

# Notes on linear algebra

Danny Nygård Hansen

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## 1 • Linear equations and matrices

### 1.1. Linear equations

Throughout we let  $\mathbb{F}$  denote an arbitrary field and  $R$  a commutative ring. Let  $m$  and  $n$  be positive integers. A *linear equation in  $n$  unknowns* is an equation on the form

$$l: a_1x_1 + \cdots + a_nx_n = b,$$

where  $a_1, \dots, a_n, b \in \mathbb{F}$ . A *solution* to  $l$  is an element  $v = (v_1, \dots, v_n) \in \mathbb{F}^n$  such that

$$a_1v_1 + \cdots + a_nv_n = b.$$

A *system of linear equations in  $n$  unknowns* is a tuple  $L = (l_1, \dots, l_m)$ , where each  $l_i$  is a linear equation in  $n$  unknowns. An element  $v \in \mathbb{F}^n$  is a *solution* to  $L$  if it is a solution to each linear equation  $l_1, \dots, l_m$ .

Let  $L$  and  $L'$  be systems of linear equations in  $n$  unknowns. We say that  $L$  and  $L'$  are *solution equivalent* if they have the same solutions. Furthermore, we say that they are *combination equivalent* if each equation in  $L'$  is a linear combination of the equations in  $L$ , and vice versa. Clearly, if  $L$  and  $L'$  are combination equivalent they are also solution equivalent, but the converse does not hold.

### 1.2. Matrices

It is well-known that a system of linear equations is equivalent to a matrix equation on the form  $Ax = b$ , where  $A \in \text{Mat}_{m,n}(\mathbb{F})$ ,  $x \in \mathbb{F}^n$  and  $b \in \mathbb{F}^m$ . Recall the *elementary row operations* on  $A$ :

- (1) multiplication of one row of  $A$  by a nonzero scalar,
- (2) addition to one row of  $A$  a scalar multiple of another (different) row, and

- (3) interchange of two rows of  $A$ .

If  $e$  is an elementary row operation, we write  $e(A)$  for the matrix obtained when applying  $e$  to  $A$ . Clearly each elementary row operation  $e$  has an ‘inverse’, i.e. an elementary row operation  $e'$  such that  $e'(e(A)) = e(e'(A)) = A$ . Two matrices  $A, B \in \text{Mat}_{m,n}(\mathbb{F})$  are called *row-equivalent* if  $A$  is obtained by applying a finite sequence of elementary row operations to  $B$  (and vice versa, though this need not be assumed since each elementary row operation has an inverse).

Clearly, if  $A, B \in \text{Mat}_{m,n}(\mathbb{F})$  are row-equivalent, then the systems of equations  $Ax = 0$  and  $Bx = 0$  are combination equivalent, hence have the same solutions.

#### DEFINITION 1.1

A matrix  $H \in \text{Mat}_{m,n}(\mathbb{F})$  is called *row-reduced* if

- (i) the first nonzero entry of each nonzero row in  $H$  is 1, and
- (ii) each column of  $H$  containing the leading nonzero entry of some row has all its other entries equal 0.

If  $H$  is row-reduced, it is called a *row-reduced echelon matrix* if it also has the following properties:

- (iii) Every row of  $H$  only containing zeroes occur below every row which has a nonzero entry, and
- (iv) if rows  $1, \dots, r$  are the nonzero rows of  $H$ , and if the leading nonzero entry of row  $i$  occurs in column  $k_i$ , then  $k_1 < \dots < k_r$ .

An *elementary matrix* is a matrix obtained by applying a single elementary row operation to the identity matrix  $I$ . It is easy to show that if  $e$  is an elementary row operation and  $E = e(I) \in \text{Mat}_m(\mathbb{F})$ , then  $e(A) = EA$  for  $A \in \text{Mat}_{m,n}(\mathbb{F})$ . If  $B \in \text{Mat}_{m,n}(\mathbb{F})$ , then  $A$  and  $B$  are row-equivalent if and only if  $A = PB$ , where  $P \in \text{Mat}_m(\mathbb{F})$  is a product of elementary matrices.

