

## Chapter 9

# Spectral Theorems in Euclidean and Hermitian Spaces

### 9.1 Normal Linear Maps

Let  $E$  be a real Euclidean space (or a complex Hermitian space) with inner product  $u, v \mapsto \langle u, v \rangle$ .

In the real Euclidean case, recall that  $\langle -, - \rangle$  is bilinear, symmetric and positive definite (i.e.,  $\langle u, u \rangle > 0$  for all  $u \neq 0$ ).

In the complex Hermitian case, recall that  $\langle -, - \rangle$  is sesquilinear, which means that it linear in the first argument, semilinear in the second argument (i.e.,  $\langle u, \mu v \rangle = \bar{\mu} \langle u, v \rangle$ ),  $\langle v, u \rangle = \overline{\langle u, v \rangle}$ , and positive definite (as above).

In both cases we let  $\|u\| = \sqrt{\langle u, u \rangle}$  and the map  $u \mapsto \|u\|$  is a *norm*.

Recall that every linear map,  $f: E \rightarrow E$ , has an *adjoint*  $f^*$  which is a linear map,  $f^*: E \rightarrow E$ , such that

$$\langle f(u), v \rangle = \langle u, f^*(v) \rangle,$$

for all  $u, v \in E$ .

Since  $\langle -, - \rangle$  is symmetric, it is obvious that  $f^{**} = f$ .

**Definition 9.1.1** Given a Euclidean (or Hermitian) space,  $E$ , a linear map  $f: E \rightarrow E$  is *normal* iff

$$f \circ f^* = f^* \circ f.$$

A linear map  $f: E \rightarrow E$  is *self-adjoint* if  $f = f^*$ , *skew self-adjoint* if  $f = -f^*$ , and *orthogonal* if  $f \circ f^* = f^* \circ f = \text{id}$ .

Our first goal is to show that for every *normal* linear map  $f: E \rightarrow E$  (where  $E$  is a Euclidean space), there is an *orthonormal basis* (w.r.t.  $\langle -, - \rangle$ ) such that the matrix of  $f$  over this basis has an especially nice form:

It is a *block diagonal matrix* in which the blocks are either one-dimensional matrices (i.e., single entries) or two-dimensional matrices of the form

$$\begin{pmatrix} \lambda & \mu \\ -\mu & \lambda \end{pmatrix}$$

This normal form can be further refined if  $f$  is self-adjoint, skew self-adjoint, or orthogonal.

As a first step, we show that  $f$  and  $f^*$  have the same kernel when  $f$  is normal.

**Lemma 9.1.2** *Given a Euclidean space  $E$ , if  $f: E \rightarrow E$  is a normal linear map, then  $\text{Ker } f = \text{Ker } f^*$ .*

The next step is to show that for *every linear map*  $f: E \rightarrow E$ , there is some subspace  $W$  of dimension 1 or 2 such that  $f(W) \subseteq W$ .

When  $\dim(W) = 1$ ,  $W$  is actually an eigenspace for some real eigenvalue of  $f$ .

Furthermore, when  $f$  is normal, there is a subspace  $W$  of dimension 1 or 2 such that  $f(W) \subseteq W$  and  $f^*(W) \subseteq W$ .

The difficulty is that the eigenvalues of  $f$  are not necessarily real. One way to get around this problem is to *complexify* both the vector space  $E$  and the inner product  $\langle -, - \rangle$ .

First, we need to embed a real vector space  $E$  into a complex vector space  $E_{\mathbb{C}}$ .

A fancy way to define  $E_{\mathbb{C}}$  is to use tensor products and to set

$$E_{\mathbb{C}} = \mathbb{C} \otimes_{\mathbb{R}} E.$$

However, we can also define  $E_{\mathbb{C}}$  directly as follows:

**Definition 9.1.3** Given a real vector space  $E$ , let  $E_{\mathbb{C}}$  be the structure  $E \times E$  under the addition operation

$$(u_1, u_2) + (v_1, v_2) = (u_1 + v_1, u_2 + v_2),$$

and multiplication by a complex scalar  $z = x + iy$  defined such that

$$(x + iy) \cdot (u, v) = (xu - yv, yu + xv).$$

It is easily shown that the structure  $E_{\mathbb{C}}$  is a complex vector space.

