

defn: Let V be a vector space over \mathbb{F} (\mathbb{R}, \mathbb{C}). A function $f: V \times V \rightarrow \mathbb{F}$, denoted $f(u, v) \leftrightarrow \langle u, v \rangle$, and satisfies

- i) $\langle u, v \rangle = \overline{\langle v, u \rangle}$
- ii) $\langle \alpha u, v \rangle = \alpha \langle u, v \rangle$
- iii) $\langle u+v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$
- iv) $\langle u, u \rangle \geq 0$ $\forall u \in V$ and $\langle u, u \rangle = 0 \Rightarrow u = 0$.

- Implications:
- ① $\langle 0, 0 \rangle = 0$
 - ② $\langle u, \alpha v \rangle = \bar{\alpha} \langle u, v \rangle$
 - ③ $\langle u, v \rangle = 0$ for all $v \in V \Rightarrow u = 0$.

Examples: ① \mathbb{R}^3 $\langle x, y \rangle = y^T x = x_1 y_1 + x_2 y_2 + x_3 y_3$

② \mathbb{R}^n $\langle x, y \rangle = y^T x = \sum_{i=1}^n x_i y_i$

③ \mathbb{C}^n $\langle x, y \rangle = y^* x = \sum_{i=1}^n x_i \bar{y}_i$

④ $A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$

$\langle x, y \rangle = y^T A x$

$\langle (i, 0), (i, 0) \rangle$

\mathbb{R}^2 positive definite matrices

$= [y_1, y_2] \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

$(i, 0) \cdot \overline{(i, 0)}$

$= 2x_1 y_1 + y_1 x_2 + y_2 x_1 + 2x_2 y_2$

$\checkmark \langle x, y \rangle = \langle y, x \rangle$, $\langle \alpha x, y \rangle = y^T A (\alpha x) = \alpha y^T A x = \alpha \langle x, y \rangle$

$\langle x+y, z \rangle = z^T A (x+y) = z^T A x + z^T A y = \langle x, z \rangle + \langle y, z \rangle$

$\langle x, x \rangle \geq 0$ $\forall x$ and $\langle x, x \rangle = 0 \Rightarrow x = 0$

$\langle x, x \rangle = 2x_1^2 + 2x_1 x_2 + 2x_2^2 = \text{Try yourself}$

$\boxed{n^2 + n + 1}$

$$\textcircled{3} \quad \mathcal{L}([a, b], \mathbb{C}) \quad \langle f, g \rangle = \int_a^b f(x) \overline{g(x)} dx.$$

$$\stackrel{\mathbb{R}}{=} \rightarrow \langle f, g \rangle = \int_a^b f(x) \overline{g(x)} dx. \quad \checkmark$$

$$\boxed{\langle f, f \rangle \geq 0} \quad \langle f, f \rangle = \int_a^b \underbrace{|f(x)|^2}_{\geq 0} dx$$

$$\int_a^b h(x) dx \geq 0 \quad \text{and} \quad |h(x)| \geq 0 \quad \forall x$$

$$\textcircled{4} \quad M_{n \times n}(\mathbb{C}) \quad \langle A, B \rangle = \text{tr}(B^* A)$$

$$\begin{aligned} \text{tr}(B^* A) &= (B^* A)_{11} + (B^* A)_{22} + \dots + (B^* A)_{nn} \\ &= (B^*)_{11} a_{11} + (B^*)_{12} a_{21} + \dots + (B^*)_{21} a_{12} + (B^*)_{22} a_{22} + \dots \\ &\approx \overline{b_{11}} a_{11} + \overline{b_{21}} a_{21} + \dots + \overline{b_{n1}} a_{n1} + \overline{b_{12}} a_{12} + \dots \\ &\Leftrightarrow \text{Standard Inner Product / Dot Product.} \end{aligned}$$

defn: Length / Norm of $x \in V, \langle \cdot, \cdot \rangle \in \mathbb{R}$ Inner Product

$$\|x\| = \sqrt{\langle x, x \rangle} \quad \checkmark$$

Standard Inner Product / Dot product

$$x \in \mathbb{R}^n \quad \sqrt{x_1^2 + x_2^2 + \dots + x_n^2} = \|x\|$$

$$\boxed{\cos \theta = \frac{x \cdot y}{\|x\| \cdot \|y\|}}$$

Thm: Cauchy-Schwarz Inequality \checkmark

Let V be an Inner Product Space over \mathbb{F} . Then for every

$$x, y \in V \quad \boxed{|\langle x, y \rangle| \leq \|x\| \cdot \|y\|} \quad \text{The equality holds}$$

if and only if $y = \alpha x$ for some scalar $\alpha \in \mathbb{F}$.

