

# Linear Algebra II Normal Linear Transformation

Ming-Hsuan Kang



We will show that the extra structure "inner product" is also very powerful in the study of linear transformation.

## Symmetric and Hermitian



For  $A \in M_n(\mathbb{R})$  and  $\vec{v}, \vec{w} \in \mathbb{R}^n$ , we have

$$\langle \vec{v}, A\vec{w} \rangle = \vec{v}^t A\vec{w} = (A^t \vec{v})^t \vec{w} = \langle A^t \vec{v}, \vec{w} \rangle.$$

Here  $A^t$  is the transpose of A. When  $A^t = A$ , A is called symmetric.

For  $A \in M_n(\mathbb{C})$  and  $\vec{v}, \vec{w} \in \mathbb{C}^n$ , we have

$$\langle \vec{v}, A\vec{w} \rangle = \vec{v}^t \overline{A\vec{w}} = (\overline{A}^t \vec{v})^t \overline{\vec{w}} = \langle \overline{A}^t \vec{v}, \vec{w} \rangle = \langle A^* \vec{v}, \vec{w} \rangle.$$

Here  $A^* := \overline{A}^t$  is the conjugate transpose of A. When  $A^* = A$ , A is called Hermitian.

## **Examples**



### Example

 $A = \begin{pmatrix} 1 & 2 \\ 2 & 3 \end{pmatrix}$  is a real symmetric matrix.

## **E**xample

 $A = \begin{pmatrix} 1+i & 1-2i \\ 1-2i & 3 \end{pmatrix}$  is a complex symmetric matrix but not a Hermitian matrix.

## Example

$$A = \begin{pmatrix} 1 & 1+2i \\ 1-2i & 3 \end{pmatrix}$$
 is a Hermitian matrix.

# **Self-adjoint Linera Transformation**



For a linear transformation T on an inner product space  $(V, \langle \cdot, \cdot \rangle)$ , a linear transformation  $T^*$  on V is called the adjoint of T if

$$\langle \vec{v}, T\vec{w} \rangle = \langle T^*\vec{v}, \vec{w} \rangle$$

for all  $\vec{v}, \vec{w} \in V$ . When  $T = T^*$ , T is called self-adjoint. We will show later that the existence and uniqueness of  $T^*$  later.

**Remark.** We also have  $\langle T\vec{v}, \vec{w} \rangle = \langle \vec{v}, T^*\vec{w} \rangle$ . (Exercise)

# **Self-Adjoint Linear Transformation**



#### Theorem

Let T be a self-adjoint linear transformation on a complex inner product space V. Then

- 1. All eigenvalues of T are real.
- 2. T admits a set of eigenvectors which forms an orthonormal basis of V. (Especially, T is diagonalizable.)

Let  $\vec{v}$  be an eigenvector corresponding to an eigenvalue of  $\lambda$  of T. Then

$$\frac{\lambda}{\langle \vec{v}, \vec{v} \rangle} = \langle \lambda \vec{v}, \vec{v} \rangle = \langle T \vec{v}, \vec{v} \rangle = \langle \vec{v}, T \vec{v} \rangle = \langle \vec{v}, \lambda \vec{v} \rangle = \frac{\lambda}{\lambda} \langle \vec{v}, \vec{v} \rangle.$$

Since  $\langle \vec{v}, \vec{v} \rangle \neq 0$ ,  $\lambda = \overline{\lambda}$  and  $\lambda$  is real.

#### **Proof**



Let us prove the part (2) by induction on  $n = \dim V$ .

When n = 1, it is obvious.

Suppose the statement holds for the case of dimension < n.

Let  $\vec{v}$  be an eigenvector corresponding to an eigenvalue  $\lambda$  and W be the orthogonal compliment of  $F\vec{v}$ . Then  $V=F\vec{v}\oplus W$ . Claim that W is T-invariant.

#### **Proof**



For  $\vec{w} \in W$ ,

$$\langle T\vec{w}, \vec{v} \rangle = \langle \vec{w}, T\vec{v} \rangle = \langle \vec{w}, \lambda \vec{v} \rangle = \bar{\lambda} \langle \vec{w}, \vec{v} \rangle = 0.$$

Therefore,  $T(W) \subseteq W$  and W is T-invariant.

Note that

$$(T|_{W})^{*} = (T^{*})|_{W} = T|_{W}.$$

Thus  $T|_W$  is also self-adjoint. By induction, there exists a set of eigenvectors  $\beta$  which forms an orthonormal basis of W. Then  $\{\vec{v}\} \cup \beta$  is the desired set.

# Real Symmetric Matrix



## **Corollary**

Let A be a real symmetric matrix.

- All eigenvalues of A are real.
- A is diagonalizable over  $\mathbb{R}$ .
- A admits a set of eigenvectors which forms an orthonormal basis.

