

- HW 3 Solutions posted later today
- Practice final posted later this weekend
- Reminder to do your Student Evaluations of Teaching (SETs), open today

Def: An orthogonal basis for a subspace W of \mathbb{R}^n is a basis for W that's also an orthogonal set.

Theorem:

Let $\{\vec{u}_1, \dots, \vec{u}_p\}$ be an orthogonal basis for a subspace W of \mathbb{R}^n .

Let $\vec{y} \in W$. Then, there is the unique expansion in this basis given by

$$\vec{y} = \left(\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \right) \vec{u}_1 + \dots + \left(\frac{\vec{y} \cdot \vec{u}_p}{\vec{u}_p \cdot \vec{u}_p} \right) \vec{u}_p.$$

proof: Since $\vec{y} \in W$ & $\{\vec{u}_1, \dots, \vec{u}_p\}$ is a basis for W , know there exists unique c_1, \dots, c_p st.

$$\vec{y} = c_1 \vec{u}_1 + \dots + c_p \vec{u}_p$$

Dot product this with \vec{u}_j ($j=1, \dots, p$)

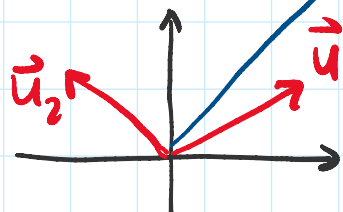
$$\begin{aligned} \vec{y} \cdot \vec{u}_j &= (c_1 \vec{u}_1 + \dots + c_p \vec{u}_p) \cdot \vec{u}_j \\ &= c_j \underbrace{\vec{u}_j \cdot \vec{u}_j}_{\substack{\dots \\ \dots}} \Rightarrow c_j = \frac{\vec{y} \cdot \vec{u}_j}{\vec{u}_j \cdot \vec{u}_j} \end{aligned}$$

$$\frac{\vec{y} \cdot \vec{u}_j}{\|\vec{u}_j\|^2} > 0 \Rightarrow c_j = \frac{\vec{y} \cdot \vec{u}_j}{\vec{u}_j \cdot \vec{u}_j}$$

$j=1, \dots, p \quad \square$

ex Let $\vec{x} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$. Verify the above $\vec{e}_1, \vec{e}_2, \vec{e}_3$ formula w/ orthogonal basis $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

$$\begin{aligned} & \frac{\vec{x} \cdot \vec{e}_1}{\vec{e}_1 \cdot \vec{e}_1} \vec{e}_1 + \frac{\vec{x} \cdot \vec{e}_2}{\vec{e}_2 \cdot \vec{e}_2} \vec{e}_2 + \frac{\vec{x} \cdot \vec{e}_3}{\vec{e}_3 \cdot \vec{e}_3} \vec{e}_3 \\ &= \frac{x}{1} \vec{e}_1 + \frac{y}{1} \vec{e}_2 + \frac{z}{1} \vec{e}_3 = x \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + y \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + z \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \vec{x}. \end{aligned}$$

ex  $\vec{u}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \vec{u}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow \vec{u}_1 \cdot \vec{u}_2 = 0$

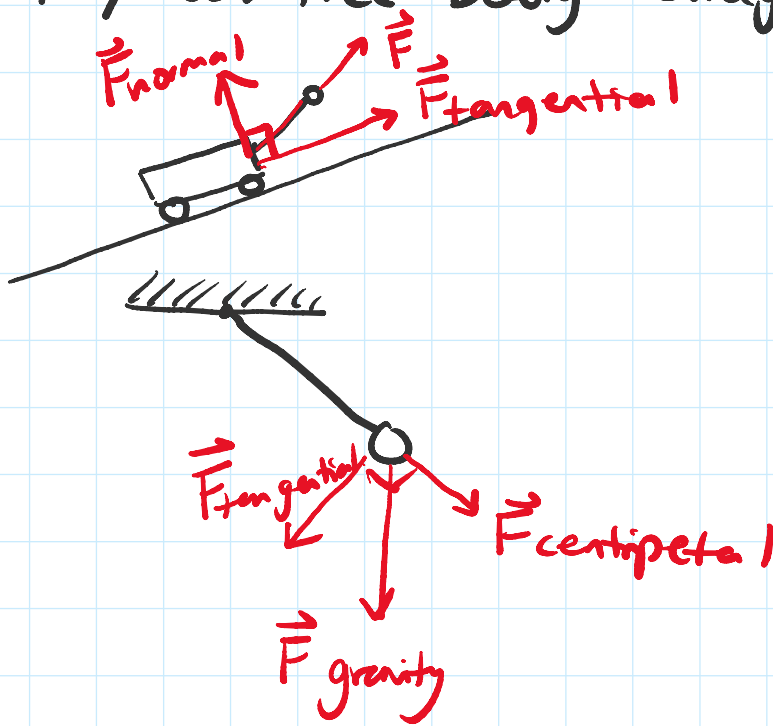
$$\vec{y} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

