

§3.1 Orthogonal vectors and subspaces

$\mathbb{R}^n$: length of a vector $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$

$$\|x\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

$$\|x\|^2 = x_1^2 + x_2^2 + \dots + x_n^2 = x^T x$$

recall: dot product $x \cdot y = x_1 y_1 + x_2 y_2 + \dots + x_n y_n = x^T y$
 $= \|x\| \|y\| \cos \theta$



two vectors are orthogonal iff $x \cdot y = 0$

A set of vectors $\{v_1, \dots, v_k\}$ is orthogonal if $v_i \cdot v_j = 0$ for all $i \neq j$.

Lemma orthogonal vectors are linearly independent.

Proof suppose $c_1 v_1 + c_2 v_2 + \dots + c_k v_k = 0$

then $(c_1 v_1 + \dots + c_k v_k) \cdot v_i = 0 \cdot v_i = 0$

$$c_1 \underbrace{v_1 \cdot v_i}_0 + \dots + c_i \underbrace{v_i \cdot v_i}_{\|v_i\|^2 > 0} + \dots + c_k \underbrace{v_k \cdot v_i}_0$$

$$c_i \|v_i\|^2 = 0 \Rightarrow c_i = 0 \Rightarrow \text{all } c_i = 0 \quad \square$$

Special bases $\{v_1, \dots, v_k\}$ orthogonal basis if $v_i \cdot v_j = 0$ $i \neq j$.

$\{v_1, \dots, v_k\}$ orthonormal basis if $v_i \cdot v_j = 0$ $i \neq j$ and $\|v_i\| = 1$

Examples $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ $\left\{ \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}, \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \right\}$

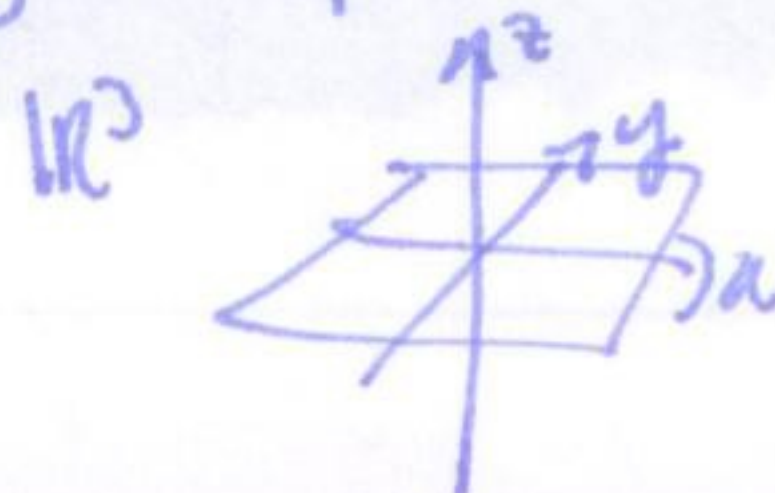
Orthogonal subspaces

V, W subspaces, are orthogonal if $v \cdot w = 0$ for all $v \in V, w \in W$.

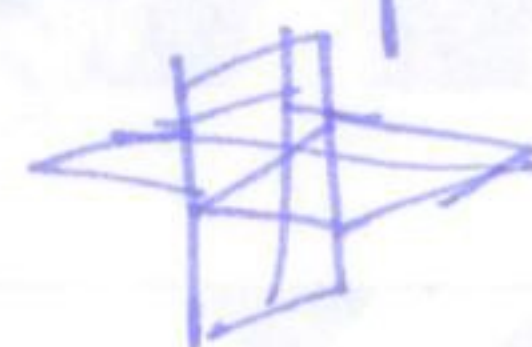
Examples $\{0\}$ orthogonal to every subspace.



V basis $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$
 W basis $\left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$



non-example



Thm 3C (Fundamental theorem of orthogonality) $A \text{ } m \times n$

The row space is orthogonal to the null space/kernel $\in \mathbb{R}^n$

The column space is orthogonal to the left nullspace/cokernel $\in \mathbb{R}^m$

Proof ① suppose $x \in N(A)$ then $Ax = 0$ $A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

i.e. x is orthogonal to all the rows of A , and so all linear combinations

$$\text{of the rows } (a_1 r_1 + a_2 r_2 + \dots + a_m r_m) \cdot x = a_1 \underbrace{r_1 \cdot x}_0 + a_2 \underbrace{r_2 \cdot x}_0 + \dots + a_m \underbrace{r_m \cdot x}_0 = 0$$

so $N(A) \perp \text{row}(A)$.

suppose $x \in N(A^T)$ so $x^T A = 0$ $[x_1 \dots x_n] A = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

i.e. x is orthogonal to the columns of A , and so all linear combinations of the cols., so $N(A^T) \perp \text{col}(A)$.

