

## 10.2. Wednesday for MAT3040

**Reviewing.** Consider the mapping

$$\begin{aligned}\phi : & \quad V \rightarrow V^* \\ \text{with } & \quad \phi(\mathbf{v}) = \phi_{\mathbf{v}} \\ \text{where } & \quad \phi_{\mathbf{v}}(\mathbf{w}) = \langle \mathbf{v}, \mathbf{w} \rangle\end{aligned}$$

The Riesz Representation Theorem claims that

1.  $\phi$  is a  $\mathbb{R}$ -linear transformation.
2.  $\phi$  is injective.
3. If  $\dim(V) < \infty$ , then  $\phi$  is an isomorphism.

*Proof for Claim (2).* Consider the equality  $\phi(\mathbf{v}) = \phi_{\mathbf{v}} = 0_{V^*}$ , which implies

$$\phi_{\mathbf{v}}(\mathbf{w}) = \langle \mathbf{v}, \mathbf{w} \rangle = 0, \forall \mathbf{w} \in V$$

By the non-degeneracy property,  $\mathbf{v} = 0_{\mathbf{v}}$ , i.e.,  $\phi$  is injective. ■

*Proof for Claim (3).* Since  $\dim_{\mathbb{R}}(V) = \dim_{\mathbb{R}}(V^*)$ , and  $\phi$  is injective as a  $\mathbb{R}$ -linear transformation, we imply  $\phi$  is an isomorphism from  $V$  to  $V^*$ , where  $V, V^*$  are treated as vector spaces over  $\mathbb{R}$ . ■

### 10.2.1. Orthogonal Complement

**Definition 10.5** [Orthogonal Complement] Let  $U \leq V$  be a subspace of an inner product space. Then the **orthogonal complement** of  $U$  is

$$U^{\perp} = \{\mathbf{v} \in V \mid \langle \mathbf{v}, \mathbf{u} \rangle = 0, \forall \mathbf{u} \in U\}$$

The analysis for orthogonal complement for vector spaces over  $\mathbb{C}$  is quite similar as what we have studied in MAT2040.

**Proposition 10.1** 1.  $U^\perp$  is a subspace of  $V$

2.  $U \cap U^\perp = \{0\}$
3.  $U_1 \subseteq U_2$  implies  $U_2^\perp \subseteq U_1^\perp$ .

*Proof.* 1. Suppose that  $\mathbf{v}_1, \mathbf{v}_2 \in U^\perp$ , where  $a, b \in K$  ( $K = \mathbb{C}$  or  $\mathbb{R}$ ), then for all  $\mathbf{u} \in U$ ,

$$\begin{aligned}\langle a\mathbf{v}_1 + b\mathbf{v}_2, \mathbf{u} \rangle &= \bar{a}\langle \mathbf{v}_1, \mathbf{u} \rangle + \bar{b}\langle \mathbf{v}_2, \mathbf{u} \rangle \\ &= \bar{a} \cdot 0 + \bar{b} \cdot 0 = 0\end{aligned}$$

Therefore,  $a\mathbf{v}_1 + b\mathbf{v}_2 \in U^\perp$ .

2. Suppose that  $\mathbf{u} \in U \cap U^\perp$ , then we imply  $\langle \mathbf{u}, \mathbf{u} \rangle = 0$ . By the positive-definiteness of inner product,  $\mathbf{u} = \mathbf{0}$ .
3. The statement (3) is easy.

■

**Proposition 10.2** 1. If  $\dim(V) < \infty$  and  $U \leq V$ , then  $V = U \oplus U^\perp$

2. If  $U, W \leq V$ , then

$$\begin{aligned}(U + W)^\perp &= U^\perp \cap W^\perp \\ (U \cap W)^\perp &\supseteq U^\perp + W^\perp \\ (U^\perp)^\perp &\supseteq U\end{aligned}$$

Moreover, if  $\dim(V) < \infty$ , then these are equalities.

*Proof.* 1. Suppose that  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  forms a basis for  $U$ , and by basis extension, we obtain  $\{\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{v}_{k+1}, \dots, \mathbf{v}_n\}$  is a basis for  $V$ .

By Gram-Schmidt Process, any finite basis induces an orthonormal basis.

Therefore, suppose that  $\{\mathbf{e}_1, \dots, \mathbf{e}_k\}$  forms an orthonormal basis for  $U$ , and  $\{\mathbf{e}_{k+1}, \dots, \mathbf{e}_n\}$  forms an orthonormal basis for  $U^\perp$ .

It's easy to show  $V = U + U^\perp$  using orthonormal basis.