Write \vec{y} as a lin comb of \vec{u}_1, \vec{u}_2

$$\begin{aligned} \vec{y} &= \frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 + \frac{\vec{y} \cdot \vec{u}_2}{\vec{u}_2 \cdot \vec{u}_2} \vec{u}_2 \\ &= \frac{5}{2} \vec{u}_1 + \frac{1}{2} \vec{u}_2 \quad \left(\frac{5}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix} \checkmark \right) \end{aligned}$$

$$= \frac{5}{2} u_1 + \frac{1}{2} u_2 \quad \left(\frac{5}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix} \checkmark \right)$$

ex Physics: free body diagrams



Def:

- A vector \vec{u} is a unit vector when $\|\vec{u}\| = \sqrt{\vec{u} \cdot \vec{u}} = 1$
- An orthonormal set is an orthogonal set of unit vectors.

ex standard basis \mathbb{R}^n is orthonormal.

An orthonormal set of vectors $\{\vec{u}_1, \dots, \vec{u}_p\}$ satisfies:
$$\vec{u}_i \cdot \vec{u}_j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

Gives $\vec{y} \in W = \text{span}\{\vec{u}_1, \dots, \vec{u}_p\}$ subspace of \mathbb{R}^n

Given $\vec{y} \in W = \text{span} \{ \vec{u}_1, \dots, \vec{u}_p \}$ subspace of \mathbb{R}^n

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

Let $\{ \vec{u}_1, \dots, \vec{u}_n \}$ be an orthonormal basis for \mathbb{R}^n .

Let

$$U = [\vec{u}_1 \ \dots \ \vec{u}_n] \quad U \text{ is invertible, claim } U^{-1} = U^T$$

Observe

$$U^T U = \begin{bmatrix} \vec{u}_1^T \\ \vdots \\ \vec{u}_n^T \end{bmatrix} [\vec{u}_1 \ \dots \ \vec{u}_n]$$

$$= \begin{bmatrix} \vec{u}_1^T \vec{u}_1 & \vec{u}_1^T \vec{u}_2 & \dots & \vec{u}_1^T \vec{u}_n \\ \vdots & \vec{u}_2^T \vec{u}_2 & & \vdots \\ \vdots & & \ddots & \vdots \\ \vec{u}_n^T \vec{u}_1 & \dots & \dots & \vec{u}_n^T \vec{u}_n \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \dots & 0 & 1 \end{bmatrix} = I \Rightarrow U^T = U^{-1}$$

such $n \times n$ matrices w/ orthonormal columns
(or equivalently whose transpose is its inverse)
are called orthogonal matrices.
(bad nomenclature \uparrow should be orthonormal)

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(bad nomenclature \uparrow should be orthonormal)

consider linear system $U\vec{x} = \vec{y}$
where U is an orthogonal matrix.

$$\Rightarrow \vec{x} = U^T \vec{y}$$

$$U\vec{x} = \vec{y} \Leftrightarrow \vec{y} = x_1 \vec{u}_1 + \dots + x_n \vec{u}_n$$

$$\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} \vec{u}_1^T \\ \vdots \\ \vec{u}_n^T \end{bmatrix} \vec{y} = \begin{bmatrix} \vec{u}_1^T \vec{y} \\ \vdots \\ \vec{u}_n^T \vec{y} \end{bmatrix} = \begin{bmatrix} \vec{y} \cdot \vec{u}_1 \\ \vdots \\ \vec{y} \cdot \vec{u}_n \end{bmatrix}$$

Theorem: [The Spectral Theorem]

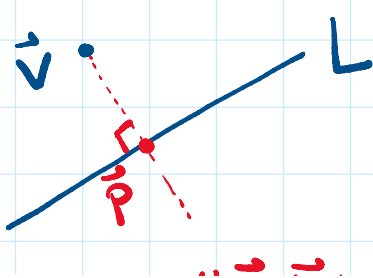
a matrix A is orthogonally diagonalizable,
i.e., there exists an orthogonal matrix U
and a diagonal matrix D s.t.

