

EE 205003 Session 18

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Ch 4.3 Least Squares Approximations

Least Square Approximation

Issue : It often happens that $A\mathbf{x} = \mathbf{b}$ has no solution
($m > n$, $\mathbf{C}(A)$ only spans a small part of \mathbb{R}^m .
If $\mathbf{b} \notin \mathbf{C}(A)$, no solution)

Q: Do we stop here ?

No! measurement includes noise.

Instead, we try to find the "best solution"

To repeat : we cannot always get error $\mathbf{e} = \mathbf{b} - A\mathbf{x}$ down to zero

When $\mathbf{e} = \mathbf{0}$, \mathbf{x} is exact solution to $A\mathbf{x} = \mathbf{b}$

When \mathbf{e} is as small as possible,

$\hat{\mathbf{x}}$ is the least square solution, or "best solution"

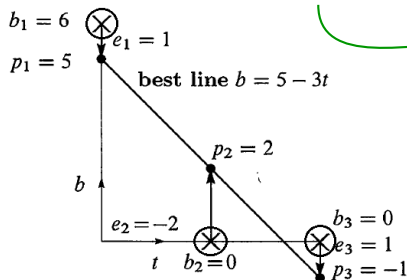
($\mathbf{p} = A\hat{\mathbf{x}}$ is the projection of \mathbf{b} onto $\mathbf{C}(A)$.)

To find the "best solution", we solve $A^T A \hat{\mathbf{x}} = A^T \mathbf{b}$.)

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Ex : Fitting a line (linear regression)

Find the closest line through $(0, 6)$, $(1, 0)$, $(2, 0)$



$$(b = C + Dt)$$

If $(0, 6)$ on the line,
 $C + D \cdot 0 = 6$

If $(1, 0)$ on the line,
 $C + D \cdot 1 = 0$

If $(2, 0)$ on the line,
 $C + D \cdot 2 = 0$

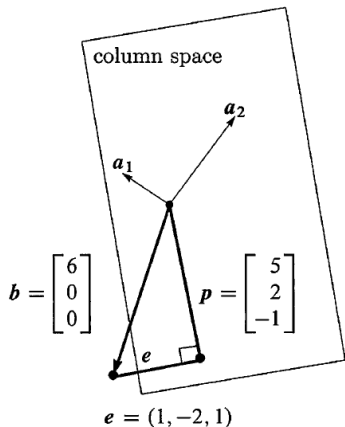
errors = vertical distances to line

3 equations, 2 unknowns : $A\mathbf{x} = \mathbf{b}$

$$\Rightarrow A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \mathbf{x} = \begin{bmatrix} C \\ D \end{bmatrix} \mathbf{b} = \begin{bmatrix} 6 \\ 0 \\ 0 \end{bmatrix}$$

(not in $\mathbf{C}(A)$, no solution)

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By Geometry

Every Ax lies on the plane $C(A)$
want to find the point closest to b
 \Rightarrow The nearest point is projection

$$P = A\hat{x}$$

Normal equation : $A^T A \hat{x} = A^T b$

(same as Ex3 in session 17, we already computed $\hat{x} = \begin{bmatrix} 5 \\ -3 \end{bmatrix}$)

$\Rightarrow b = 5 - 3t$ is the "best" line

(linear regression works if no outlier)

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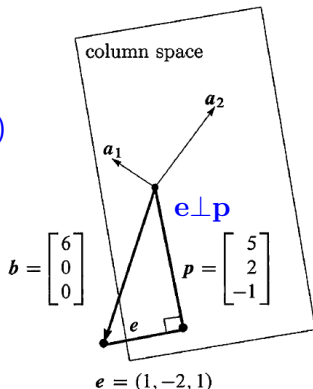
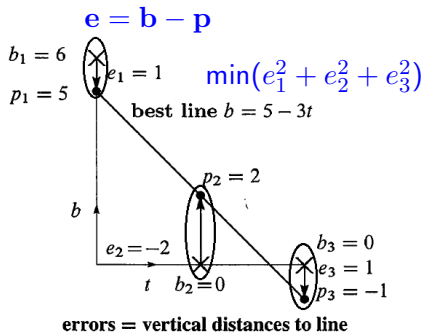



Figure 4.6: **Best line and projection: Two pictures, same problem.** The line has heights $p = (5, 2, -1)$ with errors $e = (1, -2, 1)$. The equations $A^T A \hat{x} = A^T b$ give $\hat{x} = (5, -3)$. The best line is $b = 5 - 3t$ and the projection is $p = 5a_1 - 3a_2$.