② suppose $x \in N(A)$ so $Ax = 0$
 $v \in \text{row}(A)$ so $v = A^T z$ for some z .

$$\text{so } v^T x = (A^T z)^T x = z^T A x = z \cdot 0 = 0. \quad \square$$

Defⁿ let $V \subset \mathbb{R}^n$ be a subspace, then the collection of all vectors perpendicular to V is a subspace called the orthogonal complement $V^\perp$ say " V perp"

Thm 3D (Fundamental theorem of linear algebra, part II)

$$N(A) = \text{row}(A)^\perp \quad (\subset \mathbb{R}^n)$$

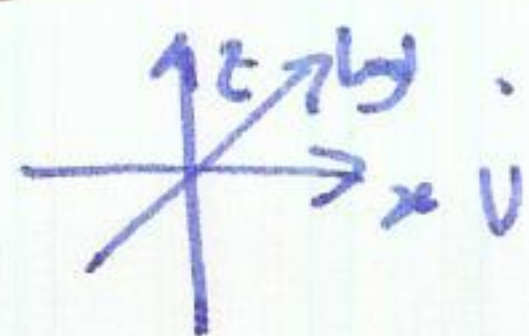
$$N(A^T) = \text{col}(A)^\perp \quad (\subset \mathbb{R}^m)$$

Proof recall: $\left. \begin{array}{l} \dim(N(A)) = n - r \\ \dim(\text{row}(A)) = r \end{array} \right\}$ orthogonal vectors independent!
 so $\dim(\text{span}(N(A), \text{row}(A))) \geq n$

similarly: $\left. \begin{array}{l} \dim(N(A^T)) = m - r \\ \dim(\text{col}(A)) = r \end{array} \right\}$ so span all of $\mathbb{R}^m$.

in each case, any orthogonal vector to $\text{row}(A)$ must lie in $N(A)$
 to $\text{col}(A)$ must lie in $N(A^T)$. $\square$

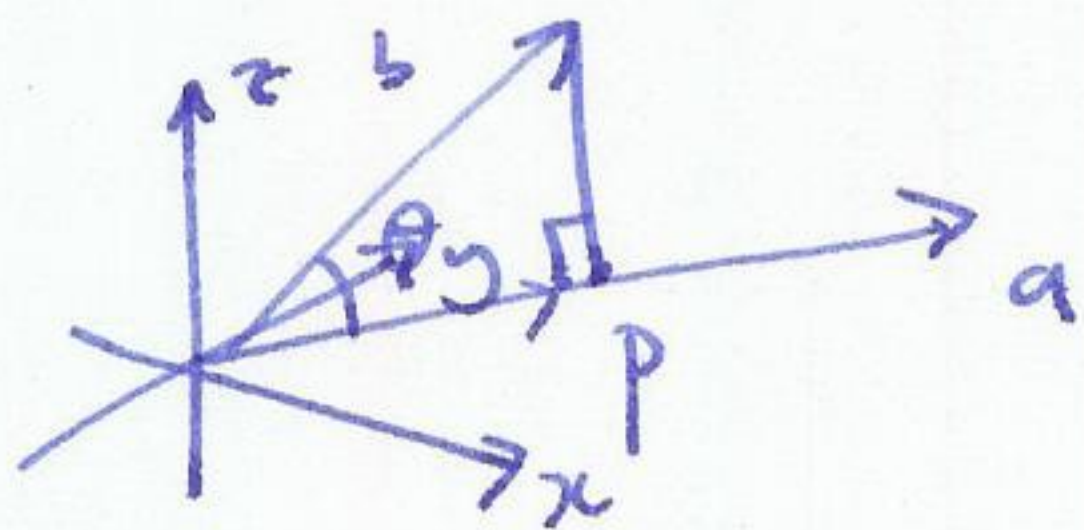
Remark v, w can be perpendicular, but not span $\mathbb{R}^n$.
 but if $V = W^\perp$ then $\dim(V) + \dim(W) = n$.



