

Lecture Orthogonal Complements

Overview of Lecture

- In this lecture, we introduce the concept of the **orthogonal complement** of a subset in an inner product space. Given a subset S of V , the orthogonal complement S^\perp consists of all vectors in V that are orthogonal to every vector in S .
- We will explore basic properties of orthogonal complements, including how they form subspaces and simple examples in \mathbb{R}^3 .
- A major result is the **Orthogonal Decomposition Theorem (Theorem 6.6)**, which states that any vector in V can be uniquely written as the sum of a vector in W and a vector in W^\perp , where W is a subspace of V . The component in W is called the **orthogonal projection** of the vector onto W .
- Understanding orthogonal projections will be important for later applications, such as solving **least squares problems** in Section 6.3.

Definition.

Let S be a nonempty subset of an inner product space V .

The **orthogonal complement** of S , denoted S^\perp , is the set of all vectors in V that are orthogonal to every vector in S . That is,

$$S^\perp = \{x \in V : \langle x, y \rangle = 0 \text{ for all } y \in S\}.$$

Exercise 1. Show S^\perp is a subspace of V for any subset S of V .

ANS: Let $x, z \in S^\perp$, $c \in F$. we want to show $cxtz \in S^\perp$

Note $\langle cxtz, y \rangle = c \langle x, y \rangle + \langle z, y \rangle = c \cdot 0 + 0 = 0$, $\forall y$ yes.

Also $0 \in S^\perp$. Thus S^\perp is a subspace of V .

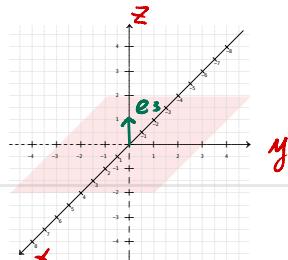
Example 2. Verify that $\{0\}^\perp = V$ and $V^\perp = \{0\}$ for any inner product space V .

ANS: By def, $\{0\}^\perp = \{x \in V \mid \langle x, y \rangle = 0 \text{ for all } y \in \{0\}\}$.

It's straightforward to see $V \subseteq \{0\}^\perp$ Thus $\{0\}^\perp = V$.

We leave $V^\perp = \{0\}$ as exercise.

Example 3. If $V = \mathbb{R}^3$ and $S = \{e_3\}$, then S^\perp equals the xy -plane.



$$S^\perp = \{x \in \mathbb{R}^3 \mid \langle x, e_3 \rangle = 0\}$$

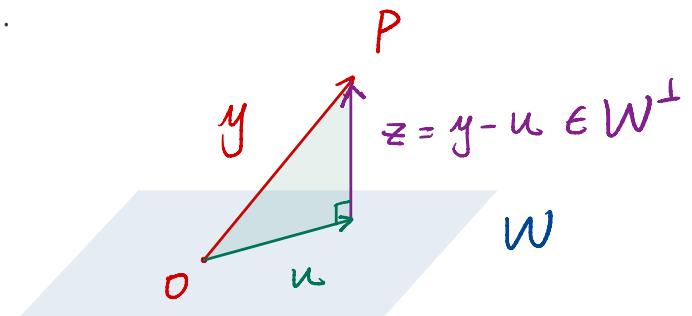
$$\langle e_1, e_3 \rangle = \langle e_2, e_3 \rangle = 0.$$

Discussion: Distance from a Point to a Subspace

Suppose we are given a point P in \mathbb{R}^3 and a plane W through the origin.

We want to find the **distance from P to the plane W** .

Let y be the vector from the origin O to the point P . The goal is to find the vector $u \in W$ that is **closest** to y . This means we want to minimize the distance $\|y - u\|$.



From the geometry, the difference $z = y - u$ is the vector from the point on the plane to the point P , and it is **perpendicular to the plane W** .

Therefore, $z \in W^\perp$, and we obtain the decomposition:

$$y = u + z, \quad \text{with } u \in W, z \in W^\perp.$$

This gives both a **geometric** and **algebraic** motivation for studying orthogonal complements. Finding the vector in W closest to y is the same as projecting y orthogonally onto W .

(See the above figure a visual representation of this idea.)

- To find the point in a subspace W closest to a vector y , we decompose y into a sum of vectors from W and W^\perp . The following theorem gives a general method for doing this when W is finite-dimensional.

Theorem 6.6. (Orthogonal Decomposition)

Let W be a finite-dimensional subspace of an inner product space V , and let $y \in V$. Then there exist unique vectors $u \in W$ and $z \in W^\perp$ such that $y = u + z$.

Furthermore, if $\{v_1, v_2, \dots, v_k\}$ is an orthonormal basis for W , then

$$u = \sum_{i=1}^k \langle y, v_i \rangle v_i.$$

Proof: Let $\{v_1, \dots, v_k\}$ be an orthonormal basis for W .

and let $u = \sum_{i=1}^k \langle y, v_i \rangle v_i$ as give above.

Consider $z = y - u$.

Then $u \in W$ since u is a linear combination of basis element for W .

7. Let β be a basis for a subspace W of an inner product space V , and let $z \in V$. Prove that $z \in W^\perp$ if and only if $\langle z, v \rangle = 0$ for every $v \in \beta$.

Next, we show $z \in W^\perp$. By HW #5 Question 1, we show z is orthogonal to each $v_j \in \{v_1, \dots, v_k\}$

Then

$$\begin{aligned}\langle z, v_j \rangle &= \left\langle y - \sum_{i=1}^k \langle y, v_i \rangle v_i, v_j \right\rangle \\ &= \langle y, v_j \rangle - \sum_{i=1}^k \langle \langle y, v_i \rangle v_i, v_j \rangle \xrightarrow{\begin{cases} 0, & i \neq j \\ 1, & i=j \end{cases}} \\ &= \langle y, v_j \rangle - \sum_{i=1}^k \langle y, v_i \rangle \underbrace{\langle v_i, v_j \rangle}_{=0} \\ &= \langle y, v_j \rangle - \langle y, v_j \rangle \\ &= 0, \quad \forall v_j \in \{v_1, \dots, v_k\}.\end{aligned}$$

Thus $z \perp v_j, j=1, \dots, k$

Finally, we show the uniqueness of the decomposition.

