

Lecture 2 Orthogonal Vectors and Matrices

Fang Zhu

February 17, 2023

1 Prerequisite

Before giving the solutions, I would like to prove some basic conclusions about this lecture

Lemma 1.1. *Given an non-singular matrix \mathbf{A} , then \mathbf{A}^{-1} is unique*

Proof. Suppose that we have two inverse matrices \mathbf{C} and \mathbf{B} w.r.t \mathbf{A} . By the definition of inverse.

$$\mathbf{B} = \mathbf{B}\mathbf{I} = \mathbf{B}(\mathbf{A}\mathbf{C}) = (\mathbf{B}\mathbf{A})\mathbf{C} = \mathbf{I}\mathbf{C} = \mathbf{C},$$

and hence we can conclude that $\mathbf{B} = \mathbf{C}$. □

Lemma 1.2. *Given an non-singular matrix $\mathbf{A} \in \mathbb{R}^{m \times m}$, its hermitian conjugate \mathbf{A}^* is also non-singular.*

Proof.

$$\mathbf{A}^{-1}\mathbf{A} = \mathbf{A}\mathbf{A}^{-1} = \mathbf{I},$$

We can apply the hermitian conjugate to both sides of the equation above:

$$\mathbf{A}^*(\mathbf{A}^{-1})^* = (\mathbf{A}^{-1})^*\mathbf{A}^* = \mathbf{I}$$

Hence we can get that \mathbf{A}^* is non-singular. □

Lemma 1.3. *Give a non-singular matrix \mathbf{A} and its hermitian conjugate \mathbf{A}^* , we have*

$$(\mathbf{A}^*)^{-1} = (\mathbf{A}^{-1})^*$$

Proof. By lemma 1.2, \mathbf{A}^* is non-singular and it's clear that the inverse is $(\mathbf{A}^{-1})^*$. However, we can get that $(\mathbf{A}^*)^{-1}$ is also the inverse of \mathbf{A}^* by definition. Futher, by lemma 1.1, we have

$$(\mathbf{A}^*)^{-1} = (\mathbf{A}^{-1})^*$$

which is exactly what we need to prove. □

Lemma 1.4. *Given two pure imaginary number di, bi , then*

$$(1 - di)(1 - bi) \neq 0.$$

Proof. LHS equals

$$1 - bd - (b + d)i, \tag{*}$$

if $(*) = 0$, then we have

$$\begin{aligned} 1 - bd &= 0 \\ b + d &= 0, \end{aligned}$$

which means that

$$-b^2 = 1,$$

since $b \in \mathbb{R}$, the equation above cannot be true, and hence $(1 - di)(1 - bi) \neq 0$. \square

2 Solutions

2.1 Exercise 2.1

Proof. Without loss of generality, we assume that \mathbf{A} is upper-triangular. By the *ex. 1.3*, we can conclude that \mathbf{A}^{-1} is also upper-triangular. It is clear that $\mathbf{A}^* = \mathbf{A}^{-1}$ since \mathbf{A} is unitary. Then \mathbf{A}^* is also an upper-triangular matrix, which is

$$\mathbf{A}_{i,j}^* = \bar{a}_{ji} = a_{ij} = 0, \quad (\forall i > j),$$

Hence, the matrix \mathbf{A} is diagonal. The same follows if \mathbf{A} is lower-triangular. \square

2.2 Exercise 2.3(a)

Let \mathbf{x} be an eigenvector of matrix \mathbf{A} w.r.t. the eigenvalue λ , then

$$\mathbf{Ax} = \lambda\mathbf{x},$$

multiplying both sides by \mathbf{x}^* , we get that

$$\mathbf{x}^* \mathbf{Ax} = \lambda \mathbf{x}^* \mathbf{x} = \lambda \|\mathbf{x}\|^2, \tag{♠}$$

then

$$\lambda = \frac{\mathbf{x}^* \mathbf{Ax}}{\|\mathbf{x}\|^2} = \frac{\mathbf{x}^* \mathbf{A}^* \mathbf{x}}{\|\mathbf{x}\|^2} = \frac{(\mathbf{x}^* \mathbf{Ax})^*}{\|\mathbf{x}\|^2} = \bar{\lambda},$$

which means that λ is real.

