

# WORKSHEET #15, 10/16/07

MATH 54, FALL 2007

1. Suppose I have a list of data  $b_1, b_2, \dots, b_n$  and I want to approximate it by a number  $c$  (you could think of this as a constant function) such that  $\sum_i (b_i - c)^2$  is smallest.

(a) Let  $\vec{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$  and  $\vec{x} = [x]$ . Reinterpret the above problem in terms of finding the least-

squares solution to  $A\vec{x} = \vec{b}$ . What is the matrix  $A$ ?

(b) Find (in terms of the  $b_i$ 's) the least squares solution  $x^* = [c]$ . Is it what you would expect (i.e. does it make sense intuitively)?

Note: In this case the error  $\sum_i (b_i - c)^2$  is called the variance of the set of data and the distance from  $A[c]$  to  $\vec{b}$  (i.e.  $\|\vec{b} - A[c]\|$ ) is  $\sigma = \sqrt{\sum_i (b_i - c)^2}$ , which is known as the standard deviation.

2. (a) Write down the matrix for the transformation  $T$  from  $P_1$  to  $\mathbb{R}^4$  given by  $T(p(x)) = \begin{bmatrix} p(-1) \\ p(0) \\ p(1) \\ p(2) \end{bmatrix}$

with respect to the bases  $\beta := (1, x)$  for  $P_1$  and the standard basis  $\gamma := (e_1, e_2, e_3, e_4)$  for  $\mathbb{R}^4$ .

(b) Find the “best fit” (i.e. least-squares solution) line to the data  $(-1, 1), (0, 2), (1, 2), (2, 5)$ . (Hint: Is the matrix from part (a) relevant?)

(c) Make a sketch of your data and your best fit line.

3. Is  $\ker(A) = \ker(A^T A)$  necessarily true? (Hint: Can you use  $\ker(A^T) = \text{im}(A)^\perp$ ?)

4. (a) Show that if the columns of  $A$  are orthogonal and of unit length, then the least squares solution to  $A\vec{x} = \vec{b}$  is simply given by  $\vec{x}^* = A^T \vec{b}$ .

(b) Relate this to orthogonal projection to the subspace spanned by the columns of  $A$ .