

## 1 Inner Products

An **inner product**  $\langle \cdot, \cdot \rangle$  on a vector space  $V$  over  $\mathbb{R}$  is a function that takes in two vectors and outputs a scalar, such that  $\langle \cdot, \cdot \rangle$  is symmetric, linear, and positive-definite.

- Symmetry:  $\langle \vec{u}, \vec{v} \rangle = \langle \vec{v}, \vec{u} \rangle$
- Scaling:  $\langle c\vec{u}, \vec{v} \rangle = c\langle \vec{u}, \vec{v} \rangle$  and  $\langle \vec{u}, c\vec{v} \rangle = c\langle \vec{u}, \vec{v} \rangle$
- Additivity:  $\langle \vec{u} + \vec{v}, \vec{w} \rangle = \langle \vec{u}, \vec{w} \rangle + \langle \vec{v}, \vec{w} \rangle$  and  $\langle \vec{u}, \vec{v} + \vec{w} \rangle = \langle \vec{u}, \vec{v} \rangle + \langle \vec{u}, \vec{w} \rangle$
- Positive-definite:  $\langle \vec{u}, \vec{u} \rangle \geq 0$  with  $\langle \vec{u}, \vec{u} \rangle = 0$  if and only if  $\vec{u} = \vec{0}$

For two vectors,  $\vec{u}, \vec{v} \in \mathbb{R}^n$ , the standard inner product is  $\langle \vec{u}, \vec{v} \rangle = \vec{u}^T \vec{v}$ . We define the **norm**, or the magnitude, of a vector  $\vec{v}$  to be  $\|\vec{v}\| = \sqrt{\langle \vec{v}, \vec{v} \rangle} = \sqrt{\vec{v}^T \vec{v}}$ . For any non-zero vector, we can *normalize*, i.e., set its magnitude to 1 while preserving its direction, by dividing the vector by its norm  $\frac{\vec{v}}{\|\vec{v}\|}$ .

### Orthogonality and Orthonormality

The inner product lets us define the angle between two vectors through the equation

$$\langle \vec{u}, \vec{v} \rangle = \|\vec{u}\| \|\vec{v}\| \cos \theta \quad (1)$$

Notice that if the angle  $\theta$  between two vectors is  $\pm 90^\circ$ , the inner product  $\langle \vec{u}, \vec{v} \rangle = 0$ .

Therefore, we define two vectors  $\vec{u}$  and  $\vec{v}$  to be **orthogonal** to each other if  $\langle \vec{u}, \vec{v} \rangle = \vec{u}^T \vec{v} = 0$ . A set of vectors is orthogonal if any two vectors in this set are orthogonal to each other.

Furthermore, we define two vectors  $\vec{u}$  and  $\vec{v}$  to be **orthonormal** to each other if they are orthogonal to each other and their norms are 1. A set of vectors is orthonormal if any two vectors in this set are orthogonal to each other and every vector has a norm of 1. In fact, for any two vectors  $\vec{u}$  and  $\vec{v}$  in an orthonormal set,

$$\langle \vec{u}, \vec{v} \rangle = \vec{u}^T \vec{v} = \begin{cases} 1, & \text{if } \vec{u} = \vec{v} \\ 0, & \text{otherwise} \end{cases}.$$

### Unitary Matrices

An **orthogonal** or **unitary** matrix is a square matrix whose columns are orthonormal with respect to the inner product. To avoid any confusion, we will often refer to these matrices as **orthonormal matrices**.

$$U = [\vec{u}_1 \quad \cdots \quad \vec{u}_n], \quad \vec{u}_i^T \vec{u}_j = \begin{cases} 1, & \text{if } i = j \\ 0, & \text{otherwise} \end{cases}$$

Note that  $U^T U = U U^T = I$ , so the inverse of a unitary matrix is its transpose  $U^{-1} = U^T$ .

