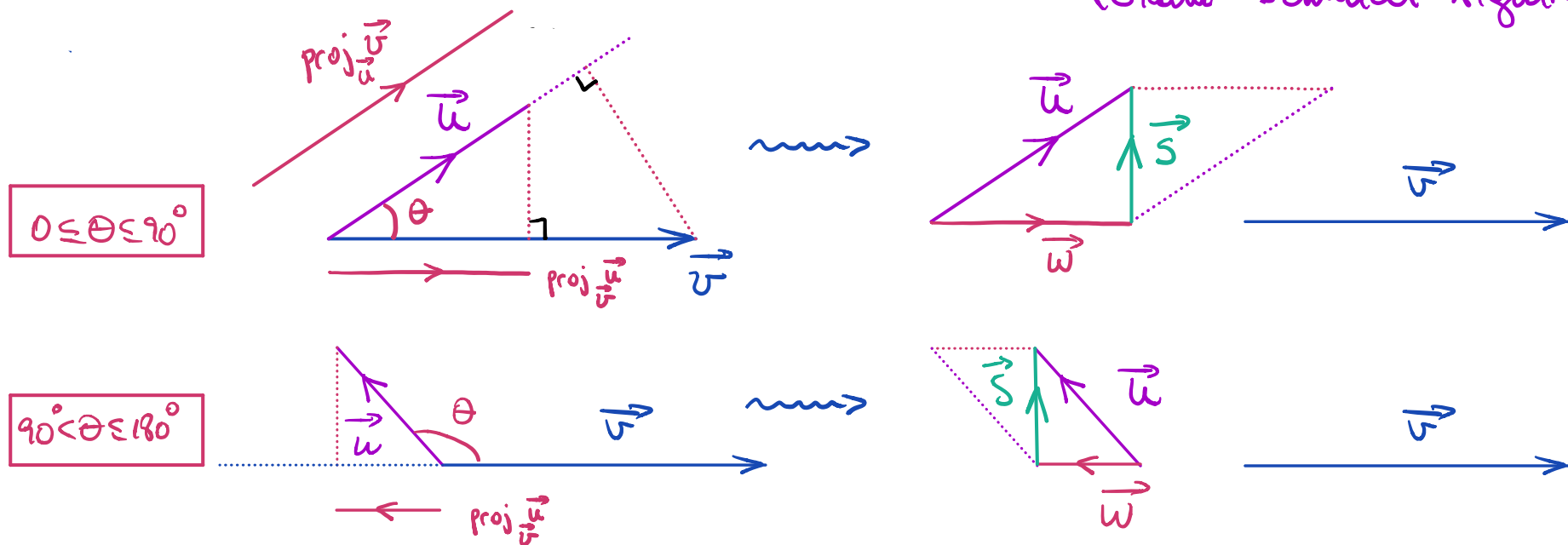


Lecture 18: § 3.6 Orthogonal bases for subspaces

TODAY: Discuss inner products (generalizing dot products on \mathbb{R}^n) & find bases that are well-behaved with respect to these inner products. (Gram-Schmidt Algorithm)



$$\vec{u} = \underbrace{\text{proj}_{\vec{v}} \vec{u}}_{\vec{w}} + \underbrace{(\vec{u} - \text{proj}_{\vec{v}} \vec{u})}_{\vec{s}} \quad \text{where } \vec{w} \parallel \vec{v} \text{ and } \vec{s} \perp \vec{v}$$

Key: \vec{w} & \vec{s} are the only vectors with these properties. (L11)

Basis: $\{\vec{v}, \vec{u}\} \rightsquigarrow \{\vec{v}, \vec{s}\} \quad \vec{v} \perp \vec{s} \quad (\text{like } e_1 \text{ \& } e_2)$

Inner Products of Subspaces V of \mathbb{R}^n

Def: An inner product for V is a function $\langle, \rangle : V \times V \longrightarrow \mathbb{R}$ assigning a number $\langle \vec{u}, \vec{v} \rangle$ to each pair of vectors $\vec{u}, \vec{v} \in V$ & satisfying the following properties:

- ① $\langle \vec{u}, \vec{u} \rangle \geq 0$ for all \vec{u} ; $\langle \vec{u}, \vec{u} \rangle = 0$ if & only if $\vec{u} = \vec{0}$
- ② $\langle \vec{u}, \vec{v} \rangle = \langle \vec{v}, \vec{u} \rangle$ (Symmetric)
- ③ $\langle \alpha \vec{u}, \vec{v} \rangle = \alpha \langle \vec{u}, \vec{v} \rangle = \langle \vec{u}, \alpha \vec{v} \rangle$ for any scalar α .
- ④ $\langle \vec{u} + \vec{w}, \vec{v} \rangle = \langle \vec{u}, \vec{v} \rangle + \langle \vec{w}, \vec{v} \rangle$
 $\langle \vec{u}, \vec{v} + \vec{w} \rangle = \langle \vec{u}, \vec{v} \rangle + \langle \vec{u}, \vec{w} \rangle$

Main example $\langle \vec{u}, \vec{v} \rangle = \vec{u} \cdot \vec{v} = \begin{matrix} 1 \times n \\ \times n \end{matrix} \begin{matrix} n \times 1 \end{matrix}$ in \mathbb{R}^n (dot product)

Q: Why? A: Inner products define norms on \mathbb{R}^n :

$$\|\vec{v}\|_{\langle, \rangle} = \sqrt{\langle \vec{v}, \vec{v} \rangle} \quad (\text{depends on } \langle, \rangle)$$

Example 2: Pick a $n \times n$ symmetric matrix of rank n

(symmetric + invertible Eg $Q = I_n$)

Set $\langle \vec{u}, \vec{v} \rangle := \vec{u}^T Q \vec{v} = \underbrace{[u_1 \dots u_n]}_{1 \times n} \underbrace{Q}_{n \times n} \underbrace{\begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}}_{n \times 1} = \text{a number}$

Check the 4 properties for an inner product:

③ $\langle \alpha \vec{u}, \vec{v} \rangle = \alpha \langle \vec{u}, \vec{v} \rangle = \langle \vec{u}, \alpha \vec{v} \rangle$ for any scalar α

$(\alpha \vec{u})^T Q \vec{v} = \alpha (\vec{u}^T Q \vec{v}) = \vec{u}^T Q \alpha \vec{v}$ ✓
 ↳ true for matrices

④ $\langle \vec{u} + \vec{w}, \vec{v} \rangle = \langle \vec{u}, \vec{v} \rangle + \langle \vec{w}, \vec{v} \rangle$ & $\langle \vec{u}, \vec{v} + \vec{w} \rangle = \langle \vec{u}, \vec{v} \rangle + \langle \vec{u}, \vec{w} \rangle$

Also true from matrix multiplication properties ✓

② $\langle \vec{u}, \vec{v} \rangle \stackrel{?}{=} \langle \vec{v}, \vec{u} \rangle$ (Symmetric)

$\langle \vec{v}, \vec{u} \rangle = \underbrace{\vec{v}^T Q \vec{u}}_{\text{in } \mathbb{R}} = (\vec{v}^T Q \vec{u})^T = \vec{u}^T Q^T \vec{v} = \langle \vec{u}, \vec{v} \rangle$
 ↳ transpose rules $Q^T = Q$

① $\langle \vec{u}, \vec{u} \rangle \geq 0$ for all \vec{u}

⚠ NOT always true!

Condition: Q needs to be positive definite (eg all eigenvalues > 0)

Ex $Q = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ gives $\langle \begin{bmatrix} x \\ y \end{bmatrix}, \begin{bmatrix} x \\ y \end{bmatrix} \rangle = x^2 - y^2$
 So $\langle [1], [1] \rangle = 0$ but $[1] \neq \vec{0}$.

