

# Supplement: Symmetric and Hermitian Matrices

## A Bunch of Definitions

**Definition:** A real  $n \times n$  matrix  $A$  is called *symmetric* if  $A^T = A$ .

**Definition:** A complex  $n \times n$  matrix  $A$  is called *Hermitian* if  $A^* = A$ , where  $A^* = \overline{A^T}$ , the conjugate transpose.

**Definition:** A complex  $n \times n$  matrix  $A$  is called *normal* if  $A^*A = AA^*$ , i.e. commutes with its conjugate transpose.

It is quite a surprising result that these three kinds of matrices are *always diagonalizable*; and moreover, one can construct an orthonormal basis (in standard inner product) for  $\mathbb{R}^n/\mathbb{C}^n$ , consisting of eigenvectors of  $A$ . Hence the matrix  $P$  that gives diagonalization  $A = PDP^{-1}$  will be orthogonal/unitary, namely:

**Definition:** An  $n \times n$  real matrix  $P$  is called *orthogonal* if  $P^T P = I_n$ , i.e.  $P^{-1} = P^T$ .

**Definition:** An  $n \times n$  complex matrix  $P$  is called *unitary* if  $P^* P = I_n$ , i.e.  $P^{-1} = P^*$ .

Diagonalization using these special kinds of  $P$  will have special names:

**Definition:** A matrix  $A$  is called *orthogonally diagonalizable* if  $A$  is similar to a diagonal matrix  $D$  with an orthogonal matrix  $P$ , i.e.  $A = PDP^T$ .

A matrix  $A$  is called *unitarily diagonalizable* if  $A$  is similar to a diagonal matrix  $D$  with a unitary matrix  $P$ , i.e.  $A = PDP^*$ .

Then we have the following big theorems:

**Theorem:** Every real  $n \times n$  symmetric matrix  $A$  is orthogonally diagonalizable

**Theorem:** Every complex  $n \times n$  Hermitian matrix  $A$  is unitarily diagonalizable.

**Theorem:** Every complex  $n \times n$  normal matrix  $A$  is unitarily diagonalizable.

To prove the above results, it is convenient to introduce the concept of *adjoint operator*, which allows us to discuss effectively the “transpose” operation in a general inner product space.

## The Adjoint Operator

Let  $V$  be an  $n$ -dimensional inner product space and let  $T : V \rightarrow V$  be a linear operator. We find out that under the inner product operation, the action of  $T : \mathbf{v} \mapsto T(\mathbf{v})$  can be replaced/represented by another inner product action using a suitably chosen vector.

**Lemma 1:** Let  $\mathbf{w} \in V$  be a given vector. Then there is a unique vector  $\mathbf{w}^* \in V$  such that:

$$\langle T(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, \mathbf{w}^* \rangle, \quad \text{for every } \mathbf{v} \in V. \quad (*)$$

**Proof:** Let  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be an orthonormal basis for  $V$ . The following  $\mathbf{w}^*$  is what we want:

$$\mathbf{w}^* = \overline{\langle T(\mathbf{u}_1), \mathbf{w} \rangle} \mathbf{u}_1 + \dots + \overline{\langle T(\mathbf{u}_p), \mathbf{w} \rangle} \mathbf{u}_p = \sum_{i=1}^n \overline{\langle T(\mathbf{u}_i), \mathbf{w} \rangle} \mathbf{u}_i.$$

Now, for  $j = 1, \dots, n$ , we check  $(*)$  for basis vector  $\mathbf{u}_j$  first:

$$\begin{aligned} \langle \mathbf{u}_j, \mathbf{w}^* \rangle &= \langle \mathbf{u}_j, \sum_{i=1}^n \overline{\langle T(\mathbf{u}_i), \mathbf{w} \rangle} \mathbf{u}_i \rangle = \sum_{i=1}^n \langle T(\mathbf{u}_i), \mathbf{w} \rangle \langle \mathbf{u}_j, \mathbf{u}_i \rangle \\ &= \langle T(\mathbf{u}_j), \mathbf{w} \rangle \langle \mathbf{u}_j, \mathbf{u}_j \rangle = \langle T(\mathbf{u}_j), \mathbf{w} \rangle. \end{aligned}$$

