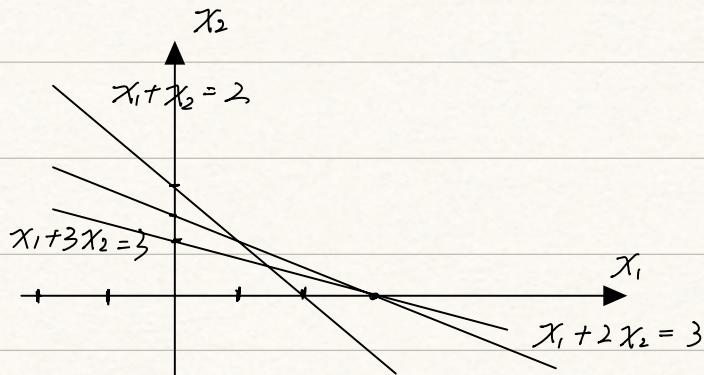


Homework 5

1. (a) $\begin{cases} x_1 + x_2 = 2 \\ x_1 + 2x_2 = 3 \\ x_1 + 3x_2 = 3 \end{cases}$

From last two equations, we can get $\begin{cases} x_1 = 3 \\ x_2 = 0 \end{cases}$

Since $x_1 + x_2 = 3 \neq 2$.



Thus, there exists not a single $(x_1, x_2) \in \mathbb{R}^2$ that satisfies the three linear equations all at the same time

1b) $\begin{cases} x_1 + x_2 = 2 \\ x_1 + 2x_2 = 3 \\ x_1 + 3x_2 = 3 \end{cases} \Rightarrow Ax = b \text{ where } A = \begin{pmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{pmatrix}, x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, b = \begin{pmatrix} 2 \\ 3 \\ 3 \end{pmatrix}$

$$\min_x \|Ax - b\|^2$$

$$A^T A x^* = A^T b \Rightarrow x^* = (A^T A)^{-1} A^T b$$

Since, $A^T A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 3 & 3 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{pmatrix} = \begin{pmatrix} 3 & 6 \\ 6 & 14 \end{pmatrix}$, $\det(A^T A) = 42 - 36 = 6$

