

The Gram–Schmidt Orthogonalization Process

Linear Algebra

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This work was initially funded by an Institutional OER Grant from the Colorado Department of Higher Education (CDHE). For similar OER materials in other courses funded by this project in the Department of Mathematical and Statistical Sciences at the University of Colorado Denver, visit <https://github.com/CU-Denver-MathStats-OER>

Introduction

- ▶ We have seen that an orthogonal basis for a subspace W of \mathbb{R}^n is particularly nice.
 - ▶ For \mathbf{y} in \mathbb{R}^n , we compute the **orthogonal projection** of \mathbf{y} onto W by

$$\hat{\mathbf{y}} = \text{proj}_W \mathbf{y} = c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2 + \dots + c_p \mathbf{u}_p \quad \text{with weights } c_i = \frac{\mathbf{y} \cdot \mathbf{u}_i}{\mathbf{u}_i \cdot \mathbf{u}_i}.$$

- ▶ When we choose an **orthonormal basis**, the calculations are even simpler.

$$\hat{\mathbf{y}} = \text{proj}_W \mathbf{y} = (UU^T)\mathbf{y}.$$

- ▶ Recall $\hat{\mathbf{y}}$ is a useful vector to find as it is the **best approximation** to \mathbf{y} in the subspace W .

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Example

Let $W = \text{Span}\{\mathbf{x}_1, \mathbf{x}_2\}$ denote the subspace of \mathbb{R}^3 where $\mathbf{x}_1 = \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}$ and $\mathbf{x}_2 = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$.

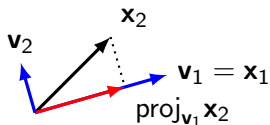
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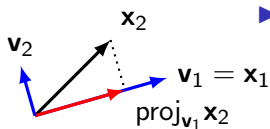


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- ▶ Let $\mathbf{v}_1 = \mathbf{x}_1$.
- ▶ Let \mathbf{v}_2 be the orth proj of \mathbf{x}_2 onto the orth compl of $Y = \text{Span}\{\mathbf{v}_1\}$.

By the **Orth Decomp Thm**, $\mathbf{x}_2 = \text{proj}_Y \mathbf{x}_2 + \mathbf{z}$, where \mathbf{z} is in Y^\perp .



$$\text{proj}_Y \mathbf{x}_2 = \text{proj}_{\mathbf{v}_1} \mathbf{x}_2 = \left(\frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 = \frac{15}{45} \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}.$$

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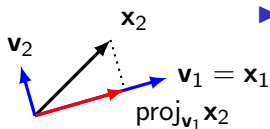
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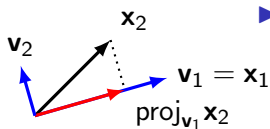


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- ▶ Thus $\{\mathbf{v}_1, \mathbf{v}_2\} = \left\{ \begin{bmatrix} 3 \\ 6 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} \right\}$ is an **orthogonal basis** for W .

The Gram–Schmidt Orthogonalization Process

Given a **basis** $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_p\}$ for a nonzero subspace W of \mathbb{R}^n , define

$$\mathbf{v}_1 = \mathbf{x}_1$$

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$$\begin{aligned}\mathbf{v}_1 &= \mathbf{x}_1 & Y_1 &= \text{Span}\{\mathbf{v}_1\} \\ \mathbf{v}_2 &= \text{proj}_{Y_1^\perp} \mathbf{x}_2 = \mathbf{x}_2 - \text{proj}_{Y_1} \mathbf{x}_2 = \mathbf{x}_2 - \left(\frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 & Y_2 &= \text{Span}\{\mathbf{v}_1, \mathbf{v}_2\} \\ \mathbf{v}_3 &= \text{proj}_{Y_2^\perp} \mathbf{x}_3 = \mathbf{x}_3 - \text{proj}_{Y_2} \mathbf{x}_3 = \mathbf{x}_3 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_2}{\mathbf{v}_2 \cdot \mathbf{v}_2} \right) \mathbf{v}_2 & Y_3 &= \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}\end{aligned}$$

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$$\vdots$$

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Theorem

Then $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$ is an **orthogonal basis** for W .

