

Week 14 Tuesday.

Recall. a basis S for V is

1) linearly independent

2) $\text{span}(S) = V$

Q: Given V , $S = \langle \vec{e}_1, \dots, \vec{e}_n \rangle$. $\vec{b} \in V$.

$$A = \begin{pmatrix} \downarrow & \downarrow & \dots & \downarrow \\ \vec{e}_1 & \vec{e}_2 & & \vec{e}_n \end{pmatrix}$$

$$A \cdot \vec{x} = \vec{b}$$

\uparrow
coefficients.

$$V = \mathbb{R}^n \quad S = \left\langle \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \dots, \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix} \right\rangle. \quad \vec{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix} \in \mathbb{R}^n.$$

$$\vec{x} = \vec{b}$$

Orthonormal basis for \mathbb{R}^n .

$$S = \{ \vec{v}_1, \dots, \vec{v}_n \}$$

S is orthonormal basis if

1) S is basis

2) $\vec{v}_i \cdot \vec{v}_j = 0$ ($\Leftrightarrow \vec{v}_i \perp \vec{v}_j$) if $i \neq j$

3) $\vec{v}_i \cdot \vec{v}_i = 1$

Suppose S is orthonormal. and $\vec{b} = \sum_i b_i \vec{v}_i$

$$\vec{b} \cdot \vec{v}_j = \left(\sum_{i=1}^n b_i \vec{v}_i \right) \cdot \vec{v}_j = \sum_{i=1}^n b_i \vec{v}_i \cdot \vec{v}_j = b_j \vec{v}_j \cdot \vec{v}_j = b_j \cdot \|\vec{v}_j\|^2$$

Now. $C = \begin{pmatrix} \vec{v}_1 \\ \vec{v}_2 \\ \vdots \\ \vec{v}_n \end{pmatrix}$ $C^T = \begin{pmatrix} \downarrow \vec{v}_1 \\ \downarrow \vec{v}_2 \\ \downarrow \\ \vdots \\ \downarrow \vec{v}_n \end{pmatrix}$

$$C \cdot C^T = \begin{pmatrix} \vec{v}_1 \cdot \vec{v}_1 & \vec{v}_1 \cdot \vec{v}_2 & \dots & \vec{v}_1 \cdot \vec{v}_n \\ \vdots & \ddots & \ddots & \vdots \\ \vec{v}_i \cdot \vec{v}_i & \dots & \dots & \dots \\ \vdots & \dots & \dots & \vdots \\ \vec{v}_n \cdot \vec{v}_1 & \dots & \dots & \vec{v}_n \cdot \vec{v}_n \end{pmatrix} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & 1 \end{pmatrix} = I.$$

Def (Orthogonal Matrix) If $A \in M_{n \times n}(\mathbb{R})$

we say A is an orthogonal matrix if $A \cdot A^T = I$.

$$A \cdot \vec{x} = \vec{b} \Leftrightarrow \vec{x} = A^{-1} \vec{b} = A^T \cdot \vec{b}$$

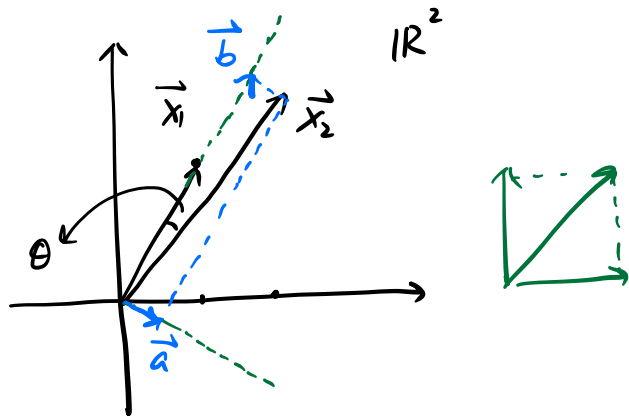
Computational Task:

Given $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n \in \mathbb{R}^n$ (not necessarily orthogonal).

construct an orthonormal basis $\vec{v}_1, \dots, \vec{v}_n$?

eg. $V = \mathbb{R}^2$

$$\vec{x}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad \vec{x}_2 = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$



$\frac{\vec{x}_1}{\|\vec{x}_1\|}$: length 1, unit vector

$$\vec{x}_2 \cdot \vec{x}_1 = \|\vec{x}_2\| \cdot \|\vec{x}_1\| \cdot \cos \theta$$

\vec{b} is the projection of \vec{x}_2 along \vec{x}_1 .

$$\|\vec{b}\| = \|\vec{x}_2\| \cdot \cos \theta = \frac{\vec{x}_2 \cdot \vec{x}_1}{\|\vec{x}_1\|}$$

$$\vec{b} = \|\vec{b}\| \cdot \frac{\vec{x}_1}{\|\vec{x}_1\|} = \left(\frac{\vec{x}_2 \cdot \vec{x}_1}{\|\vec{x}_1\| \cdot \|\vec{x}_1\|} \right) \vec{x}_1$$

$$= \left(\frac{\vec{x}_2 \cdot \vec{x}_1}{\vec{x}_1 \cdot \vec{x}_1} \right) \cdot \vec{x}_1$$

$$\vec{a} = \vec{x}_2 - \left(\frac{\vec{x}_2 \cdot \vec{x}_1}{\vec{x}_1 \cdot \vec{x}_1} \right) \vec{x}_1 \perp \vec{x}_1$$

Plug in: $\vec{b} = \left(\frac{8}{5} \right) \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 8/5 \\ 16/5 \end{pmatrix}$

$$\vec{x}_2 \cdot \vec{x}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \cdot \begin{pmatrix} 2 \\ 3 \end{pmatrix} = 1 \times 2 + 2 \times 3 = 8$$

$$\vec{x}_1 \cdot \vec{x}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 5$$

$$\vec{a} = \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 8/5 \\ 16/5 \end{pmatrix} = \begin{pmatrix} 2/5 \\ -1/5 \end{pmatrix} \quad \|\vec{a}\| = \frac{1}{\sqrt{5}}$$

$$\vec{x}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad \vec{w}_1 = \frac{1}{\sqrt{5}} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{pmatrix}$$

$$\vec{a} = \begin{pmatrix} 2/5 \\ -1/5 \end{pmatrix} \quad \vec{w}_2 = \begin{pmatrix} 2/\sqrt{5} \\ -1/\sqrt{5} \end{pmatrix}$$

$\{\vec{w}_1, \vec{w}_2\}$ is an orthonormal basis.

$\vec{r} = \begin{pmatrix} 21 \\ 13 \end{pmatrix}$ what is coefficients of \vec{r} in $\{\vec{w}_1, \vec{w}_2\}$.

$$A = \begin{pmatrix} \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ 2/\sqrt{5} & -1/\sqrt{5} \end{pmatrix}$$

$$A \cdot \vec{x} = \vec{r} \quad \vec{x} = A^{-1} \vec{r} \\ = A^T \vec{r}$$

$$A^T = \begin{pmatrix} \longrightarrow \vec{w}_1 \\ \longrightarrow \vec{w}_2 \end{pmatrix}$$

$$r_1 = \vec{v}_1 \cdot \vec{r} = \begin{pmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{pmatrix} \cdot \begin{pmatrix} 21 \\ 13 \end{pmatrix} = \frac{47}{\sqrt{5}}$$

$$r_2 = \vec{v}_2 \cdot \vec{r} = \begin{pmatrix} 2/\sqrt{5} \\ -1/\sqrt{5} \end{pmatrix} \cdot \begin{pmatrix} 21 \\ 13 \end{pmatrix} = \frac{29}{\sqrt{5}}$$