$$A = UDU^{-1} = UDU^T$$

if and only if A is symmetric ($A = A^T$) \square

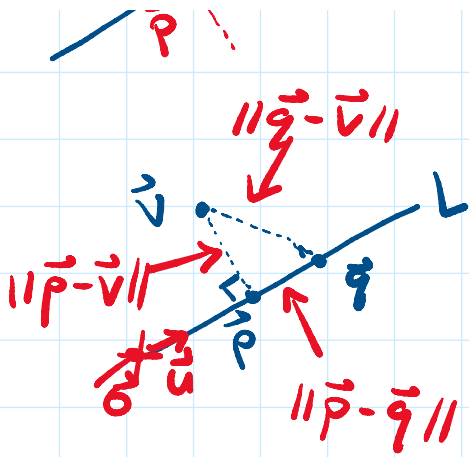
[Orthogonal Projections & Decompositions]

ex/ Consider



What is the point on L
closest to \vec{v} ?

\vec{p} is the point on L closest



\vec{p} is the point on L closest to \vec{v} .

$\vec{q} \neq \vec{p}$, \vec{q} on L

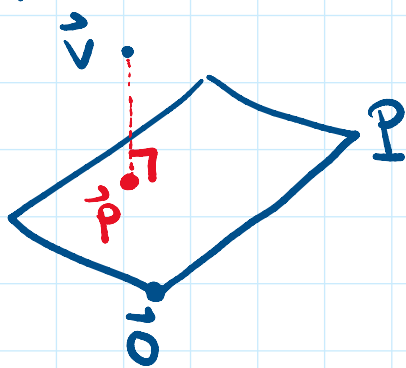
$$\|\vec{q} - \vec{v}\|^2 = \|\vec{p} - \vec{v}\|^2 + \underbrace{\|\vec{p} - \vec{q}\|^2}_{> 0}$$

$$> \|\vec{p} - \vec{v}\|^2$$

i.e. $\|\vec{p} - \vec{v}\| < \|\vec{q} - \vec{v}\|$ for all \vec{q} on L not equal to \vec{p} .

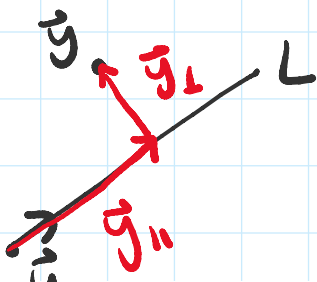
we call \vec{p} the orthogonal projection of \vec{v} onto $L = \text{span}\{\vec{u}\}$. $\text{proj}_L \vec{v}$.

ex plane P in \mathbb{R}^3



$$\vec{p} = \text{proj}_P \vec{v}$$

Going back to first example; let $\vec{u} \in \mathbb{R}^n$
let $L = \text{span}\{\vec{u}\}$. Let $\vec{y} \in \mathbb{R}^n$. $\vec{u} \neq \vec{0}$



Can decompose

$$\vec{y} = \vec{y}_{\parallel} + \vec{y}_{\perp}$$

~~\vec{u}~~ $\vec{y}_{||}$

$$U = \begin{matrix} J_{||} \\ L^{\perp} \end{matrix} \quad \begin{matrix} J_{\perp} \\ L^{\perp} \end{matrix}$$

know $\vec{y}_{||} = \alpha \vec{u}$ for some $\alpha \in \mathbb{R}$. Enforce

$$\vec{y}_{\perp} = \vec{y} - \vec{y}_{||} \in L^{\perp}$$

$$\begin{aligned} 0 &= \vec{y}_{\perp} \cdot \vec{u} = (\vec{y} - \vec{y}_{||}) \cdot \vec{u} = \vec{y} \cdot \vec{u} - \vec{y}_{||} \cdot \vec{u} \\ &= \vec{y} \cdot \vec{u} - \alpha \underbrace{\vec{u} \cdot \vec{u}}_{\neq 0} \Rightarrow \alpha = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}} \end{aligned}$$

$$\Rightarrow \vec{y}_{||} = \left(\frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}} \right) \vec{u}$$

$$\vec{y}_{\perp} = \vec{y} - \vec{y}_{||}$$

Thm: [Orthogonal Decomposition Theorem]

Let W be a subspace of \mathbb{R}^n .

