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§5.2 Gram-Schmidt Process and QR Factorization

Example 1. Find an orthogonal basis for the subspace

$$W = \operatorname{Span}\left\{ \vec{w}_1 = \begin{bmatrix} 1\\2\\3 \end{bmatrix}, \vec{w}_2 = \begin{bmatrix} 4\\5\\6 \end{bmatrix} \right\}$$



Example 2. Find an orthogonal basis for the subspace

$$V = \operatorname{Span} \left\{ \vec{v}_1 = \begin{bmatrix} 1\\1\\2\\0 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 1\\-1\\0\\0 \end{bmatrix}, \vec{v}_3 = \begin{bmatrix} 1\\2\\6\\0 \end{bmatrix} \right\}$$



The **Gram-Schmidt process** is an algorithm that produces an orthogonal (or orthonormal) basis for any subspace W of \mathbb{R}^n by starting with any basis for W.

Theorem. [Gram-Schmidt (Orthogonalize)]

Let W be a subspace of \mathbb{R}^n and let $\vec{b}_1, \dots, \vec{b}_p$ be a basis for W. Define vectors $\vec{v}_1, \dots, \vec{v}_p$ as

$$\vec{v}_{1} = \vec{b}_{1}$$

$$\vec{v}_{2} = \vec{b}_{2} - \frac{\vec{b}_{2} \cdot \vec{v}_{1}}{\vec{v}_{1} \cdot \vec{v}_{1}} \vec{v}_{1}$$

$$\vec{v}_{3} = \vec{b}_{3} - \frac{\vec{b}_{3} \cdot \vec{v}_{1}}{\vec{v}_{1} \cdot \vec{v}_{1}} \vec{v}_{1} - \frac{\vec{b}_{3} \cdot \vec{v}_{2}}{\vec{v}_{2} \cdot \vec{v}_{2}} \vec{v}_{2}$$

$$\vdots$$

$$\vec{v}_{p} = \vec{b}_{p} - \frac{\vec{b}_{p} \cdot \vec{v}_{1}}{\vec{v}_{1} \cdot \vec{v}_{1}} \vec{v}_{1} - \frac{\vec{b}_{p} \cdot \vec{v}_{2}}{\vec{v}_{2} \cdot \vec{v}_{2}} \vec{v}_{2} - \dots - \frac{\vec{b}_{p} \cdot \vec{v}_{p-1}}{\vec{v}_{p-1} \cdot \vec{v}_{p-1}} \vec{v}_{p-1}$$

Then $\{\vec{v}_1, \ldots, \vec{v}_p\}$ is an orthogonal basis for W and

$$\operatorname{Span}\{\dot{b_1},\cdots,\dot{b_k}\}=\operatorname{Span}\{\vec{v_1},\cdots,\vec{v_k}\}$$

for and $k = 1, \ldots, p$.

From the formulas we can see that $\vec{v}_2 = \vec{b}_2^{\perp}$ relative to \vec{b}_1 .

 $\vec{v}_3 = \vec{b}_3^{\perp}$ relative to $\text{Span}(\vec{b}_1, \vec{b}_2)$. $\vec{v}_i = \vec{b}_i^{\perp}$ relative to $\text{Span}(\vec{b}_1, \vec{b}_2, ..., \vec{b}_{i-1})$.

Theorem. [Gram-Schmidt (Normalize)]

If $\{\vec{v}_1, \ldots, \vec{v}_p\}$ is an orthogonal basis for W, then $\{\vec{u}_1, \ldots, \vec{u}_p\}$ is an orthonormal basis for W, where, $\vec{u}_i = \frac{\vec{v}_i}{||\vec{v}_i||}$ for $i = 1, \ldots, p$.

Basis $\xrightarrow{\text{orthogonalize}}$ Orthogonal basis $\xrightarrow{\text{normalize}}$ Orthonormal basis.

Example 3. Using Gram-Schmidt Process to find an orthonormal basis for

$$V = \operatorname{Span}\left\{ \vec{v}_1 = \begin{bmatrix} 1\\1\\2 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 2\\4\\6 \end{bmatrix}, \vec{v}_3 = \begin{bmatrix} 1\\2\\6 \end{bmatrix} \right\}.$$

$$\{\vec{v}_{1}, \vec{v}_{2}, \vec{v}_{3}\} \text{ is already a basis for } V \text{ since it is independent.}$$

$$\vec{b}_{1}^{\prime} = \vec{V}_{1} = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}$$

$$\vec{b}_{2}^{\prime} = \vec{V}_{2}^{\prime} - \left(\frac{\vec{V}_{2}^{\prime} \cdot \vec{b}_{1}}{\vec{b}_{1} \cdot \vec{b}_{1}}\right) \vec{b}_{1}^{\prime} = \begin{bmatrix} \frac{2}{4} \\ \frac{2}{6} \end{bmatrix} - \frac{18}{6} \begin{bmatrix} 1}{2} \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

