

## FINDING ORTHONORMAL BASES - GRAM-SCHMIDT

Suppose we want an orthonormal basis for  $\text{span}\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\}$ . We can build one using Gram-Schmidt procedure.

### PROCEDURE

Step 0  $\vec{q}_1 = \frac{\vec{a}_1}{\|\vec{a}_1\|}$  (normalizing to unit length)

Step 1  $\vec{q}_2 = \frac{\vec{a}_2 - P_{\text{span}\{\vec{q}_1\}} \vec{a}_2}{\|\vec{a}_2 - P_{\text{span}\{\vec{q}_1\}} \vec{a}_2\|}$  We have  $P_{\text{span}\{\vec{q}_1\}} \vec{a}_2 = \vec{p}_1 = (\vec{q}_1 \cdot \vec{a}_2) \vec{q}_1$

Step 2  $\vec{q}_3 = \frac{\vec{a}_3 - P_{\text{span}\{\vec{q}_1, \vec{q}_2\}} \vec{a}_3}{\|\vec{a}_3 - P_{\text{span}\{\vec{q}_1, \vec{q}_2\}} \vec{a}_3\|}$  where  $P_{\text{span}\{\vec{q}_1, \vec{q}_2\}} \vec{a}_3 = \vec{p}_2 = (\vec{q}_1 \cdot \vec{a}_3) \vec{q}_1 + (\vec{q}_2 \cdot \vec{a}_3) \vec{q}_2$

...  
continue

EXAMPLE: Find an orthonormal basis for the column space of  $A = \begin{bmatrix} 1 & -1 & 4 \\ 1 & 4 & -2 \\ 1 & 4 & 2 \\ 1 & -1 & 0 \end{bmatrix}$

Step 0  $r_{11} = \|\vec{a}_1\| = \left\| \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \right\| = 2.$

$$\vec{q}_1 = \frac{\vec{a}_1}{r_{11}} = \frac{\vec{a}_1}{\|\vec{a}_1\|} \text{ or } \vec{q}_1 = \begin{pmatrix} 1/2 \\ 1/2 \\ 1/2 \\ 1/2 \end{pmatrix}.$$

Step 1  $r_{12} = (\vec{a}_2 \cdot \vec{q}_1) = \vec{q}_1^T \vec{a}_2 = (1/2, 1/2, 1/2, 1/2) \begin{pmatrix} -1 \\ 4 \\ 4 \\ -1 \end{pmatrix} = 3$

Now,  $p_1 = r_{12} \vec{q}_1 = 3\vec{q}_1$

we have  $\vec{a}_2 - p_1 = \begin{pmatrix} -1 \\ 4 \\ 4 \\ -1 \end{pmatrix} - \begin{pmatrix} 3/2 \\ 3/2 \\ 3/2 \\ 3/2 \end{pmatrix} = \begin{pmatrix} -5/2 \\ 5/2 \\ 5/2 \\ -5/2 \end{pmatrix}$

$r_{22} = \|\vec{a}_2 - p_1\| = 5$

$$\vec{q}_2 = \frac{\vec{a}_2 - p_1}{\|\vec{a}_2 - p_1\|} = \frac{1}{5} \begin{pmatrix} -5/2 \\ 5/2 \\ 5/2 \\ -5/2 \end{pmatrix} \text{ or}$$

$$\vec{q}_2 = \begin{pmatrix} -1/2 \\ 1/2 \\ 1/2 \\ -1/2 \end{pmatrix}$$

Recall:

$$\vec{q}_{k+1} = \frac{\vec{a}_{k+1} - P_{\text{span}\{\vec{a}_1, \dots, \vec{a}_k\}} \vec{a}_{k+1}}{\left\| \vec{a}_{k+1} - P_{\text{span}\{\vec{a}_1, \dots, \vec{a}_k\}} \vec{a}_{k+1} \right\|}$$

where  $P_{\text{span}\{\vec{a}_1, \dots, \vec{a}_k\}} = \vec{p}_k$   
 $= (\vec{a}_{k+1} \cdot \vec{a}_1) \vec{a}_1 + \dots + (\vec{a}_{k+1} \cdot \vec{a}_k) \vec{a}_k$   
 $\qquad \qquad \qquad r_{1,k+1} \qquad \qquad \qquad r_{k,k+1}$

Step 3  $r_{13} = (\vec{a}_3 \cdot \vec{q}_1) = \vec{q}_1^T \vec{a}_3 = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right) \begin{pmatrix} 4 \\ -2 \\ 2 \\ 0 \end{pmatrix} = 2$