It is also immediate that

$$(0, v) = i(v, 0),$$

and thus, identifying  $E$  with the subspace of  $E_{\mathbb{C}}$  consisting of all vectors of the form  $(u, 0)$ , we can write

$$(u, v) = u + iv.$$

Given a vector  $w = u + iv$ , its *conjugate*  $\bar{w}$  is the vector  $\bar{w} = u - iv$ .

Given a linear map  $f: E \rightarrow E$ , the map  $f$  can be extended to a linear map  $f_{\mathbb{C}}: E_{\mathbb{C}} \rightarrow E_{\mathbb{C}}$  defined such that

$$f_{\mathbb{C}}(u + iv) = f(u) + if(v).$$

Next, we need to extend the inner product on  $E$  to an inner product on  $E_{\mathbb{C}}$ .

The inner product  $\langle -, - \rangle$  on a Euclidean space  $E$  is extended to the Hermitian positive definite form  $\langle -, - \rangle_{\mathbb{C}}$  on  $E_{\mathbb{C}}$  as follows:

$$\begin{aligned} \langle u_1 + iv_1, u_2 + iv_2 \rangle_{\mathbb{C}} \\ = \langle u_1, u_2 \rangle + \langle v_1, v_2 \rangle + i(\langle u_2, v_1 \rangle - \langle u_1, v_2 \rangle). \end{aligned}$$

Then, given any linear map  $f: E \rightarrow E$ , it is easily verified that the map  $f_{\mathbb{C}}^*$  defined such that

$$f_{\mathbb{C}}^*(u + iv) = f^*(u) + if^*(v)$$

for all  $u, v \in E$ , is the *adjoint* of  $f_{\mathbb{C}}$  w.r.t.  $\langle -, - \rangle_{\mathbb{C}}$ .

Assuming again that  $E$  is a Hermitian space, observe that Lemma 9.1.2 also holds.

**Lemma 9.1.4** *Given a Hermitian space  $E$ , for any normal linear map  $f: E \rightarrow E$ , a vector  $u$  is an eigenvector of  $f$  for the eigenvalue  $\lambda$  (in  $\mathbb{C}$ ) iff  $u$  is an eigenvector of  $f^*$  for the eigenvalue  $\bar{\lambda}$ .*

The next lemma shows a very important property of normal linear maps: eigenvectors corresponding to distinct eigenvalues are orthogonal.

**Lemma 9.1.5** *Given a Hermitian space  $E$ , for any normal linear map  $f: E \rightarrow E$ , if  $u$  and  $v$  are eigenvectors of  $f$  associated with the eigenvalues  $\lambda$  and  $\mu$  (in  $\mathbb{C}$ ) where  $\lambda \neq \mu$ , then  $\langle u, v \rangle = 0$ .*

We can also show easily that the eigenvalues of a self-adjoint linear map are real.

**Lemma 9.1.6** *Given a Hermitian space  $E$ , the eigenvalues of any self-adjoint linear map  $f: E \rightarrow E$  are real.*

Given any subspace  $W$  of a Hermitian space  $E$ , recall that the *orthogonal*  $W^\perp$  of  $W$  is the subspace defined such that

$$W^\perp = \{u \in E \mid \langle u, w \rangle = 0, \text{ for all } w \in W\}.$$

Recall that  $E = W \oplus W^\perp$  (construct an orthonormal basis of  $E$  using the Gram–Schmidt orthonormalization procedure). The same result also holds for Euclidean spaces.



The following lemma provides the key to the induction that will allow us to show that a normal linear map can be diagonalized. It actually holds for any linear map.

**Lemma 9.1.7** *Given a Hermitian space  $E$ , for any linear map  $f: E \rightarrow E$ , if  $W$  is any subspace of  $E$  such that  $f(W) \subseteq W$  and  $f^*(W) \subseteq W$ , then  $f(W^\perp) \subseteq W^\perp$  and  $f^*(W^\perp) \subseteq W^\perp$ .*

The above Lemma *also holds for Euclidean spaces*. Although we are ready to prove that for every normal linear map  $f$  (over a Hermitian space) there is an orthonormal basis of eigenvectors, we now return to real Euclidean spaces.