#### PROPOSITION 1.2

*Every matrix in  $\text{Mat}_{m,n}(\mathbb{F})$  is row-equivalent to a unique row-reduced echelon matrix.*

**PROOF.** The usual Gauss–Jordan elimination algorithm proves existence. If  $H, K \in \text{Mat}_{m,n}(\mathbb{F})$  are row-equivalent row-reduced echelon matrices, we claim that  $H = K$ . We prove this by induction in  $n$ . If  $n = 1$  then this is obvious, so assume that  $n > 1$ . Let  $H_1$  and  $K_1$  be the matrices obtained by deleting the  $n$ th

column in  $H$  and  $K$  respectively. Then  $H_1$  and  $K_1$  are also row-equivalent<sup>1</sup> and row-reduced echelon matrices, so by induction  $H_1 = K_1$ . Thus if  $H$  and  $K$  differ, they must differ in the  $n$ th column.

Let  $H_2$  be the matrix obtained by deleting columns in  $H$ , only keeping those columns containing pivots, as well as keeping the  $n$ th column. Define  $K_2$  similarly. Thus we have deleted the same columns in  $H$  and  $K$ , so  $H_2$  and  $K_2$  are also row-equivalent. Say that the number of columns in  $H_2$  and  $K_2$  is  $r + 1$ , and write the matrices on the form

$$H_2 = \begin{pmatrix} I_r & h \\ 0 & h' \end{pmatrix} \quad \text{and} \quad K_2 = \begin{pmatrix} I_r & k \\ 0 & k' \end{pmatrix},$$

where  $h, k \in \mathbb{F}^r$  and  $h', k' \in \mathbb{F}^{m-r}$  are column vectors. Since  $H_2$  and  $K_2$  are row-equivalent, the systems  $H_2x = 0$  and  $K_2x = 0$  are solution equivalent. If  $h' = 0$ , then  $H_2x = 0$  has the solution  $(-h, 1)$ . But this is also a solution to  $K_2x = 0$ , so  $h = k$  and  $k' = 0$ . If  $h' \neq 0$ , then  $H_2x = 0$  only has the trivial solution. But then  $K_2x = 0$  also only has the trivial solution, and hence  $k' \neq 0$ . But that must be because both  $H_2$  and  $K_2$  has a pivot in the rightmost column, so also in this case  $H_2 = K_2$ .  $\square$

### 1.3. Invertible matrices

Notice that elementary matrices are invertible, since elementary row operations are invertible.

#### LEMMA 1.3

If  $A \in \text{Mat}_n(\mathbb{F})$ , then the following are equivalent:

- (i)  $A$  is invertible,
- (ii)  $A$  is row-equivalent to  $I_n$ ,
- (iii)  $A$  is a product of elementary matrices, and
- (iv) the system  $Ax = 0$  has only the trivial solution  $x = 0$ .

**PROOF.** (i)  $\Leftrightarrow$  (ii): Let  $H \in \text{Mat}_n(\mathbb{F})$  be a row-reduced echelon matrix that is row-equivalent to  $A$ . Then  $H = PA$ , where  $P \in \text{Mat}_n(\mathbb{F})$  is a product of elementary matrices. Then  $A = P^{-1}H$ , so  $A$  is invertible if and only if  $H$  is. But the only invertible row-reduced echelon matrix is the identity matrix, so (i) and (ii) are equivalent.

<sup>1</sup> It should be obvious that deleting columns preserves row-equivalence, but we give a more precise argument: If  $P \in \text{Mat}_m(\mathbb{F})$  is a product of elementary matrices and  $a_1, \dots, a_n \in \mathbb{F}^m$  are the columns in  $A$ , then the columns in  $PA$  are  $Pa_1, \dots, Pa_n$ . Thus elementary row operations are applied to each column independently of the other columns.

(ii)  $\Rightarrow$  (iii): As above, there exists a product  $P$  of elementary matrices such that  $I_n = PA$ , so  $A = P^{-1}$ .

(iii)  $\Rightarrow$  (i): This is obvious since elementary matrices are invertible.

(ii)  $\Leftrightarrow$  (iv): If  $A$  and  $I_n$  are row-equivalent, then the systems  $Ax = 0$  and  $I_n x = 0$  have the same solutions. Conversely, assume that  $Ax = 0$  only has the trivial solution. If  $H \in \text{Mat}_{m,n}(\mathbb{F})$  is a row-reduced echelon matrix that is row-equivalent to  $A$ , then  $Hx = 0$  has no nontrivial solution. Thus if  $r$  is the number of nonzero rows in  $H$ , then  $r \geq n$ . But then  $r = n$ , so  $H$  must be the identity matrix.  $\square$

#### PROPOSITION 1.4

Let  $A \in \text{Mat}_n(\mathbb{F})$ . Then the following are equivalent:

- (i)  $A$  is invertible,
- (ii)  $A$  has a left inverse, and
- (iii)  $A$  has a right inverse.

