



**HACETTEPE UNIVERSITY
ENGINEERING FACULTY
ELECTRICAL AND ELECTRONICS
ENGINEERING PROGRAM**

2023-2024
SPRING SEMESTER

ELE708
NUMERICAL METHODS IN ELECTRICAL ENGINEERING

HW3

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1) Exercises

1.a) 3.1

3.1) if a vertical beam has a downward force applied its lower end, the amount by which it stretches will be proportional to the magnitude of the force. Thus the total length of y :

$$y = x_1 + x_2 \cdot t$$

$\begin{matrix} \swarrow & \searrow \\ \text{force applied} & \text{proportionality constant} \\ \text{original length} & \end{matrix}$

t	10	15	20
y	11,6	11,85	12,25

a) Set up overdetermined 3×2 system

b) Is the system consistent? If not calculate each possible values, Is there a reason to select a μ

c) Set up normal equation and solve it to obtain the least square solution. Compare with b.

a) $y = x_1 + x_2 \cdot t = \begin{bmatrix} 1 & t \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

$$A \cdot x = \begin{bmatrix} 1 & 10 \\ 1 & 15 \\ 1 & 20 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 11,6 \\ 11,85 \\ 12,25 \end{bmatrix} = b$$

b) No, It is not consistent, the equation looks like linear but the increase in y and t is not linear.

Solving by using 1 and 2 equation $\Rightarrow \begin{bmatrix} 1 & 10 \\ 1 & 15 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 11,6 \\ 11,85 \end{bmatrix} \Rightarrow \begin{matrix} x_1 = 11,1 \\ x_2 = 0,05 \end{matrix}$

Solving by using 1 and 3 equation $\Rightarrow \begin{bmatrix} 1 & 10 \\ 1 & 20 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 11,6 \\ 12,25 \end{bmatrix} \Rightarrow \begin{matrix} x_1 = 10,95 \\ x_2 = 0,065 \end{matrix}$

Solving by using 2 and 3 equation $\Rightarrow \begin{bmatrix} 1 & 15 \\ 1 & 20 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 11,85 \\ 12,25 \end{bmatrix} \Rightarrow \begin{matrix} x_1 = 10,65 \\ x_2 = 0,08 \end{matrix}$

The average of $x_1 = (11,1 + 10,95 + 10,65)/3 = 10,9$

The average of $x_2 = (0,05 + 0,065 + 0,08)/3 = 0,065$ > closest pair is the second

-c)

$$A^T \cdot A \cdot x = A^T \cdot b$$

$$\begin{bmatrix} 1 & 1 & 1 \\ 10 & 15 & 20 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 34,55 \\ 45,725 \end{bmatrix} \quad \left| \quad \begin{bmatrix} 1 & 1 & 1 \\ 10 & 15 & 20 \end{bmatrix} \cdot \begin{bmatrix} 11,6 \\ 11,85 \\ 12,25 \end{bmatrix} = \begin{bmatrix} 35,7 \\ 539,75 \end{bmatrix} \right.$$

$$\begin{bmatrix} 3 & 45 \\ 45 & 725 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 35,7 \\ 539,75 \end{bmatrix}$$

\downarrow
 $x_1 = 10,925$
 $x_2 = 0,065$ > The average values we computed in part b is similar.

1.b) 3.4

3.4) In fitting straight line $y = x_0 + x_1 \cdot t$ to the three data points $(t_i, y_i) = (0,0), (1,0), (1,1)$ is the least square solution is unique and why?

As it mentioned, "3.2 Existing and Uniqueness", the solution is unique if and only if A has full column rank.

$$A \cdot x \approx b \quad \Rightarrow \quad \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \approx \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

\downarrow $s_3 = s_3 - s_2$

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow \text{rank}(A) = 2 \quad \text{and} \quad n = 2$$

so its okay, it has unique solution.

