

## Gram-Schmidt Process

Gram-Schmidt is an algorithm that takes a set of linearly independent vectors  $\{\vec{s}_1, \dots, \vec{s}_n\}$  and generates an orthonormal set of vectors  $\{\vec{q}_1, \dots, \vec{q}_n\}$  that span the same vector space as the original set. Concretely,  $\{\vec{q}_1, \dots, \vec{q}_n\}$  satisfy the following:

- $\forall 0 < k \leq n, \text{span}(\{\vec{s}_1, \dots, \vec{s}_k\}) = \text{span}(\{\vec{q}_1, \dots, \vec{q}_k\})$
- $\{\vec{q}_1, \dots, \vec{q}_n\}$  is an orthonormal set of vectors

### Definition: Orthonormal

A set of vectors  $\{\vec{v}_1, \dots, \vec{v}_n\}$  is **orthonormal** if all the vectors are mutually orthogonal to each other and all are of unit length. That is:

- **Orthogonal:** For all pairs of vectors  $\vec{v}_i, \vec{v}_j$  where  $i \neq j$ ,  $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ . For real vectors, this means  $\vec{v}_i^T \vec{v}_j = 0$ .
- **Normalized:** For all  $i$ ,  $\|\vec{v}_i\| = 1$ . (This implies that  $\|\vec{v}_i\| = \langle \vec{v}_i, \vec{v}_i \rangle = 1$ .)

The Gram-Schmidt algorithm works by first finding the unit vector  $\vec{q}_1$  such that  $\text{span}(\{\vec{q}_1\}) = \text{span}(\{\vec{s}_1\})$ . Subsequently, the unit vector  $\vec{q}_2$  is calculated such that  $\langle \vec{q}_1, \vec{q}_2 \rangle = 0$  and  $\text{span}(\{\vec{q}_1, \vec{q}_2\}) = \text{span}(\{\vec{s}_1, \vec{s}_2\})$ . This is continued through  $n$  vectors, resulting in the orthonormal set of vectors  $\{\vec{q}_1, \dots, \vec{q}_n\}$  that span the same vector space as  $\{\vec{s}_1, \dots, \vec{s}_n\}$ .

How is this done? Finding  $\vec{q}_1$  is straightforward, since it is the first vector in our new set, and therefore we must only satisfy  $\|\vec{q}_1\| = 1$  and  $\text{span}(\{\vec{q}_1\}) = \text{span}(\{\vec{s}_1\})$ . Since  $\text{span}(\{\vec{s}_1\})$  is a one dimensional vector space, the unit vector that spans the same vector space would just be the unit vector in the same direction as  $\vec{s}_1$ . We have

$$\vec{q}_1 = \frac{\vec{s}_1}{\|\vec{s}_1\|}. \quad (1)$$

Calculating  $\vec{q}_2$  requires that we satisfy:

- Spanning the same vector space as original set:  $\text{span}(\{\vec{q}_1, \vec{q}_2\}) = \text{span}(\{\vec{s}_1, \vec{s}_2\})$
- Orthogonal to previous vectors:  $\langle \vec{q}_1, \vec{q}_2 \rangle = 0$
- Normalized:  $\|\vec{q}_2\| = 1$

Using the vector  $\vec{q}_1$  that we calculated above, we notice that

$$\text{span}(\{\vec{q}_1, \vec{s}_2\}) = \text{span}(\{\vec{s}_1, \vec{s}_2\}),$$

satisfying the first condition. However,  $\vec{q}_1$  and  $\vec{s}_2$  are not necessarily orthogonal.

We know from EE 16A that the following subspaces are equivalent for any pair of linearly independent vectors  $\vec{v}_1, \vec{v}_2$ :

- $\text{span}(\vec{v}_1, \vec{v}_2)$
- $\text{span}(\vec{v}_1, \alpha\vec{v}_2)$
- $\text{span}(\vec{v}_1, \vec{v}_1 + \vec{v}_2)$
- $\text{span}(\vec{v}_1, \vec{v}_1 - \vec{v}_2)$
- $\text{span}(\vec{v}_1, \vec{v}_2 - \alpha\vec{v}_1)$

Let us choose vector

$$\vec{z}_2 = \vec{s}_2 - \alpha\vec{q}_1,$$

which will also have the same span as  $\{\vec{q}_1, \vec{s}_2\}$  (and therefore the same span as  $\{\vec{s}_1, \vec{s}_2\}$ ).