$$\vec{b}_{3}^{\prime} = \vec{V}_{3}^{\prime} - \left(\frac{\vec{V}_{3}^{\prime} \cdot \vec{b}_{1}}{\vec{b}_{1} \cdot \vec{b}_{1}}\right) \vec{b}_{1} - \frac{\vec{V}_{3}^{\prime} \cdot \vec{b}_{2}}{\vec{b}_{2} \cdot \vec{b}_{2}} \end{bmatrix} \vec{b}_{2}$$

$$= \begin{bmatrix} \frac{1}{2} \\ \frac{1}{6} \end{bmatrix} - \frac{15}{6} \begin{bmatrix} 1 \\ \frac{2}{2} \end{bmatrix} - \frac{1}{2} \begin{bmatrix} -1 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$\{\vec{b}_{1}, \vec{b}_{2}, \vec{b}_{3}\} \text{ is an orthogonal basis for } V.$$

$$\vec{u}_{1}^{\prime} = \begin{bmatrix} \vec{b}_{1} \\ \vec{b}_{1} \end{bmatrix} = \begin{bmatrix} \sqrt{t}_{1} \\ \sqrt{t}_{1} \\ \sqrt{t}_{2} \end{bmatrix} \qquad \vec{u}_{2}^{\prime} = \frac{\vec{b}_{2}}{\vec{b}_{2} \cdot \vec{b}_{1}} = \begin{bmatrix} \sqrt{t}_{2} \\ \sqrt{t}_{1} \\ \sqrt{t}_{2} \end{bmatrix}$$

$$\{\vec{n}_{1}, \vec{u}_{2}, \vec{u}_{3}\} \text{ is an orthogonal basis for } V.$$

Since \vec{v}_i is already used in the question, in the solution, we changed the formulas. Another (better) way is rewrite the letters in the question as \vec{b}_i .

Let us rewrite the solution of Example 3 here.

Rewrite Example 3 Using Gram-Schmidt Process to find an orthonormal basis for

$$V = \operatorname{Span}\left\{\vec{b}_1 = \begin{bmatrix}1\\1\\2\end{bmatrix}, \vec{b}_2 = \begin{bmatrix}2\\4\\6\end{bmatrix}, \vec{b}_3 = \begin{bmatrix}1\\2\\6\end{bmatrix}\right\}.$$

$$\begin{split} \vec{v}_1 &= \vec{b}_1 = \begin{bmatrix} 1\\1\\2 \end{bmatrix} \\ \vec{v}_2 &= \vec{v}_2 - \left(\frac{\vec{b}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1}\right) \vec{v}_1 = \begin{bmatrix} 2\\4\\6 \end{bmatrix} - \frac{18}{6} \begin{bmatrix} 1\\1\\2 \end{bmatrix} = \begin{bmatrix} -1\\1\\0 \end{bmatrix} \\ \vec{v}_3 &= \vec{b}_3 - \left(\frac{\vec{b}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1}\right) \vec{v}_1 - \left(\frac{\vec{b}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2}\right) \vec{v}_2 = \begin{bmatrix} 1\\2\\6 \end{bmatrix} - \frac{15}{6} \begin{bmatrix} 1\\1\\2 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} -1\\1\\0 \end{bmatrix} = \begin{bmatrix} -1\\-1\\1 \end{bmatrix} \\ \{\vec{v}_1, \vec{v}_2, \vec{v}_3\} \text{ is an orthogonal basis for } V. \\ \vec{u}_1 &= \frac{\vec{v}_1}{||\vec{v}_1||} = \begin{bmatrix} 1/\sqrt{6}\\1/\sqrt{6}\\2/\sqrt{6} \end{bmatrix}; \ \vec{u}_2 &= \frac{\vec{v}_2}{||\vec{v}_2||} = \begin{bmatrix} -1/\sqrt{2}\\1/\sqrt{2}\\0 \end{bmatrix}; \ \vec{u}_3 &= \frac{\vec{v}_3}{||\vec{v}_3||} = \begin{bmatrix} -1/\sqrt{3}\\-1/\sqrt{3}\\1/\sqrt{3} \end{bmatrix} \\ \{\vec{u}_1, \vec{u}_2, \vec{u}_3\} \text{ is an orthonormal basis for } V. \end{split}$$

