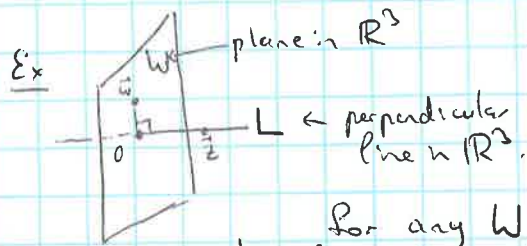


Orthogonal Complements

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If $\vec{z} \in \mathbb{R}^n$ is orthogonal to every vector in $W \subset \mathbb{R}^n$ (a subspace), then \vec{z} is orthogonal to W .

Set of all vectors in \mathbb{R}^n orthogonal to W - "orthogonal complement of W ", W^\perp - notation



$$W^\perp = L, L^\perp = W$$

- For any W
- \vec{x} is in W^\perp iff \vec{x} is orthog. to every vector in a set which spans W .
 - W^\perp is a subspace of \mathbb{R}^n .

Ex for A $m \times n$ mat., $\text{Nul } A$ and $\text{Row } A \subset \mathbb{R}^n$ - orthogonal complements of each other

$\text{Nul } A^T$ and $\text{Col } A \subset \mathbb{R}^m$ - orthogonal complements of each other.

• for $W \subset \mathbb{R}^n$, $\boxed{\dim W + \dim W^\perp = n}$

6.2 Orthogonal sets

A set of vectors $\{\vec{u}_1, \dots, \vec{u}_p\}$ in \mathbb{R}^n is an orthogonal set, if $\vec{u}_i \cdot \vec{u}_j = 0$ for each pair $i \neq j$.

Ex: $\vec{u}_1 = \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}$, $\vec{u}_2 = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}$, $\vec{u}_3 = \begin{bmatrix} -1/2 \\ 2 \\ 7/2 \end{bmatrix}$ Q: show that $\{\vec{u}_1, \vec{u}_2, \vec{u}_3\}$ is an orthog. set

Sol: $\vec{u}_1 \cdot \vec{u}_2 = 3(-1) + 1 \cdot 2 + 1 \cdot 1 = 0$, $\vec{u}_1 \cdot \vec{u}_3 = 3(-1/2) + 1 \cdot 2 + 1 \cdot 7/2 = 0$, $\vec{u}_2 \cdot \vec{u}_3 = (-1)(-1/2) + 2(-2) + 1 \cdot 7/2 = 0$

THM If $S = \{\vec{u}_1, \dots, \vec{u}_p\}$ an orthog set of nonzero vectors in \mathbb{R}^n , then S is lin. indep. and hence a basis for $\text{Span } S$.

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• an orthogonal basis for a subspace $W \subset \mathbb{R}^n$ is a basis for W which is also an orthogonal set.

THM* Let $\{\vec{u}_1, \dots, \vec{u}_p\}$ be an orthog. basis for $W \subset \mathbb{R}^n$. For each $\vec{y} \in W$, weights in

$$\vec{y} = c_1 \vec{u}_1 + \dots + c_p \vec{u}_p \text{ are given by } \boxed{c_j = \frac{\vec{y} \cdot \vec{u}_j}{\vec{u}_j \cdot \vec{u}_j}}, j=1 \dots p$$

(Indeeds $\vec{y} \cdot \vec{u}_i = c_1 \vec{u}_1 \cdot \vec{u}_i + \dots + c_i \vec{u}_i \cdot \vec{u}_i + \dots + c_p \vec{u}_p \cdot \vec{u}_i \Rightarrow c_i = \frac{\vec{y} \cdot \vec{u}_i}{\vec{u}_i \cdot \vec{u}_i}$ and similarly for other c_j)

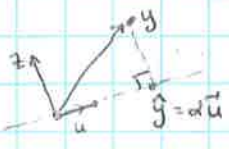
Ex: $S = \{\vec{u}_1, \vec{u}_2, \vec{u}_3\}$ from Ex* is a basis for \mathbb{R}^3 , $\vec{y} = \begin{bmatrix} 6 \\ 1 \\ -8 \end{bmatrix}$ Q: Express \vec{y} as a lin. comb. of vectors in S

Sol: $\vec{y} = \frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 + \frac{\vec{y} \cdot \vec{u}_2}{\vec{u}_2 \cdot \vec{u}_2} \vec{u}_2 + \frac{\vec{y} \cdot \vec{u}_3}{\vec{u}_3 \cdot \vec{u}_3} \vec{u}_3 = \vec{u}_1 - 2\vec{u}_2 - 2\vec{u}_3$

← did not need to solve the lin. system to compute the weights!