If $y = u + z = u' + z'$, where, $u' \in W, z' \in W^\perp$.

Then $u - u' \in W \cap W^\perp$, and $z - z' \in W \cap W^\perp$

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

Thus $u = u'$ and $z = z'$ and the decomposition of $y = u + z$ is unique.

Corollary.

In the notation of **Theorem 6.6**, the vector u is the unique vector in W that is "closest" to y ; that is, for any $x \in W$,

$$\|y - x\| \geq \|y - u\|,$$

and this inequality is an equality if and only if $x = u$.

Proof: Let $y = u + z$ as in Thm 6.6. where $u \in W$, $z \in W^\perp$

Let $x \in W$. then

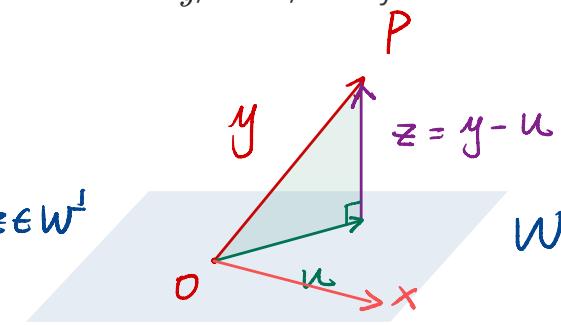
$$\begin{aligned} \|y - x\|^2 &= \|u + z - x\|^2 = \|u - x + z\|^2 = \|u - x\|^2 + \|z\|^2 + 2\operatorname{Re} \langle u - x, z \rangle \\ &= \|u - x\|^2 + \|z\|^2 \geq \|z\|^2 = \|y - u\|^2. \Rightarrow \|y - x\| \geq \|y - u\| \end{aligned}$$

(by proof of
Thm 6.2 (d))

If $\|y - x\| = \|y - u\|$ then the above inequality is an equality.

So we have $\|u - x\|^2 + \|z\|^2 = \|z\|^2 \Rightarrow \|u - x\|^2 = 0$

$\Rightarrow u = x$. If $x = u$, then $\|y - x\| = \|y - u\|$.



Remark.

The vector u in the corollary is called the **orthogonal projection** of y onto W . We will see the importance of orthogonal projections in the application to least squares in Section 6.3.

Theorem 6.7

Suppose that $S = \{v_1, v_2, \dots, v_k\}$ is an orthonormal set in an n -dimensional inner product space V . Then:

- (a) S can be extended to an orthonormal basis $\{v_1, v_2, \dots, v_k, v_{k+1}, \dots, v_n\}$ for V .
- (b) If $W = \operatorname{span}(S)$, then $S_1 = \{v_{k+1}, v_{k+2}, \dots, v_n\}$ is an orthonormal basis for W^\perp .
- (c) If W is any subspace of V , then $\dim(V) = \dim(W) + \dim(W^\perp)$.

Proof: (a) From linear algebra 1, we know S can be extended to an ordered basis.

$$S' = \{v_1, v_2, \dots, v_k, w_{k+1}, \dots, w_n\}.$$

Then we can apply GS process to S' to get \bar{S}

By HW #4 Q4, the vectors v_1, \dots, v_k remains the same as in S. Moreover \bar{S} spans V. Normalize the vectors in \bar{S} , we obtain an orthonormal basis $\{v_1, \dots, v_k, v_{k+1}, \dots, v_n\}$ for V.

(b) As $S_1 = \{v_{k+1}, \dots, v_n\}$ is an orthogonal set, by Corollary 2 in Lecture 4, we know S_1 is a linearly independent set.

Then we show $\text{Span}(S_1) = W^\perp$

By construction of S_1 , we know $\forall x \in \text{Span}(S_1)$,

$x \in W^\perp$. Thus $\text{Span}(S_1) \subseteq W^\perp$. Then we show

$W^\perp \subseteq \text{Span}(S_1)$. Let $x \in W^\perp$, then as $x \in V$, we

have $x = \sum_{i=1}^n \langle x, v_i \rangle v_i$. Since $x \in W^\perp$, we know

$\langle x, v_i \rangle = 0$, if $1 \leq i \leq k$. Thus $x = \sum_{i=k+1}^n \langle x, v_i \rangle v_i \in \text{Span}(S_1)$

Therefore S_1 is a basis of W^\perp .

(c) If W is a subspace of V of dimension k, then by (a) and (b),
 $\dim(V) = n = k + (n-k) = \dim(W) + \dim(W^\perp)$.

Recall
Definition. Basis

A subset β of a vector space V is called a **basis** if it satisfies two properties:

- β is linearly independent.
- β spans V (i.e., $\text{span}(\beta) = V$).

If β is a basis for V, we also say that the vectors in β form a **basis** for V.

Example 4.

Let $W = \text{span}(\{e_1, e_2\})$ in \mathbb{F}^3 .

Then $x = (a, b, c) \in W^\perp$ if and only if

- $0 = \langle x, e_1 \rangle = a$ and
- $0 = \langle x, e_2 \rangle = b$.

Thus, $x = (0, 0, c)$, and therefore $W^\perp = \text{span}(\{e_3\})$.

One can also deduce the same result by noting that $e_3 \in W^\perp$ and, from part (c), that $\dim(W^\perp) = 3 - 2 = 1$.