In general, this is called Gram-Schmidt process

$$\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n \in \mathbb{R}^m \quad (m > n)$$

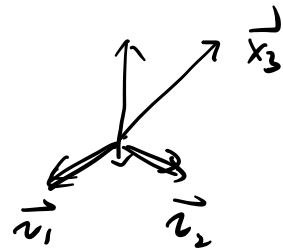
$$\vec{v}_1 = \vec{x}_1$$

$$\vec{v}_2 = \vec{x}_2 - \left(\frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \right) \vec{v}_1$$

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

⋮

$$\vec{v}_n = \vec{x}_n - \frac{\vec{x}_n \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \cdot \vec{v}_1 - \dots - \frac{\vec{x}_n \cdot \vec{v}_{n-1}}{\vec{v}_{n-1} \cdot \vec{v}_{n-1}} \cdot \vec{v}_{n-1}$$



Normalize each \vec{v}_i by $\frac{\vec{v}_i}{\|\vec{v}_i\|}$, we get orthonormal basis.

Consequence of Gram-Schmidt: QR factorization

$$\left(\begin{array}{c} \downarrow \vec{x}_1 \\ \downarrow \vec{x}_2 \end{array} \right) = \left(\begin{array}{c} \downarrow \vec{v}_1 \\ \downarrow \vec{v}_2 \end{array} \right) \cdot \begin{pmatrix} 1 & & \\ & \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} & \\ & & 1 \end{pmatrix} = \left(\begin{array}{c} \downarrow \frac{\vec{v}_1}{\|\vec{v}_1\|} \\ \downarrow \frac{\vec{v}_2}{\|\vec{v}_2\|} \end{array} \right) \cdot \begin{pmatrix} \|\vec{v}_1\| & & \\ & \frac{\vec{x}_2 \cdot \vec{v}_1 \cdot \|\vec{v}_1\|}{\vec{v}_1 \cdot \vec{v}_1} & \\ & & \|\vec{v}_2\| \end{pmatrix}$$

↑ orthogonal ↑ upper triangular

" general invertible

matrix

matrix

$$\text{eg. } \begin{pmatrix} 1 & 2 \\ 2 & 3 \end{pmatrix} = \begin{pmatrix} 1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & -1/\sqrt{5} \end{pmatrix} \cdot \begin{pmatrix} \sqrt{5} & 8/\sqrt{5} \\ 0 & 1/\sqrt{5} \end{pmatrix}$$

Ex. \mathbb{R}^4

$$\vec{x}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} \quad \vec{x}_2 = \begin{pmatrix} 0 \\ -2 \\ 1 \\ 1 \end{pmatrix} \quad \vec{x}_3 = \begin{pmatrix} 2 \\ 3 \\ 0 \\ 1 \end{pmatrix}$$

Gram-Schmidt process:

$$\vec{x}_3 \cdot \vec{v}_2 = -1 - 6 + \frac{1}{2} = -\frac{13}{2}$$

$$\vec{v}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

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

$$\vec{v}_1 \cdot \vec{v}_2 = \frac{1}{4} + 4 + 1 + \frac{1}{4}$$

$$= \begin{pmatrix} 0 \\ -2 \\ 1 \\ 1 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -1/2 \\ -2 \\ 1 \\ 1/2 \end{pmatrix}$$

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

$$= \begin{pmatrix} 2 \\ 3 \\ 0 \\ 1 \end{pmatrix} - \frac{3}{2} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} - \frac{-13/2}{11/2} \cdot \begin{pmatrix} -1/2 \\ -2 \\ 1 \\ 1/2 \end{pmatrix}$$

$$= \begin{pmatrix} 2 - 3/2 - \frac{13}{22} \\ 3 - 26/11 \\ 13/11 \\ 1 - 3/2 + \frac{13}{22} \end{pmatrix} = \begin{pmatrix} -1/11 \\ 7/11 \\ 13/11 \\ 1/11 \end{pmatrix}$$