**PROOF.** If  $A$  has a left inverse, then  $Ax = 0$  has no nontrivial solution, so  $A$  is invertible. If  $A$  has a right inverse  $B \in \text{Mat}_n(\mathbb{F})$ , i.e.  $AB = I$ , then  $B$  has a left inverse and is thus invertible. But then  $A$  is the inverse of  $B$  and hence is itself invertible.  $\square$

## 2 • Bases and coordinates

### 2.1. Bases

Let  $V$  be a vector space. A *Hamel basis* for  $V$  is a linearly independent set  $\mathcal{V} \subseteq V$  that spans  $V$ , i.e. for every  $v \in V$  there exist unique (up to ordering)  $v_1, \dots, v_n \in \mathcal{V}$  and  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$  such that  $v = \sum_{i=1}^n \alpha_i v_i$ . In other words, a Hamel basis is a maximal linearly independent subset of  $V$ .

If  $V$  is an inner product space, a subset  $\mathcal{O}$  of  $V$  is said to be *orthogonal* if  $v \neq w$  implies  $v \perp w$  for  $v, w \in \mathcal{O}$ . Furthermore, if every element of  $\mathcal{O}$  is a unit vector, then  $\mathcal{O}$  is called *orthonormal*. If  $\mathcal{V}$  is a Hamel basis for  $V$  that is also an orthogonal/orthonormal set, then  $\mathcal{V}$  is called an *orthogonal/orthonormal Hamel basis*. If every vector in an orthogonal set  $\mathcal{O}$  is nonzero, then  $\mathcal{O}$  gives rise to an orthonormal set by dividing each vector by its norm. This clearly preserves the span of every subset of  $\mathcal{O}$ ; in particular, if  $\mathcal{O}$  is an orthogonal Hamel basis then this modification yields an orthonormal Hamel basis.

There is also another notion of basis in an inner product space: A maximal orthonormal subset of  $V$  is called a *Hilbert basis*. An orthonormal Hamel basis is thus a Hilbert basis, but not vice-versa. For instance, the ‘standard basis’ of  $\mathbb{R}^{\mathbb{N}}$  consisting of sequences  $e_n = (0, \dots, 0, 1, 0, \dots)$  with a 1 in the  $n$ th place and zeros elsewhere is a Hilbert basis for the space  $l^2$ , but it is not a Hamel basis for  $l^2$  since its linear span is the *coproduct*  $\mathbb{R}^{\oplus \mathbb{N}}$ , i.e. the subspace of  $l^2$  of sequences with finitely many nonzero elements.

Zorn’s lemma can be used to show that every vector space has a Hamel basis, and that every inner product space has a Hilbert basis (every inner product space of course also has a Hamel basis). However, not every inner product space has an *orthonormal* Hamel basis. For instance, let  $\mathcal{H}$  be an infinite-dimensional Hilbert space, let  $\mathcal{O}$  be an infinite orthonormal subset of  $\mathcal{H}$ , and let  $(e_n)_{n \in \mathbb{N}}$  be a sequence of distinct elements from  $\mathcal{O}$ . Then the sum of the series  $\sum_{n=1}^{\infty} e_n/n$  lies in  $\mathcal{H}$  by completeness, but this cannot be expressed as a finite linear combination of elements in  $\mathcal{O}$ , since the terms in the sum become arbitrarily small.

The argument above in particular shows that the (Hamel) dimension of an infinite-dimensional Hilbert space  $\mathcal{H}$  is uncountable. For if  $\mathcal{I}$  is any countable, linearly independent collection of elements from  $\mathcal{H}$ , then the Gram–Schmidt process yields an orthonormal collection  $\mathcal{O} \subseteq \mathcal{H}$  with  $\text{span } \mathcal{O} = \text{span } \mathcal{I}$ . But the above shows that  $\mathcal{O}$  cannot span  $\mathcal{H}$ , so neither can  $\mathcal{I}$ .



Let  $V$  be a vector space. A *series* of subspaces  $U_i$  of  $V$  is a finite or infinite decreasing sequence

$$V = U_0 \supsetneq U_1 \supsetneq U_2 \supsetneq \cdots.$$

If the sequence is finite, then the *length* of the series is the number of strict inclusions. If the sequence is infinite, then we say that the length of the series is  $\infty$ . The maximal length of a series of subspaces of  $V$  is denoted  $l(V)$ .