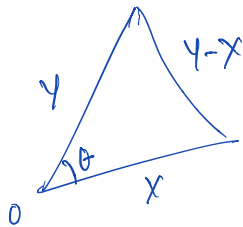
$$0 \leq \langle x - \alpha y, x - \alpha y \rangle = \langle x, x \rangle - \alpha \langle y, x \rangle - \bar{\alpha} \langle x, y \rangle + \alpha \bar{\alpha} \langle y, y \rangle$$

Take $\alpha = \frac{\langle x, y \rangle}{\|y\|^2}$, $\|y\| \neq 0$, $y \neq 0$.

$$0 \leq \|x\|^2 - \frac{\langle x, y \rangle \langle y, x \rangle}{\|y\|^2} - \frac{\langle y, x \rangle \langle x, y \rangle}{\|y\|^2} + \frac{\langle x, y \rangle \langle y, x \rangle}{\|y\|^2 \cdot \|y\|^2} \cdot \|y\|^2$$

$$\Rightarrow \|x\|^2 \geq \frac{|\langle x, y \rangle|^2}{\|y\|^2} \Rightarrow |\langle x, y \rangle|^2 \leq \|x\|^2 \cdot \|y\|^2$$

$$-1 \leq \frac{\langle x, y \rangle}{\|x\| \cdot \|y\|} \leq 1$$



Over Real Inner Product

$$\begin{aligned} \cos \theta &= \frac{\|x\|^2 + \|y\|^2 - \|y-x\|^2}{2 \|x\| \cdot \|y\|} \\ &= \frac{\|x\|^2 + \|y\|^2 - (\|x\|^2 + \|y\|^2 - 2\langle x, y \rangle)}{2 \|x\| \cdot \|y\|} \\ &= \frac{\langle x, y \rangle}{\|x\| \cdot \|y\|} \end{aligned}$$

$$\|y-x\|^2 = \langle y-x, y-x \rangle = \langle y, y \rangle - \langle x, y \rangle - \langle y, x \rangle + \langle x, x \rangle$$

Defn: Let V be an Inner Product space. Two vectors x and y are said to be orthogonal/perpendicular if $\langle x, y \rangle = 0$.

IMP: $0 \perp x$ for all $x \in V$. $x \perp y$ (Notation)

Defn: Let V be a vector space over \mathbb{F} . A fn. $f: V \rightarrow \mathbb{R}$ is called a 'Norm' if it satisfies the following Notation
 $\|\cdot\|$

i) $\|x\| \geq 0 \quad \forall x \in V$ and $\|x\| = 0 \Rightarrow x = 0$.

ii) $\|\alpha x\| = |\alpha| \cdot \|x\| \quad \forall \alpha \in \mathbb{F}, x \in V$.

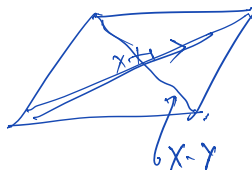
iii) $\|x+y\| \leq \|x\| + \|y\| \leftarrow$ Triangle Inequality.

$d(x, y) \Leftrightarrow$ distance between x & y
 $= \|x - y\|$

Thm: Let V be a Normed space (There is a well-defined Norm) over \mathbb{F} . Then this norm comes from an Inner Product if and only if

Parallelogram Law holds

$$\|x+y\|^2 + \|x-y\|^2 = 2(\|x\|^2 + \|y\|^2) \quad \checkmark$$



Example: $\mathbb{R}^2 \quad \|x\|_1 = \|(x_1, x_2)\|_1 := |x_1| + |x_2|$
 \leftarrow defn of norm.

$x = (1, 0), y = (0, 1)$

$x+y = (1, 1), x-y = (1, -1) \quad \text{LHS} = \|(1, 1)\|^2 + \|(1, -1)\|^2$
 $= (1+1)^2 + (1+1) = 2^2 + 2 = 8$

RHS = $2(\|(1, 0)\|^2 + \|(0, 1)\|^2)$
 $= 2(1^2 + 1^2) = \underline{\underline{4}}$

$$\begin{cases} x = (x_1, x_2, \dots, x_n) \\ y = (y_1, y_2, \dots, y_n) \end{cases} \quad y_i, x_i \in \{0, 1\} \pmod{2}$$

$$d(x, y) = |\{i \mid x_i \neq y_i\}| \text{ gives a distance in } \mathbb{Z}_2^n$$

So, if $n=4$ then for $x = (1, 1, 1, 1)$
 $x \cdot x = 0$. So, there are vectors which are self-orthogonal.

