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## Solution 5

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### Question 1: Least-Squares formulation, Normal equation

a) Given three points  $P_i = (x_i, y_i)$ ,  $i = 1, 2, 3$ , wherein

$x_i$	0	1	2
$y_i$	5.41	5.17	5.93

Determine a linear function  $y = f(x) = ax + b$ , so that the sum of the error squares in the y-direction

$$\sum_{i=1}^3 |f(x_i) - y_i|^2 \quad \text{is minimized}$$

**Solution :** We would like to minimize  $\sum_{i=1}^3 (f(x_i) - y_i)^2$

$$ax_1 + b - y_1 = r_1$$

$$ax_2 + b - y_2 = r_2$$

$$ax_3 + b - y_3 = r_3$$

We can reduce the problem to linear-system  $Ap = c$ , and solve the problem for least-squares using Normal Equations i.e.  $p = (A^T A)^{-1} A^T c$

$$A = \begin{pmatrix} x_1 & 1 \\ x_2 & 1 \\ x_3 & 1 \end{pmatrix}; \quad c = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix}; \quad p = \begin{pmatrix} a \\ b \end{pmatrix}$$

$$p = (A^T A)^{-1} A^T c \implies p = (0.26, 5.243).$$

So  $f(x) = ax + b$ , where **a = 0.26**, **b = 5.243**

b) Consider the matrix  $A$  ( $3 \times 2$ ) and vector  $b$  ( $3 \times 1$ ), given by

$$A = \begin{pmatrix} 1 & 1 \\ \epsilon & 0 \\ 0 & \epsilon \end{pmatrix} \quad \text{and} \quad b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

i) Write the least-square form which leads to the solution of linear system  $Ax = b$ .

ii) Check if the method of Normal Equations is stable for matrix  $A$  and  $0 < \epsilon \ll 1$ .

**Solution :** i)  $x = (A^T A)^{-1} A^T b$

ii)  $(A^T A)$  is singular, so the method Normal equations is unstable.

## Question 2: Least-squares, QR decomposition and SVD

Following are the velocity measurements  $f(t)$  in  $ms^{-1}$  from the pitot-tube of a descending airplane at time  $t_i$ ,  $i = 1, 2, \dots, 10$ .

$t_i$	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0
$f_i$	100	34	17	12	9	6	5	4	4	2

We express the unknown function  $f(t) = \sum_{j=1}^4 \lambda_j \phi_j(t)$  as the linear combination of known functions  $\phi_j(t)$ ,  $j = 1, 2, 3, 4$ , given by,

$$\phi_1(t) = \frac{1}{t}, \quad \phi_2(t) = \frac{1}{t^2}, \quad \phi_3(t) = e^{-(t-1)}, \quad \phi_4(t) = e^{-2(t-1)}$$

Determine the coefficients  $\lambda_j$ , the linear combination such that

$$\sum_{i=1}^{10} |f(t_i) - f_i|^2 \quad \text{is minimized}$$

- a) using normal equations
- b) using the QR decomposition of the matrix A
- c) by means of singular value decomposition (SVD)  
(MATLAB command `[U, S, V] = svd(A)` )

**Solution :** Set up the linear system  $Ax = f$  for least-square solution, i.e

$$t = [0.1 \quad 0.2 \quad 0.3 \quad \dots \quad 1.0]^T$$

$$f = [100 \quad 34 \quad 17 \quad \dots \quad 12]^T$$

$$A = [1.0/t \quad 1.0/t^2 \quad e^{-(t-1)} \quad e^{-2(t-1)}]$$

a) Method of normal equations  $x = (A^T A)^{-1} A^T f$ ,

$$\implies \lambda_1 = 4.0591, \lambda_2 = 0.6140, \lambda_3 = -2.5315, \lambda_4 = 0.7058$$

b)  $A = QR \implies QRx = b \implies Rx = Q^T b$

$$\implies \lambda_1 = 4.0591, \lambda_2 = 0.6140, \lambda_3 = -2.5315, \lambda_4 = 0.7058$$

c)  $A = USV^T \implies x = VS^{-1}U^T b$

$$\implies \lambda_1 = 4.0591, \lambda_2 = 0.6140, \lambda_3 = -2.5315, \lambda_4 = 0.7058$$

### Question 3: SVD decomposition by hand

Given the matrices A and B,

$$A = \begin{pmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{pmatrix}$$

- a) Find the rank and singular values of the matrix A and matrix B.
- b) Find the SVD decomposition of the matrix A and B. Also comment on the uniqueness of the decomposition.

**Solution :**

a) Rank(A) = 2

$$A^T A = \begin{pmatrix} 3 & 1 & 2 \\ 1 & 1 & 0 \\ 2 & 0 & 2 \end{pmatrix} \quad ; \quad \lambda_1 = \sqrt{3 + \sqrt{3}}, \lambda_2 = \sqrt{3 - \sqrt{3}}, \lambda_3 = 0$$

Rank(B) = 2

$$B^T B = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{pmatrix} \quad ; \quad \lambda_1 = \sqrt{2}, \lambda_2 = \sqrt{2}$$

Note  $\lambda$  are the singular values (i.e., square root of the Eigen values of the product with transpose).

b) Refer to the following tutorial:

[http://web.mit.edu/be.400/www/SVD/Singular\\_Value\\_Decomposition.htm](http://web.mit.edu/be.400/www/SVD/Singular_Value_Decomposition.htm)

Only the singular value matrix  $S$  is always unique for the SVD.