Example 4. Using Gram-Schmidt Process to find an orthonormal basis for

$$V = \operatorname{Span} \left\{ \vec{b}_1 = \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix}, \vec{b}_2 = \begin{bmatrix} 0\\1\\1\\1 \end{bmatrix} \right\}.$$

$$\vec{v}_1 = \vec{b}_1 = \begin{bmatrix} 1\\1\\1\\1\\1 \end{bmatrix}$$
$$\vec{v}_2 = \vec{b}_2 - \left(\frac{\vec{b}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1}\right) \vec{v}_1 = \begin{bmatrix} 0\\1\\1\\1\\1 \end{bmatrix} - \frac{3}{4} \begin{bmatrix} 1\\1\\1\\1\\1 \end{bmatrix} = \begin{bmatrix} -3/4\\1/4\\1/4\\1/4 \end{bmatrix}$$

 $\{\vec{v}_1, \vec{v}_2\}$ is an orthogonal basis for V.

$$\vec{u}_1 = \frac{\vec{v}_1}{||\vec{v}_1||} = \begin{bmatrix} 1/2\\1/2\\1/2\\1/2 \end{bmatrix}; \ \vec{u}_2 = \frac{\vec{v}_2}{||\vec{v}_2||} = \begin{bmatrix} -3/\sqrt{12}\\1/\sqrt{12}\\1/\sqrt{12}\\1/\sqrt{12}\\1/\sqrt{12} \end{bmatrix};$$

 $\{\vec{u}_1,\vec{u}_2\}$ is an orthonormal basis for V.

• Note that the formula for computing \vec{v}_i for any $i = 2, 3, \ldots, p$ can be written as

$$\vec{v}_{i} = \vec{b}_{i} - \text{proj}_{\vec{v}_{1}}(\vec{b}_{i}) - \text{proj}_{\vec{v}_{2}}(\vec{b}_{i}) - \dots - \text{proj}_{\vec{v}_{i-1}}(\vec{b}_{i})$$
$$= \vec{b}_{i} - \text{proj}_{\text{Span}\{\vec{v}_{1},\dots,\vec{v}_{i-1}\}}(\vec{b}_{i}).$$

So, $\vec{v}_i = \vec{b}_i^{\perp}$ respect to $\operatorname{Span}\{\vec{v}_1, \ldots, \vec{v}_{i-1}\}$.

• This formula is *inductive* in that the computation of \vec{v}_i relies on the vectors $\vec{v}_1, \ldots, \vec{v}_{i-1}$.

QR-Factorization.

QR-Factorization is the matrix version of the Gram-Schmidt process.

Recall the Gram-Schmidt process:

Basis $\mathscr{B} = \{\vec{b}_1, \dots, \vec{b}_p\} \xrightarrow{\text{orthogonalize}} \text{Orthogonal basis } \mathscr{V} = \{\vec{v}_1, \dots, \vec{v}_p\} \xrightarrow{\text{normalize}} \text{Orthonormal basis } \mathscr{U} = \{\vec{u}_1, \dots, \vec{u}_p\}.$

Given a vector in W, let's compare their coordinates:



Each matrix defines an isomorphism. So, M = QR.

Here $M = [\vec{b}_1 \ \dots \ \vec{b}_p]$ and $Q = [\vec{u}_1, \dots, \vec{u}_p]$.

Theorem.

Given a $n \times p$ matrix $M = [\vec{b}_1 \dots \vec{b}_p]$ with independent columns. There is a unique decomposition

M = QR

where, $Q = [\vec{u}_1, \dots, \vec{u}_p]$ has orthonormal columns and R is an $p \times p$ upper triangular matrix with

 $r_{ii} = ||\vec{v}_i||$ for $i = 1, \dots, p$ and $r_{ij} = \vec{u}_i \cdot \vec{b}_j$ for i < j.

In particular when p = 3, the upper triangular matrix R is

$$R = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ 0 & r_{22} & r_{23} \\ 0 & 0 & r_{33} \end{bmatrix} = \begin{bmatrix} ||\vec{v}_1|| & \vec{u}_1 \cdot \vec{b}_2 & \vec{u}_1 \cdot \vec{b}_3 \\ 0 & ||\vec{v}_2|| & \vec{u}_2 \cdot \vec{b}_3 \\ 0 & 0 & ||\vec{v}_3|| \end{bmatrix}$$

Proof(for p = 3): From Gram-Schmidt process, write \vec{b}_i as linear combinations of \vec{u}_i .

