

Orthogonality

Recall that we define u and v to be orthogonal if $\langle u, v \rangle = 0$.

Defn If $W \subseteq V$ is a subspace, we define

$$W^\perp = \{ u \in V : \langle u, w \rangle = 0 \text{ for all } w \in W \}$$

the orthogonal complement to W .

Lemma If W is a subspace of V , then so is W^\perp .

proof (a) Let $u, v \in W^\perp$.

Then $\langle v, w \rangle = 0$ and $\langle u, w \rangle = 0$ for all $w \in W$

Thus $\langle u+v, w \rangle = \langle u, w \rangle + \langle v, w \rangle = 0 + 0 = 0$ for all $w \in W$.

so $u+v \in W^\perp$

← additivity

(b) Suppose $u \in W^\perp$ and $c \in \mathbb{R}$.

Then $\langle u, w \rangle = 0$ for all $w \in W$

← homogeneity

Thus $\langle cu, w \rangle = c \langle u, w \rangle = c \cdot 0 = 0$ for all $w \in W$.

So $cu \in W^\perp$ □

Examples (1) In \mathbb{R}^3 , orthogonal \equiv perpendicular

So if W is a plane, then W^\perp is a line, and vice versa.

(2) $\{0\}^\perp = V$ and $V^\perp = \{0\}$

(3) In \mathbb{R}^4 , let $W = \{ (a, b, c, d) \in \mathbb{R}^4 \mid a=0 \text{ and } c=0 \}$

Then $W^\perp = \{ (a, b, c, d) \in \mathbb{R}^4 \mid b=0 \text{ and } d=0 \}$

proof, Let $A = \{ (a, b, c, d) \in \mathbb{R}^4 \mid b=0 \text{ and } d=0 \}$

$A \subseteq W^\perp$ Suppose $u \in A$. Then $u = (u_1, 0, u_3, 0)$

Then let $w \in W$. Then $w = (0, w_2, 0, w_4)$

So $\langle u, w \rangle = 0$

Thus $u \in W^\perp$

$W^\perp \subseteq A$ Let $u \in W^\perp$. Then for all $w \in W$, $\langle u, w \rangle = 0$.

Write $u = (u_1, u_2, u_3, u_4)$

Pick some $w = (0, \frac{0}{w_2}, 0, \frac{1}{w_4}) \in W$

Then $0 = \langle u, w \rangle = \cancel{u_2 w_2} + \cancel{u_4 w_4} = u_4$

Also, $(0, 1, 0, 0) \in W$. so $0 = \langle u, (0, 1, 0, 0) \rangle = u_2$

Hence $u \in A \quad \square$

Example Let $V = P_2$. Consider $x^2 + 2x + 1 \in P_2$.

Then $x^2 + 2x + 1$ is orthogonal to $x^2 - x + 1$

Example Let A be any $n \times m$ matrix. Suppose $u \in \text{Nullspace}(A)$

$A u = 0$ Let v_1, \dots, v_n be the rows of A .

$$\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} \begin{pmatrix} u \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

Then $v_i \cdot u = 0$ for every $1 \leq i \leq n$

Then if $\sum_{i=1}^n c_i v_i$ is any element of the row space of A ,

$$\begin{aligned} \text{Then } u \cdot \left(\sum_{i=1}^n c_i v_i \right) &= \sum_{i=1}^n (u \cdot (c_i v_i)) = \sum_{i=1}^n c_i (u \cdot v_i) \\ &= \sum_{i=1}^n c_i \cdot 0 = 0 \end{aligned}$$

Thus u is orthogonal to every vector in the row space,

so $u \in (\text{Row space}(A))^\perp$

Hence $\text{Nullspace}(A) \subseteq (\text{Row space}(A))^\perp$!!

Claim $(\text{Row space}(A))^\perp \subseteq \text{Nullspace}(A)$

proof Suppose $u \in (\text{Row space}(A))^\perp$

Thus u is orthogonal to every vector in the row space.

$$\text{So } Au = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} \begin{pmatrix} u \end{pmatrix} = \begin{pmatrix} v_1 \cdot u \\ v_2 \cdot u \\ \vdots \\ v_n \cdot u \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

Thus $u \in \text{Nullspace}(A) \quad \square$

Conclusion ^(Thm) $(\text{Row space}(A))^\perp = \text{Nullspace}(A)$

[Also, since $\text{Column space}(A^T) = \text{Row space}(A)$
we get $(\text{Column space}(A^T))^\perp = (\text{Row space}(A))^\perp = \text{Nullspace}(A)$]

Lemma Let V be an n dimensional vector space, and let W be a subspace.

(a) $(W^\perp)^\perp = W$. (b) Thus $W = U^\perp \Leftrightarrow U = W^\perp$

(c) Also $\dim W^\perp = n - \dim W$

proof (c) Let v_1, \dots, v_r be a basis for W .