Orthogonal bases for \mathbb{R}^n

Fix $\langle, \rangle =$ dot product

Def: \vec{u} & \vec{v} in \mathbb{R}^n are orthogonal if $\langle \vec{u}, \vec{v} \rangle = 0$. Write $\vec{u} \perp \vec{v}$

Def: A set of vectors $S = \{\vec{v}_1, \dots, \vec{v}_p\}$ in \mathbb{R}^n is orthogonal if $\vec{v}_i \perp \vec{v}_j$ for each pair with $i \neq j$

Examples: $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ is orthogonal.

$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$ —————. $v_1 \cdot v_2 = v_1 \cdot v_3 = v_2 \cdot v_3 = 0$

Q: Why do we care?

Theorem: Orthogonal set NOT containing $\vec{0}$ are ALWAYS l.i.

Why? Write $a_1 \vec{v}_1 + \dots + a_p \vec{v}_p = \vec{0}$ & show $a_1 = \dots = a_p = 0$

$$\left. \begin{aligned} 0 &= \vec{v}_1 \cdot \vec{0} = \vec{v}_1 \cdot (a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_p \vec{v}_p) \\ &= a_1 \underbrace{\vec{v}_1 \cdot \vec{v}_1}_{\|\vec{v}_1\|^2 \neq 0} + a_2 \underbrace{\vec{v}_1 \cdot \vec{v}_2}_{=0} + \dots + a_p \underbrace{\vec{v}_1 \cdot \vec{v}_p}_{=0} \end{aligned} \right\} \begin{array}{l} \text{This forces } a_1 = 0 \\ \text{(rest also 0 by the} \\ \text{same idea)} \quad \square \end{array}$$

Fix $W \neq \{\vec{0}\}$ a subspace of \mathbb{R}^n with basis $B = \{\vec{w}_1, \dots, \vec{w}_p\}$

Def We say B is an orthogonal basis if B is an orthogonal set

orthonormal basis if B is orthogonal
& $\|\vec{w}_1\| = \|\vec{w}_2\| = \dots = \|\vec{w}_p\| = 1$ (norm 1 vectors)

Examples ① $W = \mathbb{R}^n$ $B = \left\{ \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \right\}$ orthonormal basis

② $W = \text{Sp} \left(\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \right) = \text{plane } (x - z = 0)$ ($\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \times \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} -3 \\ 0 \\ 3 \end{bmatrix}$)
or $\dim W = 2$

• $B = \{\vec{v}_1, \vec{v}_2\}$ is an orthogonal basis • $\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = 1 - 2 + 1 = 0$
• basis since it spans W & $\#B = 2 = \dim W$

• B is NOT orthonormal $\|\vec{v}_1\| = \sqrt{1^2 + (-1)^2 + 1^2} = \sqrt{3}$ & $\|\vec{v}_2\| = \sqrt{1^2 + 2^2 + 1^2} = \sqrt{6}$

\rightsquigarrow New orthonormal basis: $\left\{ \frac{1}{\sqrt{3}} \vec{v}_1, \frac{1}{\sqrt{6}} \vec{v}_2 \right\}$

In general: $B = \{\vec{v}_1, \dots, \vec{v}_p\}$ orthogonal basis for W \rightsquigarrow $B' = \left\{ \frac{\vec{v}_1}{\|\vec{v}_1\|}, \dots, \frac{\vec{v}_p}{\|\vec{v}_p\|} \right\}$
is orthonormal basis for W .

Main advantage of orthonormal bases: fast to write down coordinates!

Theorem: Fix $B = \{ \vec{v}_1, \dots, \vec{v}_p \}$ orthonormal basis for $\mathbb{V} \neq \{ \vec{0} \}$ in \mathbb{R}^n .

Then, for each \vec{v} in \mathbb{V} $[\vec{v}]_B = \begin{bmatrix} \vec{v} \cdot \vec{v}_1 \\ \vdots \\ \vec{v} \cdot \vec{v}_p \end{bmatrix}$ in \mathbb{R}^p ,

meaning $\vec{v} = \underbrace{(\vec{v} \cdot \vec{v}_1)}_{\text{scalar}} \vec{v}_1 + \underbrace{(\vec{v} \cdot \vec{v}_2)}_{\text{scalar}} \vec{v}_2 + \dots + \underbrace{(\vec{v} \cdot \vec{v}_p)}_{\text{scalar}} \vec{v}_p$.