So, for a general  $\mathbf{v} \in V$ , by expressing  $\mathbf{v} = c_1 \mathbf{u}_1 + \dots + c_n \mathbf{u}_n = \sum_{j=1}^n c_j \mathbf{u}_j$ , we have:

$$\begin{aligned} \langle T(\mathbf{v}), \mathbf{w} \rangle &= \langle \sum_{j=1}^n c_j T(\mathbf{u}_j), \mathbf{w} \rangle = \sum_{j=1}^n c_j \langle T(\mathbf{u}_j), \mathbf{w} \rangle \\ &= \sum_{j=1}^n c_j \langle \mathbf{u}_j, \mathbf{w}^* \rangle = \langle \sum_{j=1}^n c_j \mathbf{u}_j, \mathbf{w}^* \rangle = \langle \mathbf{v}, \mathbf{w}^* \rangle. \end{aligned}$$

For the uniqueness of  $\mathbf{w}^*$ , let  $\mathbf{w}' \in V$  be another vector with the same property, namely:

$$\langle T(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, \mathbf{w}^* \rangle = \langle \mathbf{v}, \mathbf{w}' \rangle, \quad \text{for every } \mathbf{v} \in V.$$

Then we take difference:

$$\langle \mathbf{v}, \mathbf{w}^* - \mathbf{w}' \rangle = 0, \quad \text{for every } \mathbf{v} \in V.$$

In particular, this equality should be valid for  $\mathbf{v} = \mathbf{w}^* - \mathbf{w}' \in V$ . Thus we have:

$$\langle \mathbf{w}^* - \mathbf{w}', \mathbf{w}^* - \mathbf{w}' \rangle = 0 \quad \Rightarrow \quad \|\mathbf{w}^* - \mathbf{w}'\| = 0 \quad \Rightarrow \quad \mathbf{w}^* = \mathbf{w}' \quad \square$$

**Definition:** Let  $T : V \rightarrow V$  be a linear operator. For each  $\mathbf{w} \in V$ , we define  $T^*(\mathbf{w}) := \mathbf{w}^*$ , where  $\mathbf{w}^*$  is the unique vector obtained in Lemma 1. This  $T^*$  is called the *adjoint* of  $T$ .

**Lemma 2:** The adjoint operator  $T^* : V \rightarrow V$  is linear.

**Proof:** Straightforward checking. Let  $\mathbf{w}_1, \mathbf{w}_2 \in V$  and  $c, d \in \mathbf{C}$ . Then for every  $\mathbf{v} \in V$ , first by definition of  $T^*$  we have:

$$\langle T(\mathbf{v}), (c\mathbf{w}_1 + d\mathbf{w}_2) \rangle = \langle \mathbf{v}, T^*(c\mathbf{w}_1 + d\mathbf{w}_2) \rangle.$$

But on the other hand:

$$\begin{aligned} \langle T(\mathbf{v}), (c\mathbf{w}_1 + d\mathbf{w}_2) \rangle &= \bar{c} \langle T(\mathbf{v}), \mathbf{w}_1 \rangle + \bar{d} \langle T(\mathbf{v}), \mathbf{w}_2 \rangle \\ &= \bar{c} \langle \mathbf{v}, T^*(\mathbf{w}_1) \rangle + \bar{d} \langle \mathbf{v}, T^*(\mathbf{w}_2) \rangle \\ &= \langle \mathbf{v}, cT^*(\mathbf{w}_1) + dT^*(\mathbf{w}_2) \rangle \end{aligned}$$

The above two equalities are valid for every  $\mathbf{v} \in V$ . So by the same uniqueness proof as in Lemma 1, we obtain:

$$T^*(c\mathbf{w}_1 + d\mathbf{w}_2) = cT^*(\mathbf{w}_1) + dT^*(\mathbf{w}_2),$$

and thus  $T^*$  is linear.  $\square$

**Theorem 1:** Let  $T, U$  be linear operators on  $V$  and  $k \in \mathbb{C}$ . Then:

- (i)  $(T + U)^* = T^* + U^*$ ;
- (ii)  $(kT)^* = \bar{k}T^*$ ;
- (iii)  $(U \circ T)^* = T^* \circ U^*$ ;
- (iv)  $(T^*)^* = T$ .