If we have real Inner product space

$$0 \leq \langle x - \alpha y, x - \alpha y \rangle = \langle x, x \rangle - \alpha \langle y, x \rangle - \alpha \langle x, y \rangle + \alpha^2 \langle y, y \rangle$$

$$= \langle x, x \rangle - 2\alpha \langle x, y \rangle + \alpha^2 \langle y, y \rangle.$$

A quadratic expression in α and is non-negative for each $\alpha \Rightarrow$ Discriminant ≤ 0

$$\Rightarrow 4 \langle x, y \rangle^2 - 4 \langle x, x \rangle \cdot \langle y, y \rangle \leq 0 \Rightarrow \text{Cauchy-Schwarz Inequality}$$

Polar Identity

$$4 \langle x, y \rangle = \|x+y\|^2 - \|x-y\|^2 + i(\|x+iy\|^2 - \|x-iy\|^2)$$

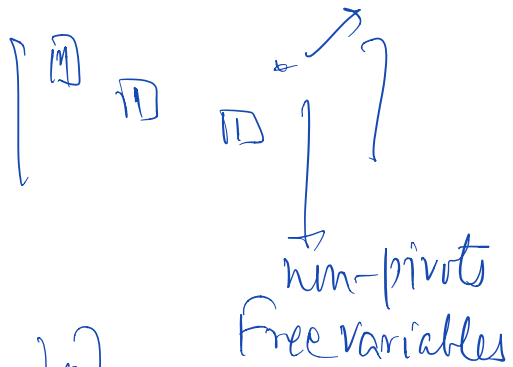
Complex Inner Product

$$\frac{\operatorname{Re} \langle x, y \rangle}{\|x\| \|y\|}$$

angle between x & y

Rank-Nullity Thm.

$AX=0$



$$\left[\begin{array}{cccccc|c} 1 & 2 & -3 & 0 & 0 & 3 & 4 & 0 \\ 0 & 0 & 0 & 1 & 0 & -4 & -3 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \end{array} \right] \rightsquigarrow$$

$x_5 = -x_7, x_4 = 4x_6 + 3x_7$
 $x_1 = -2x_2 + 3x_3 - 3x_6 - 4x_7$

$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_6 \\ x_7 \end{bmatrix} = \begin{bmatrix} \text{Free var} \\ \text{Free var} \\ \text{Free var} \\ \text{Free var} \\ \text{Free var} \end{bmatrix}$
 $= U_1 \rightarrow$ 1st Free variable = 1, rest as 0

$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} U_2 \rightarrow$ 2nd as 1, rest as 0

$y = y_2 U_1 + y_3 U_2 + y_4 U_3 + y_7 U_4$
 $\Rightarrow y \in \text{Null}(A)$

$xz=0$

$y = \sum_{i=1}^r y_i U_i + \sum_{j=r+1}^n y_j U_j$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

The i ,th compo correspond to first free variable

$$\begin{aligned} \langle y, \alpha w \rangle &= \overline{\langle \alpha w, y \rangle} = \overline{\alpha \langle w, y \rangle} \\ &= \overline{\alpha} \overline{\langle w, y \rangle} = \overline{\alpha} \langle y, w \rangle \end{aligned}$$

Defn: V as an IPS (Inner Product Space) over \mathbb{F} . Then x is said to be orthogonal/perpendicular to Y if $\langle x, Y \rangle = 0$.

Defn: Let S be a subset of V . Then the orthogonal complement of S in V , denoted S^\perp , equals

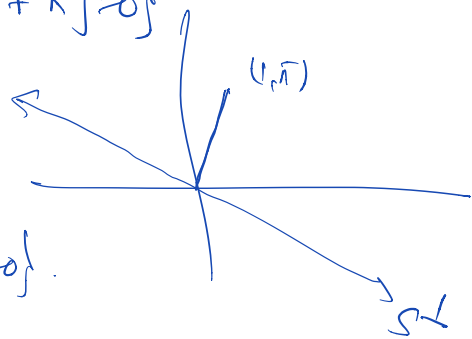
$$S^\perp = \{x \in V \mid \langle x, s \rangle = 0 \ \forall s \in S\}.$$

Example: \mathbb{R}^2 , $S = \{(1, \pi)\}$. $S^\perp = ?$

$$\begin{aligned} S^\perp &= \{(x, y) \in \mathbb{R}^2 \mid \langle (1, \pi), (x, y) \rangle = 0\} \\ &= \{(x, y) \in \mathbb{R}^2 \mid x + \pi y = 0\} \end{aligned}$$

\mathbb{R}^3 , $S = \{(1, 1, 1)\}$

$$S^\perp = \{(x, y, z) \in \mathbb{R}^3 \mid x + y + z = 0\}.$$



$S \subseteq V$, $S = \{x, y\}$, $S^\perp \stackrel{?}{=} x^\perp \cap y^\perp$

$(S(S)^\perp)^\perp$ and S^\perp ?

Thm: Let S be a subset of a v.s. V (IPS) over \mathbb{F} . Then S^\perp is a subspace of V over \mathbb{F} .

