

# Linear Algebra

vectors and matrices

$v = \begin{bmatrix} v_1 \\ \vdots \\ v_d \end{bmatrix}$   $\rightarrow$  reals  
 $\hookrightarrow$  column vector

is equivalent to a  $d$ -dimensional point ( $\mathbb{R}^d$ )

$$v^T = [v_1, \dots, v_d]$$

$\hookrightarrow$  row vector

} both will be stored as an array

$$A^{n \times d} = \begin{bmatrix} -a_1^T \\ \vdots \\ -a_n^T \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1d} \\ A_{21} & A_{22} & \dots & A_{2d} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \dots & A_{nd} \end{bmatrix}$$

$A_{ij}$   $\rightarrow$  col index  
 $\hookrightarrow$  row index

$b_i$

$$A = \begin{bmatrix} 1 & \dots & 1 \\ b_1 & \dots & b_k \\ 1 & \dots & 1 \end{bmatrix}$$

$$a_1 + a_2 = (2+4, 4+1) = (6, 5)$$

$$\text{when } a_1 = (2, 4)$$

$$a_2 = (4, 1)$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 + 2x_2 \\ 3x_1 + 4x_2 \end{bmatrix} \rightarrow \text{dot product perspective of matrix vector multiplication}$$

$A \quad x$

$$\underline{2 \times 2} \times \underline{2 \times 1} = 2 \times 1 \text{ matrix}$$

"Linear combination of columns" perspective

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ 4 \end{bmatrix} = \begin{bmatrix} x_1 \\ 3x_1 \end{bmatrix} + \begin{bmatrix} 2x_2 \\ 4x_2 \end{bmatrix} = \begin{bmatrix} x_1 + 2x_2 \\ 3x_1 + 4x_2 \end{bmatrix}$$

ex)

$$2x_1 + 5x_2 - 7x_3 = 10$$

$$-x_1 + 8x_2 - 2x_3 = -2$$

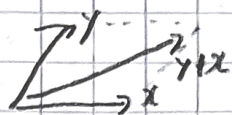
$$\begin{bmatrix} 2 & 5 & -7 \\ -1 & 8 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 10 \\ -2 \end{bmatrix}$$

$$A \cdot x = b$$

we are trying to find a linear combination of the three 2dim columns of  $A$  such that this linear combination is equal to  $b$



$$x, y \in \mathbb{R}^d \quad x+y = \begin{bmatrix} x_1 \\ \vdots \\ x_d \end{bmatrix} + \begin{bmatrix} y_1 \\ \vdots \\ y_d \end{bmatrix} = \begin{bmatrix} x_1+y_1 \\ \vdots \\ x_d+y_d \end{bmatrix}$$



$$A^{n \times d} + B^{n \times d} = C^{n \times d} \text{ where } C_{iy} = A_{iy} + B_{iy} \text{ for all } 1 \leq i \leq n$$

$$C^{n \times m} = A^{n \times d} B^{d \times m} \text{ where } C_{iy} = \sum_{k=1}^d A_{ik} \cdot B_{ky} = a_i^T \cdot b_y \quad 1 \leq i \leq n, 1 \leq y \leq m$$

$$\begin{bmatrix} -a_1^T \\ -a_n^T \end{bmatrix} \begin{bmatrix} b_1 \\ b_y \\ b_m \end{bmatrix}$$

$A \cdot B \neq B \cdot A$  (matrix multiplication is not commutative)

$(AB)C = A(BC)$  associative

$A \cdot (B+C) = AB + AC$  distributive

Vector-vector product

$$x^T y = (x_1 \dots x_d) \begin{pmatrix} y_1 \\ \vdots \\ y_d \end{pmatrix} = x_1 y_1 + \dots + x_d y_d = \sum_{i=1}^d x_i y_i = \langle x, y \rangle \Rightarrow \in \mathbb{R}, \text{ inner product}$$

$$x^T y = (\text{length}(x)) (\text{length}(y)) \cdot \cos(\theta) \quad \leftarrow \text{Euclidean product / } l_2$$

$$\hookrightarrow \|x\|_2 = \sqrt{x_1^2 + \dots + x_d^2} = \sqrt{\langle x, x \rangle} = \sqrt{x^T x}$$

$\hookrightarrow$  generalizes Pythagoras theorem

- orthogonal if two vectors if  $A \cdot B = 0$

$$\begin{pmatrix} x_1 y_1 & \dots & x_1 y_d \\ \vdots & & \vdots \\ x_d y_1 & \dots & x_d y_d \end{pmatrix} \rightarrow d \times d \Rightarrow \text{outer product, } \in \mathbb{R}^{d \times d}$$

$$\|x\|_2^2 = x^T \cdot x = \sum_{i=1}^d x_i^2$$

$l_p$  norms (well-defined for our purposes for  $p \in [1, \infty)$ )

$$\|x\|_p = \left( \sum_{i=1}^d |x_i|^p \right)^{1/p} \quad \left[ \begin{array}{c} \|x\|_p^p \\ \uparrow \\ \text{denotes} \end{array} \right]$$

