



Definition of orthogonality

Orthogonal subspaces :

if $\underbrace{w^T w = 0}$ for $\forall v \in V, \forall w \in W$

then V and W are orthogonal.



Fundamental Theorem of linear algebra

$$C(A^T) \perp N(A)$$

row space

nullspace.

proof 1

$$C(A^T) = \{ A^T y \in \mathbb{R}^n \mid y \in \mathbb{R}^m \}$$

$$N(A) = \{ x \in \mathbb{R}^n \mid Ax = 0 \}$$

Let x be any vector in $N(A)$.

$$\rightarrow x \in N(A)$$

$$\Rightarrow Ax = \begin{pmatrix} \text{row 1} \\ \vdots \\ \text{row } m \end{pmatrix} x = 0$$

$$\Rightarrow \left\{ \begin{array}{l} (\text{row 1}) \cdot x = 0 \\ \vdots \\ (\text{row } m) \cdot x = 0 \end{array} \right. \rightarrow C(A^T) \perp N(A)$$

proof 2

Let $x \in N(A)$, $A^T y \in C(A^T)$.

$$x^T (A^T y) = (Ax)^T y = 0^T y = 0$$

↑

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(Ex)

$$A = \begin{pmatrix} 1 & 3 & 4 \\ 5 & 2 & 7 \end{pmatrix} \rightarrow R = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

$$\left\{ \begin{array}{l} x = \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} \in N(A) \end{array} \right.$$

$$\left\{ \begin{array}{l} A^T y = \begin{pmatrix} 1 & 5 \\ 3 & 2 \\ 4 & 7 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x + 5y \\ 3x + 2y \\ 4x + 7y \end{pmatrix} \end{array} \right.$$

$$\Rightarrow x^T (A^T y) = 0$$

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④ Another fundamental theorem of linear algebra

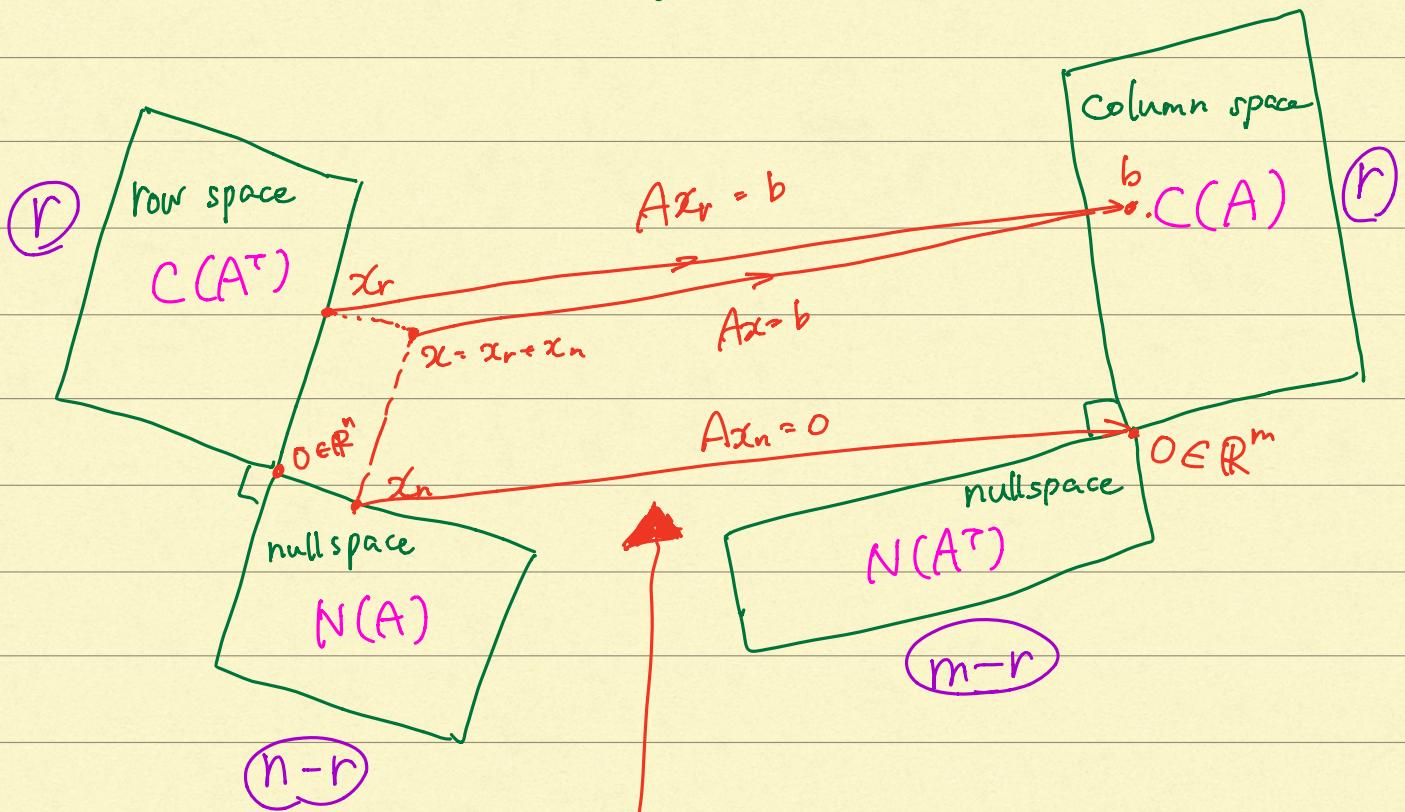
$C(A) \perp N(A^T)$
column space nullspace.



