

Thm An $m \times n$ matrix U has o/n columns iff $U^T U = I$

Thm Let U be an $m \times n$ matrix with o/n columns; let $\vec{x}, \vec{y} \in \mathbb{R}^n$. Then:

$$a) \|U\vec{x}\| = \|\vec{x}\| \quad b) (U\vec{x}) \cdot (U\vec{y}) = \vec{x} \cdot \vec{y} \quad c) U\vec{x} \perp U\vec{y} \text{ iff } \vec{x} \perp \vec{y}$$

i.e. mapping $\mathbb{R}^n \rightarrow \mathbb{R}^m$ preserves lengths and orthogonality
 $\vec{x} \mapsto U\vec{x}$

If $m=n$, square mat. U with o/n columns is called an orthogonal matrix.
 U is orthogonal iff $U^{-1} = U^T$.

Ex: matrix of rotation by angle φ : $U = \begin{bmatrix} \cos \varphi & -\sin \varphi \\ \sin \varphi & \cos \varphi \end{bmatrix}$

$$U^{-1} = \frac{1}{\cancel{\cos^2 \varphi + \sin^2 \varphi}} \begin{bmatrix} \cos \varphi & \sin \varphi \\ -\sin \varphi & \cos \varphi \end{bmatrix} = U^T \Rightarrow U \text{ is an orthogonal matrix.}$$

Thm If $\{\vec{u}_1, \dots, \vec{u}_p\}$ is an o/n basis for $W \subset \mathbb{R}^n$, then

$$(i) \text{proj}_W \vec{v} = (\vec{v} \cdot \vec{u}_1) \vec{u}_1 + \dots + (\vec{v} \cdot \vec{u}_p) \vec{u}_p.$$

(ii) If $U = [\vec{u}_1, \dots, \vec{u}_p]$ then $\boxed{\text{proj}_W \vec{v} = U U^T \vec{v}}$ for all $\vec{v} \in \mathbb{R}^n$
 matrix of projection onto W .

Gram-Schmidt process

(2)

Problem: $W \subset \mathbb{R}^n$ find an orthogonal basis for W .
 $\text{Span}(\vec{x}_1, \dots, \vec{x}_p)$

basis but not orthogonal

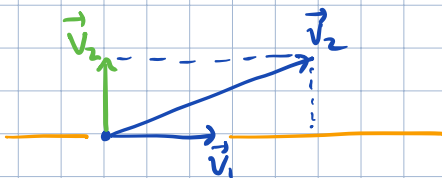
Ex: $\vec{x}_1 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$, $\vec{x}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ $W = \text{span}(\vec{x}_1, \vec{x}_2) \subset \mathbb{R}^3$

Q: find an orthogonal basis

Sol: set $\vec{v}_1 = \vec{x}_1$; \vec{v}_2 in W , lin. indep. from \vec{v}_1 and \perp to \vec{v}_1

$$\vec{x}_2 = \text{proj}_{\vec{v}_1} \vec{x}_2 + \text{perp}_{\vec{v}_1} \vec{x}_2$$

$\perp \vec{v}_1$ \vec{v}_2



Explicitly: $\text{proj}_{\vec{v}_1} \vec{x}_2 = \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 = \frac{-1}{5} \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} -1/5 \\ -2/5 \\ 0 \end{bmatrix}$

$$\text{perp}_{\vec{v}_1} \vec{x}_2 = \vec{x}_2 - \text{proj}_{\vec{v}_1} \vec{x}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} - \begin{bmatrix} -1/5 \\ -2/5 \\ 0 \end{bmatrix} = \begin{bmatrix} -4/5 \\ 2/5 \\ 1 \end{bmatrix}$$

So: $\left\{ \vec{v}_1 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} -4/5 \\ 2/5 \\ 1 \end{bmatrix} \right\}$ - orthog. basis for W .

} rescale

$$\left\{ \vec{v}_1 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \vec{v}_2' = \begin{bmatrix} -4 \\ 2 \\ 5 \end{bmatrix} \right\} \text{ - more convenient, also orthogonal}$$

Generally Let $W = \text{span}(\underbrace{\vec{x}_1, \dots, \vec{x}_p}_{\text{basis}})$

went to construct an orthog. basis for W , $\{\vec{v}_1, \dots, \vec{v}_p\}$

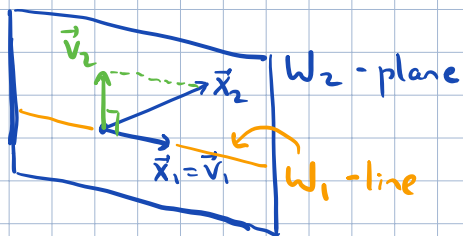
Step 1: set $\vec{v}_1 = \vec{x}_1$, $W_1 = \text{span}(\vec{x}_1) = \text{span}(\vec{v}_1)$

Step 2: $W_2 = \text{span}(\vec{x}_1, \vec{x}_2)$

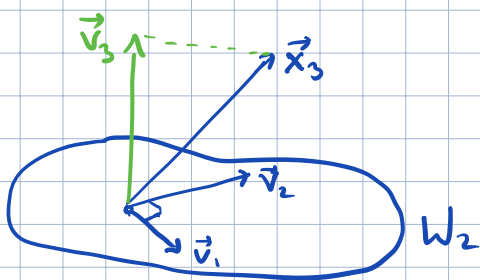
orthog. basis:

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

$$\vec{v}_2 = \text{perp}_{W_1} \vec{x}_2 = \vec{x}_2 - \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1$$



Step 3: $W_3 = \text{span}(\vec{x}_1, \vec{x}_2, \vec{x}_3)$ orthog. basis: \vec{v}_1, \vec{v}_2 - already constructed



$\vec{v}_3 = \text{perp}_{W_2} \vec{x}_3$ $\{\vec{v}_1, \vec{v}_2\}$ - orthog. basis

$$= \vec{x}_3 - \frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2$$

...

