

# 1 Section 6.1: Inner Product, Length and Orthogonality

**Definition.** Let

$$\vec{u} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} \quad \text{and} \quad \vec{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

be two vectors in  $R^n$ . The inner product, denoted by  $\vec{u} \cdot \vec{v}$ , is defined as

$$\vec{u} \cdot \vec{v} = \vec{u}^T \vec{v} = [u_1, u_2, \dots, u_n] \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = u_1 v_1 + u_2 v_2 + \dots + u_n v_n.$$

**Example 6.1.1.** Let

$$\vec{u} = \begin{bmatrix} 3 \\ -5 \\ 1 \\ 2 \end{bmatrix}, \quad \vec{v} = \begin{bmatrix} 2 \\ 1 \\ 0 \\ -1 \end{bmatrix}.$$

Then

$$\vec{u} \cdot \vec{v} = [3 \quad -5 \quad 1 \quad 2] \begin{bmatrix} 2 \\ 1 \\ 0 \\ -1 \end{bmatrix} = 3 \cdot 2 - 5 \cdot 2 + 1 \cdot 0 + 2 \cdot (-1) = -1.$$

**Properties of Inner Product:**

(a)  $\vec{u} \cdot \vec{v} = \vec{v} \cdot \vec{u}$ , (b)  $(\vec{u} + \vec{v}) \cdot \vec{w} = \vec{u} \cdot \vec{w} + \vec{v} \cdot \vec{w}$ , (c)  $(c\vec{u}) \cdot \vec{v} = c(\vec{u} \cdot \vec{v})$ , (d)  $\vec{u} \cdot \vec{u} \geq 0$ , and  $\vec{u} \cdot \vec{u} = 0$  iff  $\vec{u} = 0$ .

**Definition.** Length (or magnitude) of a vector  $\vec{u}$ , denoted by  $\|\vec{u}\|$ , is defined as

$$\|\vec{u}\| = \sqrt{\vec{u} \cdot \vec{u}} = \sqrt{u_1^2 + u_2^2 + \dots + u_n^2}.$$

**Example 6.1.2.** For the same  $\vec{u}$  as in the previous example,

$$\|\vec{u}\| = \sqrt{3^2 + (-5)^2 + 1^2 + 2^2} = \sqrt{39}.$$

Properties (continues): (e)  $\|c\vec{u}\| = |c| \|\vec{u}\|$ .

**Example 6.1.3.** For the same  $\vec{u}$  as in the previous example,

$$\|-2\vec{u}\| = \sqrt{(-6)^2 + (10)^2 + (-2)^2 + (-4)^2} = \sqrt{156} = 2\sqrt{39} = 2\|\vec{u}\|.$$

**Definition.** Distance between two vectors  $\vec{u}$  and  $\vec{v}$ , denoted by  $dist(\vec{u}, \vec{v})$ , is defined as

$$dist(\vec{u}, \vec{v}) = \|\vec{u} - \vec{v}\| = \sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2 + \dots + (u_n - v_n)^2}.$$

**Example 6.1.4.** Let

$$\vec{u} = \begin{bmatrix} 3 \\ -5 \\ 1 \\ 2 \end{bmatrix}, \quad \vec{v} = \begin{bmatrix} 2 \\ 1 \\ 0 \\ -1 \end{bmatrix}.$$

Then,

$$dist(\vec{u}, \vec{v}) = \|\vec{u} - \vec{v}\| = \left\| \begin{bmatrix} 3 \\ -5 \\ 1 \\ 2 \end{bmatrix} - \begin{bmatrix} 2 \\ 1 \\ 0 \\ -1 \end{bmatrix} \right\| = \left\| \begin{bmatrix} 1 \\ -6 \\ 1 \\ 3 \end{bmatrix} \right\| = \sqrt{47}.$$

**Definition.** The angle between two vectors  $\vec{u}$  and  $\vec{v}$ , denoted by  $\langle \vec{u}, \vec{v} \rangle$ , is defined by

$$\cos \langle \vec{u}, \vec{v} \rangle = \frac{\vec{u} \cdot \vec{v}}{\|\vec{u}\| \|\vec{v}\|}, \quad 0 \leq \langle \vec{u}, \vec{v} \rangle \leq \pi.$$

Two vectors  $\vec{u}$  and  $\vec{v}$  are said to be orthogonal to each other if  $\langle \vec{u}, \vec{v} \rangle = \pi/2$ , or  $\vec{u} \cdot \vec{v} = 0$ . We use the notation  $\vec{u} \perp \vec{v}$  when  $\vec{u}$  and  $\vec{v}$  are orthogonal.

