

Example 23, 25, 27

$$A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 1 & -1 \end{pmatrix}$$

~~$$A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 1 & -1 \end{pmatrix}$$~~

$$\text{range}(A) = \begin{pmatrix} \beta_1 + \beta_2 \\ \beta_1 \\ \beta_1 - \beta_2 \end{pmatrix}, \quad \text{Null}(A^T) = \begin{pmatrix} \beta_3 \\ -2\beta_3 \\ \beta_3 \end{pmatrix}$$

~~$$A \in \mathbb{R}^{3 \times 2}$$~~

with $\beta_1, \beta_2 \in \mathbb{R}$

with $\beta_3 \in \mathbb{R}$

$$Ax = b = \begin{pmatrix} 2 \\ -2 \\ 0 \end{pmatrix}$$

$$\begin{cases} \beta_1 = 0 \\ \beta_2 = 1 \end{cases} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \in \text{range}(A)$$

$$\beta_3 = 1 \quad \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix} \in \text{Null}(A^T)$$

$$b = b_R + b_N$$

$b \in \text{Range}(A)$ iff $b_N = 0$

$b \in \text{Null}(A^T)$ iff $b_R = 0$

$$\begin{pmatrix} 2 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} + \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$$

$$Ax = b_R = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

$$\begin{aligned} x_1 + x_2 &= 1 \\ x_1 &= 0, \quad x_2 = 1 \end{aligned}$$

$$\begin{aligned} x_1 &= 0 \\ x_2 &= 1 \end{aligned}$$

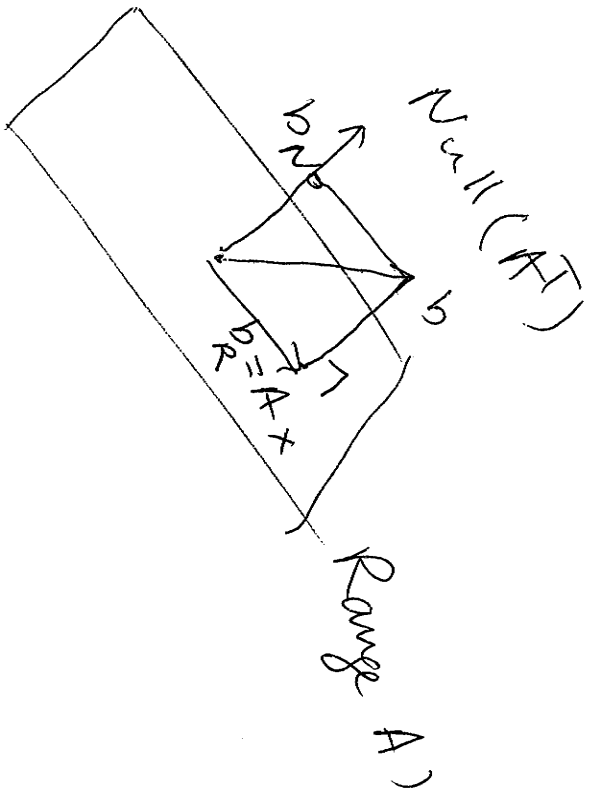
$$\|b - Ax\|_2$$

$$A \text{ } m \times n$$

$b \notin \text{Range}(A)$

otherwise for some

$$x \quad Ax = b$$



$$b = b_N + b_R \quad \text{given } b, b_R \text{ \& } b_N \text{ are unique}$$
$$b_N \in \text{Null}(A^T), \quad b_R \in \text{Range}(A)$$

So solve for $Ax = b_R$ where \uparrow

This will always have a solution.

If the columns of A are linearly independent

then $Ax = b_R$ has a unique solution

$$\text{otherwise if } Ay = b_R \quad \text{then we set } A(n-y) = 0$$
$$\Rightarrow x-y = 0$$

In many areas of practical application we have to solve

$$Ax = b_i \quad \text{for } i=1, \dots, N$$

with large N

For efficiency we proceed as follows:

Assum $Ax = b$ has no exact solution; find $b = b_R + b_N$

Solve $Ax = b_R$

$$Ax = b = b_R + b_N \Rightarrow Ax - b_R = b_N$$

$$A^T(Ax - b_R) = A^T b_N = 0$$

(since $A^T b_N = 0$)

$$A^T Ax = A^T b_R = A^T b$$

Find L such that $A^T A = LL^T$ and use the

trick of solving $LL^T x = y$ to find all solutions:
 $y = A^T b$

$y = f(x)$ f x_0 minimizes $f(x)$

then x_0 minimizes

$$y^2 = (f(x))^2$$

$$2f(x) f'(x) = 0 \Rightarrow f'(x) = 0$$