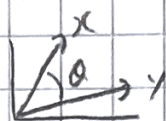
$$\|x\|_\infty = \max_{1 \leq i \leq d} |x_i|$$

$$\text{Let } x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

$$\textcircled{Q} \quad x^T y = x_1 y_1 + x_2 y_2$$

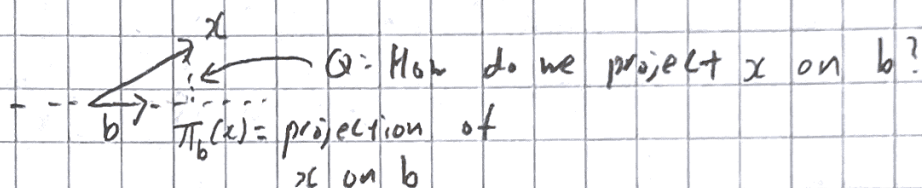
$$\|x\| = \sqrt{x_1^2 + x_2^2}$$

$$\|y\| = \sqrt{y_1^2 + y_2^2}$$



$$\cos \theta = \frac{x_1 y_1 + x_2 y_2}{\sqrt{x_1^2 + x_2^2} \cdot \sqrt{y_1^2 + y_2^2}} \quad (\text{projections})$$

② How do we project? (why does trigonometry play a role?)

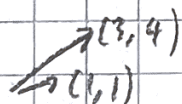


$$\text{ex) } x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \text{ then } \pi_b(x) = \pi_{\begin{pmatrix} 1 \\ 0 \end{pmatrix}} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ 0 \end{pmatrix}$$

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

This example is easy, because  $b = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \rightarrow$  standard orthonormal basis for  $\mathbb{R}^2$

$(e_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, e_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}, e_n = \begin{pmatrix} 0 \\ \vdots \\ 1 \end{pmatrix})$  is standard basis for  $\mathbb{R}^n$



(clearly,  $\pi_b(x) = \lambda \cdot b$  for some  $\lambda \in \mathbb{R}$ )

$$x - \pi_b(x) \perp b \quad \rightarrow \quad (x - \pi_b(x))^T b = 0$$

$$(x - \pi_b(x))^T b = 0$$

$$x^T b = (\pi_b(x))^T \cdot b$$

$$\lambda b^T b = x^T b$$

$$\lambda = \frac{x^T b}{b^T b} = \frac{b^T x}{\|b\|_2^2}$$

$$\text{So } \pi_b(x) = \lambda b = \left( \frac{b^T x}{b^T b} \right) b = \left[ \left( \frac{b \cdot b^T}{b^T \cdot b} \right) \right] x$$

projection matrix



Let  $P = \frac{bb^T}{b^T b}$

$$P^2 = \left( \frac{bb^T}{b^T b} \right) \left( \frac{bb^T}{b^T b} \right) = \frac{b(b^T b)b^T}{(b^T b)(b^T b)} = \frac{bb^T}{b^T b} = P$$

$x^T b = x_1 b_1 + \dots + x_n b_n = b^T x$

$$b = \left( \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$b^T b = \frac{1}{2} + \frac{1}{2} = 1$$

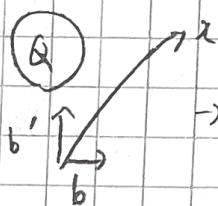
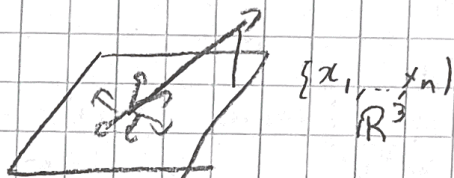
$$bb^T = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

$$P = \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix} \quad P_x = \begin{pmatrix} 1/2 x_1 + 1/2 x_2 \\ 1/2 x_1 + 1/2 x_2 \end{pmatrix}$$

$$P^2 = P \rightarrow \frac{bb^T}{b^T b}$$

$$\hookrightarrow (P \cdot x) = \pi_b(x)$$

$$\pi_b(\pi_b(x)) = \pi_b(x)$$



How to find  $b'$ ?

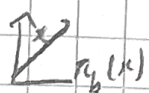
Idea 1

find  $b'$ .  $b = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$  then  $b' = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$  and repeat

$$P' = \frac{(b'b')^T}{b'^T b'} \quad \pi_{b'}(x) = P' \cdot x$$

Idea 2

$$(I - P)x = x - Px = x - \pi_b(x)$$



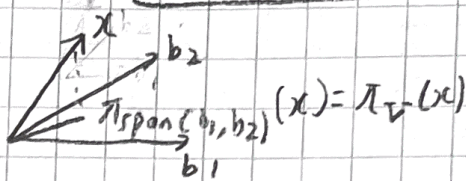
Claim

$$P' = I - P$$

$$P = \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix} \text{ (on } b = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix})$$

$$I - P = \begin{pmatrix} 1/2 & -1/2 \\ -1/2 & 1/2 \end{pmatrix} = \frac{(b')(b')^T}{(b')^T(b')} = \frac{b'b'^T}{1}$$

projection on general space



$$U = \text{span}(b_1, b_2)$$

$$\pi_U(x) = \lambda_1 b_1 + \dots + \lambda_m b_m \text{ for some } \lambda_1, \dots, \lambda_m$$

$$= \begin{bmatrix} b_1 & \dots & b_m \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_m \end{bmatrix} = Bx\lambda$$

$$\begin{aligned} (I - P)^2 &= (I - P)(I - P) \\ &= I^2 - 2P + P^2 \\ &= I^2 - 2P + P \\ &= I^2 - P = I - P \end{aligned}$$

Projection property

$$x^2 = x$$