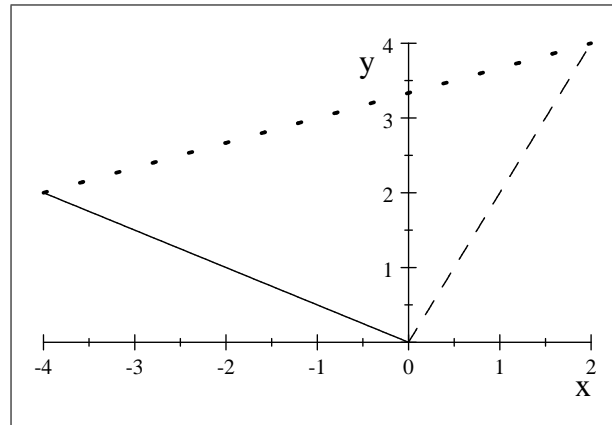
**Example 6.1.5.** For the same  $\vec{u}$  and  $\vec{v}$  as in the previous example, find the angle  $\langle \vec{u}, \vec{v} \rangle$ .

**Solution:**

$$\begin{aligned} \cos \langle \vec{u}, \vec{v} \rangle &= \frac{\vec{u} \cdot \vec{v}}{\|\vec{u}\| \|\vec{v}\|} = \frac{-1}{\sqrt{39}\sqrt{4+1+1}} = -0.065, \\ \langle \vec{u}, \vec{v} \rangle &= \arccos(-0.065) = 1.6358(\text{rad}) = 93.724^\circ \end{aligned}$$

**Theorem (Pythagorean)** In  $R^n$ ,  $\vec{u} \perp \vec{v}$  iff

$$\|\vec{u} - \vec{v}\|^2 = \|\vec{u}\|^2 + \|\vec{v}\|^2 \quad \text{or} \quad \|\vec{u} + \vec{v}\|^2 = \|\vec{u}\|^2 + \|\vec{v}\|^2.$$



solid line =  $\vec{u}$ , dash line  $\vec{v}$ , dot line =  $\vec{u} + \vec{v}$

**Proof.** We verify using direct computation and the fact  $\vec{u} \cdot \vec{v} = 0$  :

$$\begin{aligned} LHS &= \|\vec{u} - \vec{v}\|^2 = (\vec{u} - \vec{v}) \cdot (\vec{u} - \vec{v}) \\ &= \vec{u} \cdot \vec{u} - \vec{u} \cdot \vec{v} - \vec{v} \cdot \vec{u} + \vec{v} \cdot \vec{v} \\ &= \vec{u} \cdot \vec{u} + \vec{v} \cdot \vec{v} = \|\vec{u}\|^2 + \|\vec{v}\|^2 = RHS. \end{aligned}$$

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**Definition.** Let  $W$  be a subspace of  $R^n$ . A vector  $\vec{u}$  is said to be orthogonal to  $W$ , denoted by  $\vec{u} \perp W$ , if  $\vec{u}$  is orthogonal to every vector in  $W$ , i.e.,

$$\vec{u} \cdot \vec{w} = 0 \text{ for any } \vec{w} \in W \text{ (}\vec{w} \in W \text{ means } \vec{w} \text{ belongs to } W\text{)}.$$

We call the subspace

$$W^\perp = \{\vec{v} \mid \vec{v} \perp W\}$$

the orthogonal complement space of  $W$ .

**Example 6.1.6.** In  $R^2$ , let  $W$  be a line passing through the origin. Then, its orthogonal complement,  $W^\perp$  is the line passing through the origin and perpendicular to  $W$ . In  $R^3$ , (a) let  $W$  be a line passing through the origin, then  $W^\perp$  is the plane passing through the origin and perpendicular to  $W$ ; (b) let  $W$  be a plane passing through the origin, then  $W^\perp$  is the line passing through the origin and perpendicular to  $W$ .

**Example 6.1.7.** Let

$$W = \text{Span} \left\{ \vec{u} = \begin{bmatrix} 1 \\ -2 \\ 1 \\ 2 \end{bmatrix}, \vec{v} = \begin{bmatrix} 2 \\ 1 \\ -8 \\ -1 \end{bmatrix} \right\}$$

Find and describe  $W^\perp$ .

