

## Orthogonality and Least Squares

### 4.1 Inner Product

Let  $u = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$  and  $v = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$  be any two vectors in  $\mathbb{R}^n$ . Then the number  $u^T v$  is called the

**inner product** of  $u$  and  $v$ . This inner product is also commonly known as **dot product** and denoted by  $\mathbf{u.v}$ .

#### 4.1.1 Properties of Inner Product

1.  $u.v = v.u$
2.  $(u + v).w = u.w + v.w$
3.  $(\alpha u).v = \alpha(u.v) = u.(cv)$
4.  $u.u \geq 0$  and  $u.u = 0 \iff u = 0$

#### 4.1.2 The Length of a Vector

The length of a vector  $v$  is called the **norm** of  $v$ .

It is denoted by  $\|v\|$  and defined by  $\|v\| = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$  so that,  $\|v\|^2 = v.v$ . There are several kinds of norms actually, this particular norm is called **Euclidean norm**. For any scalar  $\alpha$ ,  $\|\alpha v\| = |\alpha|\|v\|$ . A vector whose length is unity is called a **unit vector**. If we divide a nonzero vector  $v$  by its length, we obtain a unit vector  $u$ . This process is called **normalizing** of the vector  $v$ .

#### Distance between vectors

For  $u$  and  $v$  in a vector space  $V$ , the distance between them is written as  $\text{dist}(u, v)$  and is defined as  $\text{dist}(u, v) = \|u - v\|$ .

### 4.1.3 Exercise

1. Compute the following for the given vectors:

$$u = \begin{bmatrix} -1 \\ 2 \end{bmatrix}, v = \begin{bmatrix} 4 \\ 6 \end{bmatrix}, w = \begin{bmatrix} 3 \\ -1 \\ 5 \end{bmatrix}, x = \begin{bmatrix} 6 \\ -2 \\ 3 \end{bmatrix}$$

$$\text{a). } u \cdot u, \quad v \cdot u, \quad \frac{v \cdot u}{u \cdot u}, \quad \|v\| \qquad \text{b). } w \cdot w, \quad x \cdot w, \quad \frac{x \cdot w}{w \cdot w}, \quad \|x\|$$

2. Find the distance between  $u$  and  $v$ , and  $w$  and  $x$ .
3. Use matrix product and transpose definition to verify, property-2 and 3 of the inner product.
4. Explain why  $u \cdot u \geq 0$ . When is  $u \cdot u = 0$ ?

## 4.2 Orthogonal Vectors

The two vectors  $u$  and  $v$  are orthogonal vectors if their dot product is zero, i.e.  $u \cdot v = 0$ . Observe that the zero vector is orthogonal to every vector as  $0^T v = 0$  for all  $v$ .

**Theorem 1** (The Pythagorean Theorem)

Two vectors  $u$  and  $v$  are orthogonal if and only if  $\|u + v\|^2 = \|u\|^2 + \|v\|^2$ .

### 4.2.1 Orthogonal Complement

- If a vector  $z$  is orthogonal to every vectors in a subspace  $W$  then,  $z$  is said to be orthogonal to  $W$ .
- The set of all vectors that are orthogonal to  $W$  is called the **orthogonal complement** of  $W$ . It is denoted by  $W^\perp$ .  $W^\perp = \{z : \forall v \in W \ z \cdot v = 0\}$

**Theorem 2** 1. A vector  $x$  is in  $W^\perp$  if and only if  $x$  is orthogonal to every vector in a set that spans  $W$ .

2.  $W^\perp$  is also a subspace.
3. Row space is orthogonal complement of the Null space for a matrix.

### 4.2.2 Exercise

1. Verify parallelogram law:  $\|u + v\|^2 = 2\|u\|^2 + 2\|v\|^2$ .
2. Suppose  $y$  is orthogonal to  $u$  and  $v$ . Show that  $y$  is orthogonal to every  $w$  in  $\text{Span}\{u, v\}$ .
3. Let  $W$  be a subspace of  $\mathbb{R}^n$ , then show that  $W^\perp$  a subspace of  $\mathbb{R}^n$ .
4. Show that if  $x$  is in both  $W$  and  $W^\perp$ , then  $x = 0$ .

### 4.2.3 Orthogonal Sets

A set of vectors  $\{u_1, \dots, u_p\}$  in a vector space  $V$  is said to be **orthogonal set** if each pair of distinct vectors from the set is orthogonal, i.e. for all  $u_i, u_j \in V$  we have  $u_i \cdot u_j = 0$  whenever  $i \neq j$ .

**Definition 1.** An orthogonal basis for a vector space  $V$  is a basis for  $V$  that is an orthogonal set.

**Theorem 1.** Let  $\{u_1, \dots, u_p\}$  be an orthogonal basis for a subspace  $W$  of  $\mathbb{R}^n$ . For each  $y$  in  $W$ , the coordinates of  $y$  with respect to the orthogonal basis :  $y = c_1 u_1 + \dots + c_n u_n$  are given by  $c_j = \frac{y \cdot u_j}{u_j \cdot u_j}$ .

### 4.2.4 Exercise

1. Determine whether the following sets of vectors are orthogonal or not.

$$(a) \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} -5 \\ -2 \\ 1 \end{bmatrix}$$

2. Show that  $\{u_1, u_2\}$  is an orthogonal basis of  $\mathbb{R}^2$  and find the coordinates of  $x$  in terms of this basis.

$$(a) u_1 = \begin{bmatrix} 2 \\ 3 \end{bmatrix}, u_2 = \begin{bmatrix} 6 \\ 4 \end{bmatrix}, x = \begin{bmatrix} 9 \\ -7 \end{bmatrix}$$

$$(b) u_1 = \begin{bmatrix} 3 \\ 1 \end{bmatrix}, u_2 = \begin{bmatrix} 2 \\ -6 \end{bmatrix}, x = \begin{bmatrix} -6 \\ 3 \end{bmatrix}$$

- (c) Compute the orthogonal projection of  $\begin{bmatrix} 1 \\ 7 \end{bmatrix}$  onto the line through  $\begin{bmatrix} -4 \\ 2 \end{bmatrix}$  and the origin.

- (d) Compute the distance of  $y = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$  to the line through  $u = \begin{bmatrix} 8 \\ 6 \end{bmatrix}$  and the origin.

### 4.2.5 Orthogonal Projection

The coordinate  $c_j$  of  $y$  in Theorem 1 is actually orthogonal projection of  $y$  into the vector  $u_j$ . This can be generalized. For any given vector  $u$ . The orthogonal projection of a vector  $y$  on  $u$  is given by the formula  $\hat{y} = \frac{y \cdot u}{u \cdot u} u$ .

Or, it can be derived as follows using the inner-product.  $(y - \alpha u)$  and  $u$  are orthogonal so,  $(y - \alpha u) \cdot u = 0$ . This gives us  $\alpha = \frac{y \cdot u}{u \cdot u}$ . For two dimensional vectors, another orthogonal

component  $z$  can be easily obtained as by subtracting the projection from the vector  $y$ .  
 $z = y - \hat{y}$ .

### 4.3 Orthonormal Sets

A set  $\{u_1, \dots, u_p\}$  is an orthonormal set if it is an orthogonal set of unit vectors. And a basis of orthonormal set is a orthonormal basis. The simplest orthonormal basis is  $\{e_1, \dots, e_n\}$  for  $\mathbb{R}^n$ .

Matrices whose columns form an orthonormal set are important in applications and in computer algorithms for matrix computations.

**Theorem 2.** An  $m \times n$  matrix  $U$  has orthonormal columns if and only if  $U^t U = I$