In the proposition below, we write  $\dim V = \infty$  if the dimension of  $V$  is infinite.

#### PROPOSITION 2.1

Let  $V$  be a vector space. Then  $\dim V = l(V)$ .

**PROOF.** First assume that  $V$  is finite-dimensional, and let  $\mathcal{V} = (v_1, \dots, v_n)$  be a basis for  $V$ . Then there is a series

$$V = \text{span}(v_1, \dots, v_n) \supsetneq \text{span}(v_1, \dots, v_{n-1}) \supsetneq \cdots \supsetneq \text{span}(v_1) \supsetneq 0$$

of subspaces of  $V$ , so  $\dim V \leq l(V)$ . Conversely, let

$$V = U_0 \supsetneq U_1 \supsetneq U_2 \supsetneq \cdots$$

be a series of subspaces of  $V$ . If the series ends with 0, remove it. Hence all subspaces in the series are nontrivial. Then choose for each  $i$  an element  $v_i \in U_i \setminus U_{i+1}$ , and collect them in a set  $\mathcal{I}$ . It is clear that  $\mathcal{I}$  is linearly independent, hence finite. Thus the series is also finite with length  $|\mathcal{I}| - 1$ . Adding back 0 to the series we obtain a series that is at least as long as the original sequence, and that is of length  $|\mathcal{I}| \leq \dim V$ . Since the sequence was arbitrary,  $l(V) \leq \dim V$ .

Next assume that  $V$  is infinite-dimensional. Then  $V$  contains a sequence  $(v_i)_{i \in \mathbb{N}}$  that is linearly independent, so the series

$$V \supseteq \text{span}\{v_i \mid i \in \mathbb{N}\} \supsetneq \text{span}\{v_i \mid i \geq 2\} \supsetneq \text{span}\{v_i \mid i \geq 3\} \supsetneq \cdots$$

is infinite, and  $l(V) = \infty$ . Conversely, assume that  $V$  has an infinite series. As above we construct a linearly independent set  $\mathcal{I}$  whose size equals the length of the sequence. Thus  $V$  contains an infinite linearly independent set, so  $\dim V = \infty$ .  $\square$

## 2.2. Coordinate maps and matrices

For  $A \in \text{Mat}_{m,n}(\mathbb{F})$  we define the map  $M_A: \mathbb{F}^n \rightarrow \mathbb{F}^m$  by  $M_A v = Av$ .

### PROPOSITION 2.2

Let  $(e_1, \dots, e_n)$  be the standard basis for  $\mathbb{F}^n$ . The map

$$\begin{aligned} \mathcal{M}: \mathcal{L}(\mathbb{F}^n, \mathbb{F}^m) &\rightarrow \text{Mat}_{m,n}(\mathbb{F}), \\ T &\mapsto (Te_1 \mid \cdots \mid Te_n), \end{aligned}$$

is a linear isomorphism with inverse  $A \mapsto M_A$ . The matrix  $\mathcal{M}(T)$  is called the standard matrix representation of  $T$ . If  $T: \mathbb{F}^n \rightarrow \mathbb{F}^m$  and  $S: \mathbb{F}^m \rightarrow \mathbb{F}^l$  are linear maps, then

- (i)  $Tv = \mathcal{M}(T)v$  for all  $v \in \mathbb{F}^n$ .
- (ii)  $\mathcal{M}(\text{id}_{\mathbb{F}^n}) = I$ .
- (iii)  $\mathcal{M}(S \circ T) = \mathcal{M}(S)\mathcal{M}(T)$ .
- (iv)  $T$  is invertible if and only if  $\mathcal{M}(T)$  is invertible, in which case  $\mathcal{M}(T^{-1}) = \mathcal{M}(T)^{-1}$ .

**PROOF.** The map  $A \mapsto M_A$  is clearly linear, so to prove the first point it suffices to show that this is the inverse of  $\mathcal{M}$ . Let  $T \in \mathcal{L}(\mathbb{F}^n, \mathbb{F}^m)$ . Then

$$M_{\mathcal{M}(T)} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} = \mathcal{M}(T) \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} = (Te_1 \mid \cdots \mid Te_n) \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} = \sum_{i=1}^n \alpha_i Te_i = T \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$$

for  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$ . Conversely, for  $A \in \text{Mat}_{m,n}(\mathbb{F})$  we have

$$\mathcal{M}(M_A) = (M_A e_1 \mid \cdots \mid M_A e_n) = (A e_1 \mid \cdots \mid A e_n) = A,$$

since  $A e_i$  is the  $i$ th column of  $A$ . We prove the remaining claims:

*Proof of (i):* Simply notice that  $Tv = M_{\mathcal{M}(T)}v = \mathcal{M}(T)v$ .