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Then $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$ is an **orthogonal basis** for W . In addition,

$$\text{Span}\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\} = Y_k = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\} \quad \text{for } 1 \leq k \leq p.$$

Example

Find an **orthogonal basis** for the subspace $W = \text{Span} \left\{ \begin{bmatrix} 1 \\ -4 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 7 \\ -7 \\ -4 \\ 1 \end{bmatrix} \right\}$ in \mathbb{R}^4 .

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Then we find $\mathbf{v}_2 = \mathbf{x}_2 - \left(\frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 = \begin{bmatrix} 7 \\ -7 \\ -4 \\ 1 \end{bmatrix} - \left(\frac{36}{18} \right) \begin{bmatrix} 1 \\ -4 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 1 \\ -4 \\ -1 \end{bmatrix}$.

Thus an **orthogonal basis** for W is $\left\{ \begin{bmatrix} 1 \\ -4 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 5 \\ 1 \\ -4 \\ -1 \end{bmatrix} \right\}$.

Example

Consider the matrix $A = \begin{bmatrix} 3 & -5 & 1 \\ 1 & 1 & 1 \\ -1 & 5 & -2 \\ 3 & -7 & 8 \end{bmatrix}$. Find an **orthogonal basis** for the column space of A .

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First we find a basis for the column space by checking the reduced row echelon form of A :

$$\begin{bmatrix} 3 & -5 & 1 \\ 1 & 1 & 1 \\ -1 & 5 & -2 \\ 3 & -7 & 8 \end{bmatrix} \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}, \text{ so a basis for Col } A \text{ is } \left\{ \begin{bmatrix} 3 \\ 1 \\ -1 \\ 3 \end{bmatrix}, \begin{bmatrix} -5 \\ 1 \\ 5 \\ -7 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -2 \\ 8 \end{bmatrix} \right\}.$$

Next we use the Gram–Schmidt Process to convert the basis into an **orthogonal** basis.

Example, continued

$$\text{Col } A = \text{Span} \left\{ \begin{bmatrix} 3 \\ 1 \\ -1 \\ 3 \end{bmatrix}, \begin{bmatrix} -5 \\ 1 \\ 5 \\ -7 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -2 \\ 8 \end{bmatrix} \right\}. \text{ We first define } \mathbf{v}_1 = \begin{bmatrix} 3 \\ 1 \\ -1 \\ 3 \end{bmatrix}.$$

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Then we continue with Gram–Schmidt process:

$$\mathbf{v}_2 = \mathbf{x}_2 - \left(\frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 = \begin{bmatrix} -5 \\ 1 \\ 5 \\ -7 \end{bmatrix} - \left(\frac{-40}{20} \right) \begin{bmatrix} 3 \\ 1 \\ -1 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ 3 \\ -1 \end{bmatrix}$$

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Normalizing the Basis

For the 4×3 matrix $A = \begin{bmatrix} 3 & -5 & 1 \\ 1 & 1 & 1 \\ -1 & 5 & -2 \\ 3 & -7 & 8 \end{bmatrix}$, we therefore have

► An orthogonal basis for $\text{Col } A$ is $\left\{ \begin{bmatrix} 3 \\ 1 \\ -1 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 3 \\ 3 \\ -1 \end{bmatrix}, \begin{bmatrix} -3 \\ 1 \\ 1 \\ 3 \end{bmatrix} \right\}$.

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► An **orthonormal basis** for $\text{Col } A$ is $\left\{ \begin{bmatrix} 3/\sqrt{20} \\ 1/\sqrt{20} \\ -1/\sqrt{20} \\ 3/\sqrt{20} \end{bmatrix}, \begin{bmatrix} 1/\sqrt{20} \\ 3/\sqrt{20} \\ 3/\sqrt{20} \\ -1/\sqrt{20} \end{bmatrix}, \begin{bmatrix} -3/\sqrt{20} \\ 1/\sqrt{20} \\ 1/\sqrt{20} \\ 3/\sqrt{20} \end{bmatrix} \right\}$.