What should  $\alpha$  be if we would like  $\vec{q}_1$  and  $\vec{z}_2 = \vec{s}_2 - \alpha\vec{q}_1$  to be orthogonal to each other? We know from working with Orthogonal Matching Pursuit (OMP), that  $\vec{s}_2 - \text{proj}_{\vec{q}_1}(\vec{s}_2)$  will be orthogonal to  $\vec{q}_1$ , where

$$\text{proj}_{\vec{q}_1}(\vec{s}_2) = \frac{\vec{q}_1^T \vec{s}_2}{\|\vec{q}_1\|^2} \vec{q}_1$$

is the projection of  $\vec{s}_2$  onto  $\vec{q}_1$ . This makes sense because the projection of  $\vec{s}_2$  onto  $\vec{q}_1$  provides the component of  $\vec{s}_2$  that is along  $\vec{q}_1$ . Subtracting off this component from  $\vec{s}_2$  will only leave components of  $\vec{s}_2$  that are orthogonal to  $\vec{q}_1$ .

Therefore, if we set

$$\alpha\vec{q}_1 = \text{proj}_{\vec{q}_1}(\vec{s}_2),$$

the resulting

$$\vec{z}_2 = \vec{s}_2 - \text{proj}_{\vec{q}_1}(\vec{s}_2) = \vec{s}_2 - \frac{\vec{q}_1^T \vec{s}_2}{\|\vec{q}_1\|^2} \vec{q}_1,$$

will be orthogonal to  $\vec{q}_1$ .

To back up this intuition, let's solve for  $\vec{z}_2$  algebraically using the definition of orthogonality:

$$\vec{q}_1^T \vec{z}_2 = 0 \tag{2}$$

$$\vec{q}_1^T (\vec{s}_2 - \alpha\vec{q}_1) = 0 \tag{3}$$

$$\vec{q}_1^T \vec{s}_2 - \alpha \|\vec{q}_1\|^2 = 0 \tag{4}$$

$$\alpha = \frac{\vec{q}_1^T \vec{s}_2}{\|\vec{q}_1\|^2} \tag{5}$$

$$\rightarrow \vec{z}_2 = \vec{s}_2 - \frac{\vec{q}_1^T \vec{s}_2}{\|\vec{q}_1\|^2} \vec{q}_1 \tag{6}$$

Now we normalize  $\vec{z}_2$  to complete the process of finding the  $\vec{q}_2$  which satisfies all three conditions above:

$$\vec{q}_2 = \frac{\vec{z}_2}{\|\vec{z}_2\|}$$

In the question below, you will work through how this methodology leads to the Gram-Schmidt algorithm for calculating the orthonormal set  $\{\vec{q}_1, \dots, \vec{q}_n\}$  from  $n$  linearly independent vectors  $\{\vec{s}_1, \dots, \vec{s}_n\}$ .

# Questions

## 1. Gram-Schmidt Algorithm

Now let's see how we can do this with a set of three linearly independent vectors  $\{\vec{s}_1, \vec{s}_2, \vec{s}_3\}$ .

- Find unit vector  $\vec{q}_1$  such that  $\text{span}(\{\vec{q}_1\}) = \text{span}(\{\vec{s}_1\})$ .
- Given  $\vec{q}_1$  from the previous step, find  $\vec{q}_2$  such that  $\text{span}(\{\vec{q}_1, \vec{q}_2\}) = \text{span}(\{\vec{s}_1, \vec{s}_2\})$  and  $\vec{q}_2$  is orthogonal to  $\vec{q}_1$ .
- Now given  $\vec{q}_1$  and  $\vec{q}_2$  in the previous steps, find  $\vec{q}_3$  such that  $\text{span}(\{\vec{q}_1, \vec{q}_2, \vec{q}_3\}) = \text{span}(\{\vec{s}_1, \vec{s}_2, \vec{s}_3\})$ .
- Let's extend this algorithm to  $n$  linearly independent vectors. That is, given an input  $\{\vec{s}_1, \dots, \vec{s}_n\}$ , write the algorithm to calculate the orthonormal set of vectors  $\{\vec{q}_1, \dots, \vec{q}_n\}$ , where  $\text{span}(\{\vec{s}_1, \dots, \vec{s}_n\}) = \text{span}(\{\vec{q}_1, \dots, \vec{q}_n\})$ . *Hint: How would you calculate the  $i^{\text{th}}$  vector,  $\vec{q}_i$ ?*

## 2. The Order of Gram-Schmidt

- If we are performing the Gram-Schmidt method on a set of vectors, does the order in which we take the vectors matter? Consider the set of vectors

$$\left\{ \vec{v}_1, \vec{v}_2, \vec{v}_3 \right\} = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\} \quad (7)$$

Perform Gram-Schmidt on these vectors first in the order  $\vec{v}_1, \vec{v}_2, \vec{v}_3$ . and then in the order  $\vec{v}_3, \vec{v}_2, \vec{v}_1$ . Do you get the same answer?