Construct the $n \times n$ matrix A with $v_1, \dots, v_r, 0, \dots, 0$ as rows.

Then $\text{Rowspace}(A) = W$, so $W^\perp = (\text{Rowspace}(A))^\perp = \text{Nullspace}(A)$

So by the Rank-Nullity Theorem:

$$\dim(\text{Rowspace}(A)) + \dim(\text{Nullspace}(A)) = n$$
$$\Rightarrow \dim W + \dim W^\perp = n$$

(a) Next we see that $(W^\perp)^\perp \supseteq W$

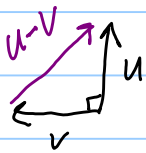
(Why? Let $u \in W$. Then u is orthogonal to every $v \in W^\perp$.
Hence $u \in (W^\perp)^\perp$)

$$\text{Next } \dim (W^\perp)^\perp = n - \dim(W^\perp) = n - (n - \dim(W)) = \dim W$$

Therefore, since W is a subspace of $(W^\perp)^\perp$ and they have the same dimension, they must be equal.

(b) $W = W^\perp \Rightarrow W^\perp = (W^\perp)^\perp \Rightarrow W^\perp = W$ by part a \square

Pythagorean Theorem If u and v are orthogonal, then $\|u\|^2 + \|v\|^2 = \|u-v\|^2$



proof Suppose $\langle u, v \rangle = 0$

$$\text{Then } \|u-v\|^2 = \langle u-v, u-v \rangle$$

$$= \langle u, u-v \rangle - \langle v, u-v \rangle$$

$$= \langle u, u \rangle - \underbrace{\langle u, v \rangle - \langle v, u \rangle}_{\text{both } 0} + \langle v, v \rangle$$

$$= \langle u, u \rangle + \langle v, v \rangle$$

$$= \|u\|^2 + \|v\|^2 \quad \square$$

Orthonormal Basis

Definition Suppose $\mathcal{B} = \{v_1, \dots, v_n\}$ is a basis for V

(A) If $\langle v_i, v_j \rangle = 0$ for every $1 \leq i \leq n$ and $1 \leq j \leq n$ where $i \neq j$, then we say $\{v_1, \dots, v_n\}$ is pairwise orthogonal

In this case, we say \mathcal{B} is an orthogonal basis.

(B) If \mathcal{B} is an orthogonal basis, and if $\|v_i\| = 1$ for every $1 \leq i \leq n$, then we say \mathcal{B} is an orthonormal basis

Example In \mathbb{R}^n , the standard basis is orthonormal

Example $\{(1,1), (1,-1)\}$ is an orthogonal basis for \mathbb{R}^2 .

Then $\left\{ \frac{(1,1)}{\sqrt{2}}, \frac{(1,-1)}{\sqrt{2}} \right\}$ is an orthonormal basis for \mathbb{R}^2

↑
This is the normalization of this

Proposition Suppose V is a finite-dimensional real inner product space, and that $\{v_1, \dots, v_n\}$ is an orthogonal basis. Let $u \in V$.

$$\text{Then } u = \frac{\langle u, v_1 \rangle}{\|v_1\|^2} v_1 + \frac{\langle u, v_2 \rangle}{\|v_2\|^2} v_2 + \dots + \frac{\langle u, v_n \rangle}{\|v_n\|^2} v_n$$

Thus if $\{v_1, \dots, v_n\}$ is an orthonormal basis,

$$u = \langle u, v_1 \rangle v_1 + \langle u, v_2 \rangle v_2 + \dots + \langle u, v_n \rangle v_n$$

Proof Let $u \in V$. Since $\{v_1, \dots, v_n\}$ is a basis, we can write

$$u = c_1 v_1 + c_2 v_2 + \dots + c_n v_n \text{ for some } c_1, \dots, c_n \in \mathbb{R}.$$

Let $1 \leq i \leq n$. Then

$$\begin{aligned} \langle u, v_i \rangle &= \langle c_1 v_1 + c_2 v_2 + \dots + c_n v_n, v_i \rangle \\ &= c_1 \underbrace{\langle v_1, v_i \rangle}_{=0} + c_2 \underbrace{\langle v_2, v_i \rangle}_{=0} + \dots + c_i \underbrace{\langle v_i, v_i \rangle}_{\neq 0} + \dots + c_n \underbrace{\langle v_n, v_i \rangle}_{=0} \end{aligned}$$

$$= c_i \langle v_i, v_i \rangle$$

$$= c_i \|v_i\|^2$$

$$\text{Thus } c_i = \frac{\langle u, v_i \rangle}{\|v_i\|^2} \text{ as desired } \square$$

