

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 α 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^{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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