**Proof:** Directly from definitions. For example, the checking for (iv):

Let  $\mathbf{v} \in V$  be any vector. Then by definition:

$$\langle (T^*)^*(\mathbf{v}), \mathbf{u} \rangle = \langle \mathbf{v}, T^*(\mathbf{u}) \rangle = \langle T(\mathbf{v}), \mathbf{u} \rangle, \quad \text{for every } \mathbf{u} \in V.$$

Hence  $(T^*)^*(\mathbf{v}) = T(\mathbf{v})$  for any  $\mathbf{v} \in V$  and thus  $(T^*)^* = T$ . □

This adjoint operator  $T^*$ , when using matrix representation with an orthonormal basis  $\mathcal{B}$ , has a simple relationship with the original linear operator  $T$ .

**Theorem 2:** Let  $\mathcal{B} = \{\mathbf{u}_1, \dots, \mathbf{u}_p\}$  be an orthonormal basis of  $V$ , and let  $T$  be a linear operator in  $V$ . Then the matrix representations of  $T$  and  $T^*$  relative to the orthonormal basis  $\mathcal{B}$  are given by:

$$[T]_{\mathcal{B}} = \left[ \langle T(\mathbf{u}_j), \mathbf{u}_i \rangle \right] \quad \text{and} \quad [T^*]_{\mathcal{B}} = [T]_{\mathcal{B}}^*.$$

**Remark:**  $\mathcal{B}$  must be orthonormal!

**Proof:** First we consider the  $j$ -th column of  $[T]_{\mathcal{B}}$ , i.e.  $[T(\mathbf{u}_j)]_{\mathcal{B}}$ . Its entries are the  $\mathcal{B}$ -coordinates of  $T(\mathbf{u}_j)$ , which are exactly the coefficients in the linear combination:

$$T(\mathbf{u}_j) = a_{1j}\mathbf{u}_1 + \dots + a_{nj}\mathbf{u}_j.$$

Since  $\mathcal{B}$  is orthonormal, the  $i$ -th coefficient in the above linear combination can be computed effectively as:

$$\langle T(\mathbf{u}_j), \mathbf{u}_i \rangle = a_{1j} \langle \mathbf{u}_1, \mathbf{u}_i \rangle + \dots + a_{nj} \langle \mathbf{u}_n, \mathbf{u}_i \rangle = a_{ij}.$$

Thus the  $(i, j)$ -th entry of  $[T]_{\mathcal{B}}$  is given by  $a_{ij} = \langle T(\mathbf{u}_j), \mathbf{u}_i \rangle$ .

Similarly the  $(i, j)$ -th entry of  $[T^*]_{\mathcal{B}}$  is given by  $\langle T^*(\mathbf{u}_j), \mathbf{u}_i \rangle$ . Using the definition of adjoint operator, we have:

$$\langle T^*(\mathbf{u}_j), \mathbf{u}_i \rangle = \overline{\langle \mathbf{u}_i, T^*(\mathbf{u}_j) \rangle} = \overline{\langle T(\mathbf{u}_i), \mathbf{u}_j \rangle} = \bar{a}_{ji}.$$

So  $[T^*]_{\mathcal{B}} = [T]_{\mathcal{B}}^*$  □

**Definition:** A linear operator  $T : V \rightarrow V$  is called *self-adjoint* if  $T^* = T$ .

Thus, by Theorem 2, matrix transformation given by a symmetric/Hermitian matrix will be a self-adjoint operator on  $\mathbb{R}^n/\mathbb{C}^n$ , using the standard inner product.

Next we need to setup some technical lemmas for the proof of the main theorem.

**Lemma 3:** Let  $T$  be a self-adjoint operator on  $V$ . Then every eigenvalue of  $T$  must be real.

**Proof:** Let  $\mathbf{v} \neq \mathbf{0}$  be an eigenvector of  $T$  corresponding to eigenvalue  $\lambda$ . We consider:

$$\langle T(\mathbf{v}), \mathbf{v} \rangle = \langle \lambda \mathbf{v}, \mathbf{v} \rangle = \lambda \langle \mathbf{v}, \mathbf{v} \rangle.$$

On the other hand, since  $T^* = T$ , we also have:

$$\langle T(\mathbf{v}), \mathbf{v} \rangle = \langle \mathbf{v}, T^*(\mathbf{v}) \rangle = \langle \mathbf{v}, T(\mathbf{v}) \rangle = \langle \mathbf{v}, \lambda \mathbf{v} \rangle = \bar{\lambda} \langle \mathbf{v}, \mathbf{v} \rangle.$$

