## Learning Objectives

- Use the Gram-Schmidt process to construct an orthogonal basis for any nonzero subspace of  $\mathbb{R}^n$
- Understand how to find a QR factorization of a matrix with linearly independent columns

## The Gram-Schmidt Process

Our goal in this section is to develop a process for producing orthogonal bases for subspaces in  $\mathbb{R}^n$ .

**Example:** Suppose  $\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$  is a basis for a subspace W of  $\mathbb{R}^4$ . Construct an orthogonal basis for W.

Solution: We may as well take our first vector to be the same:

$$\mathbf{v}_1 = \mathbf{x}_1$$

We should use  $\mathbf{x}_2$  and  $\mathbf{v}_1$  to construct our second vector,  $\mathbf{v}_2$ , and we can use the Orthogonal Decomposition Theorem (Theorem 6.8) to do this so that  $\mathbf{v}_2$  and  $\mathbf{v}_1$  are orthogonal:

$$\mathbf{v}_2 = \mathbf{x}_2 - \operatorname{proj}_{\mathbf{v}_1} \mathbf{v}_1 = \mathbf{x}_2 - \frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1$$

For our third vector, we can use the same idea: we can project  $\mathbf{x}_3$  onto the plane  $W_2 = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$ and subtract that result from  $\mathbf{x}_3$  using Theorem 6.8:

$$\mathbf{v}_3 = \mathbf{x}_3 - \operatorname{proj}_{W_2} \mathbf{x}_3 = \mathbf{x}_3 - \hat{\mathbf{x}_3} = \mathbf{x}_3 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 + \frac{\mathbf{x}_3 \cdot \mathbf{v}_2}{\mathbf{v}_2 \cdot \mathbf{v}_2} \mathbf{v}_2\right) = \mathbf{x}_3 - \frac{\mathbf{x}_3 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 - \frac{\mathbf{x}_3 \cdot \mathbf{v}_2}{\mathbf{v}_2 \cdot \mathbf{v}_2} \mathbf{v}_2$$

Theorem 6.11: The Gram-Schmidt Process Given a basis  $\{\mathbf{x}_1, \dots, \mathbf{x}_p\}$  for a nonzero subspace W of  $\mathbb{R}^n$ , define  $\mathbf{v}_1 =$  $\mathbf{v}_2 =$ 

 $\mathbf{v}_3 =$ 

 $\vdots$  $\mathbf{v}_p =$ 

Then  $\{\mathbf{v}_1, \ldots, \mathbf{v}_p\}$  is an orthogonal basis for W, and





How to construct  $\mathbf{v}_2$  from  $\mathbf{x}_2$  and  $W_1 = \operatorname{Span}\{\mathbf{v}_1\} = \operatorname{Span}\{\mathbf{x}_1\}$ 

How to construct  $\mathbf{v}_3$  from  $\mathbf{x}_3$  and  $W_2 = \operatorname{Span}{\{\mathbf{v}_1, \mathbf{v}_2\}} = \operatorname{Span}{\{\mathbf{x}_1, \mathbf{x}_2\}}$ 

**Example:** Let  $W = \text{Span}\{\mathbf{x}_1, \mathbf{x}_2\}$ , where  $\mathbf{x}_1 = \begin{bmatrix} 1\\1\\0 \end{bmatrix}$  and  $\mathbf{x}_2 = \begin{bmatrix} 2\\2\\3 \end{bmatrix}$ . Construct an orthogonal basis  $\{\mathbf{v}_1, \mathbf{v}_2\}$  for W.

Solution: 
$$\mathbf{v}_1 = \mathbf{x}_1$$
, and  $\mathbf{v}_2 = \mathbf{x}_2 - \operatorname{proj}_{\mathbf{v}_1} \mathbf{x}_2 = \mathbf{x}_2 - \frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 =$   
(since  $\mathbf{v}_1 = \begin{bmatrix} 1\\1\\0 \end{bmatrix}$ ,  $\mathbf{x}_2 \cdot \mathbf{v}_1 =$  and  $\mathbf{v}_1 \cdot \mathbf{v}_1 =$  )

So, an orthogonal basis for W is:

## **Orthonormal Bases**

An orthonormal basis can be constructed from an orthogonal basis  $\{\mathbf{v}_1, \ldots, \mathbf{v}_k\}$  by

(i.e., scaling) \_\_\_\_\_.

**Example:** Let 
$$\mathbf{x}_1 = \begin{bmatrix} 3\\1\\-1\\3 \end{bmatrix}$$
,  $\mathbf{x}_2 = \begin{bmatrix} -5\\1\\5\\7 \end{bmatrix}$ , and  $\mathbf{x}_3 = \begin{bmatrix} 1\\1\\-2\\8 \end{bmatrix}$ , and let  $W = \operatorname{Span}\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$ .

(a) How could we show that  $\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$  is a basis for W, which is a subspace of  $\mathbb{R}^4$ ?

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(b) Find an orthogonal basis for W.

$$\mathbf{v}_1 = \mathbf{x}_1$$

$$\mathbf{v}_2 = \mathbf{x}_2 - rac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 =$$

$$\mathbf{v}_3 = \mathbf{x}_3 - rac{\mathbf{x}_3 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 - rac{\mathbf{x}_3 \cdot \mathbf{v}_2}{\mathbf{v}_2 \cdot \mathbf{v}_2} \mathbf{v}_2 =$$

$$\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \left\{ \begin{bmatrix} 3\\1\\-1\\3\end{bmatrix}, \begin{bmatrix} 1\\3\\-1\\-1\end{bmatrix}, \begin{bmatrix} -3\\1\\1\\3\end{bmatrix} \right\} \text{ is an orthogonal basis for } W.$$

(c) Find an orthonormal basis for W.

We need to normalize  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  to make an orthonormal basis  $\{\mathbf{v}_1', \mathbf{v}_2', \mathbf{v}_3'\}$ :

$$\mathbf{v}_{1}' = \frac{1}{\|\mathbf{v}_{1}\|} \mathbf{v}_{1} = \begin{bmatrix} 3/\sqrt{20} \\ 1/\sqrt{20} \\ -1/\sqrt{20} \\ 3/\sqrt{20} \end{bmatrix} \qquad \mathbf{v}_{2}' = \frac{1}{\|\mathbf{v}_{2}\|} \mathbf{v}_{2} = \mathbf{v}_{3}' =$$

## QR Factorization of Matrices

If an  $m \times n$  matrix A has linearly independent columns, then applying the Gram-Schmidt process (with normalizations) to the \_\_\_\_\_\_ of A produces an extremely useful factorization for A.



Note: If the columns of Q form an orthonormal basis for  $\operatorname{Col} A,$  then

$$Q^T Q =$$

Therefore, we can find  ${\cal R}$  by

$$R = IR = Q^T(QR) = Q^TA$$