$r_{23} = (\vec{a}_3 \cdot \vec{q}_2) = \vec{q}_2^T \vec{a}_3 = -2$

and  $\vec{p}_2 = \text{span}\{\vec{q}_1, \vec{q}_2\} \vec{a}_3 = r_{13} \vec{q}_1 + r_{23} \vec{q}_2 = 2 \begin{pmatrix} 1/2 \\ 1/2 \\ 1/2 \\ 1/2 \end{pmatrix} - 2 \begin{pmatrix} -1/2 \\ 1/2 \\ 1/2 \\ -1/2 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \\ 0 \\ 2 \end{pmatrix}$

we now have  $\vec{a}_3 - \vec{p}_2 = \begin{pmatrix} 4 \\ -2 \\ 2 \\ 0 \end{pmatrix} - \begin{pmatrix} 2 \\ 0 \\ 0 \\ 2 \end{pmatrix} = \begin{pmatrix} 2 \\ -2 \\ 2 \\ -2 \end{pmatrix}$

$r_{33} = \|\vec{a}_3 - \vec{p}_2\| = 4$

$\vec{q}_3 = \frac{\vec{a}_3 - \vec{p}_2}{\|\vec{a}_3 - \vec{p}_2\|} = \frac{1}{2} \begin{pmatrix} 1 \\ -1 \\ 1 \\ -1 \end{pmatrix}$

or  $\vec{q}_3 = \begin{pmatrix} 1/2 \\ -1/2 \\ 1/2 \\ -1/2 \end{pmatrix}$

$Q = \begin{bmatrix} 1/2 & -1/2 & 1/2 \\ 1/2 & 1/2 & -1/2 \\ 1/2 & 1/2 & 1/2 \\ 1/2 & -1/2 & -1/2 \end{bmatrix}$

- $Q$  is an orthogonal matrix
- Its columns are an orthonormal basis (ONB) for  $C(A)$

# QR FACTORIZATION

Any  $m \times n$  matrix with independent columns can be factored into QR.

- the matrix Q has orthonormal columns
- the square matrix R which upper triangular with positive entries on the diagonal.

In the previous problem, we have

$$A = \begin{bmatrix} 1 & -1 & 4 \\ 1 & 4 & -2 \\ 1 & 4 & 2 \\ 1 & -1 & 0 \end{bmatrix} = \underbrace{\begin{bmatrix} \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} \end{bmatrix}}_Q \underbrace{\begin{bmatrix} 2 & 3 & 2 \\ 0 & 5 & -2 \\ 0 & 0 & 4 \end{bmatrix}}_R$$

Why do we care about QR?

Consider least squares

$$(A^T A \hat{x} = A^T \vec{b}) \quad \text{Solutions}$$

use  $A = QR$

$$(QR)^T (QR) \hat{x} = (QR)^T \vec{b} \\ \Rightarrow R^T \underbrace{(Q^T Q)}_I R \hat{x} = R^T Q^T \vec{b} \Rightarrow$$

upper triangular

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

OR

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

solve by back substitution

(2) This  $4 \times 4$  Hadamard matrix is an orthogonal matrix.

$$Q = \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} = [\vec{q}_1 \quad \vec{q}_2 \quad \vec{q}_3 \quad \vec{q}_4]$$

- (a) What projection matrix  $P_4$  will project every  $\vec{b}$  in  $\mathbb{R}^4$  onto the line through  $\vec{q}_4$ ?
- (b) What projection matrix  $P_{123}$  will project every  $\vec{b}$  in  $\mathbb{R}^4$  onto the subspace spanned by  $\vec{q}_1, \vec{q}_2$  and  $\vec{q}_3$ ?
- (c) Suppose  $A$  is the  $4 \times 3$  matrix whose columns are  $\vec{q}_1, \vec{q}_2, \vec{q}_3$ .  
Find the least squares solution  $\hat{x}$  to

$$Ax = b \iff \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & -1 \\ 1 & -1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

(a) Let  $A = \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$  Now consider the projection matrix onto the  $C(A)$  given by

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

Note that  $A^T A = \begin{bmatrix} 1 & -1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix} = 4$

and  $(A^T A)^{-1} = \frac{1}{4}$

Hence  $P = A (A^T A)^{-1} A^T = \frac{1}{4} A A^T = \frac{1}{4} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & -1 & -1 & 1 \end{bmatrix}$

$$= \frac{1}{4} \begin{bmatrix} 1 & -1 & -1 & 1 \\ -1 & 1 & 1 & -1 \\ -1 & 1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$