- Now perform Gram-Schmidt on these vectors in the order  $\vec{v}_3, \vec{v}_2, \vec{v}_1$ . Do you get the same result?

## 3. Orthonormal Matrices and Projections

An orthonormal matrix,  $\mathbf{A}$ , is a matrix whose columns,  $\vec{a}_i$ , are:

- Orthogonal (ie.  $\langle \vec{a}_i, \vec{a}_j \rangle = 0$  when  $i \neq j$ )
- Normalized (ie. vectors with length equal to 1,  $\|\vec{a}_i\| = 1$ ). This implies that  $\|\vec{a}_i\|_2 = \langle \vec{a}_i, \vec{a}_i \rangle = 1$ .

- Suppose that the matrix  $\mathbf{A} \in \mathbb{R}^{N \times M}$  has linearly independent columns. The vector  $\vec{y}$  in  $\mathbb{R}^N$  is not in the subspace spanned by the columns of  $\mathbf{A}$ . What is the projection of  $\vec{y}$  onto the subspace spanned by the columns of  $\mathbf{A}$ ?
- Show if  $\mathbf{A} \in \mathbb{R}^{N \times N}$  is an orthonormal matrix then the columns,  $\vec{a}_i$ , form a basis for  $\mathbb{R}^N$ .
- When  $\mathbf{A} \in \mathbb{R}^{N \times M}$  and  $N \geq M$  (i.e. tall matrices), show that if the matrix is orthonormal, then  $\mathbf{A}^T \mathbf{A} = \mathbf{I}_{M \times M}$ .
- Again, suppose  $\mathbf{A} \in \mathbb{R}^{N \times M}$  where  $N \geq M$  is an orthonormal matrix. Show that the projection of  $\vec{y}$  onto the subspace spanned by the columns of  $\mathbf{A}$  is now  $\mathbf{A} \mathbf{A}^T \vec{y}$ .

## 4. (Optional) Orthogonal Coordinate Change Examples

(a) **Transformation from Standard Basis to another Orthonormal Basis in  $\mathbb{R}^3$**

Calculate the coordinate transformation between the following bases

$$U = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad V = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix} \quad (8)$$

i.e. find a matrix  $T$  such that  $\vec{x}_v = T\vec{x}_u$  where  $\vec{x}_u$  is the coordinates of a vector in a basis of the columns of  $U$  and  $\vec{x}_v$  is the coordinates of the same vector in the basis of the columns of  $V$ .

Draw a picture of the two different coordinate frames. Let  $\vec{x}_u = [1, 0, 0]^T$ . Compute  $\vec{x}_v$  and compare the results with your picture. Repeat this for  $\vec{x}_u = [0, 1, 0]^T$ . Are the results intuitive?

Now let  $\vec{x}_u = [1, 2, 1]^T$ . What is  $\vec{x}_v$ ? How would you verify that this is correct?

(b) **Transformation between Two Orthonormal Bases in  $\mathbb{R}^3$**

Calculate the coordinate transformation between the following bases

$$U = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix} \quad V = \begin{bmatrix} 0 & 0 & 1 \\ 1/\sqrt{2} & -1/\sqrt{2} & 0 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \end{bmatrix} \quad (9)$$

i.e. find a matrix  $T$  such that  $\vec{x}_v = T\vec{x}_u$ . Draw a picture of the two different coordinate frames. Let  $\vec{x}_u = [1, 0, 0]^T$ . Compute  $\vec{x}_v$  and compare the results with your picture. Repeat this for  $\vec{x}_u = [0, 1, 0]^T$ . Are the results intuitive?

Now let  $\vec{x}_u = [1/\sqrt{2}, 1, 1/\sqrt{2}]^T$ . What is  $\vec{x}_v$ ? How would you verify that this is correct?

## 5. (Extra Practice) Gram-Schmidt

(a) Use Gram-Schmidt to find an orthonormal basis for the following three vectors.

$$\vec{v}_1 = \begin{bmatrix} \sqrt{2} \\ -\sqrt{2} \\ 0 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \vec{v}_3 = \begin{bmatrix} \sqrt{2} \\ 0 \\ -\sqrt{2} \end{bmatrix}$$

(b) Express  $\vec{v}_1$ ,  $\vec{v}_2$ , and  $\vec{v}_3$  as vectors in the basis you found in part (a).

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