As  $\langle \mathbf{v}, \mathbf{v} \rangle \neq 0$ , we must have  $\lambda = \bar{\lambda}$ , i.e.  $\lambda$  is real.  $\square$

**Lemma 4:** Every self-adjoint operator on  $V$  has an eigenvector.

**Proof:** Take an orthonormal basis  $\mathcal{B}$  of  $V$ . Then we get a symmetric/Hermitian matrix  $A = [T]_{\mathcal{B}}$ . By the fundamental theorem of algebra,  $A$  must have an eigenvalue  $\lambda \in \mathbb{C}$ , and hence a corresponding eigenvector  $\mathbf{x} \in \mathbb{C}^n$ . In complex case we just send this  $\mathbf{x} \in \mathbb{C}^n$  back to  $\mathbf{v} \in V$  by inverse  $\mathcal{B}$ -coordinate mapping, then we will get  $T(\mathbf{v}) = \lambda \mathbf{v}$ . In real case, we apply Lemma 3 to know that this  $\lambda$  must be real. Hence  $\mathbf{x} \in \mathbb{R}^n$  and we can send it back to  $\mathbf{v} \in V$  to get  $T(\mathbf{v}) = \lambda \mathbf{v}$  again.  $\square$

**Lemma 5:** Let  $W$  be a subspace of  $V$  such that  $T(W) \subseteq W$ , i.e.  $T(\mathbf{w}) \in W$  for every  $\mathbf{w} \in W$ . Then  $T^*(W^\perp) \subseteq W^\perp$ .

**Proof:** Let  $\mathbf{z} \in W^\perp$ . Then for  $\mathbf{w} \in W$ :

$$\langle \mathbf{w}, T^*(\mathbf{z}) \rangle = \langle T(\mathbf{w}), \mathbf{z} \rangle = 0 \quad \text{as } T(\mathbf{w}) \in W \text{ and } \mathbf{z} \in W^\perp.$$

Since the above is valid for every  $\mathbf{w} \in W$ , we should have  $T^*(\mathbf{z}) \in W^\perp$ .  $\square$

**Lemma 6:** Let  $W$  be a subspace of an  $n$ -dimensional inner product space  $V$ . Then:

$$\dim W + \dim W^\perp = n = \dim V.$$

**Proof:** Let  $\{\mathbf{w}_1, \dots, \mathbf{w}_k\}$  and  $\{\mathbf{z}_1, \dots, \mathbf{z}_\ell\}$  be orthogonal bases of  $W$  and  $W^\perp$  respectively. The lemma is proved if we can show that  $S = \{\mathbf{w}_1, \dots, \mathbf{w}_k, \mathbf{z}_1, \dots, \mathbf{z}_\ell\}$  forms a basis for  $V$ .

*Spanning  $V$ :* For every  $\mathbf{v} \in V$ , we have the orthogonal decomposition of  $\mathbf{v}$  w.r.t.  $W$ :

$$\mathbf{v} = \text{proj}_W \mathbf{v} + (\mathbf{v} - \text{proj}_W \mathbf{v}), \quad \text{where } \text{proj}_W \mathbf{v} \in W \text{ and } (\mathbf{v} - \text{proj}_W \mathbf{v}) \in W^\perp.$$

Use the bases of  $W$  and  $W^\perp$  to express  $\text{proj}_W \mathbf{v} = \sum_{i=1}^k c_i \mathbf{w}_i$  and  $(\mathbf{v} - \text{proj}_W \mathbf{v}) = \sum_{j=1}^\ell d_j \mathbf{z}_j$ . Hence  $\mathbf{v}$  can be expressed as a linear combination of vectors in  $S$ .

