<u>6.3 – Orthogonal Projections</u> <u>6.4 – Gram-Schmidt Process</u>

Given a vector \mathbf{y} and a subspace W in \mathbb{R}^n there is a vector $\hat{\mathbf{y}} \in W$ such that

- 1) \hat{y} is the unique vector in ${\it W}$ for which $y\!-\!\hat{y}$ is orthogonal to ${\it W}$
- 2) $\hat{\mathbf{y}}$ is the unique vector in *W* closest to \mathbf{y}

Theorem 8 The Orthogonal Decomposition Theorem

Let W be a subspace of \mathbb{R}^n . Then each y in \mathbb{R}^n can be written uniquely in the form

$$\mathbf{y} = \widehat{\mathbf{y}} + \mathbf{z}$$
 (1)

where $\widehat{\mathbf{y}}$ is in W and \mathbf{z} is in W^{\perp} . In fact, if $\{\mathbf{u}_1,\ldots,\mathbf{u}_p\}$ is any orthogonal basis of W, then

$$\widehat{\mathbf{y}} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \dots + \frac{\mathbf{y} \cdot \mathbf{u}_p}{\mathbf{u}_p \cdot \mathbf{u}_p} \mathbf{u}_p$$
 (2)

and $\mathbf{z} = \mathbf{y} - \widehat{\mathbf{y}}$.

Ex 1: Let $W = \text{Span}\{\mathbf{u}_1, \mathbf{u}_2\}$. Write **y** as the sum of a vector in W and a vector orthogonal to W.

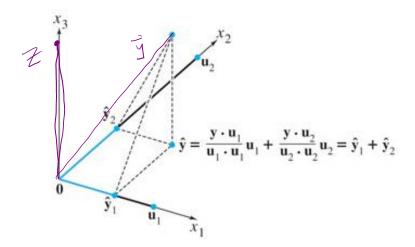
$$\begin{aligned} y &= y + z \\ \hat{y} &= \frac{y \cdot u_{1}}{u_{1} \cdot u_{1}} \quad u_{1} + \frac{y \cdot u_{2}}{u_{2} \cdot u_{2}} \quad u_{2} = \frac{1 + 9 - 10}{1 + 9 + 4} \begin{bmatrix} 1 \\ -2 \end{bmatrix} + \frac{5 + 3 + 20}{25 + 1 + 16} \begin{bmatrix} 5 \\ -1 \\ 4 \end{bmatrix} = \frac{28}{42} \begin{bmatrix} 5 \\ -1 \\ 4 \end{bmatrix} = \begin{bmatrix} 19/3 \\ -2 \\ -3 \end{bmatrix} \\ Z = y - \hat{y} = \begin{bmatrix} 1 \\ -7/3 \\ -7/3 \\ -7/3 \\ -7/3 \end{bmatrix} = \begin{bmatrix} -7/3 \\ -7/3 \\ -7/3 \\ -7/3 \\ -7/3 \\ -7/3 \end{bmatrix} \quad y = \begin{bmatrix} 19/3 \\ -7/3$$



Warnock - Class Notes

orthogonal

 $\mathbf{y} = \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix}, \mathbf{u}_1 = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}, \mathbf{u}_2 = \begin{bmatrix} 5 \\ 1 \\ 4 \end{bmatrix}$



Theorem 9 The Best Approximation Theorem

Let W be a subspace of \mathbb{R}^n , let \mathbf{y} be any vector in \mathbb{R}^n , and let $\widehat{\mathbf{y}}$ be the orthogonal projection of \mathbf{y} onto W. Then $\widehat{\mathbf{y}}$ is the closest point in W to \mathbf{y} , in the sense that

$$\|\mathbf{y} - \widehat{\mathbf{y}}\| < \|\mathbf{y} - \mathbf{v}\| \tag{3}$$

for all **v** in *W* distinct from $\widehat{\mathbf{y}}$.

Ex 2: As in Ex 1,
$$\begin{bmatrix} 10/3\\2/3\\8/3 \end{bmatrix}$$
 is the closest point in $W = \text{Span}\left\{\mathbf{u}_1 = \begin{bmatrix} 1\\3\\-2 \end{bmatrix}, \mathbf{u}_2 = \begin{bmatrix} 5\\1\\4 \end{bmatrix}\right\}$ to $\mathbf{y} = \begin{bmatrix} 1\\3\\5 \end{bmatrix}$.
Find the distance from \mathbf{y} to W

$$\left|\left|\mathbf{y} - \frac{\mathbf{y}}{\mathbf{y}}\right|\right| = \left|\left| \neq \right| \left| = \int \frac{4q}{q} + \frac{4q}{q} + \frac{4q}{q}}{q} = \begin{bmatrix} \frac{7}{3} \sqrt{3} \\ \frac{7}{3} \sqrt{3} \end{bmatrix}$$

$$\begin{bmatrix} -7/3\\7/3\\.7/3 \end{bmatrix}$$

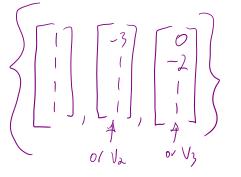
Practice Problems

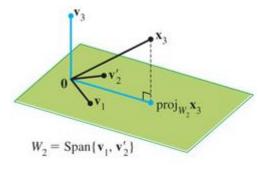
2. Let W be the subspace spanned by the **u**'s, and write **y** as the sum of a vector in W and a vector orthogonal to W.