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By Algebra

Every $\mathbf{b} = \mathbf{p} + \mathbf{e}$
 $\mathbf{p} \in \mathbf{C}(A) \quad \mathbf{e} \in \mathbf{N}(A^\top)$

orthogonal components

$A\mathbf{x} = \mathbf{b} = \mathbf{p} + \mathbf{e}$ $A\hat{\mathbf{x}} = \mathbf{p}$
(impossible) (solvable)
 (by removing \mathbf{e})

For all \mathbf{x} ,
 $\|A\mathbf{x} - \mathbf{b}\|^2 = \|A\mathbf{x} - \mathbf{p} - \mathbf{e}\|^2 = \|A\mathbf{x} - \mathbf{p}\|^2 + \|\mathbf{e}\|^2$
 $\hat{\mathbf{x}}$ makes $\|A\mathbf{x} - \mathbf{p}\|^2 = 0$ $\mathbf{p} \in \mathbf{C}(A) \quad \mathbf{e} \in \mathbf{N}(A^\top)$
this leaves the smallest possible error \mathbf{e}

Fact The least square sol. $\hat{\mathbf{x}}$ makes
 $E = \|A\mathbf{x} - \mathbf{b}\|^2$ as small as possible

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By Calculus

$$E = \|A\mathbf{x} - \mathbf{b}\|^2 = (C + D \cdot 0 - 6)^2 \\ + (C + D \cdot 1 - 0)^2 \\ + (C + D \cdot 2 - 0)^2$$

$$\frac{\partial E}{\partial C} = 2(C + D \cdot 0 - 6) + 2(C + D \cdot 1) + 2(C + D \cdot 2) = 0$$

$$\frac{\partial E}{\partial D} = 2(C + D \cdot 0 - 6)(0) + 2(C + D \cdot 1)(1) + 2(C + D \cdot 2)(2) = 0$$

$$\Rightarrow \begin{array}{l} 3C + 3D = 6 \\ 3C + 5D = 0 \end{array} \quad \begin{bmatrix} 3 & 3 \\ 3 & 5 \end{bmatrix} \text{ is } A^T A$$

$$(\text{same as } A^T A \mathbf{x} = A^T \mathbf{b}) \quad \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix}$$

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Fact The partial derivatives of $\|A\mathbf{x} - \mathbf{b}\|^2$ are zero
when $A^T A \hat{\mathbf{x}} = A^T \mathbf{b}$

Recall :

$b = 5 - 3t$ is the "best" line

$$\begin{aligned} t = 0, P_1 &= 5 - 0 = 5 \\ t = 1, P_2 &= 5 - 3 = 2 \\ t = 2, P_3 &= 5 - 6 = -1 \end{aligned} \Rightarrow \mathbf{p} = \begin{bmatrix} 5 \\ 2 \\ -1 \end{bmatrix}$$

$$\Rightarrow \mathbf{e} = \mathbf{b} - \mathbf{p} = \begin{bmatrix} 6 \\ 0 \\ 0 \end{bmatrix} - \begin{bmatrix} 5 \\ 2 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} \quad \mathbf{e} \perp \mathbf{p} \text{ (} \perp \text{ cols of } A \text{)}$$

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The Big Picture

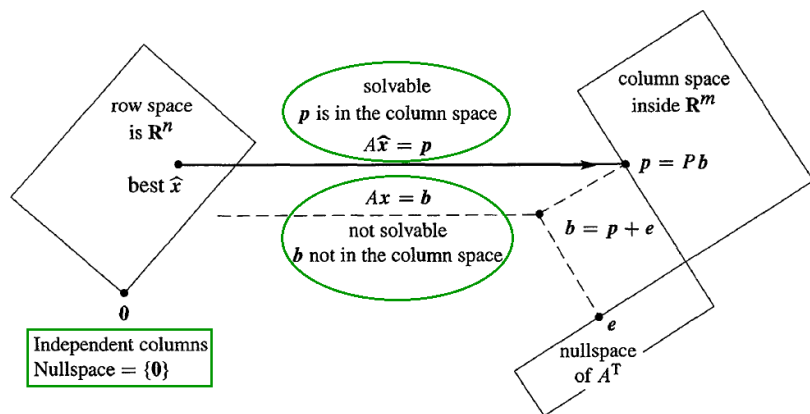


Figure 4.7: The projection $p = A\hat{x}$ is closest to b , so \hat{x} minimizes $E = \|b - Ax\|^2$.

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Recall :

If A has independent col.s, then AA^T invertible

\Rightarrow we can solve for least square solution $\hat{\mathbf{x}}$

\Rightarrow we can use linear regression to find approximate solution to unsolvable $A\mathbf{x} = \mathbf{b}$

(col.s of A is guaranteed to be independent if they are orthonormal)

(Topic for next session)