*Linearly independent:* Consider the vector equation:

$$c_1 \mathbf{w}_1 + \dots + c_k \mathbf{w}_k + d_1 \mathbf{z}_1 + \dots + d_\ell \mathbf{z}_\ell = \mathbf{0}.$$

Take inner product with  $\mathbf{w}_1$ . As  $\{\mathbf{w}_1, \dots, \mathbf{w}_k\}$  is an orthogonal set, we have  $\langle \mathbf{w}_i, \mathbf{w}_1 \rangle = 0$  for  $i \neq 1$ . On the other hand, since  $\mathbf{w}_1 \in W$  and all  $\mathbf{z}_j \in W^\perp$ , we get  $\langle \mathbf{z}_j, \mathbf{w}_1 \rangle = 0$  for all  $1 \leq j \leq \ell$ . So the above vector equation will become:

$$c_1 \|\mathbf{w}_1\|^2 + 0 + \dots + 0 = \langle \mathbf{0}, \mathbf{w}_1 \rangle = 0.$$

As  $\mathbf{w}_1 \neq \mathbf{0}$ , we get  $c_1 = 0$ . Similarly for other  $c_i$  and  $d_j$  and they are all zeros. Thus  $S$  is also linearly independent.  $\square$

Now we are ready to prove the main theorem.

### Diagonalizability of Symmetric and Hermitian Matrices

**Main Theorem:** Let  $T^* = T$  be a self-adjoint linear operator on  $V$ . Then  $V$  has an orthonormal basis consisting of eigenvectors of  $T$ .

**Proof:** We use induction on  $n = \dim V$ .

$n = 1$ : Any non-zero vector  $\mathbf{v}_1$  will be an eigenvector of  $T$  since  $V = \text{Span}\{\mathbf{v}_1\}$ . After normalization,  $\mathbf{u}_1 = \frac{\mathbf{v}_1}{\|\mathbf{v}_1\|}$ , we obtain an orthonormal basis  $\{\mathbf{u}_1\}$  of  $V$  consisting of eigenvector of  $T$ .

Now, assume the statement is true for  $\dim V = k$ . Next consider  $\dim V = k + 1$ .

By Lemma 4,  $T$  has an eigenvector  $\mathbf{u}_1$  (may assume  $\|\mathbf{u}_1\| = 1$ ) corresponding to eigenvalue  $\lambda_1$ . Let  $W = \text{Span}\{\mathbf{u}_1\}$ . Note that  $T(W) = W$ .

By Lemma 5, we have  $T^*(W^\perp) \subseteq W^\perp$ . Since  $T^* = T$ , this gives  $T(W^\perp) \subseteq W^\perp$ . In other words, we can regard  $T$  as a linear operator defined on  $W^\perp$ . Note that Lemma 6 says that  $\dim W^\perp = \dim V - \dim W = k$ , so by induction hypothesis, there is an orthonormal basis of  $W^\perp$  consisting of eigenvectors of  $T$ , say  $\{\mathbf{u}_2, \dots, \mathbf{u}_{k+1}\}$ .

Since  $\mathbf{u}_1 \in W$ ,  $\|\mathbf{u}_1\| = 1$ , and  $\{\mathbf{u}_2, \dots, \mathbf{u}_{k+1}\} \subset W^\perp$ , the combined set  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_{k+1}\}$  is again orthonormal. This will be an orthonormal basis of  $V$  consisting of eigenvectors of  $T$ .  $\square$

In the case of symmetric (or Hermitian) matrix transformation, by using such an orthonormal basis of eigenvectors to construct the matrix  $P$ , we will have the diagonalization  $A = PDP^{-1}$  with  $P^{-1} = P^T$  (or  $P^{-1} = P^*$ ).

**Remark:** To find this  $P$ , we have a more efficient method than the inductive construction in the proof of main theorem.

**Lemma 7:** Let  $T^* = T$ . Then eigenvectors of  $T$  corresponding to distinct eigenvalues are orthogonal to each other.

**Proof:** Let  $T(\mathbf{v}_1) = \lambda_1 \mathbf{v}_1$  and  $T(\mathbf{v}_2) = \lambda_2 \mathbf{v}_2$  with  $\lambda_1 \neq \lambda_2$ . Consider on the one hand:

$$\langle T(\mathbf{v}_1), \mathbf{v}_2 \rangle = \langle \lambda_1 \mathbf{v}_1, \mathbf{v}_2 \rangle = \lambda_1 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle,$$

and on the other hand:

$$\langle T(\mathbf{v}_1), \mathbf{v}_2 \rangle = \langle \mathbf{v}_1, T^*(\mathbf{v}_2) \rangle = \langle \mathbf{v}_1, T(\mathbf{v}_2) \rangle = \langle \mathbf{v}_1, \lambda_2 \mathbf{v}_2 \rangle = \bar{\lambda}_2 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle.$$

Since  $T$  is self-adjoint,  $\lambda_2$  must be real, so we obtain:

$$\lambda_1 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle = \lambda_2 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle.$$

As  $\lambda_1 \neq \lambda_2$ , we must have  $\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = 0$ .  $\square$

**Corollary:** Let  $T^* = T$  and let  $\{\mathbf{v}_{1i_1}\}, \dots, \{\mathbf{v}_{pi_p}\}$  be orthogonal sets of eigenvectors corresponding to distinct eigenvalues  $\lambda_1, \dots, \lambda_p$  of  $T$ . Then the total collection of eigenvectors  $\{\mathbf{v}_{ji_j}; 1 \leq i \leq p\}$  is again orthogonal.