*Proof of (ii):* This is obvious from the definition of  $\mathcal{M}$ .

*Proof of (iii):* Let  $v \in \mathbb{F}^n$  and notice that

$$\mathcal{M}(S \circ T)v = (S \circ T)v = S(Tv) = S(\mathcal{M}(T)v) = \mathcal{M}(S)\mathcal{M}(T)v$$

by (i). Since this holds for all  $v$ , the claim follows.

*Proof of (iv):* This follows easily from (ii) and (iii).  $\square$

Let  $V$  be a finite-dimensional  $\mathbb{F}$ -vector space. If  $\mathcal{V} = (v_1, \dots, v_n)$  is an ordered basis for  $V$ , then for every  $v \in V$  there are unique  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$  such that  $v = \sum_{i=1}^n \alpha_i v_i$ . Hence the map  $\varphi_{\mathcal{V}}: V \rightarrow \mathbb{F}^n$  given by  $\varphi_{\mathcal{V}}(v) = (\alpha_1, \dots, \alpha_n)$  is well-defined. Furthermore, it is clearly linear, and since  $\mathcal{V}$  is a basis it is also bijective, hence a linear isomorphism. The map  $\varphi_{\mathcal{V}}$  is called the *coordinate map* with respect to  $\mathcal{V}$ , and the vector  $[v]_{\mathcal{V}} = \varphi_{\mathcal{V}}(v)$  is called the *coordinate vector* of  $v$  with respect to  $\mathcal{V}$ .

Now let  $\mathcal{W}$  be another ordered basis for  $V$ . The composition  $\varphi_{\mathcal{W}, \mathcal{V}} = \varphi_{\mathcal{W}} \circ \varphi_{\mathcal{V}}^{-1}$  is called the *change of basis operator* from  $\mathcal{V}$  to  $\mathcal{W}$ , and this makes the diagram

$$\begin{array}{ccc} & & \mathbb{F}^n \\ & \nearrow \varphi_{\mathcal{V}} & \downarrow \varphi_{\mathcal{W}, \mathcal{V}} \\ V & & \mathbb{F}^n \\ & \searrow \varphi_{\mathcal{W}} & \end{array} \quad (2.1)$$

commute. Its standard matrix is denoted  ${}_{\mathcal{W}}[\square]_{\mathcal{V}}$ . This has the expected properties:

### PROPOSITION 2.3

Let  $\mathcal{V}, \mathcal{W}$  and  $\mathcal{U}$  be ordered bases for a finite-dimensional  $\mathbb{F}$ -vector space  $V$ . Then

- (i)  $[v]_{\mathcal{W}} = \varphi_{\mathcal{W}, \mathcal{V}}([v]_{\mathcal{V}})$  for all  $v \in V$ . In particular,  $[v]_{\mathcal{W}} = {}_{\mathcal{W}}[\square]_{\mathcal{V}} \cdot [v]_{\mathcal{V}}$ .
- (ii)  $\varphi_{\mathcal{V}, \mathcal{V}}$  is the identity map. In particular,  ${}_{\mathcal{V}}[\square]_{\mathcal{V}}$  is the identity matrix.
- (iii)  $\varphi_{\mathcal{U}, \mathcal{W}} \circ \varphi_{\mathcal{W}, \mathcal{V}} = \varphi_{\mathcal{U}, \mathcal{V}}$ . In particular,  ${}_{\mathcal{U}}[\square]_{\mathcal{W}} \cdot {}_{\mathcal{W}}[\square]_{\mathcal{V}} = {}_{\mathcal{U}}[\square]_{\mathcal{V}}$ .
- (iv)  $\varphi_{\mathcal{W}, \mathcal{V}}$  (resp.  ${}_{\mathcal{W}}[\square]_{\mathcal{V}}$ ) is invertible with inverse  $\varphi_{\mathcal{V}, \mathcal{W}}$  (resp.  ${}_{\mathcal{V}}[\square]_{\mathcal{W}}$ ).

**PROOF.** All claims about change of basis matrices follow by [Proposition 2.2](#) from the corresponding claims about change of basis operators.