### **Corollary**

Let A be a Hermitian matrix.

- All eigenvalues of A are real.
- A is diagonalizable.
- A admits a set of eigenvectors which forms an orthonormal(unitary) basis.

# **Existence of Adjoint Linear Transformation**



Let  $\alpha$  be an orthonormal basis of V and  $A = \operatorname{Rep}_{\alpha}(T)$ . Then for  $\vec{w}_1 = \operatorname{Rep}_{\alpha}(\vec{v}_1)$  and  $\vec{w}_2 = \operatorname{Rep}_{\alpha}(\vec{v}_2)$ ,

$$\begin{split} \langle \vec{v}_1, \, T \, \vec{v}_2 \rangle &= \langle \vec{w}_1, A \vec{w}_2 \rangle_{F^n} \\ &= \langle A^* \, \vec{w}_1, \, \vec{w}_2 \rangle_{F^n} = \langle \, T^* \, \vec{v}_1, \, \vec{v}_2 \rangle \end{split}$$

Here  $T^*$  is the linear transformation with the matrix representation  $A^*$  under the basis  $\alpha$ , which is also the adjoint of T.



Under an orthonormal basis, the conjugate transpose of the matrix representation of T is equal to the matrix representation of  $T^*$ .



A symmetric/Hermitian matrix is a matrix representation of a self-adjoint linear transform under an orthonormal basis.

# **Uniqueness of Adjoint Linear Transformation**



Suppose T' is also an adjoint of T. For any  $\vec{v} \in V$ , we will show that  $T^*\vec{v} = T'\vec{v}$ . Let  $\alpha$  be an orthonormal basis of V. Then

$$T^*\vec{v} = \sum_{\vec{w} \in \alpha} \langle T^*\vec{v}, \vec{w} \rangle \vec{w} = \sum_{\vec{w} \in \alpha} \langle \vec{v}, T\vec{w} \rangle \vec{w} = \sum_{\vec{w} \in \alpha} \langle T'\vec{v}, \vec{w} \rangle \vec{w} = T'\vec{v}.$$

Therefore,  $T^* = T'$ .

#### **Normal Linear Transformation**



A linear transformation T on an inner product space V is normal if  $T^*$  and T commute. The main theorem of this section is the following.

#### **Theorem**

A complex linear transformation is diagonalizable under some orthonormal basis if and only if it is normal.

Before proving the theorem, we need some preparations.

## **Properties of Normal Linear Transformation**



#### Theorem

Let T be a normal operator. Let  $\vec{v}$  be the eigenvector corresponding to the eigenvalue  $\lambda$  of T. Then  $\vec{v}$  be the eigenvector corresponding to the eigenvalue  $\bar{\lambda}$  of  $T^*$ .

We shall show that  $T^*\vec{v} = \bar{\lambda}\vec{v}$ . Consider

$$\begin{split} &\langle T^* \vec{v} - \bar{\lambda} \vec{v}, \, T^* \vec{v} - \bar{\lambda} \vec{v} \rangle \\ &= \langle T^* \vec{v}, \, T^* \vec{v} \rangle - \langle \bar{\lambda} \vec{v}, \, T^* \vec{v} \rangle - \langle T^* \vec{v}, \bar{\lambda} \vec{v} \rangle + \langle \bar{\lambda} \vec{v}, \bar{\lambda} \vec{v} \rangle \\ &= \langle T \vec{v}, \, T \vec{v} \rangle - \bar{\lambda} \langle T \vec{v}, \, \vec{v} \rangle - \lambda \, \langle \vec{v}, \, T \vec{v} \rangle + |\lambda|^2 \langle \vec{v}, \, \vec{v} \rangle \\ &= \langle \lambda \vec{v}, \lambda \vec{v} \rangle - \bar{\lambda} \langle \lambda \vec{v}, \, \vec{v} \rangle - \lambda \, \langle \vec{v}, \lambda \vec{v} \rangle + |\lambda|^2 \langle \vec{v}, \, \vec{v} \rangle = 0. \end{split}$$

Here the red part is the only place that we use the condition that T is normal. The proof is complete.

#### **Normal Linear Transformation**



## **Proposition**

Let T be a normal complex linear transformation T, then

- 1.  $ker(T) = ker(T^*T) = ker(TT^*) = ker(T^*)$ .
- 2.  $\ker(T) = \operatorname{Im}(T)^{\perp}$  and  $V = \ker(T) \oplus \operatorname{Im}(T)$ .
- 3.  $Im(T) = Im_{\infty}(T)$ .
- 4.  $\ker(T) = \ker_{\infty}(T)$ .
- 5. For two distinct eigenvalues  $\lambda_1$  and  $\lambda$ ,  $\ker(T \lambda_1 I) \perp \ker(T \lambda_2 I)$ .



1. 
$$ker(T) = ker(T^*T) = ker(TT^*) = ker(T^*)$$
.