3.5) Let x be the solution to the linear least square problem $Ax \approx b$, where

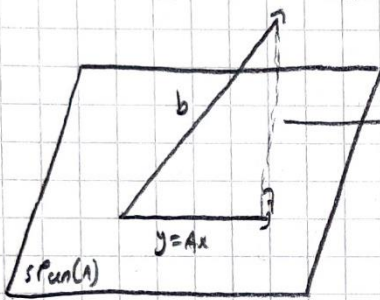
$$A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix}$$

let $r = b - Ax$ be the corresponding residual vector. Which of the following three vectors is a possible value for r ? Why?

a) $\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$

b) $\begin{bmatrix} -1 \\ -1 \\ 1 \\ 1 \end{bmatrix}$

c) $\begin{bmatrix} -1 \\ 1 \\ 1 \\ -1 \end{bmatrix}$



$$r = b - Ax$$

, the residue must be orthogonal to the A
thus,

$$0 = A^T r = A^T (b - Ax)$$

a) $A^T r = 0 \Rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 6 \end{bmatrix} \rightarrow \text{not orthogonal } \times$

b) $A^T r = 0 \Rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ -1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 4 \end{bmatrix} \rightarrow \text{only orthogonal to first column } \times$

c) $A^T r = 0 \Rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 1 \\ 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \text{fully orthogonal } \checkmark$

2) Computer Problems

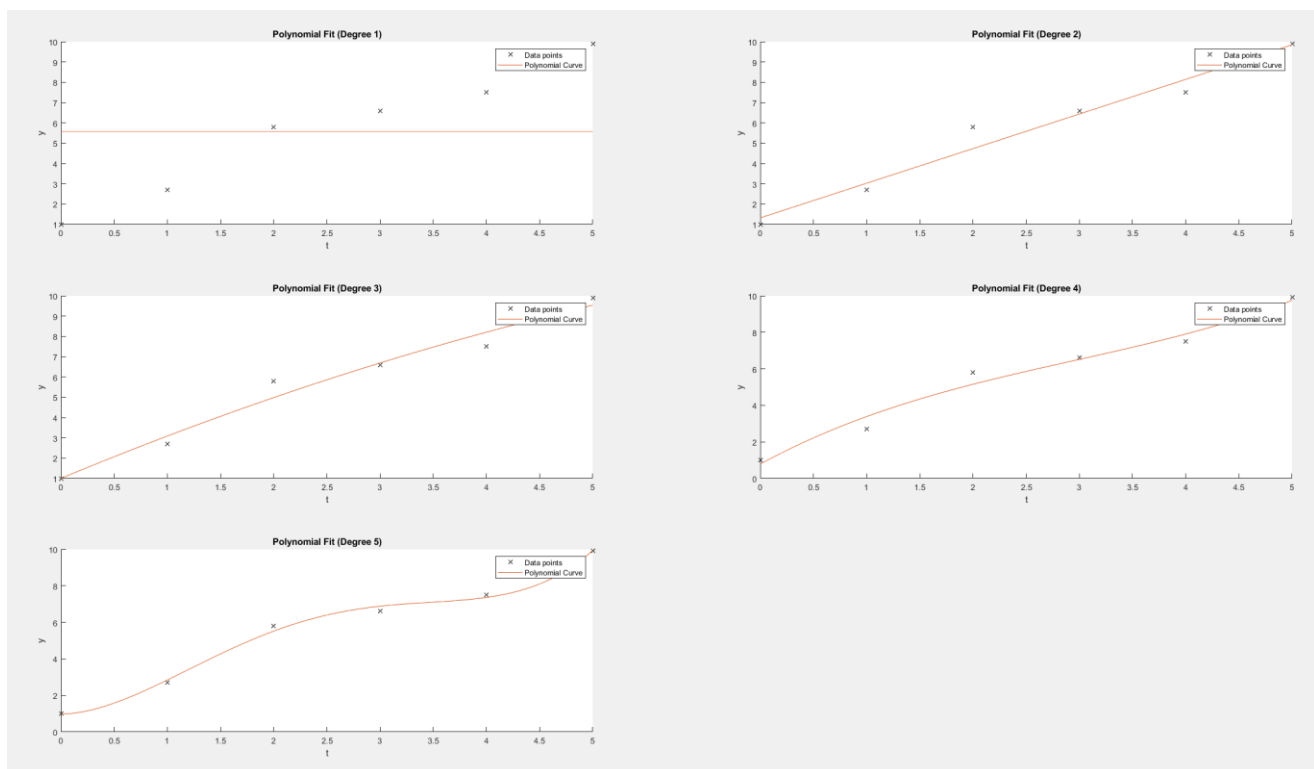
2.a) 3.3 For $n = 0, 1, \dots, 5$, fit a polynomial of degree n by least squares to the following data:

$t: \{0.0, 1.0, 2.0, 3.0, 4.0, 5.0\}$

$y: \{1.0, 2.7, 5.8, 6.6, 7.5, 9.9\}$

Make a plot of the original data points along with each resulting polynomial curve (you may make separate graphs for each curve or a single graph containing all of the curves). Which polynomial would you say captures the general trend of the data better? Obviously, this is a subjective question, and its answer depends on both the nature of the given data (e.g., the uncertainty of the data values) and the purpose of the fit. Explain your assumptions in answering.

The polynomial curves shown in figure below, in each step the curve fit the data more and more. The last curve would be the best fit yet since the data could have some error, making the curve overfit may not be the best idea. In order to avoid this I would say 3rd or 4th curve would be better choice.



2.b) 3.2 A common problem in surveying is to determine the altitudes of a series of points with respect to some reference point. The measurements are subject to error, so more observations are taken than are strictly necessary to determine the altitudes, and the resulting overdetermined system is solved in the least squares sense to smooth out errors. Suppose that there are four points whose altitudes x_1, x_2, x_3, x_4 are to be determined. In addition to direct measurements of each x_i with respect to the reference point, measurements are also taken of each point with respect to all of the others. The resulting measurements are:

$$\begin{array}{ll} x_1 = 2.95 & x_2 = 1.74 \\ x_3 = -1.45 & x_4 = 1.32 \\ x_1 - x_2 = 1.23 & x_1 - x_3 = 4.45 \\ x_1 - x_4 = 1.61 & x_2 - x_3 = 3.21 \\ x_2 - x_4 = 0.45 & x_3 - x_4 = -2.75 \end{array}$$

Set up the corresponding least squares system $Ax \approx b$ and use a library routine, or one of your own design, to solve it for the best values of the altitudes. How do your computed values compare with the direct measurements?

The answers shown in figure below. Since there was no big difference between the direct values and the values with a reference point. The answer close to direct measurements.

```
Workspace
>> CP3_2
x1 = 2.96
x2 = 1.746
x3 = -1.46
x4 = 1.314
fx >>
```

2.c) 3.4 a) Solve the following least squares problem using any method you like:

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.26 \\ 0.28 \\ 3.31 \end{bmatrix}$$

(b) Now solve the same least squares problem again, but this time use the slightly perturbed right-hand side

$$b = \begin{bmatrix} 0.27 \\ 0.25 \\ 3.33 \end{bmatrix}$$

c) Compare your results from parts a and b. Can you explain this difference?

a)

```
a):  
Normal Equation:  
    1.0000  
    1.0000  
  
SVD:  
    1.0000  
    1.0000
```

b)

```
b):  
Normal Equation:  
    7.0089  
   -8.3957  
  
SVD:  
    7.0089  
   -8.3957
```

c) In this problem we can see that it is an ill-conditioned system. The sensitivity depends on both $\text{cond}(A)$ and the angle. Even though the angle is small, the cond number increases the sensitivity.