$$\left. \begin{array}{l} Ay \in C(A) \\ x \in N(A^T) \end{array} \right\}$$

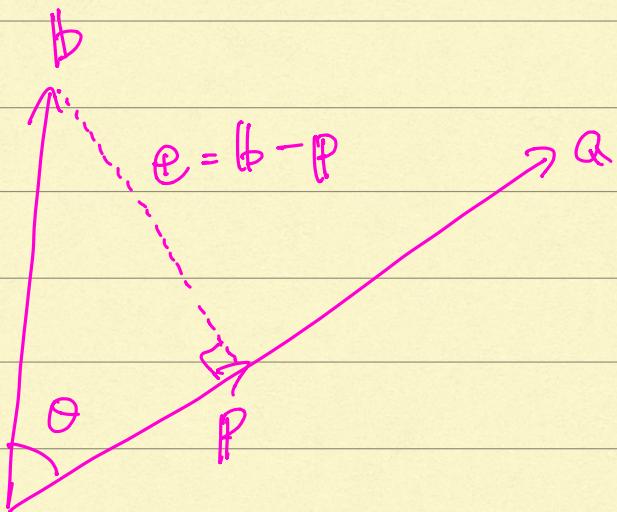
$$x^T(Ay) = (A^Tx)^T y = \mathbb{O}^T y = \mathbb{O}$$

⑤ The pairs of orthogonal subspaces



Every vector in row space moves to the column space by multiplying A !!.

Projection onto a line.



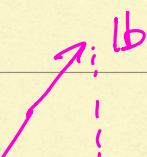
$$P = \hat{x} Q$$

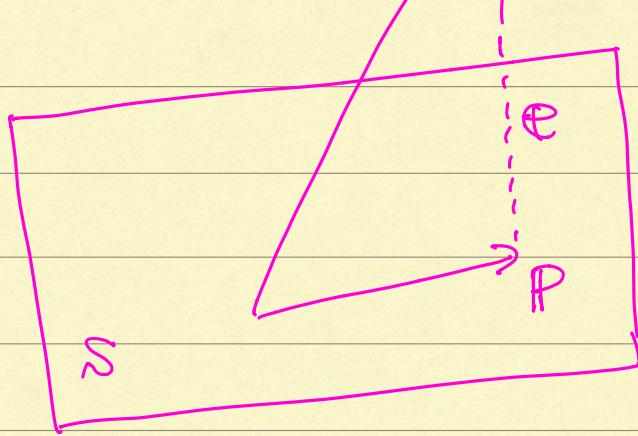
$$P^T Q = 0$$

$$\Leftrightarrow (b - \hat{x} Q)^T Q = 0$$

$$\Leftrightarrow \hat{x} = \frac{b^T Q}{Q^T Q}$$

Projection onto a Subspace





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

$$A = (q_1, \dots, q_n)$$

q_1, \dots, q_n span the subspace S



$$\left\{ \begin{array}{l} q_i^T (b - A \hat{x}) = 0 \\ \vdots \\ q_n^T (b - A \hat{x}) = 0 \end{array} \right.$$