2. (a) The reverse part  $(U + W)^\perp \supseteq U^\perp \cap W^\perp$  is trivial; for the forward part, suppose

$\mathbf{v} \in (U + W)^\perp$ , then

$$\langle \mathbf{v}, \mathbf{u} + \mathbf{w} \rangle = 0, \forall \mathbf{u} \in U, \mathbf{w} \in W$$

Taking  $\mathbf{u} \equiv \mathbf{0}$  in the equality above gives  $\langle \mathbf{v}, \mathbf{w} \rangle = 0$ , i.e.,  $\mathbf{v} \in U^\perp$ . Similarly,  $\mathbf{v} \in W^\perp$ .

- (b) Follow the similar argument as in (2a). If  $\dim(V) < \infty$ , then write down the orthonormal basis for  $U^\perp + W^\perp$  and  $(U \cap W)^\perp$ .
- (c) Follow the similar argument as in (2a). If  $\dim(V) < \infty$ , then

$$V = U^\perp \oplus (U^\perp)^\perp = U \oplus U^\perp.$$

Therefore,  $(U^\perp)^\perp = U$ .

■

**Proposition 10.3** The mapping  $\phi : V \rightarrow V^*$  maps  $U^\perp \leq V$  injectively to  $\text{Ann}(U) \leq V^*$ . If  $\dim(V) < \infty$ , then  $U^\perp \cong \text{Ann}(U)$  as  $\mathbb{R}$ -vector spaces

*Proof.* The injectivity of  $\phi$  has been shown at the beginning of this lecture. For any  $\mathbf{v} \in U^\perp$ , we imply  $\phi_{\mathbf{v}}(\mathbf{u}) = 0, \forall \mathbf{u} \in U$ , i.e.,  $\phi_{\mathbf{v}} \in \text{Ann}(U)$ .

Therefore,  $\phi(U^\perp) \leq \text{Ann}(U)$ .

Provided that  $\dim(V) < \infty$ , by (1) in proposition (10.2),

$$\dim(U) + \dim(U^\perp) = \dim(V)$$

Since  $\dim(U) + \dim(\text{Ann}(U)) = \dim(V)$ , we imply  $\dim(U^\perp) = \dim(\text{Ann}(U))$ .

Moreover,

$$\phi : U^\perp \rightarrow \text{Ann}(U)$$

is an isomorphism between  $\mathbb{R}$ -vector spaces  $U^\perp$  and  $\text{Ann}(U)$ .

■

## 10.2.2. Adjoint Map

**Motivation.** Then we study the induced mapping based on a given linear operator  $T$ , denoted as  $T'$ . This induced mapping essentially plays the similar role as taking the Hermitian for a complex matrix.

**Notation.** Previously we have studied the **adjoint** of  $T : V \rightarrow W$ , denoted as  $T^* : W^* \rightarrow V^*$ . However, from now on, we use the same terminology but with different meaning. If  $T : V \rightarrow V$  is a linear operator, then the **adjoint** of  $T$  is the linear operator  $T^* : W^* \rightarrow V^*$  defined as follows.

**Definition 10.6** [Adjoint] Let  $T : V \rightarrow V$  be a linear operator between inner product spaces. The **adjoint** of  $T$  is defined as  $T' : V \rightarrow V$  satisfying

$$\langle T'(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, T(\mathbf{w}) \rangle, \quad \forall \mathbf{w} \in V \quad (10.1)$$

**Proposition 10.4** If  $\dim(V) < \infty$ , then  $T'$  exists, and it is unique. Moreover,  $T'$  is a linear map.

*Proof.* Fix any  $\mathbf{v} \in V$ . Consider the mapping

$$\alpha_{\mathbf{v}} : \mathbf{w} \xrightarrow{T} T(\mathbf{w}) \xrightarrow{\phi_{\mathbf{v}}} \langle \mathbf{v}, T(\mathbf{w}) \rangle$$

This is a linear transformation from  $V$  to  $\mathbb{F}$ , i.e.,  $\alpha_{\mathbf{v}} \in V^*$

By Riesz representation theorem,  $\phi$  is an isomorphism from  $V$  to  $V^*$ . Moreover,  $\phi(T'(\mathbf{v})) = \alpha_{\mathbf{v}}$ . Therefore, for any  $\alpha_{\mathbf{v}} \in V^*$ , there exists a vector  $T'(\mathbf{v}) \in V$  such that

$$\phi(T'(\mathbf{v})) = \alpha_{\mathbf{v}} \in V^*$$

Or equivalently,  $\phi_{T'(\mathbf{v})}(\mathbf{w}) = \alpha_{\mathbf{v}}(\mathbf{w}), \forall \mathbf{w} \in V$ , i.e.,  $\langle T'(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, T(\mathbf{w}) \rangle$ .