Step p: $W = W_p = \text{span}(\vec{x}_1, \dots, \vec{x}_p)$

orthog. basis: $\vec{v}_1, \dots, \vec{v}_{p-1}, \vec{v}_p = \text{perp}_{W_{p-1}} \vec{x}_p = \vec{x}_p - \frac{\vec{x}_p \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \dots - \frac{\vec{x}_p \cdot \vec{v}_{p-1}}{\vec{v}_{p-1} \cdot \vec{v}_{p-1}} \vec{v}_{p-1}$

$\{\vec{v}_1, \dots, \vec{v}_{p-1}\}$ - orthog. basis

Ex: $\vec{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$, $\vec{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}$, $\vec{x}_3 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$ - basis for $W = \text{span}(\vec{x}_1, \vec{x}_2, \vec{x}_3) \subset \mathbb{R}^4$
Q: find an orthogonal basis for W .

Sol: $\vec{v}_1 = \vec{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$

$$\vec{v}_2 = \text{perp}_{\vec{v}_1} \vec{x}_2 = \vec{x}_2 - \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} -1/2 \\ 1/2 \\ 1 \\ 0 \end{bmatrix} \rightsquigarrow \text{rescale by 2} \vec{v}_2' = \begin{bmatrix} -1 \\ 1 \\ 2 \\ 0 \end{bmatrix}$$

④

$$\vec{v}_3 = \text{perp}_{\text{Span}(\vec{v}_1, \vec{v}_2)} \vec{x}_3 = \vec{x}_3 - \frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_3 \cdot \vec{v}_2'}{\vec{v}_2' \cdot \vec{v}_2'} \vec{v}_2'$$

$$= \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \frac{2}{2} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - \frac{2}{6} \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1/3 \\ -1/3 \\ 1/3 \end{bmatrix} \xrightarrow{\text{rescale}} \vec{v}_3' = \begin{bmatrix} 1 \\ -1 \\ 3 \end{bmatrix}$$

Thus, $\left\{ \vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \vec{v}_2' = \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix}, \vec{v}_3' = \begin{bmatrix} 1 \\ -1 \\ 3 \end{bmatrix} \right\}$ - orthogonal basis for W

Q: find an orthonormal basis for W

Sol: normalize $\vec{v}_1, \vec{v}_2', \vec{v}_3'$ to unit length:

$$\left\{ \vec{u}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \vec{u}_2 = \frac{1}{\sqrt{6}} \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix}, \vec{u}_3 = \frac{1}{\sqrt{12}} \begin{bmatrix} 1 \\ -1 \\ 3 \end{bmatrix} \right\} \text{ - orthonormal basis for } W.$$

QR factorization

Thm (QR factorization)

If A is an $m \times n$ matrix with lin. indep. columns, then A can be factored as $A = QR$ where Q is an $m \times n$ matrix whose columns form an orthonormal basis for $\text{col}(A)$ and R is an $n \times n$ upper-triangular invertible matrix with positive diagonal entries.

Idea: $A = [\vec{x}_1 \dots \vec{x}_n]$ $W = \text{col}(A) = \text{Span}(\vec{x}_1, \dots, \vec{x}_n) \subset \mathbb{R}^m$
 $\left\{ \begin{array}{l} \text{Gram-Schmidt + normalization} \\ \vec{u}_1, \dots, \vec{u}_n \text{ - orthonormal basis for } W \end{array} \right.$

$$\vec{x}_k = \underbrace{\text{proj}_{W_{k-1}} \vec{x}_k}_{r_{1k} \vec{u}_1 + \dots + r_{k-1,k} \vec{u}_{k-1}} + \underbrace{\vec{v}_k}_{r_{kk} \vec{u}_k} \leftarrow \text{from Gram-Schmidt}$$

$$= r_{1k} \vec{u}_1 + \dots + r_{k-1,k} \vec{u}_{k-1} + r_{kk} \vec{u}_k + 0 \cdot \vec{u}_{k+1} + \dots + 0 \cdot \vec{u}_n$$

$$\Rightarrow A = \underbrace{[\vec{u}_1 \dots \vec{u}_n]}_Q \underbrace{\begin{bmatrix} r_{11} & r_{12} & \dots & r_{1n} \\ 0 & r_{22} & & r_{2n} \\ \vdots & 0 & \ddots & \vdots \\ 0 & \vdots & \dots & 0 & r_{nn} \end{bmatrix}}_R$$

Ex: $A = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$ find QR factorization

\uparrow \uparrow \uparrow
 \vec{x}_1 \vec{x}_2 \vec{x}_3

- vectors from \mathcal{E}_x^*

Sol: $Q = [\vec{u}_1 \ \vec{u}_2 \ \vec{u}_3] = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{6} & -1/\sqrt{2} \\ 0 & 2/\sqrt{6} & 1/\sqrt{2} \\ 0 & 0 & 3/\sqrt{12} \end{bmatrix}$

normalized
G-S basis

a shortcut to get R: $A = QR \Rightarrow Q^T A = \underbrace{Q^T Q}_I R = R$

So, $R = Q^T A = \dots = \begin{bmatrix} 2 & 1/\sqrt{2} & \sqrt{2} \\ 0 & 3/\sqrt{6} & 2/\sqrt{6} \\ 0 & 0 & 4/\sqrt{12} \end{bmatrix}$