The matrix to be evaluated is

The row-sum norm of this matrix is 49 + 7 + 1 = 57. The inverse is

$$\begin{bmatrix} -0.1667 & 0.1 & 0.0667 \\ 1.5 & -1.1 & -0.4 \\ -2.3333 & 2.8 & 0.5333 \end{bmatrix}$$

The row-sum norm of the inverse is |-2.3333| + 2.8 + 0.5333 = 5.6667. Therefore, the condition number is

$$Cond[A] = 57(5.6667) = 323$$

This can be verified with MATLAB,

```
>> A = [16 4 1;4 2 1;49 7 1];
>> cond(A,inf)

ans =
  323.0000
```

#### (b) Spectral norm:

```
>> A = [16 4 1;4 2 1;49 7 1];
>> cond(A)

ans =
  216.1294
```

#### Frobenius norm:

```
>> cond(A,'fro')
ans =
  217.4843
```

11.10 The spectral condition number can be evaluated as

```
>> A = hilb(10);
>> N = cond(A)
N =
1.6025e+013
```

The digits of precision that could be lost due to ill-conditioning can be calculated as

```
>> c = log10(N)
c = 13.2048
```

Thus, about 13 digits could be suspect. A right-hand side vector can be developed corresponding to a solution of ones:

```
>> b=[sum(A(1,:)); sum(A(2,:)); sum(A(3,:)); sum(A(4,:)); sum(A(5,:));
sum(A(6,:)); sum(A(7,:)); sum(A(8,:)); sum(A(9,:)); sum(A(10,:))]
b =
    2.9290
    2.0199
    1.6032
    1.3468
    1.1682
    1.0349
    0.9307
    0.8467
```

```
0.7773
0.7188
```

The solution can then be generated by left division

```
>> x = A\b

x =

1.0000
1.0000
1.0000
0.9999
1.0003
0.9995
1.0005
0.9997
1.0001
```

The maximum and mean errors can be computed as

```
>> e=max(abs(x-1))
e =
    5.3822e-004
>> e=mean(abs(x-1))
e =
    1.8662e-004
```

Thus, some of the results are accurate to only about 3 to 4 significant digits. Because MATLAB represents numbers to 15 significant digits, this means that about 11 to 12 digits are suspect.

#### 11.11 First, the Vandermonde matrix can be set up

```
>> x1 = 4;x2=2;x3=7;x4=10;x5=3;x6=5;
\Rightarrow A = [x1<sup>5</sup> x1<sup>4</sup> x1<sup>3</sup> x1<sup>2</sup> x1 1;x2<sup>5</sup> x2<sup>4</sup> x2<sup>3</sup> x2<sup>2</sup> x2 1;x3<sup>5</sup> x3<sup>4</sup> x3<sup>3</sup>
x3^2 x3 1;x4^5 x4^4 x4^3 x4^2 x4 1;x5^5 x5^4 x5^3 x5^2 x5 1;x6^5 x6^4 x6^3
x6^2 x6 1]
A =
        1024
                     256
                                    64
                                                  16
                                    8
          32
                      16
                                                  4
                                                                2
                   2401
                                                               7
       16807
                                  343
                                                 49
      100000
                                                                             1
                                  1000
                    10000
                                                 100
                                                             10
                                                                             1
         243
                     81
                                   27
                                                 9
                                                               3
                                                                             1
                                    125
                                                 25
        3125
                      625
                                                                             1
```

The spectral condition number can be evaluated as

```
>> N = cond(A)
N =
1.4492e+007
```

The digits of precision that could be lost due to ill-conditioning can be calculated as

```
>> c = log10(N)
c = 7.1611
```

Thus, about 7 digits might be suspect. A right-hand side vector can be developed corresponding to a solution of ones:

The solution can then be generated by left division

The maximum and mean errors can be computed as

```
>> e = max(abs(x-1))
e =
          5.420774940034789e-012
>> e = mean(abs(x-1))
e =
          2.154110223528960e-012
```

Some of the results are accurate to about 12 significant digits. Because MATLAB represents numbers to about 15 significant digits, this means that about 3 digits are suspect. Thus, for this case, the condition number tends to exaggerate the impact of ill-conditioning.

**11.12** (a) The solution can be developed using your own software or a package. For example, using MATLAB,

(b) The element of the matrix that relates the concentration of Havasu (lake 4) to the loading of Powell (lake 1) is  $a_{41}^{-1} = 0.084767$ . This value can be used to compute how much the loading to Lake Powell must be reduced in order for the chloride concentration of Lake Havasu to be 75 as

$$\Delta W_1 = \frac{\Delta c_4}{a_{41}^{-1}} = \frac{100.2373 - 75}{0.084767} = 297.725$$

(c) First, normalize the matrix to give

$$[A] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & -0.91283 & 0 & 0 \\ 0 & -0.9899 & 1 & 0 \\ 0 & 0 & 1 & -0.95314 \end{bmatrix}$$

The column-sum norm for this matrix is 2. The inverse of the matrix can be computed as

$$[A]^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1.095495 & -1.09549 & 0 & 0 \\ 1.084431 & -1.08443 & 1 & 0 \\ 1.137747 & -1.13775 & 1.049165 & -1.04917 \end{bmatrix}$$

The column-sum norm for the inverse can be computed as 4.317672. The condition number is, therefore, 2(4.317672) = 8.635345. This means that less than 1 digit is suspect  $[\log_{10}(8.635345) = 0.93628]$ . Interestingly, if the original matrix is unscaled, the same condition number results.

**11.13** (a) When MATLAB is used to determine the inverse, the following error message suggests that the matrix is ill-conditioned:

```
>> inv(A)
Warning: Matrix is close to singular or badly scaled.
        Results may be inaccurate. RCOND = 1.541976e-018.
ans =
 1.0e+016 *
   -0.4504 0.9007
                     -0.4504
   0.9007
             -1.8014
                       0.9007
   -0.4504
              0.9007
                       -0.4504
The high condition number reinforces this conclusion:
>> cond(A)
ans =
  3.8131e+016
```

**(b)** However, when one of the coefficients is changed slightly, the system becomes well-conditioned:

```
>> A=[1 2 3;4 5 6;7 8 9.1];
>> inv(A)

ans =
    8.3333   -19.3333    10.0000
    -18.6667    39.6667    -20.0000
    10.0000    -20.0000    10.0000

>> cond(A)

ans =
    994.8787
```

11.14 The five simultaneous equations can be set up as

MATLAB can then be used to solve for the coefficients,

```
>> format short g

>> A=[200^4 200^3 200^2 200 1

250^4 250^3 250^2 250 1

300^4 300^3 300^2 300 1

400^4 400^3 400^2 400 1

500^4 500^3 500^2 500 1]

A =

1.6e+009 8e+006 40000 200 1
```

1

1

1

1

```
3.9063e+009 1.5625e+007
                                  62500
                                                  250
              2.7e+007
6.4e+007
                                  90000
                                                  300
     8.1e+009
    2.56e+010
                                                  400
                               1.6e+005
    6.25e+010
                 1.25e+008
                               2.5e+005
                                                  500
>> b=[0.746;0.675;0.616;0.525;0.457];
>> format long q
>> p=A\b
p =
     1.3333333333201e-012
    -4.5333333333155e-009
     5.29666666666581e-006
      -0.00317366666666649
          1.20299999999999
>> cond(A)
ans =
          11711898982423.4
```

Thus, because the condition number is so high, the system seems to be ill-conditioned. This implies that this might not be a very reliable method for fitting polynomials. Because this is generally true for higher-order polynomials, other approaches are commonly employed as will be described subsequently in Chap. 15.