Defn: A set S is said to be orthogonal if any two vectors in S are orthogonal.

$$u_i \neq 0 \quad S = \{u_1, u_2, \dots, u_k\} \subseteq V \rightarrow (\text{IPS})$$

$$S \text{ is orthogonal} \iff \langle u_i, u_j \rangle = 0 \quad i \neq j, \text{ etc}$$

$$\text{orthonormal} \iff \text{orthogonal} \ \& \ \boxed{\|u_i\| = 1 \ \forall i}$$

$$\mathbb{R}^3 \rightarrow \{e_1, e_2, e_3\}$$

$$A = \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \end{bmatrix}$$

$$\boxed{A^T A = I = A A^T}$$

orthonormal

Thm: Let $S = \{u_1, u_2, \dots, u_k\}$ be an ~~orthogonal~~ orthonormal set in an IPS V over \mathbb{F} . Then S is l.l. subset of S over \mathbb{F} .

Pf: $\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_k u_k = 0$ unknowns are $\alpha_1, \dots, \alpha_k$.

$$0 = \langle 0, u_1 \rangle = \langle \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_k u_k, u_1 \rangle$$

$$= \alpha_1 \langle u_1, u_1 \rangle + \alpha_2 \underbrace{\langle u_2, u_1 \rangle}_{0} + \dots + \alpha_k \underbrace{\langle u_k, u_1 \rangle}_{0}$$

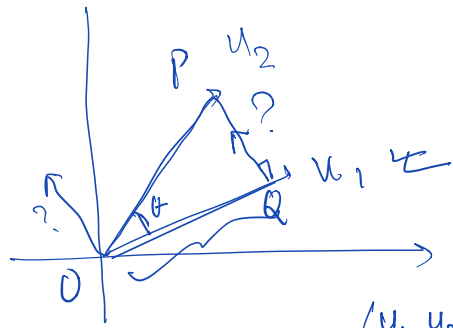
$$= \alpha_1 \|u_1\|^2$$

$$\Rightarrow \underline{\alpha_1 = 0}$$

Ques: Given a l.l. set, say $S = \{u_1, u_2, \dots, u_k\}$, can we get an orthonormal set, say $T = \{v_1, v_2, \dots\}$

such that $\boxed{LS(u_1, \dots, u_\ell) = LS(v_1, \dots, v_\ell), \quad 1 \leq \ell \leq k?}$

$v_1 = \frac{u_1}{\|u_1\|} \Rightarrow \|v_1\| = 1$.
 \leftarrow unit vector in the direction of u_1 .



$$\vec{OP} = u_2,$$

$$\cos \theta = \frac{\|\vec{OQ}\|}{\|\vec{OP}\|} \Rightarrow \|\vec{OQ}\| = \|\vec{OP}\| \cdot \cos \theta = \|u_2\| \cdot \frac{\langle u_1, u_2 \rangle}{\|u_1\| \cdot \|u_2\|}$$

$$= \frac{\langle u_1, u_2 \rangle}{\|u_1\|}$$

$\Rightarrow \vec{OQ} = \|\vec{OQ}\| \cdot$ unit vector in the direction of \vec{OQ}

$$= \frac{\langle u_1, u_2 \rangle}{\|u_1\|} \cdot \frac{u_1}{\|u_1\|} \quad \text{L2}$$

Projection of u_2 on $u_1 = \left\langle \frac{u_1}{\|u_1\|}, u_2 \right\rangle \frac{u_1}{\|u_1\|}$.

$$w_2 = u_2 - \vec{OQ} = u_2 - \left(\frac{\langle u_1, u_2 \rangle}{\|u_1\|} \right) \frac{u_1}{\|u_1\|}$$

$$\langle u_1, w_2 \rangle = \langle u_1, u_2 - \vec{OQ} \rangle = \langle u_1, u_2 - \alpha \frac{u_1}{\|u_1\|} \rangle$$

$$= \langle u_1, u_2 \rangle - \langle u_1, \alpha \frac{u_1}{\|u_1\|} \rangle,$$

$$= \langle u_1, u_2 \rangle - \alpha \langle u_1, \frac{u_1}{\|u_1\|} \rangle = \langle u_1, u_2 \rangle - \alpha \left(\|u_1\| \right)$$

$$\alpha = \left(\frac{1}{\|u_1\|} \right) \langle u_1, u_2 \rangle \quad \text{L2}$$

$$w_2 = u_2 - \alpha u_1$$

① $w_2 \perp v_1 = \frac{u_1}{\|u_1\|}$, $w_2 \neq 0$?

$$v_2 = \frac{w_2}{\|w_2\|}$$

$$w_3 = u_3 - \langle u_3, v_1 \rangle v_1 - \langle u_3, v_2 \rangle v_2$$

$$(1, 2, 3) - (1, 0, 0) - (0, 2, 0) = (0, 0, 3) \checkmark$$

$$w_3 \neq 0, \quad w_3 = 0 \Rightarrow u_3 - \alpha v_1 - \beta v_2 = 0 \\ \Rightarrow u_3 \in \text{LS}(v_1, v_2) = \text{LS}(u_1, u_2).$$

$$v_3 = \frac{w_3}{\|w_3\|} \quad \{v_1, v_2, v_3\} \text{ is an orthonormal set}$$

$$S = \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} \right\} \checkmark$$