$$\begin{split} \vec{b}_1 &= \vec{v}_1 = ||\vec{v}_1||\vec{u}_1 \\ \vec{b}_2 &= \vec{v}_2 + \frac{\vec{b}_2 \cdot \vec{v}_1}{||\vec{v}_1||^2} \vec{v}_1 = ||\vec{v}_2||\vec{u}_2 + (\vec{b}_2 \cdot \vec{u}_1)\vec{u}_1 \\ \vec{b}_3 &= \vec{b}_3 - \frac{\vec{b}_3 \cdot \vec{v}_1}{||\vec{v}_1||^2} \vec{v}_1 - \frac{\vec{b}_3 \cdot \vec{v}_2}{||\vec{v}_2||^2} \vec{v}_2 = ||\vec{v}_3||\vec{u}_3 + (\vec{b}_3 \cdot \vec{u}_1)\vec{u}_1 + (\vec{b}_3 \cdot \vec{u}_2)\vec{u}_2 \end{split}$$

So,

$$\begin{bmatrix} \vec{b}_1 & \vec{b}_2 & \vec{b}_3 \end{bmatrix} = \begin{bmatrix} \vec{u}_1 & \vec{u}_2 & \vec{u}_3 \end{bmatrix} \begin{bmatrix} ||\vec{v}_1|| & \vec{u}_1 \cdot \vec{b}_2 & \vec{u}_1 \cdot \vec{b}_3 \end{bmatrix} \begin{bmatrix} ||\vec{v}_1|| & \vec{u}_2 \cdot \vec{b}_3 \\ 0 & ||\vec{v}_2|| & \vec{u}_2 \cdot \vec{b}_3 \\ 0 & 0 & ||\vec{v}_3|| \end{bmatrix}$$

Example 5. Find the *QR*-factorization of the matrix $M = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 4 & 2 \\ 2 & 6 & 6 \end{bmatrix}$. (Use modified Example 3,)

$$\vec{b}_{1} = \begin{bmatrix} 1\\ 1\\ 2 \end{bmatrix}, \vec{b}_{2} = \begin{bmatrix} 2\\ 4\\ 6 \end{bmatrix}, \vec{b}_{3} = \begin{bmatrix} 1\\ 2\\ 6 \end{bmatrix}, \text{ and } \vec{v}_{1} = \begin{bmatrix} 1\\ 1\\ 2 \end{bmatrix}, \vec{v}_{2} = \begin{bmatrix} -1\\ 1\\ 0 \end{bmatrix}, \vec{v}_{3} = \begin{bmatrix} -1\\ -1\\ 1 \end{bmatrix}.$$
From Example 3, we have $Q = \begin{bmatrix} \frac{1}{12} & \frac{1}{12} & \frac{1}{13} \\ \frac{1}{12} & \frac{1}{12} & \frac{1}{13} \\ \frac{1}{12} & \frac{1}{12} & \frac{1}{13} \end{bmatrix}$

So $Y_{1} = \|\vec{v}_{1}^{T}\| = \int \vec{b} \quad Y_{22} = \|\vec{v}_{2}^{T}\| = \int \vec{c} \quad Y_{02} = \|\vec{v}_{1}^{T}\| = J \vec{c}$

$$Y_{22} = \|\vec{v}_{2}^{T}\| = \int \vec{c} \quad Y_{22} = \|\vec{v}_{2}^{T}\| = \int \vec{c} \quad Y_{23} = \vec{c}_{12} \cdot \vec{b}_{1} = J \vec{c}$$

$$Y_{23} = \vec{c}_{12} \cdot \vec{b}_{2} = J \vec{b}_{12} \quad Y_{23} = \vec{c}_{12} \cdot \vec{b}_{1} = J \vec{c}$$

$$\vec{c} \quad R = \begin{bmatrix} \int \vec{v}_{1} & |\vec{v}_{1}| & |\vec{v}_{1}| \\ 0 & J \equiv J \vec{c}_{1} \\ 0 & J \equiv J \vec{c}_{1} \\ 0 & J \equiv J \vec{c}_{1} \end{bmatrix}$$

Example 6. Using Gram-Schmidt Process to find the QR-factorization of the matrix M = $\begin{bmatrix} 1 & 0 \end{bmatrix}$ $\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{bmatrix}$ •

From Example 4,
$$Q = \begin{bmatrix} \frac{1}{2} & \frac{-3}{245} \\ \frac{1}{2} & \frac{1}{245} \\ \frac{1}{2} & \frac{1}{245} \end{bmatrix}$$

 $r_{11} = ||\vec{v}_1|| = |\vec{v}_2|$ $r_{22} = ||\vec{v}_2|| = |\vec{b}_{/2}|$
 $r_{12} = \vec{u}_1 \cdot \vec{b}_2 = \frac{3}{2}$
 $r_{12} = \vec{u}_1 \cdot \vec{b}_2 = \frac{3}{2}$
 $S_0 \quad R = \begin{bmatrix} 2 & \frac{3}{2} \\ 0 & \frac{1}{5} \\ 0 & \frac{1}{5} \\ \end{bmatrix}$