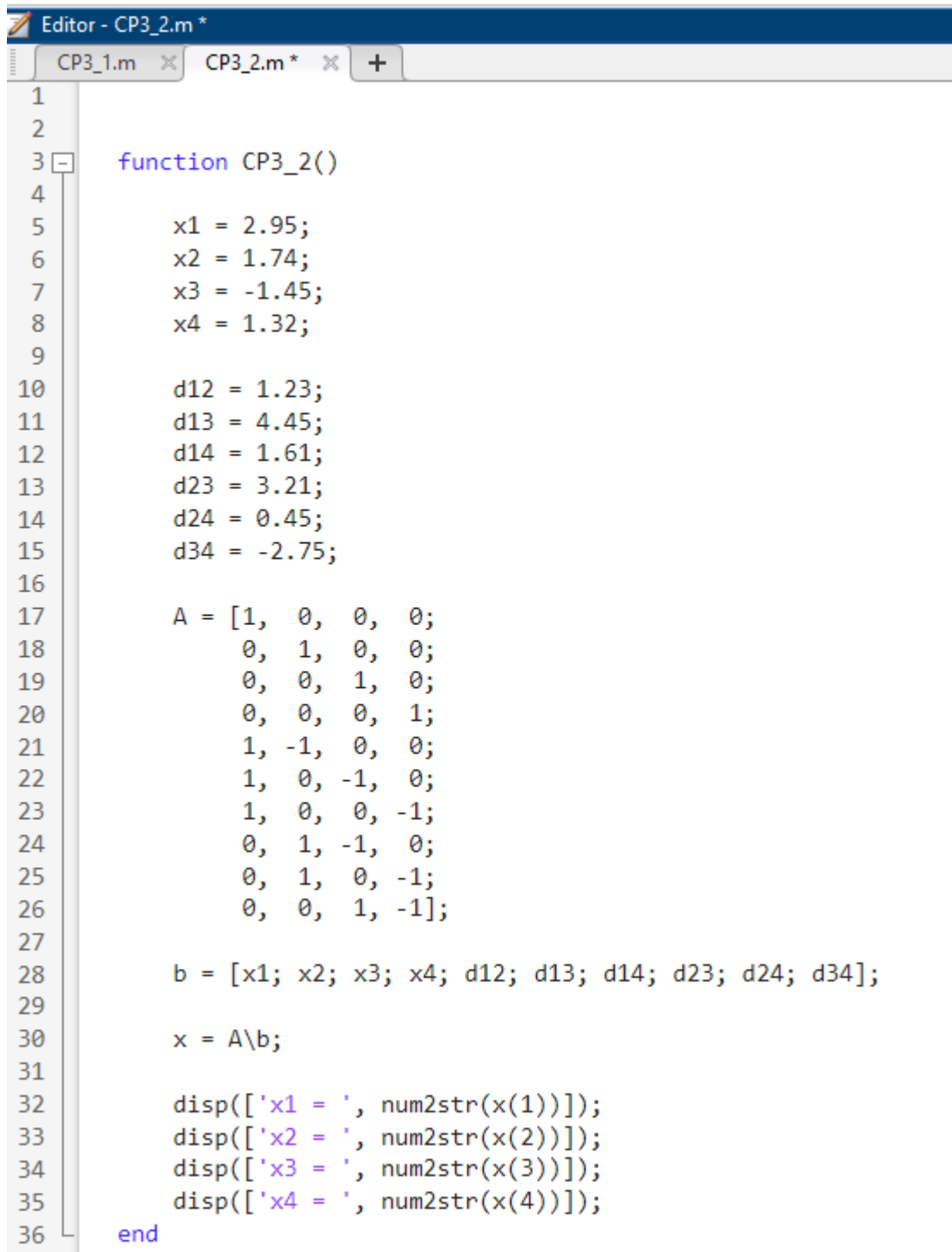
```
c)  
condA = 1097.5387  
cos_theta_1 = 1  
cos_theta_2 = 0.99998  
cos_theta = 0.01123  
delx_over_x = 7.8862  
delb_over_b = 0.01123  
bound = 0.1384
```

2.d) Codes for 1.1 and 1.4

2.d.i) 3.1

```
11 function CP3_1()
12
13     t = [0; 1; 2; 3; 4; 5];
14     y = [1; 2.7; 5.8; 6.6; 7.5; 9.9];
15     m = size(t,1);
16     A = ones(m,1);
17     p = 51;
18     ts = linspace(0,5,p)';
19
20
21     for n = 1:5
22         subplot(3,2,n);
23         hold on;
24         plot(t, y, 'kx');
25         x = A(:,1:n)\y;
26         ys(1:p,n) = x(n);
27
28         for k = 2:n
29             ys(:,n) = ys(:,n).* ts+x(n-k+1);
30         end
31
32
33         plot(ts, ys(:,n));
34         title(['Polynomial Fit (Degree ', num2str(n), ')']);
35         xlabel('t'); ylabel('y');
36         legend('Data points', 'Polynomial Curve');
37         hold off;
38         A(:,n+1) = A(:,n) .*t;
39     end
40 end
```