Orthogonal projection

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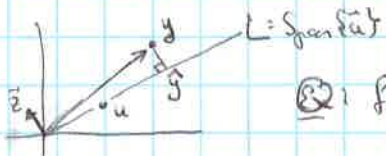
Given $\vec{u} \neq \vec{0}$ in \mathbb{R}^n , want to write $\vec{y} \in \mathbb{R}^n$

as $\vec{y} = \underbrace{\hat{y}}_{\alpha \vec{u} \text{ proj. to } \vec{u}} + \underbrace{\vec{z}}_{\text{orthog. to } \vec{u}}$ $\Rightarrow (\vec{y} - \alpha \vec{u}) \cdot \vec{u} = 0 \Rightarrow \alpha = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}}$

$\hat{y} = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}} \vec{u} =: \text{proj}_L \vec{y}$ - orthogonal projection of \vec{y} onto $L = \text{Span}\{\vec{u}\}$
 does not change if $\vec{u} \rightarrow c\vec{u}$, $c \neq 0$.

Ex: $\vec{y} = \begin{bmatrix} 7 \\ 6 \end{bmatrix}$, $\vec{u} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$. Write \vec{y} as $\hat{y} + \vec{z}$
 $\hat{y} \in \text{Span}\{\vec{u}\}$ orthog. to \vec{u} .

Sol: $\vec{y} \cdot \vec{u} = 40$, $\vec{u} \cdot \vec{u} = 20 \Rightarrow \hat{y} = \frac{40}{20} \begin{bmatrix} 4 \\ 2 \end{bmatrix} = \begin{bmatrix} 8 \\ 4 \end{bmatrix} = 2\vec{u}$, $\vec{z} = \vec{y} - \hat{y} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$. So: $\begin{bmatrix} 7 \\ 6 \end{bmatrix} = \begin{bmatrix} 8 \\ 4 \end{bmatrix} + \begin{bmatrix} -1 \\ 2 \end{bmatrix}$
 $L = \text{Span}\{\vec{u}\}$

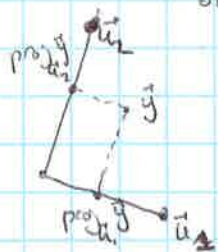


Q: find $\text{dist}(\vec{y}, L)$
 $L = \text{Span}\{\vec{u}\}$

Sol: $\text{dist}(\vec{y}, L) = \text{dist}(\vec{y}, \hat{y}) = \|\vec{y} - \hat{y}\|$
 = $\|\vec{z}\| = \sqrt{(-1)^2 + 2^2} = \sqrt{5}$
 closest point on L to \vec{y}

Ex: (THM, geom. picture)

$W = \mathbb{R}^2 = \text{Span}\{\vec{u}_1, \vec{u}_2\}$
 orthogonal



$\vec{y} \in \mathbb{R}^2$ can be written as $\vec{y} = \underbrace{\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1}_{\text{proj}_{\vec{u}_1} \vec{y}} + \underbrace{\frac{\vec{y} \cdot \vec{u}_2}{\vec{u}_2 \cdot \vec{u}_2} \vec{u}_2}_{\text{proj}_{\vec{u}_2} \vec{y}}$

So: THM decomposes \vec{y} into a sum of orthog. projection onto 1-dim subspaces (which are mutually orthogonal)

Orthonormal sets $S = \{\vec{u}_1, \dots, \vec{u}_p\}$ is orthonormal if it is an orthog. set of unit vectors.

If $W = \text{Span} S$, then S - o/n basis for W .

Ex: $\{\vec{e}_1, \dots, \vec{e}_n\}$ is an o/n basis for \mathbb{R}^n .

Ex: $\vec{v}_1 = \begin{bmatrix} 3/\sqrt{11} \\ 1/\sqrt{11} \\ 1/\sqrt{11} \end{bmatrix}$, $\vec{v}_2 = \begin{bmatrix} -1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix}$, $\vec{v}_3 = \begin{bmatrix} -1/\sqrt{66} \\ -4/\sqrt{66} \\ 7/\sqrt{66} \end{bmatrix}$ - o/n basis for \mathbb{R}^3 .

- obtained from E_n by normalizing vectors to unit length: $\vec{v}_i = \frac{1}{\|\vec{u}_i\|} \vec{u}_i$

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

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

(a) $\|U\vec{x}\| = \|\vec{x}\|$ (b) $(U\vec{x}) \cdot (U\vec{y}) = \vec{x} \cdot \vec{y}$ (c) $(U\vec{x}) \cdot (U\vec{y}) = 0$ iff $\vec{x} \cdot \vec{y} = 0$

I.e. mapping $\vec{x} \mapsto U\vec{x}$ preserves length and orthogonality.