Since the columns of a unitary matrix are orthonormal vectors, we can interpret these matrices as “rotation” and “reflection” matrices of the standard axes. This also implies that  $\|U\vec{v}\| = \|\vec{v}\|$  for any vector  $\vec{v}$ .

## 2 Spectral Theorem

Let  $A$  be an  $n \times n$  **symmetric** matrix with real entries. Then the following statements will be true.

1. All eigenvalues of  $A$  are real.
2.  $A$  has  $n$  linearly independent eigenvectors  $\in \mathbb{R}^n$ .
3.  $A$  has orthogonal eigenvectors, i.e.,  $A = V\Lambda V^{-1} = V\Lambda V^T$ , where  $\Lambda$  is a diagonal matrix and  $V$  is an orthonormal matrix. We say that  $A$  is orthogonally diagonalizable.

Recall that a matrix  $A$  is symmetric if  $A = A^T$ . Furthermore, if  $A$  is of the form  $B^T B$  for some arbitrary matrix  $B$ , then all of the eigenvalues of  $A$  are non-negative, i.e.,  $\lambda \geq 0$ .

- a) Prove the following: All eigenvalues of a symmetric matrix  $A$  are real.

*Hint:* Let  $(\lambda, \vec{v})$  be an eigenvalue/vector pair. Then  $A\vec{v} = \lambda\vec{v}$  and take the complex conjugate and transpose of both sides. Try to show that  $\bar{\lambda} = \lambda$ .

### Answer

Let  $\lambda$  be an eigenvalue of  $A$  with corresponding eigenvector  $\vec{v}$ .

$$A\vec{v} = \lambda\vec{v}$$

Then we can take the complex conjugate of both sides and use the fact that  $A$  is real.

$$A\bar{\vec{v}} = \bar{\lambda}\bar{\vec{v}}$$

Now let's take the transpose of both sides and use the fact that  $A$  is symmetric.

$$\bar{\vec{v}}^T A = \bar{\lambda}\bar{\vec{v}}^T$$

Let's right multiply by  $\vec{v}$ , to see that

$$\bar{\vec{v}}^T A\vec{v} = \bar{\lambda}\bar{\vec{v}}^T \vec{v}$$

Applying the eigenvector property once more,

$$\lambda\bar{\vec{v}}^T \vec{v} = \bar{\lambda}\bar{\vec{v}}^T \vec{v}$$

$\bar{\vec{v}}^T \vec{v}$  must be greater than 0 since  $\vec{v}$  is an eigenvector and cannot be  $\vec{0}$ . Therefore, the only possibility is  $\lambda = \bar{\lambda}$  which implies that  $\lambda$  is real.

- b) Prove the following: For any symmetric matrix  $A$ , any two eigenvectors corresponding to distinct eigenvalues of  $A$  are orthogonal.

*Hint:* Let  $\vec{v}_1$  and  $\vec{v}_2$  be eigenvectors of  $A$  with eigenvalues  $\lambda_1 \neq \lambda_2$ .

$$A\vec{v}_1 = \lambda_1\vec{v}_1$$

$$A\vec{v}_2 = \lambda_2\vec{v}_2$$

Take the transpose of the second equation and show that  $\lambda_1\langle\vec{v}_1, \vec{v}_2\rangle = \lambda_2\langle\vec{v}_1, \vec{v}_2\rangle$ .