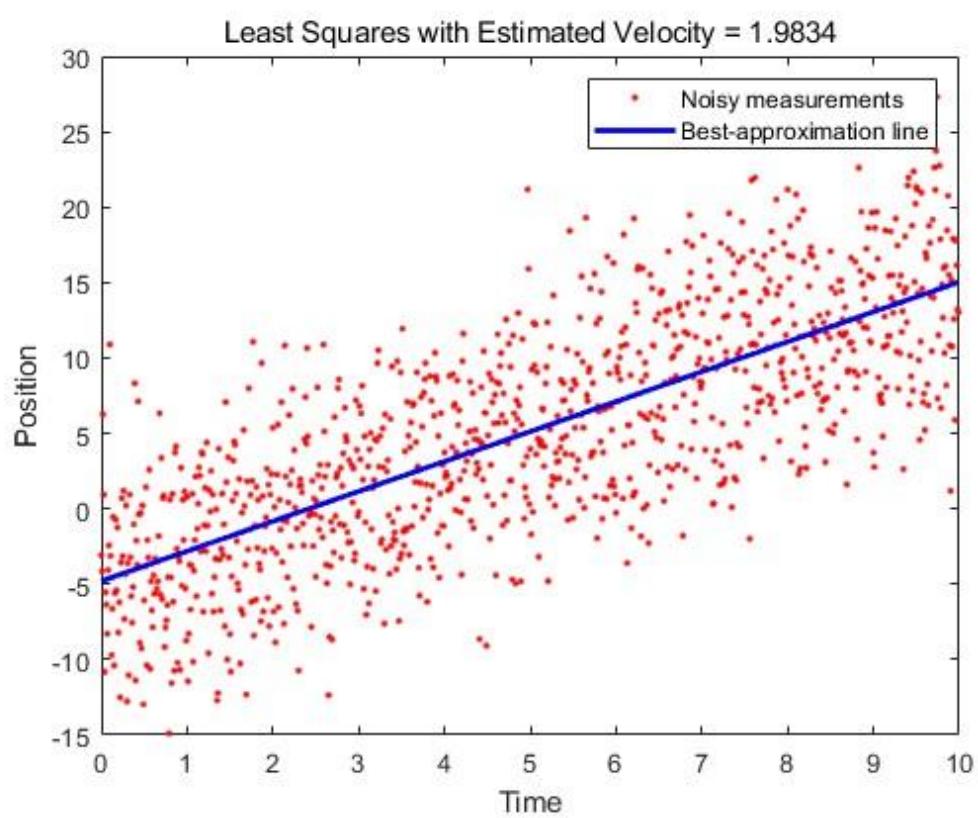
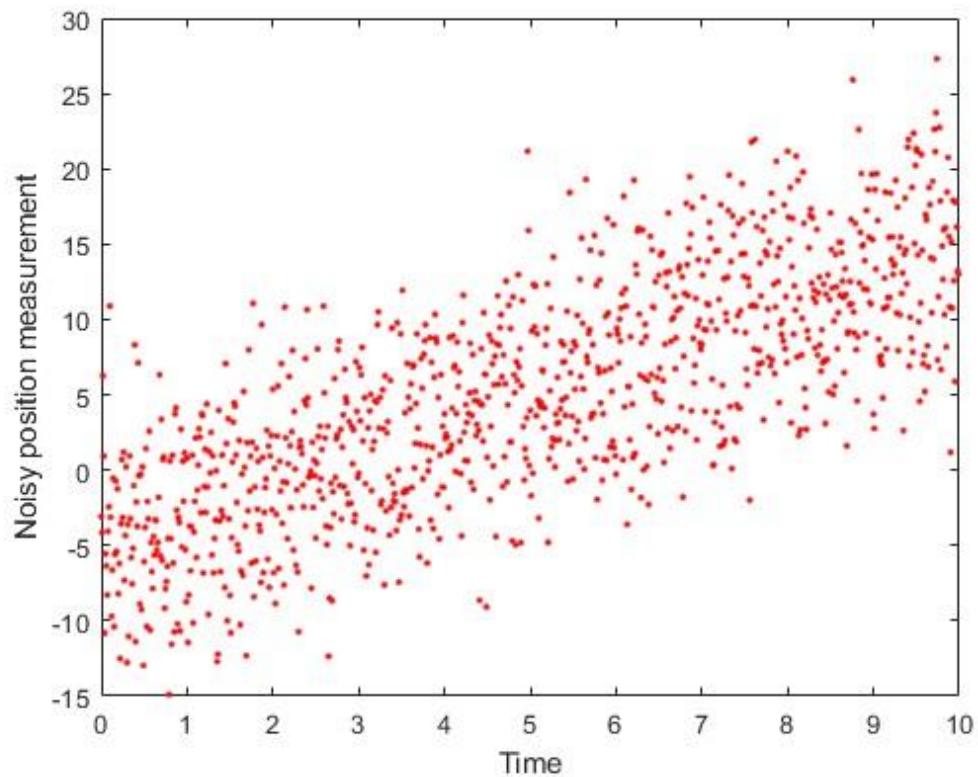
$$(A^T A)^{-1} = \frac{1}{6} \begin{pmatrix} 14 & -6 \\ -6 & 3 \end{pmatrix}$$

$$(A^T A)^{-1} A^T b = \frac{1}{6} \begin{pmatrix} 14 & -6 \\ -6 & 3 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} 2 \\ 3 \\ 3 \end{pmatrix}$$

$$= \frac{1}{6} \begin{pmatrix} 8 & -2 & -4 \\ -3 & 0 & 3 \end{pmatrix} \begin{pmatrix} 2 \\ 3 \\ 3 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 10 \\ 3 \end{pmatrix} = \begin{pmatrix} \frac{5}{3} \\ \frac{1}{2} \end{pmatrix}$$