If  $f: E \rightarrow E$  is a linear map and  $w = u + iv$  is an eigenvector of  $f_{\mathbb{C}}: E_{\mathbb{C}} \rightarrow E_{\mathbb{C}}$  for the eigenvalue  $z = \lambda + i\mu$ , where  $u, v \in E$  and  $\lambda, \mu \in \mathbb{R}$ , since

$$f_{\mathbb{C}}(u + iv) = f(u) + if(v)$$

and

$$\begin{aligned} f_{\mathbb{C}}(u + iv) &= (\lambda + i\mu)(u + iv) \\ &= \lambda u - \mu v + i(\mu u + \lambda v), \end{aligned}$$

we have

$$f(u) = \lambda u - \mu v \quad \text{and} \quad f(v) = \mu u + \lambda v,$$

from which we immediately obtain

$$f_{\mathbb{C}}(u - iv) = (\lambda - i\mu)(u - iv),$$

which shows that  $\bar{w} = u - iv$  is an eigenvector of  $f_{\mathbb{C}}$  for  $\bar{z} = \lambda - i\mu$ . Using this fact, we can prove the following lemma:

**Lemma 9.1.8** *Given a Euclidean space  $E$ , for any normal linear map  $f: E \rightarrow E$ , if  $w = u + iv$  is an eigenvector of  $f_{\mathbb{C}}$  associated with the eigenvalue  $z = \lambda + i\mu$  (where  $u, v \in E$  and  $\lambda, \mu \in \mathbb{R}$ ), if  $\mu \neq 0$  (i.e.,  $z$  is not real) then  $\langle u, v \rangle = 0$  and  $\langle u, u \rangle = \langle v, v \rangle$ , which implies that  $u$  and  $v$  are linearly independent, and if  $W$  is the subspace spanned by  $u$  and  $v$ , then  $f(W) = W$  and  $f^*(W) = W$ . Furthermore, with respect to the (orthogonal) basis  $(u, v)$ , the restriction of  $f$  to  $W$  has the matrix*

$$\begin{pmatrix} \lambda & \mu \\ -\mu & \lambda \end{pmatrix}.$$

*If  $\mu = 0$ , then  $\lambda$  is a real eigenvalue of  $f$  and either  $u$  or  $v$  is an eigenvector of  $f$  for  $\lambda$ . If  $W$  is the subspace spanned by  $u$  if  $u \neq 0$ , or spanned by  $v \neq 0$  if  $u = 0$ , then  $f(W) \subseteq W$  and  $f^*(W) \subseteq W$ .*

If  $f$  is a normal linear map, the proof of Lemma 9.1.8 shows that  $\lambda, \mu, u$ , and  $v$ , satisfy the equations

$$\begin{aligned}f(u) &= \lambda u - \mu v, \\f(v) &= \mu u + \lambda v, \\f^*(u) &= \lambda u + \mu v, \\f^*(v) &= -\mu u + \lambda v,\end{aligned}$$

From the above, it is easy to see that  $\lambda$  is an eigenvalue of  $1/2(f + f^*)$ , that  $-\mu^2$  is an eigenvalue of  $(1/2(f - f^*))^2$ , and that  $u$  and  $v$  are both eigenvectors of  $1/2(f + f^*)$  for  $\lambda$  and of  $(1/2(f - f^*))^2$  for  $-\mu^2$ .

It is easily verified that  $1/2(f + f^*)$  and  $(1/2(f - f^*))^2$  are self-adjoint.

We can finally prove our first main theorem.