Henceforce, from  $\mathbf{v}$  we have constructed  $T'(\mathbf{v})$  satisfying (10.1). Now define  $T' : V \rightarrow V$  by  $\mathbf{v} \mapsto T'(\mathbf{v})$ .

- Since the choice of  $T'(\mathbf{v})$  is unique by the injectivity of  $\phi$ ,  $T'$  is well-defined.
- Now we show  $T'$  is a linear transformation: Let  $\mathbf{v}_1, \mathbf{v}_2 \in V, a, b \in K$ . For all  $\mathbf{w} \in V$ , we have

$$\begin{aligned}
 \langle T'(a\mathbf{v}_1 + b\mathbf{v}_2), \mathbf{w} \rangle &= \langle a\mathbf{v}_1 + b\mathbf{v}_2, T(\mathbf{w}) \rangle \\
 &= \bar{a}\langle \mathbf{v}_1, T(\mathbf{w}) \rangle + \bar{b}\langle \mathbf{v}_2, T(\mathbf{w}) \rangle \\
 &= \bar{a}\langle T'(\mathbf{v}_1), \mathbf{w} \rangle + \bar{b}\langle T'(\mathbf{v}_2), \mathbf{w} \rangle \\
 &= \langle aT'(\mathbf{v}_1) + bT'(\mathbf{v}_2), \mathbf{w} \rangle
 \end{aligned}$$

Therefore,

$$\langle T'(a\mathbf{v}_1 + b\mathbf{v}_2) - [aT'(\mathbf{v}_1) + bT'(\mathbf{v}_2)], \mathbf{w} \rangle = 0, \forall \mathbf{w} \in V$$

By the non-degeneracy of inner product,

$$T'(a\mathbf{v}_1 + b\mathbf{v}_2) - [aT'(\mathbf{v}_1) + bT'(\mathbf{v}_2)] = \mathbf{0},$$

$$\text{i.e., } T'(a\mathbf{v}_1 + b\mathbf{v}_2) = aT'(\mathbf{v}_1) + bT'(\mathbf{v}_2)$$

■

■ **Example 10.2** Let  $V = \mathbb{R}^n$ ,  $\langle \cdot, \cdot \rangle$  as the usual inner product. Consider the matrix-multiplication mapping

$$T: V \rightarrow V$$

$$T(\mathbf{v}) = A\mathbf{v}$$

Then  $\langle T'(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, T(\mathbf{w}) \rangle$  implies

$$\begin{aligned}
 (T'(\mathbf{v}))^T \mathbf{w} &= \langle \mathbf{v}, A\mathbf{w} \rangle \\
 &= \mathbf{v}^T A\mathbf{w} \\
 &= (A^T \mathbf{v})^T \mathbf{w}
 \end{aligned}$$

Therefore,  $T'(\mathbf{v}) = A^T \mathbf{v}$ .

■

**Proposition 10.5** Let  $T : V \rightarrow V$  be a linear transformation,  $V$  a inner product space.

Suppose that  $\mathcal{B} = \{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  is an orthonormal basis of  $V$ , then

$$(T')_{\mathcal{B}, \mathcal{B}} = \overline{((T)_{\mathcal{B}, \mathcal{B}})^T}$$

*Proof.* Suppose that  $(T)_{\mathcal{B}, \mathcal{B}} = (a_{ij})$ , where  $T(\mathbf{e}_j) = \sum_{k=1}^n a_{kj} \mathbf{e}_k$ , then

$$\begin{aligned} \langle \mathbf{e}_i, T(\mathbf{e}_j) \rangle &= \langle \mathbf{e}_i, \sum_{k=1}^n a_{kj} \mathbf{e}_k \rangle \\ &= \sum_{k=1}^n a_{kj} \langle \mathbf{e}_i, \mathbf{e}_k \rangle \\ &= a_{ij} \end{aligned}$$

Also, suppose  $(T')_{\mathcal{B}, \mathcal{B}} = (b_{ij})$ , we imply  $T'(\mathbf{e}_j) = \sum_{k=1}^n b_{kj} \mathbf{e}_k$ , which follows that

$$\langle \mathbf{e}_i, T'(\mathbf{e}_j) \rangle = b_{ij} \implies \overline{\langle T'(\mathbf{e}_j), \mathbf{e}_i \rangle} = b_{ij} \implies \overline{\langle \mathbf{e}_j, T(\mathbf{e}_i) \rangle} = b_{ij},$$

i.e.,  $\overline{a_{ji}} = b_{ij}$ . ■

**R** Proposition (10.5) does not hold if  $\mathcal{B}$  is not an orthonormal basis.