

Harvey Mudd College Math Tutorial:

The Gram-Schmidt Algorithm

In any **inner product space**, we can choose the basis in which to work. It often greatly simplifies calculations to work in an **orthogonal** basis. For one thing, if $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is an orthogonal basis for an inner product space V , then it is a simple matter to express any vector $\mathbf{w} \in V$ as a linear combination of the vectors in S :

$$\mathbf{w} = \frac{\langle \mathbf{w}, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 + \frac{\langle \mathbf{w}, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 + \dots + \frac{\langle \mathbf{w}, \mathbf{v}_n \rangle}{\|\mathbf{v}_n\|^2} \mathbf{v}_n.$$

Given an arbitrary basis $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ for an n -dimensional inner product space V , the **Gram-Schmidt algorithm** constructs an orthogonal basis $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ for V :

That is, \mathbf{w} has coordinates

$$\begin{bmatrix} \frac{\langle \mathbf{w}, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 \\ \frac{\langle \mathbf{w}, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 \\ \vdots \\ \frac{\langle \mathbf{w}, \mathbf{v}_n \rangle}{\|\mathbf{v}_n\|^2} \mathbf{v}_n \end{bmatrix}$$

relative to the basis S .

Step 1 Let $\mathbf{v}_1 = \mathbf{u}_1$.

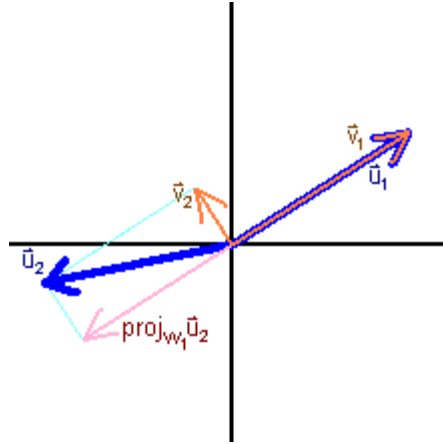
Step 2 Let $\mathbf{v}_2 = \mathbf{u}_2 - \text{proj}_{W_1} \mathbf{u}_2 = \mathbf{u}_2 - \frac{\langle \mathbf{u}_2, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1$ where W_1 is the space spanned by \mathbf{v}_1 , and $\text{proj}_{W_1} \mathbf{u}_2$ is the **orthogonal projection** of \mathbf{u}_2 on W_1 .

Step 3 Let $\mathbf{v}_3 = \mathbf{u}_3 - \text{proj}_{W_2} \mathbf{u}_3 = \mathbf{u}_3 - \frac{\langle \mathbf{u}_3, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\langle \mathbf{u}_3, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2$ where W_2 is the space spanned by \mathbf{v}_1 and \mathbf{v}_2 .

Step 4 Let $\mathbf{v}_4 = \mathbf{u}_4 - \text{proj}_{W_3} \mathbf{u}_4 = \mathbf{u}_4 - \frac{\langle \mathbf{u}_4, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\langle \mathbf{u}_4, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 - \frac{\langle \mathbf{u}_4, \mathbf{v}_3 \rangle}{\|\mathbf{v}_3\|^2} \mathbf{v}_3$ where W_3 is the space spanned by $\mathbf{v}_1, \mathbf{v}_2$ and \mathbf{v}_3 .

\vdots

Continue this process up to \mathbf{v}_n . The resulting orthogonal set $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ consists of n linearly independent vectors in V and so forms an orthogonal basis for V .



Notes

- To obtain an **orthonormal** basis for an inner product space V , use the Gram-Schmidt algorithm to construct an orthogonal basis. Then simply normalize each vector in the basis.
- For R^n with the Euclidean inner product (dot product), we of course already know of the orthonormal basis $\{(1, 0, 0, \dots, 0), (0, 1, 0, \dots, 0), \dots, (0, \dots, 0, 1)\}$. For more abstract spaces, however, the existence of an orthonormal basis is not obvious. The Gram-Schmidt algorithm is powerful in that it not only guarantees the existence of an orthonormal basis for any inner product space, but actually gives the construction of such a basis.

Example

Let $V = R^3$ with the Euclidean inner product. We will apply the Gram-Schmidt algorithm to orthogonalize the basis $\{(1, -1, 1), (1, 0, 1), (1, 1, 2)\}$.

Step 1 $\mathbf{v}_1 = (1, -1, 1)$.

$$\begin{aligned} \mathbf{v}_2 &= (1, 0, 1) - \frac{(1, 0, 1) \cdot (1, -1, 1)}{\|(1, -1, 1)\|^2} (1, -1, 1) \\ \text{Step 2} \quad &= (1, 0, 1) - \frac{2}{3} (1, -1, 1) \\ &= \left(\frac{1}{3}, \frac{2}{3}, \frac{1}{3}\right). \end{aligned}$$

$$\begin{aligned} \mathbf{v}_3 &= (1, 1, 2) - \frac{(1, 1, 2) \cdot (1, -1, 1)}{\|(1, -1, 1)\|^2} (1, -1, 1) - \frac{(1, 1, 2) \cdot (\frac{1}{3}, \frac{2}{3}, \frac{1}{3})}{\|(\frac{1}{3}, \frac{2}{3}, \frac{1}{3})\|^2} (\frac{1}{3}, \frac{2}{3}, \frac{1}{3}) \\ \text{Step 3} \quad &= (1, 1, 2) - \frac{2}{3} (1, -1, 1) - \frac{5}{2} (\frac{1}{3}, \frac{2}{3}, \frac{1}{3}) \\ &= \left(\frac{-1}{2}, 0, \frac{1}{2}\right). \end{aligned}$$

You can verify that $\left\{(1, -1, 1), \left(\frac{1}{3}, \frac{2}{3}, \frac{1}{3}\right), \left(\frac{-1}{2}, 0, \frac{1}{2}\right)\right\}$ forms an orthogonal basis for R^3 . Normalizing the vectors in the orthogonal basis, we obtain the orthonormal basis

$$\left\{\left(\frac{\sqrt{3}}{3}, \frac{-\sqrt{3}}{3}, \frac{\sqrt{3}}{3}\right), \left(\frac{\sqrt{6}}{6}, \frac{\sqrt{6}}{3}, \frac{\sqrt{6}}{6}\right), \left(\frac{-\sqrt{2}}{2}, 0, \frac{\sqrt{2}}{2}\right)\right\}.$$

0.57735 0.8164 0.4082 0.707

Key Concepts

Given an arbitrary basis $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ for an n -dimensional inner product space V , the **Gram-Schmidt algorithm** constructs an orthogonal basis $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ for V :

Step 1 Let $\mathbf{v}_1 = \mathbf{u}_1$.

Step 2 Let $\mathbf{v}_2 = \mathbf{u}_2 - \frac{\langle \mathbf{u}_2, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1$.

Step 3 Let $\mathbf{v}_3 = \mathbf{u}_3 - \frac{\langle \mathbf{u}_3, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\langle \mathbf{u}_3, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2$.

\vdots

[I'm ready to take the quiz.] [I need to review more.]
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