Then, for any $\vec{y} \in \mathbb{R}^n$, there is a unique decomposition

$$\vec{y} = \vec{y}_{||} + \vec{y}_{\perp} \quad \text{where} \quad \begin{matrix} \vec{y}_{||} \in W \\ \vec{y}_{\perp} \in W^{\perp} \end{matrix}$$

Furthermore, assume $W \neq \{\vec{0}\}$; let $\{\vec{u}_1, \dots, \vec{u}_p\}$ be an orthogonal basis for W . Then,

$$\vec{y}_{||} = \left(\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \right) \vec{u}_1 + \dots + \left(\frac{\vec{y} \cdot \vec{u}_p}{\vec{u}_p \cdot \vec{u}_p} \right) \vec{u}_p = \text{proj}_W \vec{y}$$

$$\vec{y}_{\parallel} = \left(\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \right) \vec{u}_1 + \dots + \left(\frac{\vec{y} \cdot \vec{u}_p}{\vec{u}_p \cdot \vec{u}_p} \right) \vec{u}_p = \text{proj}_W \vec{y}$$

$$\vec{y}_{\perp} = \vec{y} - \vec{y}_{\parallel}$$

proof: clearly $\vec{y}_{\parallel} \in W$.

$$\vec{y}_{\perp} = \vec{y} - \vec{y}_{\parallel} \quad \text{check } \in W^{\perp}$$

$$\Leftrightarrow \text{check } \vec{y}_{\perp} \cdot \vec{u}_j = 0 \quad j=1, \dots, p$$

$$\vec{y}_{\perp} \cdot \vec{u}_j = (\vec{y} - \vec{y}_{\parallel}) \cdot \vec{u}_j$$

$$= \vec{y} \cdot \vec{u}_j - \frac{\vec{y} \cdot \vec{u}_j}{\vec{u}_j \cdot \vec{u}_j} \vec{u}_j \cdot \vec{u}_j = 0, \quad j=1, \dots, p$$

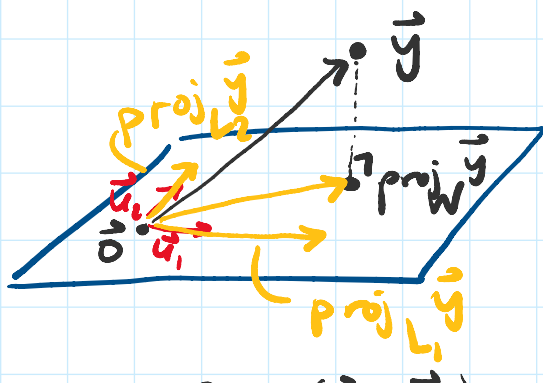
$$\Rightarrow \vec{y}_{\perp} \in W^{\perp}$$

uniqueness

$$\begin{array}{ccc} \vec{y}_{\parallel} & + & \vec{y}_{\perp} = \vec{y} = \vec{z}_{\parallel} + \vec{z}_{\perp} \\ \uparrow & & \uparrow \\ W & & W^{\perp} \end{array}$$

$$W \ni \vec{y}_{\parallel} - \vec{z}_{\parallel} = \vec{z}_{\perp} - \vec{y}_{\perp} \in W^{\perp}$$

$$\Rightarrow \vec{y}_{\parallel} = \vec{z}_{\parallel} \quad \& \quad \vec{y}_{\perp} = \vec{z}_{\perp} \quad \square$$



$$W = \text{span} \{ \vec{u}_1, \vec{u}_2 \}$$

$$L_1 = \text{span} \{ \vec{u}_1 \} \quad L_2 = \text{span} \{ \vec{u}_2 \}$$

$$\begin{aligned} \text{proj}_W \vec{y} &= \left(\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \right) \vec{u}_1 + \left(\frac{\vec{y} \cdot \vec{u}_2}{\vec{u}_2 \cdot \vec{u}_2} \right) \vec{u}_2 \\ &= \text{proj}_{L_1} \vec{y} + \text{proj}_{L_2} \vec{y} \end{aligned}$$

ex let $\vec{u}_1 = \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix}$, $\vec{u}_2 = \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$.

Let $W = \text{span} \{ \vec{u}_1, \vec{u}_2 \}$. Let $\vec{y} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$.

