

7.9. Orthonormal basis and the Gram-Schmidt Process

We can find an orthonormal basis for any vector space using Gram-Schmidt process. Such bases are very useful. Orthogonal projections can be computed using dot products. Fourier series, wavelets, and so on from these.

Orthogonal basis. Orthonormal basis

- Orthogonal basis: A basis that is an orthogonal set.
- Orthonormal basis: A basis that is an orthonormal set.
- Example 1: $\{(0,1,0), (1,0,1), (-1,0,1)\}$
- Example 2: $\{(3/7,-6/7,2/7), (2/7,3/7,6/7), (6/7,2/7,-3/7)\}$
- Example 3: The standard basis of \mathbb{R}^n .

Theorem 7.9.1 *An orthogonal set of nonzero vectors in R^n is linearly independent.*

- Proof: v_1, v_2, \dots, v_k Orthogonal set.
 - Suppose $c_1v_1 + c_2v_2 + \dots + c_kv_k = 0$.
 - Dot with v_1 . $c_1v_1 \cdot v_1 = 0$. Since v_1 has nonzero length, $c_1 = 0$.
 - Do for each v_j s. Thus all $c_j = 0$.
- Thus an orthogonal (orthonormal) set of n nonzero vectors is a basis always.

How to find these?

Orthogonal projections using orthonormal projections

- $\text{Proj}_W x = M(M^T M)^{-1} M^T(x)$.
- Recall M has columns that form a basis of W .
- Suppose we chose the orthonormal basis of W .
- $M^T M = I$ by orthonormality.
- Thus $\text{Proj}_W(x) = M M^T x$.
- $P = M M^T$.
- Example 4.

Theorem 7.9.2

(a) If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ is an orthonormal basis for a subspace W of R^n , then the orthogonal projection of a vector \mathbf{x} in R^n onto W can be expressed as

$$\text{proj}_W \mathbf{x} = (\mathbf{x} \cdot \mathbf{v}_1)\mathbf{v}_1 + (\mathbf{x} \cdot \mathbf{v}_2)\mathbf{v}_2 + \dots + (\mathbf{x} \cdot \mathbf{v}_k)\mathbf{v}_k \quad (7)$$

(b) If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ is an orthogonal basis for a subspace W of R^n , then the orthogonal projection of a vector \mathbf{x} in R^n onto W can be expressed as

$$\text{proj}_W \mathbf{x} = \frac{\mathbf{x} \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 + \frac{\mathbf{x} \cdot \mathbf{v}_2}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 + \dots + \frac{\mathbf{x} \cdot \mathbf{v}_k}{\|\mathbf{v}_k\|^2} \mathbf{v}_k \quad (8)$$

- Proof: (a) $M = [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k]$.

$$\begin{aligned} \text{proj}_W \mathbf{x} &= M(M^T \mathbf{x}) = [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k] \begin{bmatrix} \mathbf{v}_1^T \\ \mathbf{v}_2^T \\ \vdots \\ \mathbf{v}_k^T \end{bmatrix} \mathbf{x} \\ &= [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k] \begin{bmatrix} \mathbf{v}_1^T \mathbf{x} \\ \mathbf{v}_2^T \mathbf{x} \\ \vdots \\ \mathbf{v}_k^T \mathbf{x} \end{bmatrix} = (x \cdot \mathbf{v}_1)\mathbf{v}_1 + (x \cdot \mathbf{v}_2)\mathbf{v}_2 + \dots + (x \cdot \mathbf{v}_k)\mathbf{v}_k \end{aligned}$$

- Proof(b): Divide by the lengths to obtain an orthonormal basis of W . Apply (a).
- Note: Even if $W=\mathbb{R}^n$, one can use the same formula.

Theorem 7.9.3 *If P is the standard matrix for an orthogonal projection of \mathbb{R}^n onto a subspace of \mathbb{R}^n , then $\text{tr}(P) = \text{rank}(P)$.*

- Proof: $P=MM^T=v_1v_1^T+\dots+v_kv_k^T$.
 - $\text{tr}P=\text{tr}(v_1v_1^T)+\dots+\text{tr}(v_kv_k^T)=v_1 \cdot v_1+\dots+v_k \cdot v_k=k$
 - This by Formula 27 in Sec 3.1.
- Example 7: $13/49+45/49+40/49=2$ (Example 4)

Linear combinations of orthonormal basis vectors.