$u_1 \quad u_2 \quad u_3 \quad u_4$

$$u_1 \rightarrow v_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \quad \|u_1\| = \sqrt{3} \checkmark$$

$$w_2 = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} - \left\langle \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{pmatrix} \right\rangle \begin{pmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} - \frac{1}{3} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 1/3 \\ 2/3 \\ 1 \end{pmatrix} = \begin{pmatrix} 2/3 \\ 4/3 \\ 2 \end{pmatrix} \checkmark$$

$$v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \checkmark$$

$$w_3 = (2, 1, 3) - \left\langle (2, 1, 3), \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right) \right\rangle \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right) \\ - \left\langle (2, 1, 3), \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right) \right\rangle \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right) \\ = (2, 1, 3) - \frac{6}{\sqrt{3}} \cdot \frac{1}{\sqrt{3}} (1, 1, 1) - \frac{1}{\sqrt{2}} \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right)$$

$$= (2, 1, 3) - (2, 2, 2) + \left(\frac{1}{2}, 0, -\frac{1}{2}\right) = \left(\frac{1}{2}, -1, \frac{1}{2}\right)$$

$$\cancel{u_3} \quad v_3 = \frac{w_3}{\|w_3\|} = \frac{1}{\sqrt{6}}(1, -2, 1) \quad \rightarrow \left(\frac{1}{2}\right) \underline{(1, -2, 1)} \quad \leftarrow$$

$\left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right), \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right), \frac{1}{\sqrt{6}}(1, -2, 1)$

$$\left(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}}\right)$$

$$\begin{aligned} (1, 2, 1) &= (1, 2, 1) - \underbrace{\langle (1, 2, 1), v_1 \rangle}_{\alpha} v_1 - \underbrace{\langle (1, 2, 1), v_2 \rangle}_{\beta} v_2 - \underbrace{\langle (1, 2, 1), v_3 \rangle}_{\gamma} v_3 \\ &= 0. \end{aligned}$$

$$(1, 2, 1) = \alpha v_1 + \beta v_2 + \gamma v_3 \quad \text{Find } \underline{\alpha, \beta, \gamma}$$

$$\begin{aligned} \langle (1, 2, 1), v_1 \rangle &= \langle \alpha v_1 + \beta v_2 + \gamma v_3, v_1 \rangle \\ &= \alpha \langle v_1, v_1 \rangle + \beta \langle v_2, v_1 \rangle + \gamma \langle v_3, v_1 \rangle \\ &= \alpha \|v_1\|^2 = \underline{\underline{\alpha}} \end{aligned}$$

QR-algorithm

$$[u_1 \ u_2 \ u_3 \ u_4] = \begin{bmatrix} \checkmark v_1 & v_2 & v_3 \end{bmatrix} \begin{bmatrix} \checkmark a_{11} & a_{12} & a_{13} & ? \\ 0 & a_{22} & a_{23} & ? \\ 0 & & a_{33} & ? \end{bmatrix}$$

Orthogonal

Know: $LS(u_1) = LS(v_1)$

$$u_2 = a_{12}v_1 + a_{22}v_2$$

$$\underline{u_2} \in \underline{LS(v_1, v_2)} = \underline{LS(u_1, u_2)}$$

$$\begin{aligned} &\downarrow \\ &a_{11}v_1 + 0 \cdot v_2 + 0 \cdot v_3 \\ &= a_{11}v_1 = u_1 \\ &\text{for some choice} \\ &\text{of } a_{11}. \end{aligned}$$

$$x \longmapsto Ax$$

$$\begin{aligned} \|x\|^2 &= \langle x, x \rangle \\ &= x^T x \end{aligned}$$

$$\begin{aligned} \|Ax\|^2 &\rightarrow (Ax)^T (Ax) \\ &= x^T \underbrace{(A^T A)}_{\substack{|| \\ \underline{C} \\ ||}} x \end{aligned}$$

$$\{u_1, u_2, \dots, u_n\} \subset \mathbb{R}^n \iff \{u_2, u_3, u_1, u_4, \dots\}$$

$$\begin{aligned} \text{Proj}_V(u) &= \left\langle u, \frac{v}{\|v\|} \right\rangle \frac{v}{\|v\|}, \quad f = \frac{v}{\|v\|} \\ &= \langle u, f \rangle f \quad \text{f is a unit vector.} \end{aligned}$$

\mathbb{R}^n , f is a unit vector in \mathbb{R}^n

$$\begin{aligned} \text{The projection of } u \text{ on } f &\iff \text{Proj}_f(u) = \langle u, f \rangle f \\ &= (f^T u) f \\ &= f (f^T u) \\ &= (f f^T) u \end{aligned}$$

The matrix of the projection onto the line $f \iff \mathcal{L}(f)$ is given by $A = \underline{f f^T}$.

$$A^2 = (f f^T)(f f^T) = f (f^T f) f^T = f \cdot (1) \cdot f^T = f f^T = A.$$

$$\underline{A^T = A.}$$

$$f^T = \underline{\underline{(v_3, v_3, v_3)}}$$

$$A \rightarrow \begin{matrix} 3 \times 3 \\ \left[\begin{array}{c} v_3 \\ v_3 \\ v_3 \end{array} \right] \left[\begin{array}{ccc} v_3 & v_3 & v_3 \end{array} \right] \end{matrix}$$

$$\text{Proj}_f(u) = \underline{\underline{A u}}$$

Remember so that what we did during Gram-Schmidt process made sense.