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§3.2 Projections

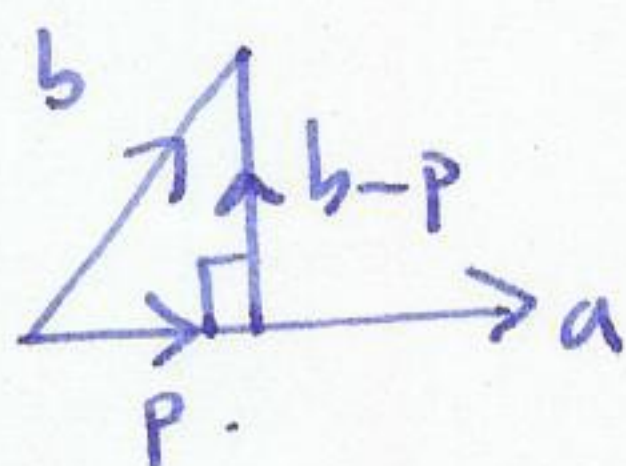
p = projection of b onto line through a .



Recall: geometric defⁿ of dot product
 $a^T b = \|a\| \|b\| \cos \theta$

key fact: $p = \frac{a^T b}{a^T a} a = \frac{a \cdot b}{a \cdot a} a$

Proof $p = \lambda a$ for some λ



$$b - p \cdot a = 0$$

$$(b - (\lambda a)) \cdot a = 0$$

$$a \cdot b - \lambda a \cdot a = 0$$

$$\lambda = \frac{a \cdot b}{a \cdot a} \quad \text{so} \quad p = \frac{a \cdot b}{a \cdot a} a \quad \square$$

Corollary Schwarz inequality

$$|a^T b| = |a \cdot b| \leq \|a\| \|b\|$$

with equality iff $a = \lambda b$.

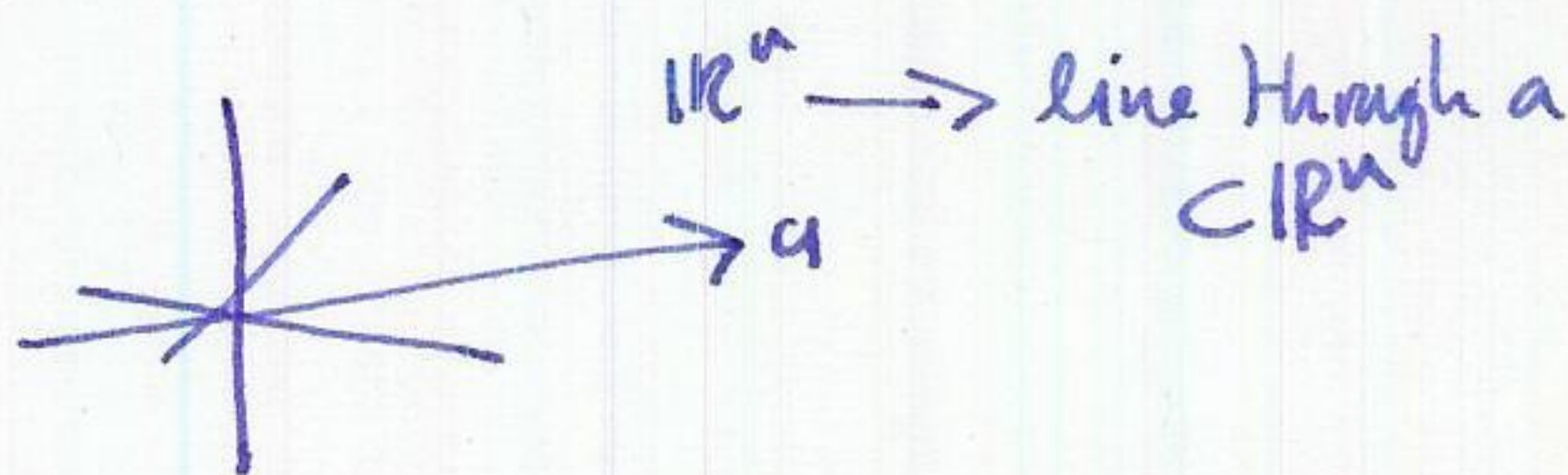
Example project $b = (1, 2, 3)$ onto $a = (3, 2, 1)$

$$p = \frac{a \cdot b}{a \cdot a} a = \frac{3+4+3}{9+2+1} (3, 2, 1) = \frac{10}{12} (3, 2, 1)$$

