

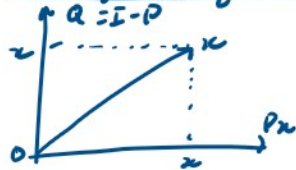
Least Squares

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Pythagoras Theorem

Let $x \in \mathbb{R}^n$, $P \rightarrow$ orthogonal projection matrix, $Q = I - P$.

$$\therefore \|x\|_2^2 = \|Px\|_2^2 + \|Qx\|_2^2$$



$$\begin{aligned} \|x\|_2^2 &= \langle x, x \rangle = \langle Ix, Ix \rangle \\ &= \langle (P+Q)x, (P+Q)x \rangle \\ &= \langle Px, Px \rangle + \langle Px, Qx \rangle + \langle Qx, Px \rangle + \langle Qx, Qx \rangle \\ &= \|Px\|_2^2 + 2\langle Px, Qx \rangle + \|Qx\|_2^2 \end{aligned}$$

$$\text{If } \langle Px, Qx \rangle = 0, \langle Px, Qx \rangle = Px^T Qx = x^T P^T Qx$$

$$P^T = P \quad \underbrace{x^T P Q x}_0 = 0 \rightarrow \|x\|_2^2 = \|Px\|_2^2 + \|Qx\|_2^2$$

1.3 Least Squares & Projection onto $R(A)$

$Ax = b$ has soln. if $b \in R(A)$. What if $b \notin R(A)$? ($Ax = \vec{b} \rightarrow$ no solution)

Find vector x s.t. $Ax \in R(A)$ & Ax is as close as possible to \vec{b} in $R(A)$

\rightarrow Find vector x_0 s.t. $Ax_0 \in R(A)$ & $\|Ax_0 - b\|$ is minimized

$\rightarrow Ax_0 = \text{proj}_{R(A)} b$, $P_{R(A)}$ is proj matrix onto $R(A)$

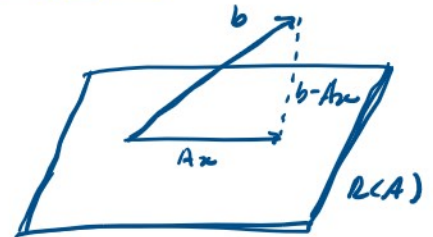
$$\vec{0}b = \underbrace{(I - P)}_{= b - Ax_0} \vec{b} \perp R(A) \rightarrow b - Ax_0 \perp R(A)$$

$$R(A)^T = N(A)^T \rightarrow Ax_0 - b \in R(A)^T = N(A)^T$$

$$\rightarrow A^T(Ax_0 - b) = 0$$

$$\rightarrow A^T Ax_0 = A^T b \rightarrow \text{Least Squares equation}$$

\therefore every x_0 satisfying eqn. = least square solution.
 $Ax_0 = \text{proj of } b \text{ onto } R(A)$



Properties

- L.S.E. always has a solution

$$\rightarrow A^T Ax = A^T b \text{ has soln. if } A^T b \in R(A^T A). R(A^T) = R(A^T A) \therefore N(A) = N(A^T A)$$

- If $A^T A$ is invertible, least square soln. is unique

$$\rightarrow A^T Ax_0 = A^T b \quad x_0 = (A^T A)^{-1} A^T b \therefore \text{unique soln.}$$

- let A be an orthogonal projection matrix ($A^2 = A$, $A^T = A$)

$$\begin{aligned} \text{LSE: } A^T Ax &= A^T b & \rightarrow & \quad A^T Ax = A^T b \\ A^T &= A & \rightarrow & \quad A^2 x = Ab \\ A^2 &= A & \rightarrow & \quad Ax = Ab \end{aligned}$$

$$\therefore x = Ab \rightarrow \text{soln. of least square eqn.}$$

$$\text{ex. } A = \begin{bmatrix} 0 & 1 \\ 0 & -1 \\ 0 & 0 \end{bmatrix} \quad b = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \quad \text{is } b \in R(A)?$$

→ no soln, for $b = Ax \therefore$ Find L.S. Soln.

$$Ax = b \rightarrow A^T Ax = A^T b$$

$$AA = \begin{bmatrix} 1 & 0 & 0 \\ 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & -1 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \quad A^T b = \begin{bmatrix} 1 & 0 & 0 \\ 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\therefore A^T Ax = A^T b \rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} x = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$\therefore \det \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} = 2 - 1 \neq 0$, $\begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \rightarrow$ invertible.

$$\therefore x_{LS} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \frac{1}{2-1} \begin{bmatrix} 2 & -1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 3 \\ -2 \end{bmatrix} \rightarrow \text{unique soln. for } A^T Ax = A^T b$$

$$\text{ls dist. b/w } Ax_{LS} \text{ \& } b : Ax_{LS} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ -2 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

$$\| \begin{bmatrix} 3 \\ 1 \end{bmatrix} - \begin{bmatrix} 1 \\ 2 \end{bmatrix} \|_2 = \sqrt{9} = 3 \quad \therefore \text{min. dist. b/w } b \text{ \& } R(A) = 3$$

or Proj. onto $R(A) : P = A(A^T A)^{-1} A^T$

$$Pb = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

- If $A^T A$ not invertible:

- Cols of $A =$ lin. dependent

Find proj. onto $R(A)$:

i) Find basis for $R(A) \rightarrow$ G.E.

ii) Write basis vectors as column vectors in matrix \tilde{A}

iii) \therefore columns of $\tilde{A} =$ lin. independent, $\tilde{A}^T \tilde{A} =$ invertible $\&$ proj. onto $R(\tilde{A})$:

$$\rightarrow P_{R(\tilde{A})} = \tilde{A}(\tilde{A}^T \tilde{A})^{-1} \tilde{A}^T$$

\therefore columns of \tilde{A} form basis for $R(A)$, $R(\tilde{A}) = R(A)$

$$\& P_{R(A)} = \tilde{A}(\tilde{A}^T \tilde{A})^{-1} \tilde{A}^T$$

Least Square equation always has a soln.

→ Prove that $N(A) = N(A^T A)$

$$N(A) \subseteq N(A^T A) : x \in N(A), \text{ then } Ax = 0 \rightarrow A A^T x = 0 \rightarrow x \in N(A^T A)$$

$$N(A^T A) \subseteq N(A) : x \in N(A^T A), \text{ then } A^T A x = 0 \rightarrow Ax \in N(A^T) = [R(A)]^\perp \\ \rightarrow Ax \in [R(A)]^\perp \quad Ax \perp R(A) \quad Ax \perp \{Ay : y \in \mathbb{R}^m\} \quad Ax \perp Ay \text{ for } y \in \mathbb{R}^m \\ Ax \perp Ax \rightarrow Ax = 0 \quad x \in N(A)$$