

**Department of Applied Physics and Applied Mathematics  
Columbia University**

**APMA S3101D: Applied Math I - Introduction to Linear Algebra  
Summer 2003**

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**Problem Set 4**

**Note: This is for those who use the **SECOND** version of the textbook. Those who  
have the third version of the book please go to ps2-v3.ps**

(Due June 26, 2003)

1. Problem Set 4.2: 4<sup>1</sup>, 5, 13, 22
2. Problem Set 4.3: 12
3. Problem Set 4.4: 1, 12, 18<sup>2</sup>, 22
4. Consider approximating the cubic  $y_3(x) = x^3 + 2x$  by a quadratic function  $y_2(x) = c_1 + c_2x + c_3x^2$  in the interval  $-2 \leq x \leq 2$ . To do this, sample the cubic at points  $x = (-2, -1, 0, 1, 2)$  to obtain corresponding values of the function  $y_3(x)$  at those points. Then perform a least squares fit of these “data” to the quadratic function  $y_2$  to obtain  $(c_1, c_2, c_3)$ .

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<sup>1</sup>Construct the projection matrices  $P_1$  and  $P_2$  onto the lines through the  $a$ 's in Problem 2. Is it true that  $(P_1 + P_2)^2 = P_1 + P_2$ ? This *would* be true if  $P_1P_2 = 0$ .

<sup>2</sup>Find orthogonal vectors  $A, B, C$  by Gram-Schmidt from  $a, b, c$ :

$$a = (1, -2, 0, 0), b = (0, 1, -1, 0), c = (0, 0, 1, -1)$$

$A, B, C$  and  $a, b, c$  are bases for the vectors perpendicular to  $d = (1, 1, 1, 1)$

5. (*Weighted least squares* (10 pts)) Compute the least squares solution  $\hat{\mathbf{x}}$  by minimizing the following *cost function* with respect to  $\mathbf{x}$ :

$$J(\mathbf{x}) = (\mathbf{y} - \mathbf{Ax})^T \mathbf{R}^{-1} (\mathbf{y} - \mathbf{Ax}) + (\mathbf{x} - \mathbf{x}_o)^T \mathbf{W}^{-1} (\mathbf{x} - \mathbf{x}_o)$$

In the above,  $\mathbf{R}$  and  $\mathbf{W}$  are invertible, symmetric matrices known as the observation noise covariance matrix and the prior covariance matrix, respectively.  $\mathbf{x}_o$  represents a *prior* guess of  $\mathbf{x}$  (i.e., our best estimate of  $\mathbf{x}$  before any observations  $\mathbf{y}$  become available) and  $\mathbf{W}$  represents the uncertainty in this guess. The matrix  $\mathbf{R}$  represents the *noisiness* of the observations.