

Linear algebra for AI & ML

$$Ax = b$$

$$A \in \mathbb{R}^{m \times n}, \quad b \in \mathbb{R}^m$$

$$r = (Ax - b) \in \mathbb{R}^m$$

$$r = \begin{pmatrix} r_1 \\ r_2 \\ \vdots \\ r_m \end{pmatrix}$$

$$\min_{x \in \mathbb{R}^n} \|r\|_2^2 = \min_{x \in \mathbb{R}^n} r_1^2 + r_2^2 + \dots + r_m^2$$

$$\boxed{Ax = b}$$

$$\|Ax - b\|$$

$$A = [a_1 \dots a_n] \Rightarrow Ax = a_1 x_1 + a_2 x_2 + \dots + a_n x_n$$

$$\|Ax - b\|_2^2 = \|a_1 x_1 + a_2 x_2 + \dots + a_n x_n - b\|_2^2$$

$$A = \begin{bmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_m^T \end{bmatrix} \Rightarrow Ax = \begin{bmatrix} a_1^T x \\ \vdots \\ a_m^T x \end{bmatrix}$$

$$\|Ax - b\|_2^2 = \left\| \begin{bmatrix} a_1^T x - b_1 \\ \vdots \\ a_m^T x - b_m \end{bmatrix} \right\|_2^2$$

$$\|Ax - b\|_2^2 = \sum_{i=1}^m (\tilde{a}_i^T x - b_i)^2 = \sum_{i=1}^m x_i^2$$

standing assumption: $A \in \mathbb{R}^{m \times n}$ such that columns of A are linearly independent.

$$f = \|x\|_2^2 = \|Ax - b\|_2^2$$

We want to $\min_{x \in \mathbb{R}^n} f(x) = \min_{x \in \mathbb{R}^n} \|Ax - b\|_2^2$

$$\nabla f(x) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix} = 0$$

$$f(x) = \|Ax - b\|_2^2 = \sum_{i=1}^m \left(\sum_{j=1}^n A_{ij} x_j - b_i \right)^2$$

$$\frac{\partial f}{\partial x_k} = \sum_{i=1}^m 2 \left(\sum_{j=1}^n A_{ij} x_j - b_i \right) (A_{ik})$$

$$= \sum_{i=1}^m 2 (A^T)_{ki} (Ax - b)_i$$

$$= 2 (A^T (Ax - b))_k$$

$$\nabla_x f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix} = 2 A^T (Ax - b)$$

\hat{x} which minimizes $\|Ax - b\|_2^2$ satisfies

$$\nabla_x f(\hat{x}) = 0$$

$$\Rightarrow 2A^T(A\hat{x} - b) = 0$$

$$\Rightarrow A^T A \hat{x} = A^T b$$

← normal eq^{ns}

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

$$= \boxed{\hat{x} = A^+ b}$$

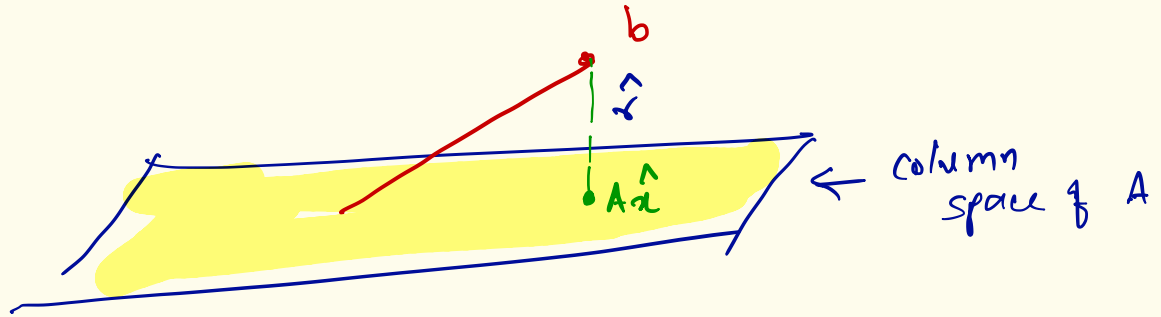
Recall: $Ax = b$ & left inverse exists.

$$CA = I$$

$$Ax = b$$

$$\text{set } \hat{x} = Cb$$

$$Ax = b$$



\hat{r} : residual vector at \hat{x}
 $= A\hat{x} - b$

\hat{r} is orthogonal to the column space of A .
 For any $z \in \mathbb{R}^n$, $Az \perp \hat{r}$
 $\Rightarrow (Az)^T \hat{r} = 0$

$$\Rightarrow z^T A^T (A\hat{x} - b) = 0$$

$$\forall z_0 \in \mathbb{R}^n$$

$$\Rightarrow A^T (A\hat{x} - b) = 0$$

$$\Rightarrow A^T A \hat{x} = A^T b$$

$$A \in \mathbb{R}^{m \times n}$$

$$b \in \mathbb{R}^m$$

Find \hat{x} s.t. $\|A\hat{x} - b\|_2^2$ is minimum.

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \\ \color{red}{a_{m+1,1}} & \color{red}{a_{m+1,2}} & \dots & \color{red}{a_{m+1,n}} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \\ \color{red}{x_{m+1}} \end{bmatrix}$$

$\color{red}{A}$

$$\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \\ \color{red}{b_{m+1}} \end{bmatrix}$$

$\color{red}{b}$

$$\hat{x}_{m+1} = (\tilde{A}^T \tilde{A})^{-1} \tilde{A} \tilde{b}$$

$$= \left(\begin{bmatrix} A \\ \tilde{a}_{m+1}^T \end{bmatrix}^T \begin{bmatrix} A \\ \tilde{a}_{m+1}^T \end{bmatrix} \right)^{-1} \begin{bmatrix} A \\ \tilde{a}_{m+1}^T \end{bmatrix}^T \begin{bmatrix} b \\ b_{m+1} \end{bmatrix}$$

$$= \left(\begin{bmatrix} A^T & \tilde{a}_{m+1} \end{bmatrix} \begin{bmatrix} A \\ \tilde{a}_{m+1}^T \end{bmatrix} \right)^{-1} \begin{bmatrix} A^T & \tilde{a}_{m+1} \end{bmatrix} \begin{bmatrix} b \\ b_{m+1} \end{bmatrix}$$

$$\hat{x}_{m+1} = (A^T A + \tilde{a}_{m+1} \tilde{a}_{m+1}^T)^{-1} (A^T b + \tilde{a}_{m+1} b_{m+1})$$

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

Matrix inversion lemma.

$$(A + UCB)^{-1} = A^{-1} - A^{-1}U(C + BA^T U)^{-1}BA^T A^{-1}$$

$$(A^T A + \tilde{a}_{m+1} \tilde{a}_{m+1}^T)^{-1} = (A^T A)^{-1} - (A^T A)^{-1} \tilde{a}_{m+1} \left(1 + \tilde{a}_{m+1}^T (A^T A)^{-1} \tilde{a}_{m+1} \right)^{-1}$$