**Proof:** Exercise.

With Lemma 7 and its corollary, we only need to produce orthonormal basis for each eigenspace, which can be done by a Gram-Schmidt process. Then the total collection will be automatically orthonormal. And it is guaranteed by the main theorem that  $A$  must be diagonalizable.

**Remark:** If  $\mathbf{v}_1, \mathbf{v}_2$  are eigenvectors of  $A$  corresponding to distinct eigenvalues, we know that  $\mathbf{v}_1 + \mathbf{v}_2$  can never be an eigenvector of  $A$ . So Gram-Schmidt process should not be applied across bases for different eigenspaces.

**Example:** Orthogonally diagonalize the following symmetric matrix:

$$A = \begin{bmatrix} 1 & 2 & -4 \\ 2 & -2 & -2 \\ -4 & -2 & 1 \end{bmatrix}.$$

**Solution:** The characteristic equation of  $A$  is:

$$\det(A - \lambda I) = -\lambda^3 + 27\lambda + 54 = -(\lambda + 3)^2(\lambda - 6) = 0.$$

So the eigenvalues are  $-3, -3, 6$ .

For the eigenvalue  $\lambda = -3$ , we solve for  $\text{Nul}(A + 3I)$ :

$$A + 3I = \begin{bmatrix} 4 & 2 & -4 \\ 2 & 1 & -2 \\ -4 & -2 & 4 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & \frac{1}{2} & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

So  $\text{Nul}(A + 3I)$  has a basis  $\{[1 \ 0 \ 1]^T, [-\frac{1}{2} \ 1 \ 0]^T\}$ . By Gram-Schmidt process, we obtain an orthonormal basis for  $\text{Nul}(A + 3I)$ :

$$\left\{ \begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \end{bmatrix}, \begin{bmatrix} -\frac{1}{\sqrt{18}} \\ \frac{4}{\sqrt{18}} \\ \frac{1}{\sqrt{18}} \end{bmatrix} \right\}.$$

For the eigenvalue  $\lambda = 6$ , we solve for  $\text{Nul}(A - 6I)$ :

$$A - 6I = \begin{bmatrix} -5 & 2 & -4 \\ 2 & -8 & -2 \\ -4 & -2 & -5 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 \end{bmatrix}.$$

So  $\text{Nul}(A - 6I)$  has a basis  $\{[-1 \ -\frac{1}{2} \ 1]^T\}$  and we obtain an orthonormal basis for  $\text{Nul}(A - 6I)$ :

$$\left\{ \begin{bmatrix} -\frac{2}{3} \\ \frac{1}{3} \\ \frac{2}{3} \end{bmatrix} \right\}.$$

We construct the orthogonal matrix  $P$  and diagonal matrix  $D$  as:

$$P = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{18}} & -\frac{2}{3} \\ 0 & \frac{4}{\sqrt{18}} & \frac{1}{3} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{18}} & \frac{2}{3} \end{bmatrix}, \quad D = \begin{bmatrix} -3 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 0 & 6 \end{bmatrix}.$$

Then one can check that  $A = PDP^T$ .

**Note:** The diagonalization  $A = PDP^T$  is not unique, as one can have different choices of orthonormal bases for those eigenspaces with dimension greater than one. For example, the above  $A$  also allows an orthogonal diagonalization  $A = QDQ^T$  with:

$$Q = \begin{bmatrix} \frac{1}{3} & -\frac{2}{3} & -\frac{2}{3} \\ \frac{2}{3} & \frac{2}{3} & -\frac{1}{3} \\ \frac{2}{3} & -\frac{1}{3} & \frac{2}{3} \end{bmatrix}.$$

### Diagonalization of Complex Normal Matrices

**Definition:** A linear operator  $T$  on  $V$  is called *normal* if  $T \circ T^* = T^* \circ T$ .

To make the proof of main theorem also work for normal operator, we need the following technical lemma.