- If w is in W , then $\text{proj}_W(w)=w$. In particular, if $W=\mathbb{R}^n$, and w any vector, we have

Theorem 7.9.4

(a) If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ is an orthonormal basis for a subspace W of \mathbb{R}^n , and if \mathbf{w} is a vector in W , then

$$\mathbf{w} = (\mathbf{w} \cdot \mathbf{v}_1)\mathbf{v}_1 + (\mathbf{w} \cdot \mathbf{v}_2)\mathbf{v}_2 + \cdots + (\mathbf{w} \cdot \mathbf{v}_k)\mathbf{v}_k \quad (11)$$

(b) If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ is an orthogonal basis for a subspace W of \mathbb{R}^n , and if \mathbf{w} is a vector in W , then

$$\mathbf{w} = \frac{\mathbf{w} \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2}\mathbf{v}_1 + \frac{\mathbf{w} \cdot \mathbf{v}_2}{\|\mathbf{v}_2\|^2}\mathbf{v}_2 + \cdots + \frac{\mathbf{w} \cdot \mathbf{v}_k}{\|\mathbf{v}_k\|^2}\mathbf{v}_k \quad (12)$$

- The above formula is very useful to find “coordinates” given an orthonormal basis.
- Example 8:

Gram-Schmidt orthogonalization process

- W a nonzero subspace $\{w_1, w_2, \dots, w_k\}$ Any basis
- We will produce orthogonal basis $\{v_1, v_2, \dots, v_k\}$
- Let $v_1 = w_1$.
- $v_2 = w_2 - \text{proj}_{W_1}(w_2) = w_2 - \frac{w_2 \cdot v_1}{\|v_1\|^2} v_1$.
 - v_2 is not zero. (Otherwise, $w_2 = \text{proj}_{W_1}(w_2) \in W_1$.)
 - $\{v_1, v_2\}$ orthogonal set. Let $W_2 = \text{Span}\{v_1, v_2\}$
- $v_3 = w_3 - \text{proj}_{W_2}(w_3) = w_3 - \frac{w_3 \cdot v_1}{\|v_1\|^2} v_1 - \frac{w_3 \cdot v_2}{\|v_2\|^2} v_2$.
- v_3 is nonzero since w_3 is not in W_2 by independence of $\{w_1, w_2, w_3\}$. v_3 is orthogonal to v_1 and v_2 .

- We obtained orthogonal set of v_1, v_2, \dots, v_l . Let $W_l = \text{Span}\{v_1, \dots, v_l\}$.
- $v_{l+1} = w_{l+1} - \text{proj}_{W_l}(w_{l+1}) = w_{l+1} - v_1(w_{l+1} \cdot v_1) / \|v_1\|^2 - \dots - v_l(w_{l+1} \cdot v_l) / \|v_l\|^2$
- Then v_{l+1} is not 0 since w_{l+1} is not in W_l .
- v_{l+1} is orthogonal to v_1, \dots, v_l .
 - $v_i \cdot (w_{l+1} - v_1(w_{l+1} \cdot v_1) / \|v_1\|^2 - \dots - v_l(w_{l+1} \cdot v_l) / \|v_l\|^2) = v_i \cdot w_{l+1} - v_i \cdot v_i (w_{l+1} \cdot v_i) / \|v_i\|^2 = 0$ for $i=1, \dots, l$.
- Finally, we achieve v_1, v_2, \dots, v_k .
- We can normalize to obtain an orthonormal basis.

- Example 9: $(0,0,0,1), (0,0,1,1), (0,1,1,1), (1,1,1,1)$.
- Example 10: $x+y+z+2t = 0, 2x+y+z+t=0$.
- Properties:

Theorem 7.9.6 *If $S = \{\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_k\}$ is a basis for a nonzero subspace of R^n , and if $S' = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ is the corresponding orthogonal basis produced by the Gram–Schmidt process, then:*

- (a) $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_j\}$ is an orthogonal basis for $\text{span}\{\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_j\}$ at the j th step.
- (b) \mathbf{v}_j is orthogonal to $\text{span}\{\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_{j-1}\}$ at the j th step ($j \geq 2$).

Extending the orthonormal set to orthonormal basis.

Theorem 7.9.7 *If W is a nonzero subspace of R^n , then:*

- (a) Every orthogonal set of nonzero vectors in W can be enlarged to an orthogonal basis for W .*
- (b) Every orthonormal set in W can be enlarged to an orthonormal basis for W .*

- Proof (a): Given v_1, \dots, v_k . Add v_{k+1} orthogonal to $\text{Span}\{v_1, \dots, v_k\}$. Add v_{k+2} orthogonal to $\text{Span}\{v_1, v_2, \dots, v_k, v_{k+1}\}$. By induction....
- Proof (b): see book