

Claim v_1, v_2, \dots, v_n are L.I. in \mathbb{R}^n iff v_1, v_2, \dots, v_n span \mathbb{R}^n

pf: " \Rightarrow " Let $A = [v_1 v_2 \dots v_n] \Rightarrow v_1, \dots, v_n$ are L.I. in \mathbb{R}^n , then $\text{rank}(A) = n$
 $\rightarrow A$ invertible $\rightarrow Ax = b$ solvable $\forall b \in \mathbb{R}^n \rightarrow v_1, \dots, v_n$ span \mathbb{R}^n
 " \Leftarrow " If v_1, \dots, v_n span \mathbb{R}^n , then $Ax = b$ solvable $\forall b \in \mathbb{R}^n$
 $\rightarrow \text{rank}(R) = \text{rank}(A) = n \rightarrow N(A) = \{0\} \rightarrow v_1, \dots, v_n$ are L.I.

Remark Any n L.I. vectors form a basis in \mathbb{R}^n . Also any n vectors that span \mathbb{R}^n are a basis for \mathbb{R}^n .

Claim If $V \perp W$, then $V \cap W = \{0\}$

pf: Suppose $v \in V \cap W$, then $v \in V$ and $v \in W$. we have $v^T v = \|v\|^2 = 0$, which implies $v = 0$

Claim $\forall x \in \mathbb{R}^n$, we can have $x = x_r + x_n$ where $x_r \in C(A^T)$ and $x_n \in N(A)$

pf: Let v_1, \dots, v_r be a basis for $C(A^T)$, w_1, \dots, w_{n-r} be a basis for $N(A)$
 Suppose $a_1 v_1 + \dots + a_r v_r + b_1 w_1 + \dots + b_{n-r} w_{n-r} = 0$, let
 $u = a_1 v_1 + \dots + a_r v_r = -b_1 w_1 - \dots - b_{n-r} w_{n-r}$. Then $u \in C(A^T)$ and $u \in N(A)$
 Since $C(A^T) \perp N(A)$, we must have $u = 0$.
 Since v_1, \dots, v_r are a basis for $C(A^T)$ and
 w_1, \dots, w_{n-r} are a basis for $N(A) \rightarrow a_1 = \dots = a_r = b_1 = \dots = b_{n-r} = 0$.
 Therefore $v_1, \dots, v_r, w_1, \dots, w_{n-r}$ are L.I. and hence form a basis for \mathbb{R}^n
 $\forall x \in \mathbb{R}^n$, we can have $x = c_1 v_1 + \dots + c_r v_r + c_{r+1} w_1 + \dots + c_n w_{n-r} = \underbrace{x_r}_{\in C(A^T)} + \underbrace{x_n}_{\in N(A)}$

Remark The decomposition $x = x_r + x_n$ is unique.

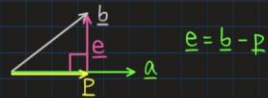
$Ax = A(x_r + x_n) = Ax_r + Ax_n = Ax_r$.
 x_n goes to 0: $Ax_n = 0$, x_r goes to the column space: $Ax_r = Ax$



Claim $\forall \underline{b}$ in the column space comes from one and only one vector in the row space

pf: Let $\underline{b} = A \underline{x}_r = A \underline{x}'_r \in C(A)$ where $\underline{x}_r, \underline{x}'_r \in C(A^T) \Rightarrow A(\underline{x}_r - \underline{x}'_r) = \underline{0}$
 $\Rightarrow \underline{x}_r - \underline{x}'_r \in N(A)$. Since $\underline{x}_r - \underline{x}'_r \in C(A^T)$, we can have $\underline{x}_r - \underline{x}'_r = \underline{0}$
yielding $\underline{x}_r = \underline{x}'_r$ (Note: $C(A^T) \perp N(A)$)

Projection



Let $P = \hat{x} \underline{a}$, Then $\underline{a}^T \underline{e} = \underline{a}^T (\underline{b} - \hat{x} \underline{a}) = 0 \Rightarrow \hat{x} = \frac{\underline{a}^T \underline{b}}{\underline{a}^T \underline{a}}$

$$\therefore P = \frac{\underline{a}^T \underline{b}}{\underline{a}^T \underline{a}} \underline{a}$$

$P = \hat{x} \underline{a} = \underline{a} \hat{x} = \underline{a} \frac{\underline{a}^T \underline{b}}{\underline{a}^T \underline{a}} = \frac{\underline{a} \underline{a}^T}{\underline{a}^T \underline{a}} \underline{b} = P \underline{b}$, where $P = \frac{\underline{a} \underline{a}^T}{\underline{a}^T \underline{a}}$ is the projection matrix

$P^n = P$ (projecting a second time doesn't change anything) $\rightarrow P^2 = \frac{\underline{a} \underline{a}^T \underline{a} \underline{a}^T}{\underline{a}^T \underline{a} \underline{a}^T \underline{a}} = \frac{\underline{a} \underline{a}^T}{\underline{a}^T \underline{a}} = P$