Thus, $x^* = \begin{pmatrix} \frac{5}{3} \\ \frac{1}{2} \end{pmatrix}$

Problem2



```

clc;
clear;

load('least_squares_data.mat')
plot(0:dt:9.99, y_meas_noise,'r.');
xlabel('Time');
ylabel('Noisy position measurement');

% Create the time vector
time = 0:dt:9.99;

% Design matrix with a column of time values and a column of ones
X = [time', ones(length(time)), 1];

% Compute the least squares solution, beta=(X^T*X)^(-1)*X^T*y
beta = (X' * X) \ (X' * y_meas_noise');

% Extract the slope (velocity) and intercept from the solution
v = beta(1);
c = beta(2);

% Plot the noisy measurements
figure;
plot(time, y_meas_noise, 'r.', 'DisplayName', 'Noisy measurements');
hold on;

% Plot the best-fit line
y_fit = v * time + c;
plot(time, y_fit, 'b-', 'LineWidth', 2, 'DisplayName', 'Best-approximation line');

% Label the axes and add a legend
xlabel('Time');
ylabel('Position');
legend('show');
title(sprintf('Least Squares with Estimated Velocity = %.4f', v));

```

3. According to the question, $A \in \mathbb{R}^{n \times m}$ is a matrix with m linearly independent columns that constitute a basis for a subspace $X \subset \mathbb{R}^n$ of dimension m , and $\text{range}(A) = \{x \in \mathbb{R}^n : x = Au \text{ for some } u \in \mathbb{R}^m\}$

$\forall x \in \text{range}(A)$, $\exists u \in \mathbb{R}^m$ s.t. $x = Au$, which means that x is a linear combination of the columns of A , where the coefficients of this combination are given by the components of u . Thus, $x \in X$

Therefore, $\text{range}(A) \subseteq X$

$\forall x \in X$, where x is the linear combination of a basis of X . Since the columns of A form a basis for X . Thus, x can be expressed as Au for some $u \in \mathbb{R}^m$. Thus, $x \in \text{range}(A)$

Therefore, $X \subseteq \text{range}(A)$

Thus, $X = \text{range}(A)$

$$4.(a) X^\perp = \{y \in \mathbb{R}^n : x^T y = 0 \text{ for all } x \in X\}$$

$$0 \in \mathbb{R}^n, x^T 0 = 0 \text{ for all } x \in X, \text{ Thus, } 0 \in X^\perp$$

Suppose $y_1, y_2 \in X^\perp$. Then, for any $x \in X$, we have $x^T y_1 = 0$, $x^T y_2 = 0$

$$x^T(y_1 + y_2) = x^T y_1 + x^T y_2 = 0 + 0 = 0, \text{ Thus, } y_1 + y_2 \in X^\perp$$

Suppose $y \in X^\perp$ and α is a scalar, $x^T y = 0$. For any $x \in X$, we have

$$x^T(\alpha y) = \alpha x^T y = 0. \text{ Thus, } \alpha y \in X^\perp$$

Therefore, X^\perp have a vector space structure and is a subspace of \mathbb{R}^n

(b) Prove that $X \subset (X^\perp)^\perp$

$\forall x \in X$, $X^\perp = \{y \in \mathbb{R}^n : x^T y = 0 \text{ for all } x \in X\}$, we know that

$$x^T y = 0, \forall y \in X^\perp$$

This means that $x \in (X^\perp)^\perp$. This implies that $X \subset (X^\perp)^\perp$

Prove that $(X^\perp)^\perp \subset X$

From the projection theorem, we know that for any $y \in \mathbb{R}^n$, we have that

$$(y - x^*) \in X^\perp \text{ where } x^* = P_X y$$

Consider any $y \in (X^\perp)^\perp$, then we must have $y^T(y - P_X y) = 0$

Note that we also have $y - P_X y \in X^\perp$ and $P_X y \in X$

$$\text{Thus, } (P_X y)^\perp (y - P_X y) = 0$$

$$0 = y^T(y - P_X y) - (P_X y)^\perp (y - P_X y) = \|y - P_X y\|_2^2 \implies y - P_X y \in X$$

Hence, any vector in $(X^\perp)^\perp$ is also in X , which means that $(X^\perp)^\perp \subset X$

Therefore, $(X^\perp)^\perp = X$

5. $X = \{x \in \mathbb{R}^3 : x_1 + x_2 + x_3 = 0\}$. This is a plane through the origin in \mathbb{R}^3

