$$\begin{aligned}
X &= \begin{cases} x_1 \\ \vdots \\ y_n \end{cases}, & y &= \begin{cases} y_1 \\ \vdots \\ y_n \end{cases} \\
&= \begin{cases} x_1 & x_2 & \dots & x_n \end{cases} \begin{cases} y_1 \\ y_2 \\ \vdots \\ y_n \end{cases} & = x^T y. \\
&= x^T y. \end{cases}$$

$$\begin{aligned}
X &= \begin{cases} x_1 & x_2 & \dots & x_n \end{cases} \begin{cases} y_1 \\ y_2 \\ \vdots \\ y_n \end{cases} & = x^T y. \end{cases}$$

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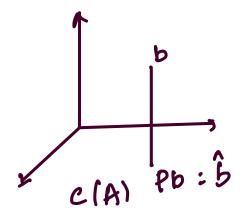
$$\begin{aligned}
X &= \begin{cases} x_1 & x_2 & \dots & x_n \end{cases} \begin{cases} x_1 \\ y_2 \\ \vdots \\ y_n \end{cases} & = x^T y. \end{cases}$$

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Given a matrix A,



P: involves computation of inverses.

Algorithm

Step1: C(A) -> orthonormal basis fu, uz, ..., un's

(u:, ug) = 0 , \* fj

 $\langle u_i, u_i \rangle = \|u_i\|^2 = 1$ 

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= 
$$\frac{(u_1, b)}{||u_1||}$$
  $\frac{(u_2, b)}{||u_2||}$   $\frac{(u_2, b)}{||u_2||}$   $\frac{(u_1, b)}{||u_2||}$   $\frac{(u_1, b)}{||u_2||}$ 

$$= \langle u_1, b \rangle u_1 + \langle u_2, b \rangle u_2 + \dots + \langle u_n, b \rangle u_n$$

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Then 
$$\hat{b} : u_1^T b u_1 + u_2^T b u_2 + \cdots + u_n^T b u_n$$

$$= u_1 (u_1^T b) + u_2 (u_2^T b) + \cdots + u_n (u_n^T b) \text{ (as } u_1^T b)$$

$$= (u_1 u_1^T + u_2 u_2^T + \cdots + u_n u_n^T) \text{ b} \text{ (s a } u_1^T b)$$

$$= [u_1 u_2 \cdots u_n] \begin{bmatrix} u_1^T \\ u_2^T \end{bmatrix} \text{ b} \text{ (vectorization)}$$

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Q. Without using matrix inverse, find projection matrix; given A = [3 -1]
0 0
4 7 Soln:  $\{w_1, w_2\} = \{(3,0,4), (-1,0,7)\}.$ Gran - Schnidt process: 1, = 10, : (3,0,4)  $v_2 = w_2 - \frac{\langle w_2, v_1 \rangle v_1}{11 v_1 11^2}$ 

= (-1,0,7) - < (-1,0,7), (3,0,4) > (3,0,4)

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$$u_{1} = \frac{v_{1}}{11 v_{2} l_{1}} = \frac{(-4, 0, 3)}{\sqrt{4 + o^{2} + 3^{2}}} = \frac{1}{5} (-4, 0, 3)$$

$$P = u_1 u_1^T + u_2 u_2^T$$

$$= \frac{1}{5} \begin{bmatrix} 3 \\ 0 \\ 4 \end{bmatrix} \frac{1}{5} \begin{bmatrix} 3 \\ 0 \\ 4 \end{bmatrix} + \frac{1}{5} \begin{bmatrix} 4 \\ 0 \\ 3 \end{bmatrix} \frac{1}{5} \begin{bmatrix} 4 \\ 0 \end{bmatrix} \frac{$$

$$\frac{Q}{A}$$
. Project  $(5,6,7)$  on  $CCA$ ); given

$$A = \begin{bmatrix} 3 & -1 \\ 0 & 0 \\ 4 & 7 \end{bmatrix}.$$