**Theorem 9.1.9** (*Main Spectral Theorem*) *Given a Euclidean space  $E$  of dimension  $n$ , for every normal linear map  $f: E \rightarrow E$ , there is an orthonormal basis  $(e_1, \dots, e_n)$  such that the matrix of  $f$  w.r.t. this basis is a block diagonal matrix of the form*

$$\begin{pmatrix} A_1 & & \cdots & \\ & A_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & A_p \end{pmatrix}$$

*such that each block  $A_i$  is either a one-dimensional matrix (i.e., a real scalar) or a two-dimensional matrix of the form*

$$A_i = \begin{pmatrix} \lambda_i & -\mu_i \\ \mu_i & \lambda_i \end{pmatrix}$$

*where  $\lambda_i, \mu_i \in \mathbb{R}$ , with  $\mu_i > 0$ .*

After this relatively hard work, we can easily obtain some nice normal forms for the matrices of self-adjoint, skew self-adjoint, and orthogonal, linear maps.

However, for the sake of completeness (and since we have all the tools to so do), we go back to the case of a Hermitian space and show that normal linear maps can be diagonalized with respect to an orthonormal basis.

**Theorem 9.1.10** *Given a Hermitian space  $E$  of dimension  $n$ , for every normal linear map  $f: E \rightarrow E$ , there is an orthonormal basis  $(e_1, \dots, e_n)$  of eigenvectors of  $f$  such that the matrix of  $f$  w.r.t. this basis is a diagonal matrix*

$$\begin{pmatrix} \lambda_1 & & \cdots & \\ & \lambda_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & \lambda_n \end{pmatrix}$$

where  $\lambda_i \in \mathbb{C}$ .

*Remark:* There is a *converse* to Theorem 9.1.10, namely, if there is an orthonormal basis  $(e_1, \dots, e_n)$  of eigenvectors of  $f$ , then  $f$  is normal. We leave the easy proof as an exercise.

## 9.2 Self-Adjoint, Skew Self-Adjoint, and Orthogonal Linear Maps

We begin with *self-adjoint* maps.

**Theorem 9.2.1** *Given a Euclidean space  $E$  of dimension  $n$ , for every self-adjoint linear map  $f: E \rightarrow E$ , there is an orthonormal basis  $(e_1, \dots, e_n)$  of eigenvectors of  $f$  such that the matrix of  $f$  w.r.t. this basis is a diagonal matrix*

$$\begin{pmatrix} \lambda_1 & & \cdots & \\ & \lambda_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & \lambda_n \end{pmatrix}$$

where  $\lambda_i \in \mathbb{R}$ .

Theorem 9.2.1 implies that if  $\lambda_1, \dots, \lambda_p$  are the distinct real eigenvalues of  $f$  and  $E_i$  is the eigenspace associated with  $\lambda_i$ , then

$$E = E_1 \oplus \cdots \oplus E_p,$$

where  $E_i$  and  $E_j$  are orthogonal for all  $i \neq j$ .

Next, we consider *skew self-adjoint* maps.

**Theorem 9.2.2** *Given a Euclidean space  $E$  of dimension  $n$ , for every skew self-adjoint linear map  $f: E \rightarrow E$ , there is an orthonormal basis  $(e_1, \dots, e_n)$  such that the matrix of  $f$  w.r.t. this basis is a block diagonal matrix of the form*

$$\begin{pmatrix} A_1 & & \cdots & \\ & A_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & A_p \end{pmatrix}$$

*such that each block  $A_i$  is either 0 or a two-dimensional matrix of the form*

$$A_i = \begin{pmatrix} 0 & -\mu_i \\ \mu_i & 0 \end{pmatrix}$$

*where  $\mu_i \in \mathbb{R}$ , with  $\mu_i > 0$ . In particular, the eigenvalues of  $f_{\mathbb{C}}$  are pure imaginary of the form  $i\mu_i$ , or 0.*

*Remark:* One will note that if  $f$  is skew self-adjoint, then  $if_{\mathbb{C}}$  is self-adjoint w.r.t.  $\langle -, - \rangle_{\mathbb{C}}$ .



By Lemma 9.1.6, the map  $if_{\mathbb{C}}$  has real eigenvalues, which implies that the eigenvalues of  $f_{\mathbb{C}}$  are pure imaginary or 0.

Finally, we consider *orthogonal* linear maps.

