

Assignment 7

1. Ans: $A = [P_1, P_2, P_3] = \begin{bmatrix} 1 & 4 & 3 \\ 2 & -2 & 4 \\ 3 & -6 & -2 \\ 4 & -7 & 1 \end{bmatrix}$

The projection matrix $P = A(A^H A)^{-1} A^H$

The best approximation $\hat{x} = A \underline{z} = A(A^H A)^{-1} A^H x \approx \begin{bmatrix} 0.8946 \\ 2.9665 \\ 4.1247 \\ 5.6996 \end{bmatrix}$

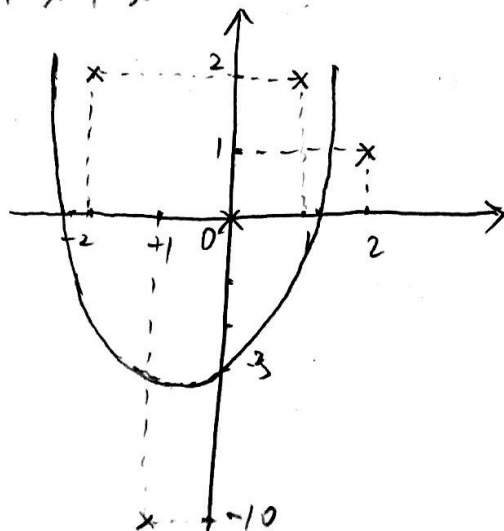
The projection of x onto the orthogonal complement of W should be error

between x and \hat{x} : $x - \hat{x} = \begin{bmatrix} 0.1054 \\ -0.9665 \\ -1.1247 \\ 1.3004 \end{bmatrix}$

2. Ans: $A = \begin{bmatrix} 1 & -2 & 4 \\ 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{bmatrix}$ $y = \begin{bmatrix} 2 \\ -10 \\ 0 \\ 2 \\ 1 \end{bmatrix}$

$\hat{z} = (A^H A)^{-1} A^H y = \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix}$

$\therefore y = -3 + x + x^2$



3. Ans: W is the subspace of polynomials with degree at most 2

$$\text{let } \underline{w}_1 = 1, \underline{w}_2 = x, \underline{w}_3 = x^2$$

For calculating Gramian Matrix:

$$\langle \underline{w}_1 | \underline{w}_1 \rangle = \int_0^1 dx = 1, \quad \langle \underline{w}_1 | \underline{w}_2 \rangle = \int_0^1 x dx = \frac{1}{2}, \quad \langle \underline{w}_1 | \underline{w}_3 \rangle = \frac{1}{3}$$

$$\langle \underline{w}_2 | \underline{w}_1 \rangle = \frac{1}{2}, \quad \langle \underline{w}_2 | \underline{w}_2 \rangle = \frac{1}{3}, \quad \langle \underline{w}_2 | \underline{w}_3 \rangle = \frac{1}{4}, \dots$$

$$\therefore G = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} \end{bmatrix}$$

$$\underline{v} = f(x) = e^x$$

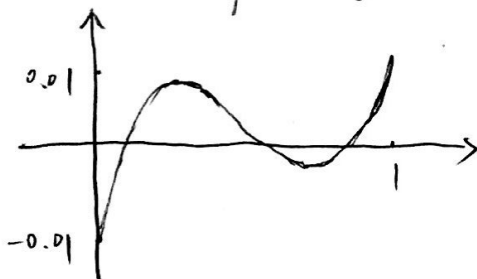
$$\therefore \langle \underline{v} | \underline{w}_1 \rangle = e^{-1}, \quad \langle \underline{v} | \underline{w}_2 \rangle = \int_0^1 x e^x dx = 1, \quad \langle \underline{v} | \underline{w}_3 \rangle = \int_0^1 x^2 e^x dx = e^{-2}$$

$$\text{we get } \underline{t} = \begin{bmatrix} e^{-1} \\ 1 \\ e^{-2} \end{bmatrix}$$

$$\therefore \underline{s} = G^{-1} \underline{t} = \begin{bmatrix} 1.01 \\ 0.65 \\ 0.64 \end{bmatrix}$$

$$\therefore \hat{f}(x) = 1.01 + 0.65x + 0.64x^2$$

$$\text{error} = f(x) - \hat{f}(x)$$



4. For showing that $I-P$ is an orthogonal projection matrix, we have to show it is idempotent and Hermitian: since P is an orthogonal projection matrix, $P^2 = P$ and $P^H = P$

$$(I-P)^2 = (I-P)(I-P) = I - IP - PI + P^2 = I - P - P + P = I - P$$

$$(I-P)^H = I^H - P^H = I - P$$

$$\text{range}(I-P) = \text{null}(P)$$

First, we show $\text{null}(P) \subseteq \text{range}(I-P)$. We have a vector \underline{v} such that $P\underline{v} = 0$. Then $(I-P)\underline{v} = \underline{v} - P\underline{v} = \underline{v}$. Thus, any \underline{v} in the nullspace of P is also in the range of $I-P$. Then, we show $\text{range}(I-P) \subseteq \text{nullspace}(P)$. If we have any $\underline{x} \in \text{range}(I-P)$, then $\underline{x} = (I-P)\underline{v}$ for some \underline{v} .