Why? Write $\vec{v} = a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_p \vec{v}_p$ a_1, \dots, a_p scalars

$$\begin{aligned} \text{Then } \vec{v} \cdot \vec{v}_1 &= (a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_p \vec{v}_p) \cdot \vec{v}_1 \\ &= a_1 (\underbrace{\vec{v}_1 \cdot \vec{v}_1}_{=\|\vec{v}_1\|^2=1}) + a_2 (\underbrace{\vec{v}_2 \cdot \vec{v}_1}_{=0 \text{ } (\vec{v}_2 \perp \vec{v}_1)}) + \dots + a_p (\underbrace{\vec{v}_p \cdot \vec{v}_1}_{=0 \text{ } (\vec{v}_p \perp \vec{v}_1)}) \end{aligned}$$

So $a_1 = \vec{v} \cdot \vec{v}_1$. Same idea gives $\begin{cases} a_2 = \vec{v} \cdot \vec{v}_2 \\ a_p = \vec{v} \cdot \vec{v}_p \end{cases}$ (use $\vec{v} \cdot \vec{v}_2$) (use $\vec{v} \cdot \vec{v}_p$)

! If B is just an orthogonal basis, then

$$\vec{v} = \frac{\vec{v} \cdot \vec{v}_1}{\|\vec{v}_1\|^2} \vec{v}_1 + \dots + \frac{\vec{v} \cdot \vec{v}_p}{\|\vec{v}_p\|^2} \vec{v}_p.$$

Gram-Schmidt Algorithm

Build orthogonal basis from any basis

- INPUT: $B = \{\vec{w}_1, \dots, \vec{w}_p\}$ basis for subspace \mathcal{V} of \mathbb{R}^n
- OUTPUT: $B' = \{\vec{u}_1, \dots, \vec{u}_p\}$ orthogonal basis for \mathcal{V} .

Routine: $\vec{u}_1 = \vec{w}_1$

$$\vec{u}_2 = \vec{w}_2 - \text{proj}_{\vec{u}_1} \vec{w}_2 = \vec{w}_2 - \frac{\vec{u}_1 \cdot \vec{w}_2}{\|\vec{u}_1\|^2} \vec{u}_1$$

$$\vec{u}_3 = \vec{w}_3 - \text{proj}_{\vec{u}_1} \vec{w}_3 - \text{proj}_{\vec{u}_2} \vec{w}_3 = \vec{w}_3 - \frac{\vec{u}_1 \cdot \vec{w}_3}{\|\vec{u}_1\|^2} \vec{u}_1 - \frac{\vec{u}_2 \cdot \vec{w}_3}{\|\vec{u}_2\|^2} \vec{u}_2$$

⋮

In general: once we build $\vec{u}_1, \dots, \vec{u}_{j-1}$, we set \vec{u}_j by

$$\begin{aligned} \vec{u}_j &= \vec{w}_j - \text{proj}_{\vec{u}_1} \vec{w}_j - \text{proj}_{\vec{u}_2} \vec{w}_j - \dots - \text{proj}_{\vec{u}_{j-1}} \vec{w}_j \\ &= \vec{w}_j - \frac{\vec{u}_1 \cdot \vec{w}_j}{\|\vec{u}_1\|^2} \vec{u}_1 - \frac{\vec{u}_2 \cdot \vec{w}_j}{\|\vec{u}_2\|^2} \vec{u}_2 - \dots - \frac{\vec{u}_{j-1} \cdot \vec{w}_j}{\|\vec{u}_{j-1}\|^2} \vec{u}_{j-1} \end{aligned}$$

↑ scalars

$$\vec{u}_j = \vec{w}_j - \frac{\vec{u}_1 \cdot \vec{w}_j}{\|\vec{u}_1\|^2} \vec{u}_1 - \frac{\vec{u}_2 \cdot \vec{w}_j}{\|\vec{u}_2\|^2} \vec{u}_2 - \dots - \frac{\vec{u}_{j-1} \cdot \vec{w}_j}{\|\vec{u}_{j-1}\|^2} \vec{u}_{j-1}$$

↑ scalars

Q: Why these scalars?