$\text{range}(A) = \{x \in \mathbb{R}^n : x = Au \text{ for some } u \in \mathbb{R}^m\}$, where u is coefficient vector for column vector of A

$$A_1 = \begin{pmatrix} -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{pmatrix}, \text{ column vector } \vec{a}_1 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \quad \vec{a}_2 = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}, \quad a_{11} + a_{12} + a_{13} = 0, \quad a_{21} + a_{22} + a_{23} = 0$$

Since both columns of A_1 satisfy the equation $x_1 + x_2 + x_3 = 0$. This means \vec{a}_1 and \vec{a}_2 are on

the plane X . Also, \vec{a}_1 and \vec{a}_2 are linearly independent. Thus, $\text{range}(A_1) = X$

$$A_2 = \begin{pmatrix} -1 & -2 \\ 1 & 1 \\ 0 & 1 \end{pmatrix}, \text{ column vector } \vec{a}_1 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \quad \vec{a}_2 = \begin{pmatrix} -2 \\ 1 \\ 1 \end{pmatrix}, \quad a_{11} + a_{12} + a_{13} = 0, \quad a_{21} + a_{22} + a_{23} = 0$$

Since both columns of A_2 satisfy the equation $x_1 + x_2 + x_3 = 0$. This means \vec{a}_1 and \vec{a}_2 are on

the plane X . Also, \vec{a}_1 and \vec{a}_2 are linearly independent. Thus, $\text{range}(A_2) = X$

Therefore, $X = \text{range}(A_1) = \text{range}(A_2)$

$$P_X = A(A^T A)^{-1} A^T$$

$$\text{Suppose } P_{X_1} = A_1(A_1^T A_1)^{-1} A_1^T$$

$$A_1^T A_1 = \begin{pmatrix} -1 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix} \begin{pmatrix} -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$$

$$(A_1^T A_1)^{-1} = \frac{1}{4-1} \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}$$

$$P_{X_1} = \begin{pmatrix} -1 & -1 \\ 0 & 0 \end{pmatrix} \frac{1}{3} \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} -1 & -1 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix}$$

$$\text{Suppose } P_{X_2} = A_2 (A_2^T A_2)^{-1} A_2^T$$

$$A_2^T A_2 = \begin{pmatrix} -1 & 1 & 0 \\ -2 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} -1 & -2 \\ 1 & 1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 3 \\ 3 & 6 \end{pmatrix}$$

$$(A_2^T A_2)^{-1} = \frac{1}{12-9} \begin{pmatrix} 6 & -3 \\ -3 & 2 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 6 & -3 \\ -3 & 2 \end{pmatrix}$$

$$P_{X_2} = A_2 (A_2^T A_2)^{-1} A_2^T = \begin{pmatrix} -1 & -2 \\ 1 & 1 \\ 0 & 1 \end{pmatrix} \frac{1}{3} \begin{pmatrix} 6 & -3 \\ -3 & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 & 0 \\ 2 & 1 & 1 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 0 & -1 \\ 3 & -1 \\ -3 & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 & 0 \\ 2 & 1 & 1 \end{pmatrix}$$

$$= \frac{1}{3} \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix}$$

$$\text{Thus, } P_{X_1} = P_{X_2}$$

Therefore, the resulting P_X in those two different cases are identical.

6. The line can be expressed as an underdetermined linear system $Ax = b$

$$A = \begin{pmatrix} 1 & 1 & 1 \\ -1 & -1 & 1 \end{pmatrix}, \quad x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}, \quad b = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \iff \begin{cases} x_1 + x_2 + x_3 = 1 \\ -x_1 - x_2 + x_3 = 0 \end{cases}$$

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

$$A A^T = \begin{pmatrix} 1 & 1 & 1 \\ -1 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & -1 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix}$$

$$(A A^T)^{-1} = \frac{1}{9-1} \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} = \frac{1}{8} \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix}$$

$$x^* = A^T (A A^T)^{-1} b = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \frac{1}{8} \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \frac{1}{8} \begin{pmatrix} 2 & -2 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \frac{1}{8} \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$

$$= \frac{1}{4} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$