**Theorem 9.2.3** *Given a Euclidean space  $E$  of dimension  $n$ , for every orthogonal linear map  $f: E \rightarrow E$ , there is an orthonormal basis  $(e_1, \dots, e_n)$  such that the matrix of  $f$  w.r.t. this basis is a block diagonal matrix of the form*

$$\begin{pmatrix} A_1 & & \cdots & \\ & A_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & A_p \end{pmatrix}$$

*such that each block  $A_i$  is either 1,  $-1$ , or a two-dimensional matrix of the form*

$$A_i = \begin{pmatrix} \cos \theta_i & -\sin \theta_i \\ \sin \theta_i & \cos \theta_i \end{pmatrix}$$

*where  $0 < \theta_i < \pi$ .*

*In particular, the eigenvalues of  $f_{\mathbb{C}}$  are of the form*

$\cos \theta_i \pm i \sin \theta_i$ , or 1, or  $-1$ .

It is obvious that we can reorder the orthonormal basis of eigenvectors given by Theorem 9.2.3, so that the matrix of  $f$  w.r.t. this basis is a block diagonal matrix of the form

$$\begin{pmatrix} I_p & & & \cdots & & \\ & -I_q & & & & \\ & & A_1 & \cdots & & \\ \vdots & & \vdots & \ddots & \vdots & \\ & & & \cdots & & A_r \end{pmatrix}$$

where each block  $A_i$  is a two-dimensional rotation matrix  $A_i \neq \pm I_2$  of the form

$$A_i = \begin{pmatrix} \cos \theta_i & -\sin \theta_i \\ \sin \theta_i & \cos \theta_i \end{pmatrix}$$

with  $0 < \theta_i < \pi$ .

The linear map  $f$  has an eigenspace  $E(1, f) = \text{Ker}(f - \text{id})$  of dimension  $p$  for the eigenvalue 1, and an eigenspace  $E(-1, f) = \text{Ker}(f + \text{id})$  of dimension  $q$  for the eigenvalue  $-1$ .

If  $\det(f) = +1$  ( $f$  is a rotation), the dimension  $q$  of  $E(-1, f)$  must be even, and the entries in  $-I_q$  can be paired to form two-dimensional blocks, if we wish.

*Remark:* Theorem 9.2.3 can be used to prove a sharper version of the Cartan-Dieudonné Theorem.

**Theorem 9.2.4** *Let  $E$  be a Euclidean space of dimension  $n \geq 2$ . For every isometry  $f \in \mathbf{O}(E)$ , if  $p = \dim(E(1, f)) = \dim(\text{Ker}(f - \text{id}))$ , then  $f$  is the composition of  $n - p$  reflections and  $n - p$  is minimal.*

The theorems of this section and of the previous section can be immediately applied to matrices.

### 9.3 Normal, Symmetric, Skew Symmetric, Orthogonal, Hermitian, Skew Hermitian, and Unitary Matrices

First, we consider real matrices.

**Definition 9.3.1** Given a real  $m \times n$  matrix  $A$ , the *transpose*  $A^\top$  of  $A$  is the  $n \times m$  matrix  $A^\top = (a_{i,j}^\top)$  defined such that

$$a_{i,j}^\top = a_{j,i}$$

for all  $i, j$ ,  $1 \leq i \leq m$ ,  $1 \leq j \leq n$ . A real  $n \times n$  matrix  $A$  is

1. *normal* iff

$$A A^\top = A^\top A,$$

2. *symmetric* iff

$$A^\top = A,$$

3. *skew symmetric* iff

$$A^\top = -A,$$

4. *orthogonal* iff

$$A A^\top = A^\top A = I_n.$$

Theorems 9.1.9 and 9.2.1–9.2.3 can be restated as follows.