← 1st the 2nd

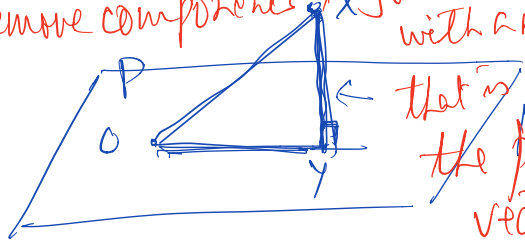
Thm! Let H be a finite dimensional vector subspace of an IPS V over \mathbb{F} . If $\{u_1, u_2, \dots, u_k\}$ is an orthonormal basis of H then for any $x \in V$

① $y = \langle x, u_1 \rangle u_1 + \langle x, u_2 \rangle u_2 + \dots + \langle x, u_k \rangle u_k \in H$
 is the ONLY closest point to x in H

② $x - y$ is orthogonal to $H \Leftrightarrow x - y \in H^\perp$

whenever you remove components of x you are left out with a new vector that is \perp to the previous vectors.

$\text{Proj}_H(x) \leftrightarrow$ projection of x on the plane H .



Pf: Take $\alpha_i = \langle x, u_i \rangle$ for $1 \leq i \leq k$.

Then $\langle x - y, u_1 \rangle = \langle x, u_1 \rangle - \langle y, u_1 \rangle$
 $= \langle x, u_1 \rangle - \langle \sum_{i=1}^k \alpha_i u_i, u_1 \rangle$
 $= \langle x, u_1 \rangle - \sum_{i=1}^k \alpha_i \langle u_i, u_1 \rangle$
 $= \langle x, u_1 \rangle - \alpha_1$
 $= \langle x, u_1 \rangle - \langle x, u_1 \rangle = \underline{\underline{0}}$

$$\|x\|^2 = \|y + x - y\|^2 = \|y\|^2 + \|x - y\|^2$$

$$\Rightarrow \|x\|^2 \geq \underbrace{\|y\|^2 = |\langle x, u_1 \rangle|^2 + |\langle x, u_2 \rangle|^2 + \dots + |\langle x, u_k \rangle|^2}_{\text{as } y \in H}$$

Bessel's Inequality.

$$\|\alpha u_1 + \beta u_2\|^2 \stackrel{L}{=} \langle \alpha u_1 + \beta u_2, \alpha u_1 + \beta u_2 \rangle$$

$$= \alpha \langle u_1, \alpha u_1 \rangle + \alpha \langle u_1, \beta u_2 \rangle + \beta \langle u_2, \alpha u_1 \rangle + \beta \langle u_2, \beta u_2 \rangle$$

$$= |\alpha|^2 \langle u_1, u_1 \rangle + \alpha \bar{\beta} \langle u_1, u_2 \rangle + \beta \bar{\alpha} \langle u_2, u_1 \rangle + |\beta|^2 \langle u_2, u_2 \rangle$$

$$= |\alpha|^2 + |\beta|^2.$$

Bessel's Inequality: H a f.d. subspace of an IPS V with orthonormal basis $\{u_1, u_2, \dots, u_k\}$. Let $x \in V$.

$$\text{Then } \|x\|^2 \geq \sum_{i=1}^k |\langle x, u_i \rangle|^2. \quad L$$

Equality holds if and only if $\underline{\underline{x \in H}}$.

$H \leftrightarrow \{u_1, u_2, \dots, u_k\}$ orthonormal.

$$\text{If } x \in H \text{ then } \|x\|^2 = \sum_{i=1}^k |\langle x, u_i \rangle|^2$$

Pythagoras Thm.

Thm: Let $A_{m \times n}$ with real entries and $b \in \mathbb{R}^m$. Then every solution of the system $A^T A x = A^T b$ is a least square solution of the system $Ax = b$.