- $\vec{u}_1 = \vec{w}_1$ so \vec{w}_1 in $\text{Sp}(\vec{u}_1)$ & \vec{u}_j in $\text{Sp}(\vec{u}_1, \vec{u}_2, \dots, \vec{u}_j)$
- \vec{u}_j in $\text{Sp}(\vec{w}_j, \vec{u}_1, \dots, \vec{u}_{j-1}) = \text{Sp}(\vec{w}_j, \vec{w}_1, \vec{w}_2, \dots, \vec{w}_{j-1})$
- $\vec{u}_j \cdot \vec{u}_1 = \vec{w}_j \cdot \vec{u}_1 - \frac{\vec{u}_1 \cdot \vec{w}_j}{\|\vec{u}_1\|^2} \|\vec{u}_1\|^2 - \frac{\vec{u}_2 \cdot \vec{w}_j}{\|\vec{u}_2\|^2} \underbrace{\vec{u}_2 \cdot \vec{u}_1}_{=0} - \dots - \frac{\vec{u}_{j-1} \cdot \vec{w}_j}{\|\vec{u}_{j-1}\|^2} \underbrace{\vec{u}_{j-1} \cdot \vec{u}_1}_{=0}$
 $= \vec{w}_j \cdot \vec{u}_1 - \vec{u}_1 \cdot \vec{w}_j = 0$

Same idea says $\vec{u}_j \cdot \vec{u}_2 = \dots = \vec{u}_j \cdot \vec{u}_{j-1} = 0$

Three consequences: ① $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_p$ are all in \mathbb{V}

② Each $\vec{w}_1, \dots, \vec{w}_p$ is in $\text{Sp}(\overbrace{\vec{u}_1, \dots, \vec{u}_p}^{= B'})$ so
 So $\mathbb{V} = \text{Sp}(\vec{w}_1, \dots, \vec{w}_p) = \text{Sp}(\vec{u}_1, \dots, \vec{u}_p)$

③ B' is an orthogonal set

B' is a basis

Example

$V = \mathbb{R}^3$

INPUT: $\{ \overset{\vec{w}_1}{\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}}, \overset{\vec{w}_2}{\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}}, \overset{\vec{w}_3}{\begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix}} \}$

OUTPUT: $\{ \vec{u}_1, \vec{u}_2, \vec{u}_3 \} = \{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 4 \end{bmatrix} \}$

• $\vec{u}_1 = \vec{w}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \implies \|\vec{u}_1\| = \vec{u}_1 \cdot \vec{u}_1 = 1^2 + 0^2 + 0^2$
 $\vec{u}_1 \cdot \vec{w}_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} = 1$

• $\vec{u}_2 = \vec{w}_2 - \text{proj}_{\vec{u}_1} \vec{w}_2 = \vec{w}_2 - \frac{\vec{u}_1 \cdot \vec{w}_2}{\|\vec{u}_1\|^2} \vec{u}_1 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} - \frac{1}{1} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}$

• $\vec{u}_3 = \vec{w}_3 - \text{proj}_{\vec{u}_1} \vec{w}_3 - \text{proj}_{\vec{u}_2} \vec{w}_3 = \vec{w}_3 - \frac{\vec{u}_1 \cdot \vec{w}_3}{\|\vec{u}_1\|^2} \vec{u}_1 - \frac{\vec{u}_2 \cdot \vec{w}_3}{\|\vec{u}_2\|^2} \vec{u}_2$
 $= \begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix} - \frac{1}{1} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - \frac{(-6)}{4} \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 3 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 4 \end{bmatrix}$

$\|\vec{u}_2\|^2 = \vec{u}_2 \cdot \vec{u}_2 = 0^2 + 2^2 + 0^2 = 4$

$\vec{u}_1 \cdot \vec{w}_3 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix} = 1$; $\vec{u}_2 \cdot \vec{w}_3 = \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix} = -6$

Q: Orthonormal basis? A: $\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \}$