

The Gram–Schmidt Orthogonalization Process

Linear Algebra

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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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- ▶ 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 .

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

- ▶ Hence $\mathbf{v}_2 = \mathbf{z} = \mathbf{x}_2 - \text{proj}_{\mathbf{v}_1} \mathbf{x}_2 = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$.

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- ▶ Since $\mathbf{v}_2 = \mathbf{x}_2 - \text{proj}_{\mathbf{v}_1} \mathbf{x}_2 = \mathbf{x}_2 + c\mathbf{x}_1$, \mathbf{v}_2 is in the subspace W .

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How can we find an **orthogonal basis** for W ?

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- ▶ Since $\mathbf{v}_2 = \mathbf{x}_2 - \text{proj}_{\mathbf{v}_1} \mathbf{x}_2 = \mathbf{x}_2 + c\mathbf{x}_1$, \mathbf{v}_2 is in the subspace W .
- ▶ 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

$$\mathbf{v}_p = \text{proj}_{Y_{p-1}^\perp} \mathbf{x}_p = \mathbf{x}_p - \text{proj}_{Y_{p-1}} \mathbf{x}_p$$

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$$Y_p = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \dots, \mathbf{v}_p\}$$

Theorem

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

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Theorem

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:

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

Example, continued

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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 $\left\{ \begin{bmatrix} 3/\sqrt{20} \\ 1/\sqrt{20} \\ -1/\sqrt{20} \\ 3/\sqrt{20} \end{bmatrix}, \begin{bmatrix} 1/\sqrt{20} \\ 3/\sqrt{20} \\ 3/\sqrt{20} \\ -1/\sqrt{20} \end{bmatrix}, \begin{bmatrix} -3/\sqrt{20} \\ 1/\sqrt{20} \\ 1/\sqrt{20} \\ 3/\sqrt{20} \end{bmatrix} \right\}$.