

See 5.3 Least Square Problems

Given an $m \times n$ matrix A and a vector $\bar{b} \in \mathbb{R}^m$
the system $A\bar{x} = \bar{b}$ may be inconsistent
ie. there may not be a $\bar{x} \in \mathbb{R}^n$ such that
 $A\bar{x} - \bar{b} = \bar{0}$

Question: How to find a $\bar{x} \in \mathbb{R}^n$ such that
 $A\bar{x}$ is closest to \bar{b} ?

eg. $A = \begin{bmatrix} 1 & 1 \\ -2 & 3 \\ 2 & -1 \end{bmatrix}$, $\bar{b} = \begin{bmatrix} 3 \\ 1 \\ 2 \end{bmatrix}$ Consider $A\bar{x} = \bar{b}$

$$\left[\begin{array}{cc|c} 1 & 1 & 3 \\ -2 & 3 & 1 \\ 2 & -1 & 2 \end{array} \right] \rightarrow \left[\begin{array}{cc|c} 1 & 1 & 3 \\ 0 & 5 & 7 \\ 0 & 2 & 3 \end{array} \right] \rightarrow \left[\begin{array}{cc|c} 1 & 1 & 3 \\ 0 & 1 & \frac{7}{5} \\ 0 & 0 & 3 - \frac{14}{5} \end{array} \right]$$

inconsistent.

Thm. If A is an $m \times n$ matrix of rank n , then
the linear system $A^T A \bar{x} = A^T \bar{b}$ has a solution \hat{x}
such that

$\|A\bar{x} - \bar{b}\|$ obtains its minimum at \hat{x} .

\hat{x} is called the least squares solution of $A\bar{x} = \bar{b}$.

$$A^T A = \begin{bmatrix} 1 & -2 & 2 \\ 1 & 3 & -1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 1 \\ -2 & 3 \\ 2 & -1 \end{bmatrix} = \begin{bmatrix} 9 & -7 \\ -7 & 11 \end{bmatrix}$$

$$A^T \bar{b} = \begin{bmatrix} 1 & -2 & 2 \\ 1 & 3 & -1 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$$

We want to consider the new system. $A^T A \vec{x} = A^T \vec{b}$

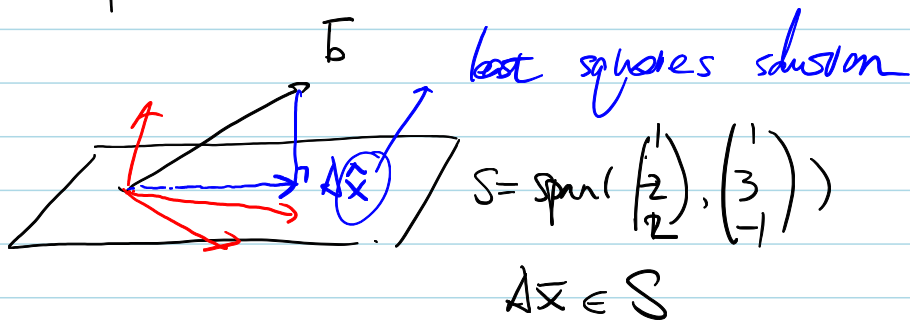
i.e.,
$$\begin{bmatrix} 9 & -7 \\ -7 & 11 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$$

$$\left[\begin{array}{cc|c} 9 & -7 & 5 \\ -7 & 11 & 4 \end{array} \right] \rightarrow \left[\begin{array}{cc|c} 9 & -7 & 5 \\ 0 & \frac{50}{9} & \frac{71}{9} \end{array} \right] \cdot \frac{7}{9}$$

$11 - \frac{49}{9} \quad 4 + \frac{35}{9}$

$$\begin{cases} 9x_1 - 7x_2 = 5 \\ \frac{50}{9}x_2 = \frac{71}{9} \end{cases} \Rightarrow \begin{cases} x_1 = \frac{83}{50} \\ x_2 = \frac{71}{50} \end{cases}$$

Geometric explanation



One can prove that

$A\hat{x}$ is the projection of \vec{b} onto the plane.

$$\hat{x} = (A^T A)^{-1} \cdot A^T \vec{b}$$

$A\hat{x} = \underbrace{A(A^T A)^{-1} A^T}_{P} \vec{b}$ is the projection of \vec{b} onto the plane

The matrix $P = A(A^T A)^{-1} A^T$ is called the projection matrix.

P satisfies

$$P^2 = P \quad \text{and} \quad P^T = P$$

Proof.

$$\begin{aligned} P^2 &= P \cdot P = A \underbrace{(A^T A)^{-1} \cdot A^T}_{I} \cdot A (A^T A)^{-1} A^T \\ &= A \cdot I \cdot (A^T A)^{-1} A^T \\ &= A \cdot (A^T A)^{-1} A^T = P \end{aligned}$$

$$\begin{aligned} P^T &= \left(A \underbrace{(A^T A)^{-1} \cdot A^T}_{I} \right)^T \\ &= (A^T)^T \cdot \left((A^T A)^{-1} \right)^T \cdot A^T \\ &= A \cdot \left((A^T A)^T \right)^{-1} \cdot A^T \\ &= A (A^T A)^{-1} A^T \\ &= P. \end{aligned}$$

we use the fact that
 $(B^{-1})^T = (B^T)^{-1}$ for any
nonsingular B .