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



$$\hat{x} = \underbrace{(A^T A)^{-1} A^T b}_{\text{solution}}$$

$$P = A \hat{x} = \underbrace{A (A^T A)^{-1} A^T b}_{\text{projection}}$$

$$\downarrow P = Pb$$

$$P = A(A^T A)^{-1} A^T$$

projection matrix

Summary

$$m = 1$$

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

$$P = A \frac{A^T b}{A^T A}$$

$$P = \frac{A A^T}{A^T A}$$

$$m > 1$$

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

$$P = A(A^T A)^{-1} A^T b$$

$$P = A(A^T A)^{-1} A^T$$

$$P = A \hat{x} : \text{projected vector}$$

$$P = P b : P \text{ is projection matrix}$$

 $A^T A$ is invertible if and only if A has linearly independent columns



Procedure

① $A^T A$ has the same nullspace as A

② When the columns of A are linearly independent, its nullspace contains only the zero vector. Then $A^T A$, with the same nullspace, is invertible.

proof of ①

(i)

$$A\mathbf{x} = \mathbf{0} \quad \Rightarrow \quad \mathbf{x} A^T$$

$$\Rightarrow A^T A \mathbf{x} = \mathbf{0}$$

$$A\mathbf{x} = \mathbf{0} \Rightarrow A^T A \mathbf{x}$$

(ii)

$$A^T A \mathbf{x} = \mathbf{0}$$

$$\Rightarrow \mathbf{x}^T A^T A \mathbf{x} = \mathbf{0}$$

$$\Rightarrow (A\mathbf{x})^T A \mathbf{x} = \mathbf{0}$$

$$\Rightarrow \|A\mathbf{x}\|^2 = \mathbf{0}$$

$$\Rightarrow A\mathbf{x} = \mathbf{0}$$

$$A^T A \mathbf{x} = \mathbf{0} \Rightarrow A\mathbf{x} = \mathbf{0}$$

$$\therefore A^T A x_4 = 0 \Leftrightarrow A x_4 = 0$$

 Orthonormal matrix Q

$$Q = (q_1, \dots, q_n) \quad (q_j \in \mathbb{R}^m)$$

q_1, \dots, q_n are orthonormal

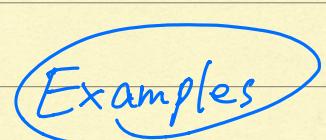
$$\Rightarrow q_i^T q_j = \begin{cases} 1 & \text{when } i = j \\ 0 & \text{when } i \neq j \end{cases}$$

 Properties

$$Q^T Q = I$$

$$\|Qx\| = \|x\| \quad \text{for every vector } x$$

$$(Qx)^T (Qy) = x^T Q^T Q y = x^T y$$

 Examples

$$\text{Rotation : } Q = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

$$\text{Permutation : } P = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}$$

Reflection: $Q = I - 2uu^T$

(u is any unit vector)

⑩ Projection Using Orthonormal Bases.

Suppose the basis vectors are actually orthonormal.

$$q_1, \dots, q_n \rightarrow f_1, \dots, f_n$$

$$A^T A \rightarrow Q^T Q = I$$

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

$$P = Q \hat{x} = Q Q^T b$$

$$P = Q Q^T b$$

⑪ The Gram-Schmidt Process

Q, B, C : linearly independent



A, B, C : orthogonal

↓

q_1, q_2, q_3 : orthonormal

Step 1

$$A = a$$

Step 2

$$B = b - \frac{A^T b}{A^T A} A \quad (\Rightarrow A^T B = 0)$$

Step 3

$$C = c - \frac{A^T c}{A^T A} A - \frac{B^T c}{B^T B} B \quad (\Rightarrow A^T C = 0, B^T C = 0)$$

Step 4

$$q_1 = \frac{a}{\|a\|} \quad q_2 = \frac{b}{\|b\|} \quad q_3 = \frac{c}{\|c\|}$$



QR Factorization

$$A = (a \ b \ c) \rightarrow Q = (q_1 \ q_2 \ q_3)$$

$$A = Q R$$

$$= (q_1 \ q_2 \ q_3) \begin{pmatrix} q_1^T a & q_1^T b & q_1^T c \\ q_2^T b & q_2^T c \end{pmatrix}$$

→ Least squares

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

$$\Leftrightarrow R^T R \hat{x} = R^T Q^T b$$

$$\Leftrightarrow R \hat{x} = Q^T b$$

$$\Leftrightarrow \hat{x} = R^{-1} Q^T b$$

easy !!