The Gram-Schmidt Process

Ex 3: Let
$$W = \operatorname{Span} \left\{ \mathbf{x}_{1} = \begin{bmatrix} 3 \\ 0 \\ -1 \end{bmatrix}, \mathbf{x}_{2} = \begin{bmatrix} 8 \\ 5 \\ -6 \end{bmatrix} \right\}$$
, construct an orthogonal basis $\{\mathbf{v}_{1}, \mathbf{v}_{2}\}$.
 $\bigvee_{1} = \chi_{1}$ find a proj of χ_{p} on to χ_{1} , then $\overline{\psi}_{2} = \chi_{2} - \overline{\chi}_{2}$
 $(\overline{z} =) \quad \overline{\psi}_{2} = \chi_{2} - \frac{\chi_{2} \cdot \chi_{1}}{\chi_{1} \cdot \chi_{1}} \times_{1} = \begin{bmatrix} \frac{5}{2} \\ -\frac{5}{2} \end{bmatrix} - \frac{24 + o + 6}{2 + o + 1} \begin{bmatrix} 3 \\ -\frac{7}{2} \end{bmatrix} = \begin{bmatrix} 5 \\ -\frac{5}{2} \end{bmatrix} - \begin{bmatrix} -\frac{7}{2} \\ -\frac{7}{2} \end{bmatrix}$
Ex 4:
Let $\mathbf{x}_{1} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{x}_{2} = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \text{ and } \mathbf{x}_{3} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}$. Then $\{\mathbf{x}_{1}, \mathbf{x}_{2}, \mathbf{x}_{3}\}$ is
clearly linearly independent and thus is a basis for a subspace W of \mathbb{R}^{4} .
Construct an orthogonal basis for W .
 $\bigvee_{l} = \chi_{2} - \frac{\chi_{3} \cdot V_{l}}{V_{l} \cdot V_{l}} \times_{l} = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix} - \frac{3}{4} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -\frac{3}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \end{bmatrix}$

$$V_{3} = X_{3} - \frac{X_{5} V_{1} V_{1}}{V_{1} V_{1}} - \frac{X_{3} V_{2}}{V_{2} V_{2}} V_{2} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} - \frac{2}{4} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \frac{2}{12} \begin{bmatrix} -3 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 - \frac{1}{3} + \frac{1}{2} \\ 0 - \frac{1}{3} - \frac{1}{6} \\ 1 - \frac{1}{2} - \frac{1}{6} \\ 1 - \frac{1}{2} - \frac{1}{6} \end{bmatrix} = \begin{bmatrix} 0 - \frac{1}{3} + \frac{1}{2} \\ 0 - \frac{1}{3} - \frac{1}{6} \\ 1 - \frac{1}{2} - \frac{1}{6} \\ 1 - \frac{1}{2} - \frac{1}{6} \end{bmatrix}$$





Theorem 11 The Gram–Schmidt Process

Given a basis $\{\mathbf{x}_1,\ldots,\,\mathbf{x}_p\}$ for a nonzero subspace W of $\mathbb{R}^n,\,$ define

$$\mathbf{v}_{1} = \mathbf{x}_{1}$$

$$\mathbf{v}_{2} = \mathbf{x}_{2} - \frac{\mathbf{x}_{2} \cdot \mathbf{v}_{1}}{\mathbf{v}_{1} \cdot \mathbf{v}_{1}} \mathbf{v}_{1}$$

$$\mathbf{v}_{3} = \mathbf{x}_{3} - \frac{\mathbf{x}_{3} \cdot \mathbf{v}_{1}}{\mathbf{v}_{1} \cdot \mathbf{v}_{1}} \mathbf{v}_{1} - \frac{\mathbf{x}_{3} \cdot \mathbf{v}_{2}}{\mathbf{v}_{2} \cdot \mathbf{v}_{2}} \mathbf{v}_{2}$$

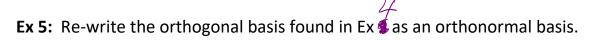
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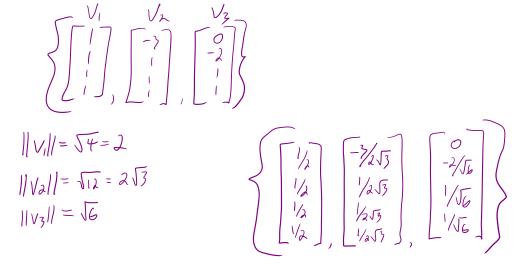
$$\mathbf{v}_{p} = \mathbf{x}_{p} - \frac{\mathbf{x}_{p} \cdot \mathbf{v}_{1}}{\mathbf{v}_{1} \cdot \mathbf{v}_{1}} \mathbf{v}_{1} - \frac{\mathbf{x}_{p} \cdot \mathbf{v}_{2}}{\mathbf{v}_{2} \cdot \mathbf{v}_{2}} \mathbf{v}_{2} - \dots - \frac{\mathbf{x}_{p} \cdot \mathbf{v}_{p-1}}{\mathbf{v}_{p-1} \cdot \mathbf{v}_{p-1}} \mathbf{v}_{p-1}$$