Conversely, every least square solution of the system $Ax = b$ is a solution of the system $A^T A x = A^T b$.

$$\boxed{Ax = b}$$

$$\|err\| = \|Ax - b\|$$

Find $x_0 \in \mathbb{R}^n$ such that $\|Ax_0 - b\|$

$$= \min_{x \in \mathbb{R}^n} \|Ax - b\|$$

Pf:

Let x_0 be a least square solution of the system $Ax = b$.

$$\Rightarrow \|Ax_0 - b\| = \min_{x \in \mathbb{R}^n} \|Ax - b\|$$

Need to show:

$$A^T A x_0 = A^T b.$$

$$\text{To show: } A^T (Ax_0 - b) = 0. \quad \checkmark$$

$$\|Ax - b\|^2 = \|Ax - Ax_0 + Ax_0 - b\|^2$$

$$\stackrel{?}{=} \|Ax - Ax_0\|^2 + \|Ax_0 - b\|^2 + 2 \langle Ax - Ax_0, Ax_0 - b \rangle$$

$$\geq \|Ax_0 - b\|^2 \text{ for all } x \in \mathbb{R}^n.$$

$$\underline{\text{Minimize}} \min_{x \in \mathbb{R}^n} \|Ax - b\|^2$$

$$= \min_{x \in \mathbb{R}^n} (Ax - b)^T (Ax - b)$$

$$= \min_{x \in \mathbb{R}^n} (x^T A^T - b^T) (Ax - b)$$

$$= \min_{x \in \mathbb{R}^n} \left(\underbrace{x^T A^T A x}_{\leftarrow} - \underbrace{x^T A^T b}_{\leftarrow} - \underbrace{b^T A x}_{\leftarrow} + b^T b \right)$$

$$\frac{d}{dx} \rightarrow \left[2A^T A x - 2A^T b \right] = 0 \rightarrow \underline{A^T A x = A^T b.}$$

$$f: \mathbb{R}^n \rightarrow \mathbb{R} \rightarrow \nabla f = \left(\frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right)$$

$$= \underbrace{x_1 (A^T A x)_1}_{\leftarrow} + \underbrace{x_2 (A^T A x)_2}_{\leftarrow} + \dots + \underbrace{x_n (A^T A x)_n}_{\leftarrow}$$

$$\begin{pmatrix} x & y \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \underbrace{a_{11} x^2 + 2a_{12} xy + a_{22} y^2}_{\leftarrow}$$

$$\frac{\partial}{\partial x} \rightarrow 2a_{11}x + 2a_{12}y$$

$$\frac{\partial}{\partial y} \rightarrow 2a_{12}x + 2a_{22}y$$

$$\begin{pmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \end{pmatrix} = 2 \begin{pmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

$AX = b \Leftrightarrow$ If No soln then look for the Best possible solution. \Updownarrow

soln of $A^T A x = A^T b.$

pseudo-inverse / Moore-Penrose Inverse $\Leftrightarrow (G)$

$$AGA = A, \quad GAG = G, \quad (AG)^T = AG$$

$$(GA)^T = GA.$$

Thm: Let V be a f.d. IPS and W a proper subspace of V . Then $V = \boxed{W \oplus W^\perp}$ \Leftarrow
 \uparrow direct sum

(\Leftrightarrow each element of V can be uniquely expressed as the sum of an element from W and an element from W^\perp)

$$\mathbb{R}^3, \quad W_1 = \{ (x, y, z) \in \mathbb{R}^3 \mid x+y=z \}$$

$$W_2 = \{ (x, y, z) \in \mathbb{R}^3 \mid x+y=0 \}$$

$$W_1 \rightarrow \{ (1, 0, 1), (0, 1, 1) \} \quad W_2 \leftrightarrow \{ (1, -1, 0), (0, 0, 1) \}$$

$$\{ \underbrace{(1, 0, 1), (0, 1, 1)}, \underbrace{(1, -1, 0), (0, 0, 1)} \} \Leftrightarrow \underline{\underline{S}}$$

W_1 is a plane, W_2 is also a plane in \mathbb{R}^3

$W_1 \cap W_2$ is a line in \mathbb{R}^3

① $\Rightarrow \{0\} \subsetneq W_1 \cap W_2$. $W_1 \cap W_2$ contains a non-zero vector.

② $\mathbb{R}^3 = \text{LS} \{ (1, 0, 1), (0, 1, 1), (1, -1, 0), (0, 0, 1) \}$.
 $= W_1 + W_2$. \Leftarrow

$$V = W_1 + W_2.$$

Any $x \in V$ can be written as

$$x = w_1 + w_2, \quad w_1 \in W_1 \\ \& w_2 \in W_2.$$

$$W \cap W^\perp = \underline{\underline{\{0\}}}.$$

$$W \subseteq V = W \oplus \underline{\underline{W^\perp}}?$$

Let $\{u_1, u_2, \dots, u_k\}$ be a basis of W .

Apply Gram-Schmidt to get $\{v_1, v_2, \dots, v_k\}$
as a basis of W . orthonormal set.

Extend $\{v_1, v_2, \dots, v_k\}$ to get a basis of V .