2.3 Exercise 2.3(b)

Let $\mathbf{x}_1, \mathbf{x}_2$ be two eigenvectors of the hermitian matrix \mathbf{A} . Denote λ_k the eigenvalue w.r.t \mathbf{x}_k ($k = 1, 2$), where $\lambda_1 \neq \lambda_2$, then

$$\begin{aligned}\lambda_2 \mathbf{x}_1^* \mathbf{x}_2 &= \mathbf{x}_1^* \mathbf{A} \mathbf{x}_2, \\ \lambda_1 \mathbf{x}_2^* \mathbf{x}_1 &= \mathbf{x}_2^* \mathbf{A} \mathbf{x}_1.\end{aligned}$$

Note that \mathbf{A} is hermitian, we can get that

$$\lambda_2 \mathbf{x}_1^* \mathbf{x}_2 = \mathbf{x}_1^* \mathbf{A} \mathbf{x}_2 = \mathbf{x}_1^* \mathbf{A}^* \mathbf{x}_2 = (\mathbf{x}_2^* \mathbf{A} \mathbf{x}_1)^* = \lambda_1^* \mathbf{x}_1^* \mathbf{x}_2,$$

then

$$(\lambda_2 - \lambda_1^*) \mathbf{x}_1^* \mathbf{x}_2 = 0 \Rightarrow \mathbf{x}_1^* \mathbf{x}_2 = 0,$$

which is exactly what we need to prove.

2.4 Exercise 2.4

Let λ be an eigenvalue of \mathbf{A} , and \mathbf{x} be the eigenvector w.r.t λ , then we have $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$ and $\|\mathbf{A}\mathbf{x}\|_2^2 = \|\lambda\mathbf{x}\|$, which is

$$\mathbf{x}^* \mathbf{A}^* \mathbf{A} \mathbf{x} = \mathbf{x}^* \|\lambda\| \lambda \mathbf{x}.$$

Since \mathbf{A} is unitary, then

$$\mathbf{x}^* \mathbf{A}^* \mathbf{A} \mathbf{x} = \mathbf{x}^* \mathbf{I} \mathbf{x} = \mathbf{x}^* \|\lambda\|_2^2 \mathbf{x}.$$

Furthermore,

$$x_1^2 + x_2^2 + \cdots + x_n^2 = \|\lambda\|_2^2 (x_1^2 + x_2^2 + \cdots + x_n^2),$$

it follows that $\|\lambda\|_2^2 = 1$ since \mathbf{x} is non-zero vector.

2.5 Exercise 2.5(a)

Let \mathbf{x} be an eigenvector of matrix \mathbf{S} w.r.t. the eigenvalue λ , then

$$\mathbf{S}\mathbf{x} = \lambda\mathbf{x}.$$

By the equation ex. 2.3(♠), we have

$$\lambda = \frac{\mathbf{x}^* \mathbf{S} \mathbf{x}}{\|\mathbf{x}\|^2} = \frac{\mathbf{x}^* (-\mathbf{S}^*) \mathbf{x}}{\|\mathbf{x}\|^2} = \frac{-(\mathbf{x}^* \mathbf{S} \mathbf{x})^*}{\|\mathbf{x}\|^2} = -\bar{\lambda},$$

then we can get that $\lambda + \bar{\lambda} = 0$, which means that λ is purely imaginary.

2.6 Exercise 2.5(b)

Assume that λ is the eigenvalue of \mathbf{S} , it follows that $1 - \lambda$ is the eigenvalue of $1 - \mathbf{S}$. Since λ is purely imaginary number, then by *Lemma 2.1.4*, we have

$$\det(1 - \mathbf{S}) = \prod_{i=1}^n (1 - \lambda_i) \neq 0,$$

where $\lambda_i, i \in \{1, 2, \dots, n\}$ are eigenvalues of \mathbf{S} . Hence we can conclude that $1 - \mathbf{S}$ is non-singular.

2.7 Exercise 2.5(c)

Assume that $\mathbf{Q} = (\mathbf{I} - \mathbf{S})^{-1}(\mathbf{I} + \mathbf{S})$, then we have

$$\mathbf{Q}\mathbf{Q}^* = (\mathbf{I} - \mathbf{S})^{-1}(\mathbf{I} + \mathbf{S})(\mathbf{I} + \mathbf{S}^*)((\mathbf{I} + \mathbf{S})^{-1})^*,$$

by *Lemma 2.1.3*, we can get that

$$\begin{aligned} \mathbf{Q}\mathbf{Q}^* &= (\mathbf{I} - \mathbf{S})^{-1}(\mathbf{I} + \mathbf{S})(\mathbf{I} - \mathbf{S})((\mathbf{I} - \mathbf{S})^{-1})^* \\ &= (\mathbf{I} - \mathbf{S})^{-1}(\mathbf{I} + \mathbf{S})(\mathbf{I} - \mathbf{S})(\mathbf{I} + \mathbf{S})^{-1} \\ &= (\mathbf{I} - \mathbf{S})^{-1}(\mathbf{I} - \mathbf{S})(\mathbf{I} + \mathbf{S})(\mathbf{I} + \mathbf{S})^{-1} \\ &= \mathbf{I}. \end{aligned}$$

Hence, we can conclude that $(1 - \mathbf{S})^{-1}(1 - \mathbf{S})$ is unitary.