Then $\{\mathbf{v}_1,\ldots,\,\mathbf{v}_p\}$ is an orthogonal basis for *W*. In addition $\mathrm{Span}\;\{\mathbf{v}_1,\ldots,\,\mathbf{v}_k\}=\mathrm{Span}\;\{\mathbf{x}_1,\ldots,\,\mathbf{x}_k\}\quad ext{for }1\leq k\leq p$

The result of this is that every nonzero subspace W in \mathbb{R}^n has an orthogonal basis.

An orthonormal basis is constructed easily by normalizing all the \mathbf{v}_k 's to unit vectors.





Practice Problems

1. Let
$$W = \text{Span} \{\mathbf{x}_1, \mathbf{x}_2\}$$
, where $\mathbf{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ and $\mathbf{x}_2 = \begin{bmatrix} 1/3 \\ 1/3 \\ -2/3 \end{bmatrix}$

Construct an orthonormal basis for W.

$$V_{1} = X_{1}$$

$$V_{2} = X_{2} - \frac{X_{2} \cdot V_{1}}{V_{1} \cdot V_{1}} V_{1} = \begin{bmatrix} \frac{1}{3} \\ \frac{1}{3} \\ -\frac{1}{3} \end{bmatrix} - \frac{\frac{1}{3} + \frac{1}{3} - \frac{2}{3}}{3} V_{1} = V_{2} = \begin{bmatrix} \frac{1}{3} \\ \frac{1}{3} \\ -\frac{2}{3} \end{bmatrix} ||X_{1}|| = \begin{bmatrix} \frac{6}{9} \\ -\frac{1}{3} \\ -\frac{1}{3} \end{bmatrix} = \frac{16}{3}$$

$$= \frac{16}{3}$$

$$\frac{X_{1}}{||X_{1}||} = \sqrt{\frac{16}{3}} \int_{-\frac{1}{3}}^{1} \frac{1}{3} \int_{-\frac{1}{3}}^{1} \frac{1}{3$$

2. Use the Gram–Schmidt process to produce an orthogonal basis for W.

$$W = \operatorname{Span} \{ \mathbf{x}_{1}, \mathbf{x}_{2}, \mathbf{x}_{3} \} \text{ where } \mathbf{x}_{1} = \begin{bmatrix} -1 \\ 3 \\ 1 \\ 1 \end{bmatrix}, \mathbf{x}_{2} = \begin{bmatrix} 6 \\ -8 \\ -2 \\ -4 \end{bmatrix}, \mathbf{x}_{3} = \begin{bmatrix} 6 \\ 3 \\ 6 \\ -3 \end{bmatrix}$$

$$\bigvee_{1} = \times_{1}$$

$$\bigvee_{d} = \times_{a} - \frac{\times_{a} \cdot \bigvee_{1}}{\bigvee_{1} \cdot \bigvee_{1}} \bigvee_{1} = \begin{bmatrix} 6 \\ -\frac{9}{-\lambda} \\ -\lambda \\ -4 \end{bmatrix} - \frac{-26}{12} \begin{bmatrix} -1 \\ 7 \\ 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 6 + 3(-1) \\ -8 + 3(3) \\ -2 + 3(1) \\ -4 + 3(1) \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \\ -1 \\ -1 \end{bmatrix}$$

$$\bigvee_{3} = \times_{3} - \frac{\times_{5} \cdot \bigvee_{1}}{\bigvee_{1} \cdot \bigvee_{1} - \frac{\times_{5} \cdot \bigvee_{2}}{\bigvee_{2} \cdot \bigvee_{2}}} \bigvee_{2} = \begin{bmatrix} 6 \\ 3 \\ -3 \\ -3 \\ -3 \end{bmatrix} - \frac{6}{12} \begin{bmatrix} -1 \\ 7 \\ 1 \\ -1 \end{bmatrix} - \frac{300}{12} \begin{bmatrix} 3 \\ -1 \\ -1 \\ -1 \\ -1 \end{bmatrix}$$

$$= \begin{bmatrix} 6 + \bigvee_{2} - 15 \\ 2 \\ -3 - \bigvee_{2} - 5 \\ -3 \\ -3 - \bigvee_{2} - 5 \\ -3 - \bigvee_{2} - 5 \\ -3 \\ -1 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \end{bmatrix} = \begin{bmatrix} 7 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \end{bmatrix}$$