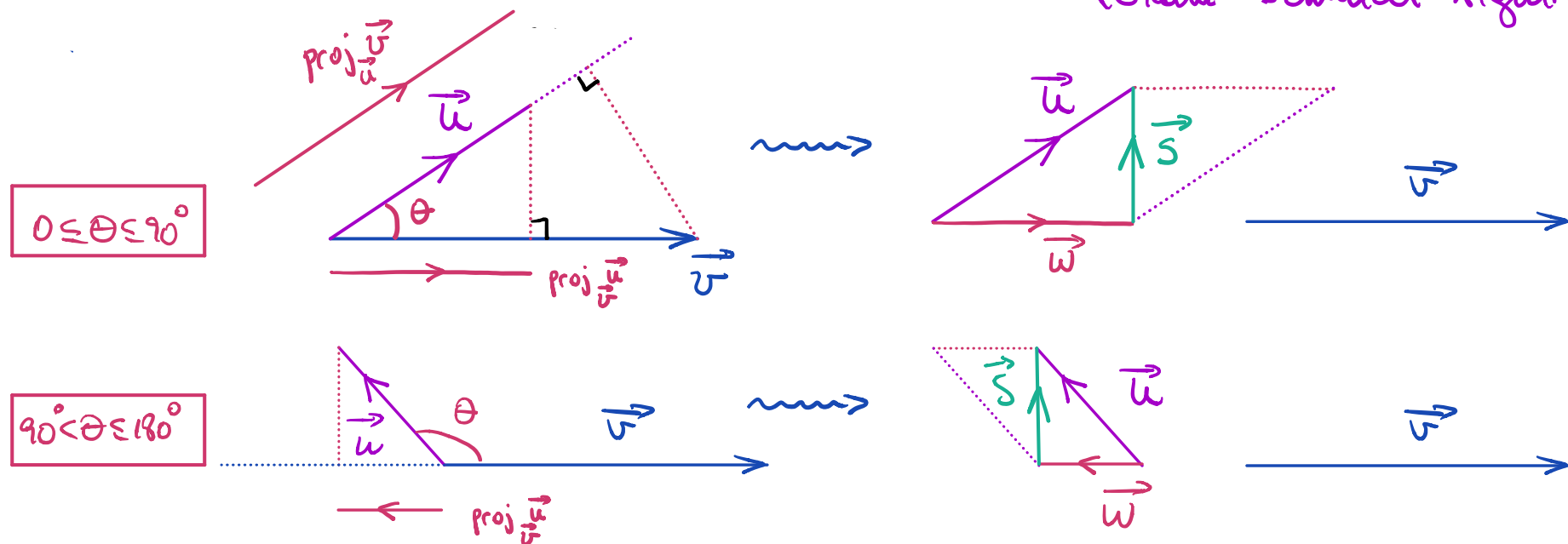


Lecture 18: § 3.6 Orthogonal bases for subspaces

TODAY: Discuss inner products (generalizing dot products in $\mathbb{R}^n$)
 & 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)

Bases: $\{\vec{v}, \vec{u}\} \rightsquigarrow \{\vec{v}, \vec{s}\} \quad \vec{v} \perp \vec{s} \quad (\text{like } \vec{e}_1 \text{ \& } \vec{e}_2)$

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

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

Example 2: Pick Q $n \times n$ symmetric matrix of rank n
(symmetric + invertible Eg $Q = I_n$)

Check the 4 properties for an inner product:

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

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

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

$$\textcircled{1} \quad \langle \vec{u}, \vec{u} \rangle \geq 0 \quad \text{for all } \vec{u} \quad ; \quad \langle \vec{u}, \vec{u} \rangle = 0 \quad \text{if \& mly if } \vec{u} = \vec{0}.$$

Orthogonal bases for $\mathbb{R}^n$

Fix $\langle, \rangle =$ dot product

Def: $\vec{u}$ & $\vec{v}$ in $\mathbb{R}^n$ are orthogonal if

Def: A set of vectors $S = \{\vec{v}_1, \dots, \vec{v}_p\}$ in $\mathbb{R}^n$ is orthogonal if

Q: Why do we care?

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

Fix $V \neq \{\vec{0}\}$ 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
orthonormal basis if

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

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

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

$$u_3 = \vec{w}_3 - \text{proj}_{\vec{u}_1} \vec{w}_3 - \text{proj}_{\vec{u}_2} \vec{w}_3 = \vec{w}_3 - \frac{\vec{u}_1 \cdot \vec{w}_3}{\|\vec{u}_1\|^2} \vec{u}_1 - \frac{\vec{u}_2 \cdot \vec{w}_3}{\|\vec{u}_2\|^2} \vec{u}_2$$

⋮

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

Q: Why these scalars?

↑ scalars

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

Example

$V = \mathbb{R}^3$

OUTPUT : $\{ \vec{u}_1, \vec{u}_2, \vec{u}_3 \} =$