**Lemma 8:** Let  $T$  be a normal operator on  $V$ . Then:

- (i)  $\mathbf{v}$  is an eigenvector of  $T$  corresponding to eigenvalue  $\lambda$   
 $\Leftrightarrow \mathbf{v}$  is an eigenvector of  $T^*$  corresponding to eigenvalue  $\bar{\lambda}$ .
- (ii) Eigenvectors corresponding to distinct eigenvalues of  $T$  are orthogonal to each other.

**Proof:** (i) First we claim that  $\|T(\mathbf{v})\| = \|T^*(\mathbf{v})\|$ .

$$\begin{aligned} \|T(\mathbf{v})\|^2 &= \langle T(\mathbf{v}), T(\mathbf{v}) \rangle = \langle \mathbf{v}, T^*T(\mathbf{v}) \rangle \\ &= \langle \mathbf{v}, TT^*(\mathbf{v}) \rangle = \langle T^*(\mathbf{v}), T^*(\mathbf{v}) \rangle = \|T^*(\mathbf{v})\|^2. \end{aligned}$$

Then for any scalar  $\lambda$ , note that the operator  $U = T - \lambda I$  is also normal with  $U^* = T^* - \bar{\lambda}I$ , so we have:

$$\|(T - \lambda I)(\mathbf{v})\| = \|(T^* - \bar{\lambda}I)(\mathbf{v})\|. \quad (*)$$

Hence:

$$\begin{aligned} &\mathbf{v} \text{ is an eigenvector of } T \text{ corresponding to eigenvalue } \lambda \\ \Leftrightarrow &(T - \lambda I)(\mathbf{v}) = \mathbf{0} \\ \Leftrightarrow &(T^* - \bar{\lambda}I)(\mathbf{v}) = \mathbf{0} \quad (\text{by } (*)) \\ \Leftrightarrow &\mathbf{v} \text{ is an eigenvector of } T^* \text{ corresponding to eigenvalue } \bar{\lambda} \end{aligned}$$

(ii) Now let  $\mathbf{v}_1, \mathbf{v}_2$  be eigenvectors of  $T$ , corresponding to distinct eigenvalues  $\lambda_1 \neq \lambda_2$  respectively. Consider on the one hand:

$$\langle T(\mathbf{v}_1), \mathbf{v}_2 \rangle = \langle \lambda_1 \mathbf{v}_1, \mathbf{v}_2 \rangle = \lambda_1 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle;$$

and on the other hand:

$$\langle T(\mathbf{v}_1), \mathbf{v}_2 \rangle = \langle \mathbf{v}_1, T^*(\mathbf{v}_2) \rangle = \langle \mathbf{v}_1, \bar{\lambda}_2 \mathbf{v}_2 \rangle = \bar{\lambda}_2 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle.$$

So we again obtain:

$$\lambda_1 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle = \bar{\lambda}_2 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle.$$

As  $\lambda_1 \neq \lambda_2$ , we must have  $\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = 0$ . □

Now, we give the proof of main theorem for normal operators.

**Main Theorem':** Let  $T$  be a normal operator on a complex inner product space  $V$ . Then  $V$  has an orthonormal basis consisting of eigenvectors of  $T$ .

**Proof:** We use induction on  $n = \dim V$ .

$n = 1$ : Same as before.

Now, assume the statement is true for  $\dim V = k$ . Next consider  $\dim V = k + 1$ .

Since  $V$  is a complex inner product space,  $T$  will have an eigenvector  $\mathbf{u}_1$  (may assume  $\|\mathbf{u}_1\| = 1$ ) corresponding to eigenvalue  $\lambda_1$ . (*For real inner product space we might get stuck at this point.*)

By Lemma 8(i),  $\mathbf{u}_1$  is also an eigenvector of  $T^*$ . So if we set  $W = \text{Span}\{\mathbf{u}_1\}$ , we have  $T^*(W) \subseteq W$ .

By Lemma 5, we have  $(T^*)^*(W^\perp) \subseteq W^\perp$ . As  $(T^*)^* = T$ , this means  $T(W^\perp) \subseteq W^\perp$ . Then we can continue the inductive argument as in the previous proof of Main Theorem. □