**Answer**

Let's apply the hint by taking the transpose of the second equation.

$$\begin{aligned} A\vec{v}_1 &= \lambda_1\vec{v}_1 \\ \vec{v}_2^T A^T &= \lambda_2\vec{v}_2^T \end{aligned}$$

Since  $A$  is symmetric,  $A^T = A$  and we can left-multiply the first equation by  $\vec{v}_2^T$  and right-multiply the second equation by  $\vec{v}_2$  to say

$$\begin{aligned} \vec{v}_2^T A\vec{v}_1 &= \lambda_1\vec{v}_2^T\vec{v}_1 \\ \vec{v}_2^T A\vec{v}_1 &= \lambda_2\vec{v}_2^T\vec{v}_1 \end{aligned}$$

This tells us that  $\lambda_1\langle\vec{v}_1, \vec{v}_2\rangle = \lambda_2\langle\vec{v}_1, \vec{v}_2\rangle$  meaning

$$(\lambda_1 - \lambda_2)\langle\vec{v}_1, \vec{v}_2\rangle = 0$$

The only way this equation can be satisfied when  $\lambda_1 \neq \lambda_2$  is for  $\langle\vec{v}_1, \vec{v}_2\rangle$  to be zero. Therefore,  $\vec{v}_1$  and  $\vec{v}_2$  must be orthogonal to each other

- c) Prove the following: For any matrix  $A$ ,  $A^T A$  is symmetric and only has non-negative eigenvalues. *Hint:* Consider the quantity  $\|A\vec{v}\|^2$ . Remember that norms are positive-definite.

**Answer**

$A^T A$  is symmetric since  $(A^T A)^T = A^T (A^T)^T = A^T A$  equals itself.

Let  $\lambda$  be an eigenvalue of  $A^T A$  with corresponding eigenvector  $\vec{v}$ .

$$A^T A\vec{v} = \lambda\vec{v}$$

We left-multiply  $\vec{v}^T$  on both sides.

$$\begin{aligned} \vec{v}^T A^T A\vec{v} &= \vec{v}^T \lambda\vec{v} \\ (A\vec{v})^T A\vec{v} &= \lambda\vec{v}^T\vec{v} \\ \|A\vec{v}\|^2 &= \lambda\|\vec{v}\|^2 \\ \lambda &= \frac{\|A\vec{v}\|^2}{\|\vec{v}\|^2} \geq 0 \end{aligned}$$

Note that  $\|\vec{v}\| \neq 0$  because we assumed that  $\vec{v}$  is an eigenvector corresponding to  $\lambda$ .

### 3 Outer Products

An **outer product**  $\otimes$  is a function that takes two vectors and outputs a **matrix**. We define  $\vec{x} \otimes \vec{y} = \vec{x}\vec{y}^T$ .

a) Let  $\vec{x} = \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix}$  and  $\vec{y} = \begin{bmatrix} 4 \\ 2 \\ -1 \end{bmatrix}$ .

(i) Compute the outer-product  $A = \vec{x}\vec{y}^T$ .

**Answer**

$$\vec{x}\vec{y}^T = \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix} \begin{bmatrix} 4 & 2 & -1 \end{bmatrix} = \begin{bmatrix} 4 & 2 & -1 \\ 12 & 6 & -3 \\ -8 & -4 & 2 \end{bmatrix}$$

(ii) What is the shape of the matrix  $A$ ?

**Answer**

$\vec{x}\vec{y}^T$  is a  $3 \times 3$  matrix.

(iii) What is the rank of  $A$ ?

**Answer**

$\text{Rank}(A) = 1$ .

b) Let  $B = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}$ .

(i) Write  $B$  as an outer-product of two vectors  $\vec{x}$  and  $\vec{y}$ .

**Answer**

$$B = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}$$

(ii) What is the rank of  $B$ ?

**Answer**

$\text{Rank}(B) = 1$ .

c) Let  $C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 \end{bmatrix}$ .

(i) Write  $C$  as a sum of outer-products:  $\vec{x}\vec{y}^T + \vec{u}\vec{w}^T$ .

**Answer**

$$C = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$$

(ii) What is the rank of  $C$ ?

**Answer**

$$\text{Rank}(C) = 2.$$

d) Let  $D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ .

(i) Write  $D$  as a sum of outer-products.

**Answer**

$$D = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}$$

(ii) What is the rank of  $D$ ?

**Answer**

$$\text{Rank}(D) = 3.$$