Apply Gram-Schmidt again:

$$\{ \underbrace{v_1, v_2, \dots, v_k}_W, \underbrace{v_{k+1}, v_{k+2}, \dots, v_n} \}$$

is an orthonormal basis of V .

$$W = \text{LS}(v_1, v_2, \dots, v_k). \quad ? \quad \underline{\underline{W^\perp = \text{LS}(v_{k+1}, \dots, v_n)}}$$

$$\text{col} \cap \text{col}^\perp = \{0\}.$$

$$\begin{aligned} v \in V \quad v &= \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n \\ &= \underbrace{(\alpha_1 v_1 + \dots + \alpha_k v_k)}_w + \underbrace{(\alpha_{k+1} v_{k+1} + \dots + \alpha_n v_n)}_{w^\perp}. \end{aligned}$$

$AX=b$ Know: $b \notin \text{Col}(A)$. \perp $A_{m \times n}$

$$\text{err} = AX - b, \quad x \in \mathbb{R}^n$$

$b \in V, \quad b \notin \text{Col}(A) \Rightarrow \exists$ a unique $y \in \text{Col}(A)$
 s.t. $\boxed{b = y + v}, \quad \underline{v} \in \text{Col}(A)^\perp \perp$

$$\underline{\text{Null}(A)} = \text{Row}(A)^\perp \quad \left(\underline{\text{Row}(A^T)} \right)^\perp \perp$$

$$\underline{AX=0} \perp$$

Claim! The vector $y \in \text{Col}(A)$ is the one that we are looking for. \Downarrow
 $\exists x_0 \in \mathbb{R}^n$ s.t. $y = \underline{Ax_0}$.

$$\begin{aligned} A^T A x_0 &= A^T y = A^T (b - v) = A^T b - A^T v \\ &= A^T b - 0 = A^T b. \end{aligned}$$

Need to show $\|AX_0 - b\| = \min_X \{ \|AX - b\| \}$.

$$\begin{aligned} \|AX - b\|^2 &= \|AX - y + y - b\|^2 = \|AX - y\|^2 + \|y - b\|^2 \\ &\quad + 2\langle AX - y, y - b \rangle, \\ &= \|AX - y\|^2 + \|y - b\|^2 \\ &\geq \underline{\underline{\|y - b\|^2}}. \end{aligned}$$

Conversely, if x_0 satisfies $A^T A x_0 = A^T b$ then
 $\|AX_0 - b\| = \min_X \{ \|AX - b\| \}$. \leftarrow yourself.

Q. Plot the curve ~~is~~

$$\left\{ (x, y, z) \in \mathbb{R}^3 \mid (x, y, z) \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = 10 \right\}.$$

$$\boxed{2x^2 + 2y^2 + 2z^2 + 2xy + 2yz + 2xz = 10}$$

$$A = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}$$

$$A \begin{bmatrix} 1/\sqrt{6} \\ 1/\sqrt{6} \\ -2/\sqrt{6} \end{bmatrix} = \begin{bmatrix} 1/\sqrt{6} \\ 1/\sqrt{6} \\ -2/\sqrt{6} \end{bmatrix}$$

$$A \begin{bmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix} = 4 \begin{bmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix}, \quad A \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \\ 0 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}$$

$$U = \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \end{bmatrix}$$

$$UU^T = I \Rightarrow \underline{\underline{U^{-1} = U^T}}$$

$$A\tilde{U} = [A U(:,1), A U(:,2), A U(:,3)]$$

$$= [4 U(:,1), U(:,2), U(:,3)] = U \begin{bmatrix} 4 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} ?$$

$$\underline{A} = U \text{Diag}(4,1,1) U^{-1} = U \text{Diag}(4,1,1) U^T.$$

$$x^T A x = 10 \Leftrightarrow x^T (U \text{Diag}(4,1,1) U^T) x = 10$$

$$\text{Define } \underline{y} = U^T x = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} \Leftrightarrow [y_1 \ y_2 \ y_3] \begin{bmatrix} 4 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = 10$$

$$\Leftrightarrow 4y_1^2 + y_2^2 + y_3^2 = 10 \Leftrightarrow \frac{y_1^2}{\frac{10}{4}} + \frac{y_2^2}{10} + \frac{y_3^2}{10} = 1$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = U^T x = \begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} & -\frac{2}{\sqrt{6}} \end{pmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{pmatrix} \frac{x+y+z}{\sqrt{3}} \\ \frac{x-y}{\sqrt{2}} \\ \frac{x+y-2z}{\sqrt{6}} \end{pmatrix}$$

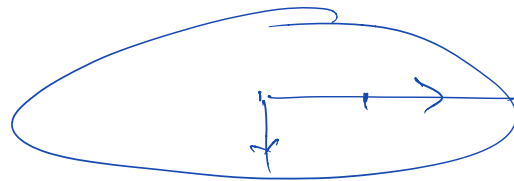
Principal Axes

$$x+y+z=0$$

$$x-y=0, \quad x+y-2z=0$$

$A_{n \times n}$

$$\{Ax : x \in \mathbb{R}^n\} \Leftrightarrow$$



$$A_{n \times n} \in M_{n \times n}(\mathbb{C})$$

\mathbb{C}^n over \mathbb{C} .

Eigenvalues & Eigenvectors