Observation projection is a linear map!

$$\frac{(x+y) \cdot a}{a \cdot a} a = \frac{x \cdot a}{a \cdot a} a + \frac{y \cdot a}{a \cdot a} a$$

$$\lambda \frac{x \cdot a}{a \cdot a} a = \lambda \frac{x \cdot a}{a \cdot a} a$$



Q: what is the matrix P corresponding to projection?

$$P: x \mapsto a \frac{a^T x}{a \cdot a}$$

$$\text{so } P = \frac{aa^T}{a \cdot a}$$

$$\begin{array}{c} a \quad a^T \\ \underbrace{1 \times 1 \quad 1 \times n}_{n \times n} \\ a^T \quad a \\ \underbrace{1 \times n \quad n \times 1}_{1 \times 1} \end{array}$$

Example project onto $(1,1,1)$

$$P = \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

Properties of projections

① P is symmetric

② $P^2 = P$

$$\left(\frac{aa^T}{a \cdot a} \right)^T = \frac{(a^T)^T a^T}{a \cdot a} = \frac{aa^T}{a \cdot a}$$

Remark (3.5) A^T can be defined as: the matrix s.t.

$$Ax \cdot y = x \cdot A^T y$$

$$\text{i.e. } (Ax)^T y = x^T A^T y = x^T (A^T y)$$

§3.3 Projections and least squares

Inconsistent equations:

$$2x = b_1$$

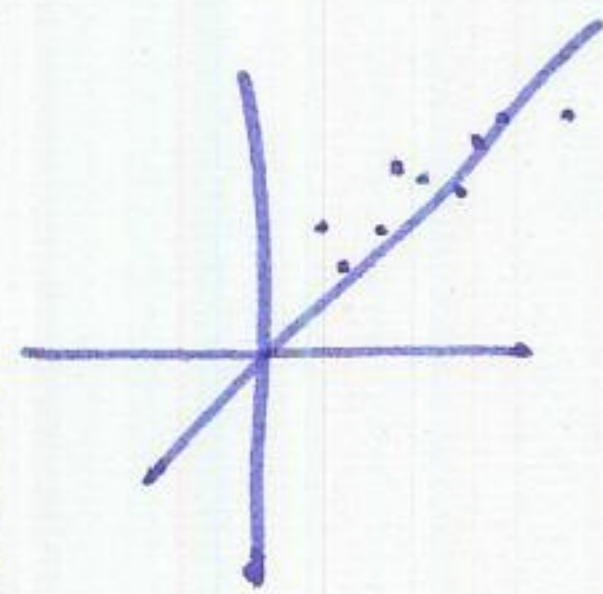
$$3x = b_2$$

$$4x = b_3$$

$$a^T x = b$$

$$a = \begin{bmatrix} 2 \\ 3 \\ 4 \end{bmatrix}$$

$$b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$



find x which minimizes error squared: $E^2 = (2x - b_1)^2 + (3x - b_2)^2 + (4x - b_3)^2$

$$\frac{dE^2}{dx} = 2[(2x - b_1)2 + (3x - b_2)3 + (4x - b_3)4] = 0$$

$$\text{minimum at } \hat{x} = \frac{2b_1 + 3b_2 + 4b_3}{2^2 + 3^2 + 4^2} = \frac{a^T b}{a^T a} \leftarrow \text{least squares solution.}$$

General case

$$ax = b$$

Note: error

$$\text{minimize } E^2 = \|ax - b\|^2 = (a_1 x - b_1)^2 + \dots + (a_n x - b_n)^2$$

$$\frac{dE^2}{dx} = 2[(a_1 x - b_1)a_1 + \dots + (a_n x - b_n)a_n] = 0 \quad \hat{x} = \frac{a^T b}{a^T a}$$

note error $ax - b \perp a$! $a \cdot (ax - b) = a \cdot a \frac{a \cdot b}{a \cdot a} - a \cdot b = 0$