$$\underline{x} = \underline{v} - P\underline{v} = -(P\underline{v} - \underline{v})$$

$$P\underline{x} = -P(P\underline{v} - \underline{v}) = -P^2\underline{v} + P\underline{v}$$

$$\therefore \underline{x} \in \text{null}(P)$$

$$\therefore \text{range}(I-P) = \text{null}(P)$$

$$= -P\underline{v} + P\underline{v} = 0$$

Similarly, we can prove $\text{null}(I-P) = \text{range}(P)$

J. Ans:

$$(a) \text{cov}(x) \triangleq E[xx^T] + E[x]E[x]^T = \begin{bmatrix} 1 & 0.7 & 0.5 \\ 0.7 & 4 & 0.2 \\ 0.5 & 0.2 & 3 \end{bmatrix}$$

Because x is zero-mean vector, $E[x]E[x]^T = 0$

$$E[xx^T] = \begin{bmatrix} E[x_1x_1^T] & E[x_2x_1^T] & E[x_3x_1^T] \\ E[x_1x_2^T] & E[x_2x_2^T] & E[x_3x_2^T] \\ E[x_1x_3^T] & E[x_2x_3^T] & E[x_3x_3^T] \end{bmatrix} = \begin{bmatrix} 1 & 0.7 & 0.5 \\ 0.7 & 4 & 0.2 \\ 0.5 & 0.2 & 3 \end{bmatrix}$$

$$\hat{x}_1 = c_1x_2 + c_2x_3$$

$$\text{let } \underline{s} = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$$

$$G \cdot \underline{s} = \underline{t}$$

$$\begin{bmatrix} E[x_2x_2^T] & E[x_3x_2^T] \\ E[x_2x_3^T] & E[x_3x_3^T] \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} E[x_1x_2^T] \\ E[x_1x_3^T] \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 4 & 0.2 \\ 0.2 & 3 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 0.7 \\ 0.5 \end{bmatrix}$$

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 4 & 0.2 \\ 0.2 & 3 \end{bmatrix}^{-1} \begin{bmatrix} 0.7 \\ 0.5 \end{bmatrix} = \begin{bmatrix} 0.167 \\ 0.156 \end{bmatrix}$$

(b) the minimum mean-square error is

$$\|x_1 - \hat{x}_1\|^2 = E[x_1 x_1^T] - \mathbf{z}^H G^{-1} \mathbf{z} = 1 - \begin{bmatrix} 0.7 \\ 0.5 \end{bmatrix}^T \begin{bmatrix} 4 & 0.2 \\ 0.2 & 3 \end{bmatrix}^{-1} \begin{bmatrix} 0.7 \\ 0.5 \end{bmatrix}$$

$$\approx 1 - 0.195 = 0.805$$

(c) $E[x] = (1, 2, 3)^T$

Then $E[x] E[x]^T = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix}$

$$E[x x^T] = \begin{bmatrix} 1 & 0.7 & 0.5 \\ 0.7 & 4 & 0.2 \\ 0.5 & 0.2 & 3 \end{bmatrix} + \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix} = \begin{bmatrix} 2 & 2.7 & 3.5 \\ 2.7 & 8 & 6.2 \\ 3.5 & 6.2 & 12 \end{bmatrix}$$

$$\hat{x}_1 = c_1 x_2 + c_2 x_3$$

$$\begin{bmatrix} 8 & 6.2 \\ 6.2 & 12 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 2.7 \\ 3.5 \end{bmatrix} \Rightarrow \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 0.186 \\ 0.196 \end{bmatrix}$$

the minimum mean-square error = $2 - \begin{bmatrix} 2.7 \\ 3.5 \end{bmatrix}^T \begin{bmatrix} 8 & 6.2 \\ 6.2 & 12 \end{bmatrix}^{-1} \begin{bmatrix} 2.7 \\ 3.5 \end{bmatrix}$

$$\approx 0.813$$

6. (1) to show $R(A) \subseteq R(U)$

assume $\underline{w} = V \Sigma^{-1} \underline{s}$

then for $\forall \underline{a} = A \underline{w} \in R(A)$

$$\underline{a} = A \underline{w} = U \Sigma V^H V \Sigma^{-1} \underline{s} = U \underline{s}$$

thus $R(A) \subseteq R(U)$

(2) to show $R(U) \subseteq R(A)$

assume $\underline{s} = \Sigma V^H \underline{w}$

then for $\forall \underline{b} = U \underline{s} \in R(U)$

$$\underline{b} = U \underline{s} = U \Sigma V^H \underline{w} = A V \Sigma^{-1} \Sigma V^H \underline{w} = A \underline{w}$$

thus $R(U) \subseteq R(A)$

Therefore, $R(U) = R(A)$

$$P_A = P_U = U(U^H U)^{-1} U^H = U U^H$$