**Solution.** We are looking for all  $\vec{x}$  in  $R^4$  such that

$$\begin{aligned} \vec{x} \cdot \vec{u} &= 0 \\ \vec{x} \cdot \vec{v} &= 0 \end{aligned}$$

or

$$\begin{bmatrix} \vec{x} \cdot \vec{u} \\ \vec{x} \cdot \vec{v} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Let  $A$  be the matrix of rows  $\vec{u}$  and  $\vec{v}$ , i.e.,

$$A = \begin{bmatrix} \vec{u}^T \\ \vec{v}^T \end{bmatrix} = \begin{bmatrix} 1 & -2 & 1 & 2 \\ 2 & 1 & -8 & -1 \end{bmatrix}.$$

Then,  $W^\perp$  consists of all vectors  $\vec{x}$  such that

$$A\vec{x} = \begin{bmatrix} \vec{u}^T \\ \vec{v}^T \end{bmatrix} \vec{x} = \begin{bmatrix} \vec{u} \cdot \vec{x} \\ \vec{v} \cdot \vec{x} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

In other words,

$$W^\perp = \text{Null}(A).$$

In general, let

$$W = \text{Span}\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_p\} \text{ be a subspace in } R^n.$$

Then

$$W^\perp = \text{Null}(A), \quad A = \left( [\vec{u}_1 \ \vec{u}_2 \ \dots \ \vec{u}_p]_{n \times p} \right)^T = \begin{bmatrix} \vec{u}_1^T \\ \vec{u}_2^T \\ \vdots \\ \vec{u}_p^T \end{bmatrix}_{p \times n}.$$

We now proceed to describe  $W^\perp = \text{Null}(A)$  using row operations.

$$\begin{aligned} A &= \begin{bmatrix} 1 & -2 & 1 & 2 \\ 2 & 1 & -8 & -1 \end{bmatrix} \xrightarrow{R_2 - 2R_1 \rightarrow R_2} \begin{bmatrix} 1 & -2 & 1 & 2 \\ 0 & 5 & -10 & -5 \end{bmatrix} \\ &\xrightarrow{R_2/5 \rightarrow R_2} \\ &\xrightarrow{R_1 + 2R_2 \rightarrow R_1} \begin{bmatrix} 1 & 0 & -1 & 1 \\ 0 & 1 & -2 & -1 \end{bmatrix}. \end{aligned}$$

So,  $x_1 = x_3 - x_4$ ,  $x_2 = 2x_3 + x_4$ . The parametric form is (with  $x_3 = s$ ,  $x_4 = t$ )

$$\vec{x} = \begin{bmatrix} x_3 - x_4 \\ 2x_3 + x_4 \\ x_3 \\ x_4 \end{bmatrix} = s \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 1 \\ 0 \\ 1 \end{bmatrix},$$

and

$$W^\perp = \text{Null}(A) = \text{Span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Now let  $A$  be a  $m \times n$  matrix that has  $m$  row vectors,

$$A = \begin{bmatrix} \vec{v}_1 \\ \vec{v}_2 \\ \vdots \\ \vec{v}_m \end{bmatrix}, \quad \vec{v}_i \text{ be a row vector, i.e., } (\vec{v}_i)^T = \vec{u}_i \text{ is a column vector.}$$

Then, the "row space" of  $A$  is defined by

$$\text{Row}(A) = \text{Span}\{\text{rows of } A\} = \text{Span}\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\},$$

where a vector is understood as a row, i.e.,  $1 \times n$  matrix. If we would like to interpret row vectors as column vectors, then

$$\begin{aligned} \text{Row}(A) &= (\text{Span}\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\})^T = \text{Span}\{(\vec{v}_1)^T, (\vec{v}_2)^T, \dots, (\vec{v}_m)^T\} \\ &= \text{Span}\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_m\} \end{aligned}$$

From the above example, we can easily see that

$$\text{Row}(A)^\perp = \text{Nul}\left(\begin{bmatrix} \vec{u}_1^T \\ \vec{u}_2^T \\ \vdots \\ \vec{u}_m^T \end{bmatrix}_{m \times n}\right) = \text{Nul}\left(\begin{bmatrix} \vec{v}_1 \\ \vec{v}_2 \\ \vdots \\ \vec{v}_m \end{bmatrix}_{m \times n}\right) = \text{Null}(A).$$

Analogously, since columns of  $A = \text{rows of } A^T$ , or

$$\text{Col}(A) = \text{Row}(A^T),$$

we have

$$\text{Col}(A)^\perp = \text{Row}(A^T)^\perp = \text{Null}(A^T).$$