2.d.ii) 3.2



The image shows a MATLAB Editor window titled "Editor - CP3_2.m *". The window has two tabs: "CP3_1.m" and "CP3_2.m *", with the latter being the active tab. The code is written in a function format, starting with "function CP3_2()". The code defines several variables: x1, x2, x3, x4, d12, d13, d14, d23, d24, d34, and a matrix A. It then calculates b as a row vector of these variables, solves the system x = A\b, and displays the first four elements of x. The code ends with the "end" keyword.

```
1
2
3 function CP3_2()
4
5     x1 = 2.95;
6     x2 = 1.74;
7     x3 = -1.45;
8     x4 = 1.32;
9
10    d12 = 1.23;
11    d13 = 4.45;
12    d14 = 1.61;
13    d23 = 3.21;
14    d24 = 0.45;
15    d34 = -2.75;
16
17    A = [1, 0, 0, 0;
18         0, 1, 0, 0;
19         0, 0, 1, 0;
20         0, 0, 0, 1;
21         1, -1, 0, 0;
22         1, 0, -1, 0;
23         1, 0, 0, -1;
24         0, 1, -1, 0;
25         0, 1, 0, -1;
26         0, 0, 1, -1];
27
28    b = [x1; x2; x3; x4; d12; d13; d14; d23; d24; d34];
29
30    x = A\b;
31
32    disp(['x1 = ', num2str(x(1))]);
33    disp(['x2 = ', num2str(x(2))]);
34    disp(['x3 = ', num2str(x(3))]);
35    disp(['x4 = ', num2str(x(4))]);
36 end
```

2.d.iii) 3.4

```

CP3_1.m  CP3_2.m  CP3_4.m  +
1  function CP3_4()
2
3      A = [0.16, 0.10;
4           0.17, 0.11;
5           2.02, 1.29;];
6
7      b1 = [0.26; 0.28; 3.31];
8      x1_ne = A \ b1;
9      [U, S, V] = svd(A, 'econ');
10     x1_svd = V * (S' * inv(S * S')) * U' * b1;
11     disp('a:');
12     disp('Normal Equation:');
13     disp(x1_ne);
14     disp('SVD:');
15     disp(x1_svd);
16
17
18
19     b2 = [0.27; 0.25; 3.33];
20     x2_ne = A \ b2;
21     [U, S, V] = svd(A, 'econ');
22     x2_svd = V * (S' * inv(S * S')) * U' * b2;
23     disp('b:');
24     disp('Normal Equation:');
25     disp(x2_ne);
26     disp('SVD:');
27     disp(x2_svd);
28
29     %pseudo_inverse_A = inv(transpose(A)*A)*transpose(A)
30
31     condA = cond(A);
32     cos_theta_1 = norm(A*x1_ne) / norm(b1);
33     cos_theta_2 = norm(A*x2_ne) / norm(b2);
34     cos_theta = norm(b2 - b1) / norm(b1);
35
36     delta_x_over_x = norm(x2_ne - x1_ne) / norm(x1_ne);
37     delta_b_over_b = norm(b2 - b1) / norm(b1);
38     bound = condA * cos_theta * delta_b_over_b;
39
40     disp('c');
41     disp(['condA = ', num2str(condA)]);
42     disp(['cos_theta_1 = ', num2str(cos_theta_1)]);
43     disp(['cos_theta_2 = ', num2str(cos_theta_2)]);
44     disp(['cos_theta = ', num2str(cos_theta)]);
45
46     disp(['delx_over_x = ', num2str(delta_x_over_x)]);
47     disp(['delb_over_b = ', num2str(delta_b_over_b)]);
48     disp(['bound = ', num2str(bound)]);
49
50
51     end

```