**Theorem 9.3.2** *For every normal matrix  $A$ , there is an orthogonal matrix  $P$  and a block diagonal matrix  $D$  such that  $A = P D P^\top$ , where  $D$  is of the form*

$$D = \begin{pmatrix} D_1 & & \cdots & \\ & D_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & D_p \end{pmatrix}$$

*such that each block  $D_i$  is either a one-dimensional matrix (i.e., a real scalar) or a two-dimensional matrix of the form*

$$D_i = \begin{pmatrix} \lambda_i & -\mu_i \\ \mu_i & \lambda_i \end{pmatrix}$$

*where  $\lambda_i, \mu_i \in \mathbb{R}$ , with  $\mu_i > 0$ .*

**Theorem 9.3.3** *For every symmetric matrix  $A$ , there is an orthogonal matrix  $P$  and a diagonal matrix  $D$  such that  $A = P D P^\top$ , where  $D$  is of the form*

$$D = \begin{pmatrix} \lambda_1 & & \cdots & \\ & \lambda_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & \lambda_n \end{pmatrix}$$

where  $\lambda_i \in \mathbb{R}$ .

**Theorem 9.3.4** *For every skew symmetric matrix  $A$ , there is an orthogonal matrix  $P$  and a block diagonal matrix  $D$  such that  $A = PD P^\top$ , where  $D$  is of the form*

$$D = \begin{pmatrix} D_1 & & \cdots & \\ & D_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & D_p \end{pmatrix}$$

*such that each block  $D_i$  is either 0 or a two-dimensional matrix of the form*

$$D_i = \begin{pmatrix} 0 & -\mu_i \\ \mu_i & 0 \end{pmatrix}$$

*where  $\mu_i \in \mathbb{R}$ , with  $\mu_i > 0$ . In particular, the eigenvalues of  $A$  are pure imaginary of the form  $i\mu_i$ , or 0.*



**Theorem 9.3.5** *For every orthogonal matrix  $A$ , there is an orthogonal matrix  $P$  and a block diagonal matrix  $D$  such that  $A = PD P^\top$ , where  $D$  is of the form*

$$D = \begin{pmatrix} D_1 & & \cdots & \\ & D_2 & \cdots & \\ \vdots & \vdots & \ddots & \vdots \\ & & \cdots & D_p \end{pmatrix}$$

*such that each block  $D_i$  is either 1,  $-1$ , or a two-dimensional matrix of the form*

$$D_i = \begin{pmatrix} \cos \theta_i & -\sin \theta_i \\ \sin \theta_i & \cos \theta_i \end{pmatrix}$$

*where  $0 < \theta_i < \pi$ .*

*In particular, the eigenvalues of  $A$  are of the form  $\cos \theta_i \pm i \sin \theta_i$ , or 1, or  $-1$ .*

We now consider complex matrices.

**Definition 9.3.6** Given a complex  $m \times n$  matrix  $A$ , the *transpose*  $A^\top$  of  $A$  is the  $n \times m$  matrix  $A^\top = (a_{i,j}^\top)$  defined such that

$$a_{i,j}^\top = a_{j,i}$$

for all  $i, j$ ,  $1 \leq i \leq m$ ,  $1 \leq j \leq n$ . The *conjugate*  $\overline{A}$  of  $A$  is the  $m \times n$  matrix  $\overline{A} = (b_{i,j})$  defined such that

$$b_{i,j} = \overline{a_{i,j}}$$

for all  $i, j$ ,  $1 \leq i \leq m$ ,  $1 \leq j \leq n$ . Given an  $n \times n$  complex matrix  $A$ , the *adjoint*  $A^*$  of  $A$  is the matrix defined such that

$$A^* = \overline{(A^\top)} = (\overline{A})^\top.$$

A complex  $n \times n$  matrix  $A$  is

1. *normal* iff

$$AA^* = A^*A,$$

2. *Hermitian* iff

$$A^* = A,$$

3. *skew Hermitian* iff

$$A^* = -A,$$

4. *unitary* iff

$$AA^* = A^*A = I_n.$$

Theorem 9.1.10 can be restated in terms of matrices as follows. We can also say a little more about eigenvalues (easy exercise left to the reader).

**Theorem 9.3.7** *For every complex normal matrix  $A$ , there is a unitary matrix  $U$  and a diagonal matrix  $D$  such that  $A = UDU^*$ . Furthermore, if  $A$  is Hermitian,  $D$  is a real matrix, if  $A$  is skew Hermitian, then the entries in  $D$  are pure imaginary or null, and if  $A$  is unitary, then the entries in  $D$  have absolute value 1.*