(1) For any  $\vec{v} \in \ker(T^*T)$ , we have

$$0 = \langle \vec{v}, T^* T \vec{v} \rangle = \langle T \vec{v}, T \vec{v} \rangle.$$

Therefore,  $T\vec{v}=\vec{0}$ , which implies that  $\ker(T)=\ker(T^*T)$ . By the same argument and the fact  $T^{**}=T$ , we also have  $\ker(T^*)=\ker(TT^*)$ . Together with the assumption  $TT^*=T^*T$ ,, we complete the proof.



(2) 
$$\ker(T) = \operatorname{Im}(T)^{\perp}$$
 and  $V = \ker(T) \oplus \operatorname{Im}(T)$ .

For any  $\vec{v} \in V$  and  $\vec{w} \in \operatorname{Im}(T)^{\perp}$ ,

$$\vec{0} = \langle \vec{w}, T(\vec{v}) \rangle = \langle T^* \vec{w}, \vec{v} \rangle$$

which implies that  $\vec{w} \in \ker(T^*) = \ker(T)$ . Therefore,  $\operatorname{Im}(T)^{\perp} \subseteq \ker(T)$ . On the other hand, by dimension formula,

$$\dim \ker(T) = n - \dim \operatorname{Im}(T) = \dim \operatorname{Im}(T)^{\perp}.$$

Therefore,  $Im(T)^{\perp} = \ker(T)$  and

$$V = \operatorname{Im}(T) \oplus \operatorname{Im}(T)^{\perp} = \operatorname{Im}(T) \oplus \ker(T).$$



(3) 
$$\operatorname{Im}(T) = \operatorname{Im}_{\infty}(T)$$
.

Note that

$$Im(T) = T(V) = T\left(\ker(T) \oplus Im(T)\right) = Im(T^2).$$

Therefore, for all  $k \geq 2$ ,

$$\operatorname{Im}(T^k) = T^{k-2} \left( \operatorname{Im}(T^2) \right) = T^{k-2} \left( \operatorname{Im}(T) \right) = \operatorname{Im}(T^{k-1}).$$

We conclude that  $\operatorname{Im}(T) = \operatorname{Im}(T^2) = \operatorname{Im}(T^3) = \cdots = \operatorname{Im}_{\infty}(T)$ .



(4) 
$$\ker(T) = \ker_{\infty}(T)$$
.

By the dimension formula and (3), we have

$$\dim \ker_{\infty}(T) = n - \dim \operatorname{Im}_{\infty}(T) = n - \dim \operatorname{Im}(T) = \dim \ker(T).$$

Together with the fact  $\ker(T) \subseteq \ker_{\infty}(T)$ , we have  $\ker(T) = \ker_{\infty}(T)$ .



(5) For two distinct eigenvalues  $\lambda_1$  and  $\lambda_2$ ,  $\ker(T - \lambda_1 I) \perp \ker(T - \lambda_2 I)$ .

For any  $\vec{v}_1 \in \ker(T - \lambda_1 I)$  and  $\vec{v}_2 \in \ker(T - \lambda_2 I)$ ,

$$\begin{split} \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle &= \langle \lambda_1 \vec{v}_1, \vec{v}_2 \rangle = \langle T \vec{v}_1, \vec{v}_2 \rangle \\ &= \langle \vec{v}_1, T^* \vec{v}_2 \rangle = \langle \vec{v}_1, \bar{\lambda}_2 \vec{v}_2 \rangle = \lambda_2 \langle \vec{v}_1, \vec{v}_2 \rangle. \end{split}$$

Here "=" is followed by the previous theorem. We conclude that  $\langle \vec{v}_1, \vec{v}_2 \rangle = 0$ .

## Normal ⇒ Diagonalizable under Orthonormal Basis



Note that if T is normal,  $T-\lambda I$  is also normal. (Here  $(T-\lambda I)^*=T^*-\bar{\lambda}I$ .) By (4) in the previous proposition, every generalized eigenspace  $\ker_{\infty}(T-\lambda I)$  is equal to the eigenspace  $\ker(T-\lambda I)$ . Therefore, V can be written as a direct sum of eigenspaces.

By (5) in the previous proposition, eigenspaces are mutually orthogonal. Therefore, we can choose an orthonormal basis in each eigenspace and their union will give an orthonormal basis of V which diagonalizes  $\mathcal{T}$ .

# Diagonalizable under Orthonormal Basis $\Rightarrow$ Normal



Let  $\alpha$  be an orthonormal basis such that  $A=\operatorname{Rep}_{\alpha}(T)$  is a diagonal matrix. Then  $A^*=\operatorname{Rep}_{\alpha}(T^*)$  is also a diagonal matrix. Since any two diagonal matrices commute, A and  $A^*$  commute. Therefore T and  $T^*$  commute.