2.8 Exercise 2.6

Proof. If \mathbf{A} is singular, there exists a vector $\mathbf{x} \in \mathbb{C} \setminus \{0\}$ such that

$$\mathbf{A}\mathbf{x} = \mathbf{x} + \mathbf{u}\mathbf{v}^*\mathbf{x} = 0,$$

then $\mathbf{x} = -\mathbf{u}(\mathbf{v}^*\mathbf{x})$ where $\mathbf{v}^*\mathbf{u}$ is scalar. Let $\mathbf{x} = t\mathbf{u}(t \in \mathbb{R})$, then we can get that

$$t\mathbf{u} + \mathbf{u}(\mathbf{v}^*t\mathbf{u}) = t\mathbf{u}(1 + \mathbf{v}^*\mathbf{u}) = 0,$$

It follows that $\mathbf{v}^*\mathbf{u} = -1$ since $\mathbf{x} = t\mathbf{u} \neq 0$. Assume that $\alpha = -1/(1 + \mathbf{v}^*\mathbf{u})$, then

$$(\mathbf{I} + \mathbf{u}\mathbf{v}^*)(\mathbf{I} + \alpha\mathbf{u}\mathbf{v}^*) = \mathbf{I}.$$

Note that we have shown that $\mathbf{v}^*\mathbf{u} = -1$ is a necessary condition of \mathbf{A} is singular. For sufficiency, we assume that $\mathbf{v}^*\mathbf{u} = -1$. Then for any $t \in \mathbb{C} \setminus \{0\}$, we have

$$\mathbf{A}\mathbf{u} = t\mathbf{u} + \mathbf{u}\mathbf{v}^*t\mathbf{u} = t\mathbf{u} + t\mathbf{u}(\mathbf{v}^*\mathbf{u}) = 0,$$

which implies that \mathbf{A} is singular. Combined, we conclude that \mathbf{A} is singular iff. $\mathbf{v}^*\mathbf{u} = -1$. In this case,

$$\text{null}(\mathbf{A}) = \{t\mathbf{u}, t \in \mathbb{R}\},$$

the linear subspace spanned by \mathbf{u} . □

2.9 Exercise 2.7

Proof. We can verify that \mathbf{H}_{k+1} is Hadamard matrix directly,

$$\begin{aligned}
 \mathbf{H}_{k+1}^T \mathbf{H}_{k+1} &= \begin{pmatrix} \mathbf{H}_k^T & \mathbf{H}_k^T \\ \mathbf{H}_k^T & -\mathbf{H}_k^T \end{pmatrix} \begin{pmatrix} \mathbf{H}_k & \mathbf{H}_k \\ \mathbf{H}_k & -\mathbf{H}_k \end{pmatrix} \\
 &= \begin{pmatrix} \mathbf{H}_k^T \mathbf{H}_k + \mathbf{H}_k^T \mathbf{H}_k & \mathbf{H}_k^T \mathbf{H}_k - \mathbf{H}_k^T \mathbf{H}_k \\ \mathbf{H}_k^T \mathbf{H}_k - \mathbf{H}_k^T \mathbf{H}_k & \mathbf{H}_k^T \mathbf{H}_k + \mathbf{H}_k^T \mathbf{H}_k \end{pmatrix} \\
 &= \begin{pmatrix} 2\mathbf{H}_k^T \mathbf{H}_k & \mathbf{0} \\ \mathbf{0} & 2\mathbf{H}_k^T \mathbf{H}_k \end{pmatrix} \\
 &= \begin{pmatrix} 2c\mathbf{I}_k & \mathbf{0} \\ \mathbf{0} & 2c\mathbf{I}_k \end{pmatrix} \\
 &= 2c \cdot \mathbf{I}_{2k}.
 \end{aligned}$$

Then we can get that $\mathbf{H}_{k+1}^T = 2c\mathbf{H}_{k+1}^{-1}$. Note that the entries of \mathbf{H}_{k+1} are also all ± 1 by the recursion formula. Hence \mathbf{H}_{k+1} is also a Hadamard matrix. \square