

## Sec 5.6 The Gram-Schmidt Orthogonalization Process

Gram-Schmidt Process: Given  $n$  linearly independent vectors  $\bar{x}_1, \dots, \bar{x}_n$  in  $\mathbb{R}^n$ , one can find an orthonormal basis of  $\mathbb{R}^n$  using  $\bar{x}_1, \dots, \bar{x}_n$

Step 1: Let  $u_1 = \frac{1}{\|\bar{x}_1\|} \cdot \bar{x}_1$

Step 2: Find the projection of  $\bar{x}_2$  onto  $u_1$ , namely  $P_1$ :

$$P_1 = \left( \frac{\langle \bar{x}_2, u_1 \rangle}{\|u_1\|^2} \right) \cdot u_1 = \langle \bar{x}_2, u_1 \rangle \cdot u_1$$

Let  $v_2 = \bar{x}_2 - P_1$   
(Then  $v_2$  is orthogonal to  $u_1$ )

Let  $u_2 = \frac{1}{\|v_2\|} \cdot v_2$ .

(Then  $\{u_1, u_2\}$  is an orthonormal set)

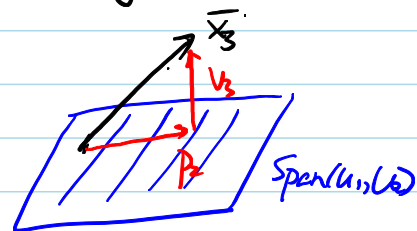
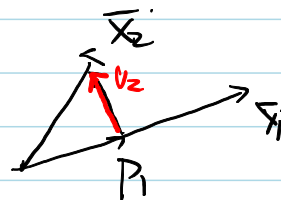
Step 3: Find the projection of  $\bar{x}_3$  onto  $\text{Span}(u_1, u_2)$ , namely,  $P_2$ :

$$P_2 = \langle \bar{x}_3, u_1 \rangle \cdot u_1 + \langle \bar{x}_3, u_2 \rangle \cdot u_2$$

Let  $v_3 = \bar{x}_3 - P_2$

and  $u_3 = \frac{1}{\|v_3\|} \cdot v_3$

(Then  $u_1, u_2, u_3$  is an orthonormal set)



Recursively, after we have an orthonormal set  $u_1, \dots, u_k$

Step (k+1)

Projection of  $\bar{x}_{k+1}$  onto  $\text{Span}(u_1, \dots, u_k)$  is given by

$$P_k = \langle \bar{x}_{k+1}, u_1 \rangle u_1 + \langle \bar{x}_{k+1}, u_2 \rangle u_2 + \dots + \langle \bar{x}_{k+1}, u_k \rangle u_k$$

Let  $v_{k+1} = \bar{x}_{k+1} - P_k$  and  $u_{k+1} = \frac{1}{\|v_{k+1}\|} \cdot v_{k+1}$

Stop at Step  $n$ , when we reach  $\bar{x}_n$  and have an orthonormal set  $u_1, u_2, \dots, u_n$ .

Remark. A key observation in G-S process is:

If  $p_k$  is the orthogonal projection of  $\bar{x}_{k+1}$  onto  $\text{Span}(u_1, \dots, u_k)$  then  $\langle \bar{x}_{k+1}, u_j \rangle = \langle p_k, u_j \rangle$  for  $j=1, \dots, k$ .

eg. Given  $\bar{x}_1 = (1, 1, 1)^T$ ,  $\bar{x}_2 = (1, 0, 1)^T$ ,  $\bar{x}_3 = (0, 2, -1)^T$ , use the G-S process to obtain an orthonormal basis for  $\mathbb{R}^3$ .

$$u_1 = \frac{1}{\|\bar{x}_1\|} \cdot \bar{x}_1 = \frac{1}{\sqrt{3}} (1, 1, 1)^T$$

$$\langle \bar{x}_2, u_1 \rangle = (1, 0, 1) \cdot \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \frac{1}{\sqrt{3}} (1+0+1) = \frac{2}{\sqrt{3}}$$

$$p_1 = \langle \bar{x}_2, u_1 \rangle u_1 = \frac{2}{\sqrt{3}} \cdot \frac{1}{\sqrt{3}} (1, 1, 1)^T = \frac{2}{3} (1, 1, 1)^T$$

$$v_2 = \bar{x}_2 - p_1 = (1, 0, 1)^T - \left(\frac{2}{3}, \frac{2}{3}, \frac{2}{3}\right)^T = \left(\frac{1}{3}, -\frac{2}{3}, \frac{1}{3}\right)^T$$

$$\|v_2\| = \sqrt{\left(\frac{1}{3}\right)^2 + \left(-\frac{2}{3}\right)^2 + \left(\frac{1}{3}\right)^2} = \sqrt{\frac{1+4+1}{9}} = \sqrt{\frac{2}{3}}$$

$$u_2 = \frac{1}{\|v_2\|} \cdot v_2 = \frac{1}{\sqrt{\frac{2}{3}}} \left(\frac{1}{3}, -\frac{2}{3}, \frac{1}{3}\right)^T = \sqrt{\frac{3}{2}} \cdot \frac{1}{3} (1, -2, 1)^T = \frac{1}{\sqrt{6}} (1, -2, 1)^T$$

$$p_2 = \langle \bar{x}_3, u_1 \rangle \cdot u_1 + \langle \bar{x}_3, u_2 \rangle \cdot u_2$$