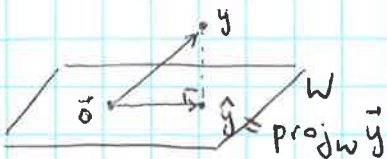
• Case $m=n$: square U with o/n columns is an orthogonal matrix. U orthogonal iff $U^{-1} = U^T$

Orthogonal projections

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Given $\vec{y} \in \mathbb{R}^n$ and $W \subset \mathbb{R}^n$, ^{there exists a unique} ~~want to find~~ $\hat{\vec{y}} \in W$ s.t. (1) $\vec{y} - \hat{\vec{y}} \perp W$

(2) $\hat{\vec{y}}$ is the closest vector in W to \vec{y}



THM (The Orthogonal Decomposition THM)

Let W be a subspace of \mathbb{R}^n . Then each $\vec{y} \in \mathbb{R}^n$ can be written uniquely as

$\vec{y} = \hat{\vec{y}} + \vec{z}$ with $\hat{\vec{y}} \in W$ and $\vec{z} \in W^\perp$. Moreover, if $\{\vec{u}_1, \dots, \vec{u}_p\}$ is any orthogonal

basis for W , then $\hat{\vec{y}} = \frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 + \dots + \frac{\vec{y} \cdot \vec{u}_p}{\vec{u}_p \cdot \vec{u}_p} \vec{u}_p$ ^(**) and $\vec{z} = \vec{y} - \hat{\vec{y}}$.

$\hat{\vec{y}} =: \text{proj}_W \vec{y}$ is the orthogonal projection of \vec{y} onto W .

Ex: $\vec{u}_1 = \begin{bmatrix} 2 \\ 5 \\ -1 \end{bmatrix}$, $\vec{u}_2 = \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$, $\vec{y} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$
 Orthog. basis for $W = \text{Span}\{\vec{u}_1, \vec{u}_2\}$

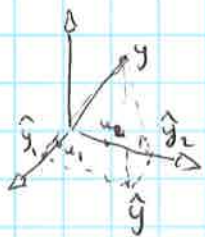
Q: Write \vec{y} as a sum of a vector in W and a vector orthogonal to W .

Sol: $\hat{\vec{y}} = \frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 + \frac{\vec{y} \cdot \vec{u}_2}{\vec{u}_2 \cdot \vec{u}_2} \vec{u}_2 = \frac{9}{30} \begin{bmatrix} 2 \\ 5 \\ -1 \end{bmatrix} + \frac{3}{6} \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -2/5 \\ 2 \\ 1/5 \end{bmatrix}$; $\vec{z} = \vec{y} - \hat{\vec{y}} = \begin{bmatrix} 7/5 \\ 0 \\ 14/5 \end{bmatrix}$

Sol: $\vec{y} = \begin{bmatrix} -2/5 \\ 2 \\ 1/5 \end{bmatrix} + \begin{bmatrix} 7/5 \\ 0 \\ 14/5 \end{bmatrix}$
 $\in W$ $\in W^\perp$

• formula (***) is the sum of projections ^{of \vec{y}} onto lines $\text{Span}\{\vec{u}_1\}, \dots, \text{Span}\{\vec{u}_p\}$

• If $\vec{y} \in W$, $\text{proj}_W \vec{y} = \vec{y}$



THM (best approximation theorem)

Let $W \subset \mathbb{R}^n$, $\vec{y} \in \mathbb{R}^n$ and $\hat{\vec{y}} = \text{proj}_W \vec{y}$. Then $\hat{\vec{y}}$ is the closest point in W to \vec{y} , i.e., $\|\vec{y} - \hat{\vec{y}}\| < \|\vec{y} - \vec{v}\|$ for all $\vec{v} \in W$, $\vec{v} \neq \hat{\vec{y}}$.

$\hat{\vec{y}}$ is the best approximation of \vec{y} by elements of W , $\|\vec{y} - \hat{\vec{y}}\|$ - "error" of the approximation.

Ex: $\hat{\vec{y}} = \begin{bmatrix} -2/5 \\ 2 \\ 1/5 \end{bmatrix}$ - closest point to $\vec{y} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ on the plane W .

$\text{dist}(\vec{y}, W) = \|\vec{y} - \hat{\vec{y}}\| = \|\vec{z}\| = \frac{7}{5} \sqrt{5} = \frac{7}{\sqrt{5}}$

dist. between \vec{y} and closest point to \vec{y} on W

14M If $\{\vec{u}_1, \dots, \vec{u}_p\}$ is an orthonormal basis for $W \subset \mathbb{R}^n$, then

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

If $U = [\vec{u}_1 \ \vec{u}_2 \ \dots \ \vec{u}_p]$, then $\boxed{\text{proj}_W \vec{y} = U U^T \vec{y}}$ for all $\vec{y} \in \mathbb{R}^n$.