Find $\text{proj}_W \vec{y}$ and the decomposition

$$\vec{y} = \underbrace{\vec{y}_{\parallel}}_W + \underbrace{\vec{y}_{\perp}}_{W^{\perp}}$$

check $\{ \vec{u}_1, \vec{u}_2 \}$ is an orthogonal basis for W
 $\vec{u}_1 \cdot \vec{u}_2 = -2 + 2 = 0$.

$$\begin{aligned} \vec{y}_{\parallel} &= \text{proj}_W \vec{y} = \left(\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \right) \vec{u}_1 + \left(\frac{\vec{y} \cdot \vec{u}_2}{\vec{u}_2 \cdot \vec{u}_2} \right) \vec{u}_2 \\ &= \frac{2}{6} \vec{u}_1 + \frac{3}{5} \vec{u}_2 = \frac{2}{6} \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} + \frac{3}{5} \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix} \end{aligned}$$

$$\vec{y}_{\perp} = \vec{y} - \vec{y}_{\parallel} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \frac{2}{6} \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} - \frac{3}{5} \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$$

check $\vec{y}_{\perp} \cdot \vec{u}_1 = 0 = \vec{y}_{\perp} \cdot \vec{u}_2$

Theorem [Best Approximation]

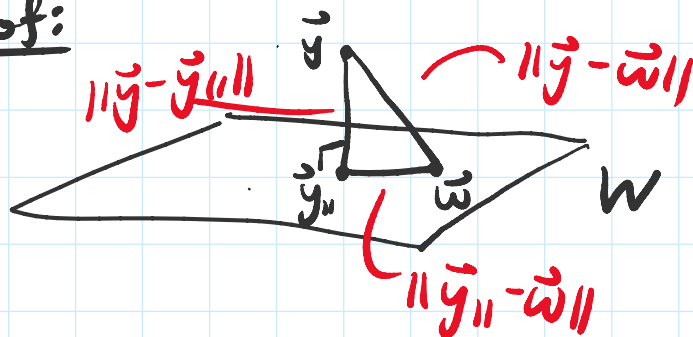
- Let W be a subspace of \mathbb{R}^n . Let $\vec{y} \in \mathbb{R}^n$.
- Then, the orthogonal projection of \vec{y} onto W ,
 $\vec{y}_{||} = \text{Proj}_W \vec{y}$, is the closest vector in

W to \vec{y} , i.e.

for any $\vec{w} \in W$ not equal to $\vec{y}_{||}$,

$$\|\vec{y} - \vec{y}_{||}\| < \|\vec{y} - \vec{w}\|.$$

proof:



Pythagorean thm:

$$\begin{aligned} \vec{y} - \vec{w} &= (\underbrace{\vec{y} - \vec{y}_{||}}_{\substack{= \vec{y}_{\perp} \\ \in W^{\perp}}}) + (\underbrace{\vec{y}_{||} - \vec{w}}_{\in W}) \end{aligned}$$

$$\begin{aligned} \Rightarrow \|\vec{y} - \vec{w}\|^2 &= \|\vec{y} - \vec{y}_{||}\|^2 + \underbrace{\|\vec{y}_{||} - \vec{w}\|^2}_{> 0} \\ &> \|\vec{y} - \vec{y}_{||}\|^2 \end{aligned}$$

□

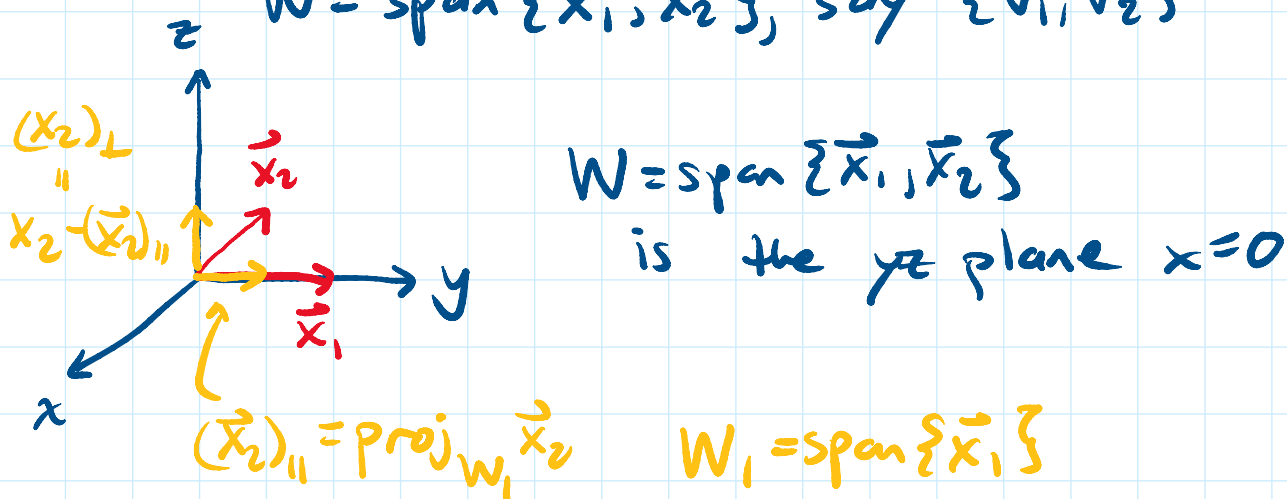
The Gram-Schmidt Process:

The Gram-Schmidt Process:

Given a basis for a subspace W of \mathbb{R}^n , how can we turn it into an orthogonal basis?

ex Consider $\vec{x}_1 = \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}$ and $\vec{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$.

Construct orthogonal basis for $W = \text{span}\{\vec{x}_1, \vec{x}_2\}$, say $\{\vec{v}_1, \vec{v}_2\}$



Take

$$\vec{v}_1 = \vec{x}_1. \quad W_1 = \text{span}\{\vec{x}_1\} = \text{span}\{\vec{v}_1\}.$$

$$\vec{v}_2 = \vec{x}_2 - \text{proj}_{W_1} \vec{x}_2$$

Claim: $\{\vec{v}_1, \vec{v}_2\}$ is a basis for W .

it suffices to show $\{\vec{v}_1, \vec{v}_2\}$ are lin. ind.
(since they're both in W)

$\vec{v}_1 \cdot \vec{v}_2 = 0$ $\vec{v}_1 = \vec{x}_1 \neq \vec{0}$

Since they're both in W_1

$$\vec{v}_1 \cdot \vec{v}_2 = 0 \quad \text{orthogonal} \quad \vec{v}_1 = \vec{x}_1 \neq 0$$

show $\vec{v}_2 \neq \vec{0}$. If $\vec{v}_2 = \vec{0}$, $\vec{x}_2 = \text{proj}_{W_1} \vec{x}_2$

$$\Rightarrow \vec{x}_2 \in W_1 \Rightarrow \vec{x}_2 \in \text{span}\{\vec{x}_1\} \\ \Rightarrow \text{contradiction.} \quad \square$$

$$\vec{v}_1 = \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}$$

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

$$= \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} - \frac{2}{4} \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Thm: [The Gram-Schmidt Process]

Let $\{\vec{x}_1, \dots, \vec{x}_p\}$ be a basis for a subspace W of \mathbb{R}^n .

Let:

$$\vec{v}_1 = \vec{x}_1 \quad W_1 = \text{span}\{\vec{x}_1\} = \text{span}\{\vec{v}_1\}$$

$$\vec{v}_2 = \vec{x}_2 - \text{proj}_{W_1} \vec{x}_2$$

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

$$W_2 = \text{span}\{\vec{x}_1, \vec{x}_2\} \\ = \text{span}\{\vec{v}_1, \vec{v}_2\}.$$

$$\vec{v}_3 = \vec{x}_3 - \text{proj}_{W_2} \vec{x}_3$$

$$\begin{aligned}
 \vec{v}_3 &= \vec{x}_3 - \text{proj}_{W_2} \vec{x}_3 \\
 &= \vec{x}_3 - \left(\frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \right) \vec{v}_1 - \left(\frac{\vec{x}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \right) \vec{v}_2
 \end{aligned}$$

$$W_k = \text{span} \{ \vec{v}_1, \dots, \vec{v}_k \}$$

$$\text{where } \vec{v}_k = \vec{x}_k - \text{proj}_{W_{k-1}} \vec{x}_k$$

$$\vec{v}_p = \vec{x}_p - \text{proj}_{W_{p-1}} \vec{x}_p$$

$$= \vec{x}_p - \left(\frac{\vec{x}_p \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \right) \vec{v}_1 - \dots - \left(\frac{\vec{x}_p \cdot \vec{v}_{p-1}}{\vec{v}_{p-1} \cdot \vec{v}_{p-1}} \right) \vec{v}_{p-1}$$

Then, $\{ \vec{v}_1, \dots, \vec{v}_p \}$ is an orthogonal basis for W w/ the additional property

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

$$\text{span} \{ \vec{x}_1, \vec{x}_2 \} = \text{span} \{ \vec{v}_1, \vec{v}_2 \}$$

↓

$$\text{span} \{ \vec{x}_1, \dots, \vec{x}_k \} = \text{span} \{ \vec{v}_1, \dots, \vec{v}_k \}$$

for $1 \leq k \leq p$.

□

