

18.06 (Spring 14) Problem Set 5

This problem set is due Thursday, March 20, 2014 by 4pm in E17-131. The problems are out of the 4th edition of the textbook. This homework has 8 questions worth 100 points in total. Please WRITE NEATLY. You may discuss with others (and your TA), but you must turn in your own writing.

1. Least squares approximation: if $b = (5, 13, 17)$ at $t = (-1, 1, 2)$ find the best line $C + Dt$, and the vectors p and e . Check that $e^T p = 0$.

Solution:

$$A = \begin{bmatrix} 1 & -1 \\ 1 & 1 \\ 1 & 2 \end{bmatrix}, x = \begin{bmatrix} C \\ D \end{bmatrix}, b = \begin{bmatrix} 5 \\ 13 \\ 17 \end{bmatrix}$$

The system is solvable and has solution $x = \begin{bmatrix} 9 \\ 4 \end{bmatrix}$. Hence the best line is $9 + 4t$, which fits perfectly giving zero error $e = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$, and $p = b$.

2. Least squares approximation: If b takes values 0, 1, 3, respectively 4 at the corners $(x, y) = \{(1, 0), (0, 1), (-1, 0), (0, -1)\}$ of a square, find the best least squares fit to b by a plane $C + Dx + Ey$.

Solution:

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{bmatrix}, x = \begin{bmatrix} C \\ D \\ E \end{bmatrix}, b = \begin{bmatrix} 0 \\ 1 \\ 3 \\ 4 \end{bmatrix}$$

The system is **not** solvable. We solve $A^T A \hat{x} = A^T b$, where

$$A^T A = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}, A^T b = \begin{bmatrix} 8 \\ -3 \\ -3 \end{bmatrix}$$

which gives

$$\hat{x} = \begin{bmatrix} 2 \\ -3/2 \\ -3/2 \end{bmatrix}$$

So the best fitting plane is $2 - 3/2x - 3/2y$. (You can also check by running `lsqlin(A, b)` in MATLAB.)

3. Consider the reflection matrix $R = I - 2P$, where P is a projection.

- (a) Show that $R^2 = I$.
- (b) For vectors v in $C(P)$, what is Rv ?
- (c) For vectors w orthogonal to $C(P)$, what is Rw ?

Solution:

- (a) $R^2 = (I - 2P)^2 = I^2 - 2PI - 2IP + 4P^2 = I - 4P + 4P = I$. (We used the fact that $P^2 = P$, since P is a projection).
 - (b) Let $v = Px \in C(P)$. Then $Rv = (I - 2P)v = v - 2Pv = v - 2P(Px) = v - 2P^2x = v - 2Px = v - 2v = -v$.
 - (c) Let $w = (I - P)x \in C(P)^\perp$. Then $Rw = (I - 2P)w = w - 2Pw = w - 2P(I - P)x = w - 2Px + 2P^2x = w - 2Px + 2Px = w$.
4. If q_1, q_2, \dots, q_n are orthonormal vectors in \mathbf{R}^n , what combination of the q 's produces a given vector v ? How do you know that the q 's are a basis for \mathbf{R}^n ?

Solution:

$$v = \sum_{i=1}^n (q_i^T v) \cdot q_i$$

The vectors are orthonormal, hence linearly independent, so they form a basis. Another way to verify is to look at the projection matrix on the span of the q_i 's. Consider the matrix with the q_i 's as column vectors $A = [q_1 | \dots | q_n]$. Then $A^T A = I$. Hence the projection matrix onto their span is $P = A(A^T A)^{-1} A^T = A I^{-1} A^T = A A^T = (A^T A)^T = I$.

5. (a) In class, you may have seen that in the projection matrix formula $P = A(A^T A)^{-1} A^T$, we always want A to have linearly independent columns. Why?
- (b) Given an $m \times n$ matrix A ($m > n$), what is the projection matrix P_{null} onto the nullspace of A^T ?

Solution:

- (a) If the columns are not linearly independent, A does not have full column rank. So $A^T A$ is not full rank, and hence not invertible.
- (b) Recall that $N(A^T) \perp C(A)$. Let P be the projection on $C(A)$. Hence $P = A(A^T A)^{-1} A^T$. The projection onto the null space of A^T is perpendicular to that on the column space of A , hence $P_{null} = I - P = I - A(A^T A)^{-1} A^T$.