$$= (0, 2, -1) \cdot \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \frac{1}{\sqrt{3}} (1, 1, 1)^T + (0, 2, -1) \cdot \frac{1}{\sqrt{6}} (1, -2, 1)^T \cdot \frac{1}{\sqrt{6}} (1, -2, 1)^T$$

$$= \frac{1}{\sqrt{3}} (2-1) \cdot \frac{1}{\sqrt{3}} (1, 1, 1)^T + \frac{1}{\sqrt{6}} (-4-1) \cdot \frac{1}{\sqrt{6}} (1, -2, 1)^T$$

$$= \frac{1}{3} (1, 1, 1)^T + \frac{5}{6} (1, -2, 1)^T$$

$$= \left(\frac{1}{3} + \frac{5}{6}, \frac{1}{3} - \frac{5}{6}, \frac{1}{3} + \frac{5}{6}\right)^T = \left(-\frac{1}{2}, 2, -\frac{1}{2}\right)^T$$

$$v_3 = \bar{x}_3 - p_2 = (0, 2, -1)^T - \left(-\frac{1}{2}, 2, -\frac{1}{2}\right)^T = \left(\frac{1}{2}, 0, -\frac{1}{2}\right)^T$$

$$\|v_3\| = \sqrt{\frac{1}{4} + \frac{1}{4}} = \frac{1}{\sqrt{2}}, \quad u_3 = \frac{1}{\|v_3\|} \cdot v_3 = \sqrt{2} \left(\frac{1}{2}, 0, -\frac{1}{2}\right)^T$$

eg. Let  $\bar{x}_1 = (2, 1, 1, 0)^T$ ,  $\bar{x}_2 = (1, -1, 0, 1)^T$

Use G-S process to find an orthonormal basis for  $\text{Span}(\bar{x}_1, \bar{x}_2)$

$$u_1 = \frac{1}{\|\bar{x}_1\|} \cdot \bar{x}_1 = \frac{1}{\sqrt{6}} (2, 1, 1, 0)^T$$

$$\langle \bar{x}_2, u_1 \rangle = (1, -1, 0, 1) \cdot \frac{1}{\sqrt{6}} (2, 1, 1, 0)^T = \frac{1}{\sqrt{6}} (2 - 1 + 0 + 0) = \frac{1}{\sqrt{6}}$$

$$p = \langle \bar{x}_2, u_1 \rangle \cdot u_1 = \frac{1}{\sqrt{6}} \cdot \frac{1}{\sqrt{6}} (2, 1, 1, 0)^T = \frac{1}{6} (2, 1, 1, 0)^T$$

$$u_2 = \bar{x}_2 - p = (1, -1, 0, 1)^T - \frac{1}{6} (2, 1, 1, 0)^T = \left(\frac{2}{3}, -\frac{7}{6}, -\frac{1}{6}, 1\right)^T$$

$$\|u_2\| = \sqrt{\left(\frac{2}{3}\right)^2 + \left(-\frac{7}{6}\right)^2 + \left(-\frac{1}{6}\right)^2 + 1} = \sqrt{\frac{16 + 49 + 1 + 36}{36}} = \sqrt{\frac{102}{36}} = \sqrt{\frac{17}{6}}$$

$$u_2 = \frac{1}{\|u_2\|} \cdot u_2 = \frac{\sqrt{6}}{\sqrt{17}} \cdot \left(\frac{2}{3}, -\frac{7}{6}, -\frac{1}{6}, 1\right)^T$$

$\{u_1, u_2\}$  is an orthonormal basis for  $\text{Span}(\bar{x}_1, \bar{x}_2)$ .

eg. (Application to the diagonalization of the symmetric matrix)

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

eigenvalue problem of  $A$ :  $p(\lambda) = \det(\lambda I - A) = -1 = \lambda^2 - 4\lambda + 3 = 0$

two eigenvalues:  $\lambda_1 = 1$ ,  $\lambda_2 = 3$

eigenvector:  $\lambda_1 = 1$ .  $A - I = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$  eigenvector:  $v_1 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

$\lambda_2 = 3$ .  $A - 3I = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix}$  eigenvector  $v_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .

$v_1 \perp v_2$  since  $\langle v_1, v_2 \rangle = 0$ .

Let  $u_1 = \frac{1}{\|v_1\|} v_1 = \frac{1}{\sqrt{2}} (1, -1)^T$ ,  $u_2 = \frac{1}{\|v_2\|} v_2 = \frac{1}{\sqrt{2}} (1, 1)^T$ ,

Then  $u_1, u_2$  is an orthonormal basis and  $Au_1 = \lambda_1 u_1$ ,  $Au_2 = \lambda_2 u_2$ .

Let  $Q = (u_1, u_2)$ . Then

$$A(u_1, u_2) = (u_1, u_2) \cdot \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix} \Leftrightarrow AQ = Q \cdot D, \text{ where } D = \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}.$$

therefore,  $Q^T \cdot A \cdot Q = D$  (is as the usual diagonalization)

Notice that  $Q$  is an orthogonal matrix, we have  $Q^T = Q^{-1}$ .

Therefore,  $Q^T \cdot A \cdot Q = D$ .

Remark. In general, one can prove that if  $A$  is (an  $n \times n$ ) symmetric matrix then one can find an orthogonal matrix  $Q$  such that  $Q^T A Q = D$ , where  $D$  is diagonal.