The claim (i) follows by commutativity of the diagram (2.1), i.e.

$$\varphi_{\mathcal{W},\mathcal{V}}([v]_{\mathcal{V}}) = (\varphi_{\mathcal{W}} \circ \varphi_{\mathcal{V}}^{-1}) \circ \varphi_{\mathcal{V}}(v) = \varphi_{\mathcal{W}}(v) = [v]_{\mathcal{W}}.$$

Claim (ii) is an immediate consequence of the definition of  $\varphi_{\mathcal{V},\mathcal{V}}$ . The remaining claims are proved similarly to (i).  $\square$

Next consider a linear map  $T: V \rightarrow W$ . If  $\mathcal{V} \in V^n$  and  $\mathcal{W} \in W^m$  are bases for  $V$  and  $W$  respectively, then the diagram

$$\begin{array}{ccc} V & \xrightarrow{\varphi_{\mathcal{V}}} & \mathbb{F}^n \\ T \downarrow & & \downarrow \varphi_{\mathcal{W}} \circ T \circ \varphi_{\mathcal{V}}^{-1} \\ W & \xrightarrow{\varphi_{\mathcal{W}}} & \mathbb{F}^m \end{array}$$

commutes. The map  $\varphi_{\mathcal{W}} \circ T \circ \varphi_{\mathcal{V}}^{-1}$  is the *basis representation* of  $T$  with respect to the bases  $\mathcal{V}$  and  $\mathcal{W}$ . This is a linear map  $\mathbb{F}^n \rightarrow \mathbb{F}^m$ , so it has a standard matrix which we denote  ${}_{\mathcal{W}}[T]_{\mathcal{V}}$ . This is called the *matrix representation* of  $T$  with respect to the bases  $\mathcal{V}$  and  $\mathcal{W}$ .

#### PROPOSITION 2.4

Let  $V$  and  $W$  be finite-dimensional  $\mathbb{F}$ -vector spaces with ordered bases  $\mathcal{V} \in V^n$  and  $\mathcal{W} \in W^m$ , respectively. The map

$$\begin{aligned} {}_{\mathcal{W}}[\cdot]_{\mathcal{V}}: \mathcal{L}(V, W) &\rightarrow \text{Mat}_{m,n}(\mathbb{F}), \\ T &\mapsto {}_{\mathcal{W}}[T]_{\mathcal{V}}, \end{aligned}$$

is a linear isomorphism. Let  $T: V \rightarrow W$  and  $S: W \rightarrow U$  be linear maps, and let  $\mathcal{U} \in U^l$  be an ordered basis for  $U$ . Then

- (i)  $[Tv]_{\mathcal{W}} = {}_{\mathcal{W}}[T]_{\mathcal{V}} \cdot [v]_{\mathcal{V}}$  for all  $v \in V$ .
- (ii) If  $\mathcal{V}'$  is another basis for  $V$ , then  ${}_{\mathcal{V}}[\text{id}_V]_{\mathcal{V}} = {}_{\mathcal{V}'}[\text{id}_V]_{\mathcal{V}}$ .
- (iii)  ${}_{\mathcal{U}}[S \circ T]_{\mathcal{V}} = {}_{\mathcal{U}}[S]_{\mathcal{W}} \cdot {}_{\mathcal{W}}[T]_{\mathcal{V}}$ .
- (iv)  $T$  is invertible if and only if  ${}_{\mathcal{W}}[T]_{\mathcal{V}}$  is invertible, in which case  ${}_{\mathcal{V}}[T^{-1}]_{\mathcal{W}} = {}_{\mathcal{W}}[T]_{\mathcal{V}}^{-1}$ .

**PROOF.** For the first claim, notice that the map  $T \mapsto \varphi_{\mathcal{W}} \circ T \circ \varphi_{\mathcal{V}}^{-1}$  is a linear isomorphism, since pre- and postcomposition with linear isomorphisms are themselves linear isomorphisms. Composing this map with  $\mathcal{M}$  yields  ${}_{\mathcal{W}}[\cdot]_{\mathcal{V}}$ , so this is a linear isomorphism by [Proposition 2.2](#).



*Proof of (i):* Notice that

$$\begin{aligned} [Tv]_{\mathcal{W}} &= (\varphi_{\mathcal{W}} \circ T)(v) \\ &= (\varphi_{\mathcal{W}} \circ T \circ \varphi_{\mathcal{V}}^{-1}) \circ \varphi_{\mathcal{V}}(v) \\ &= (\varphi_{\mathcal{W}} \circ T \circ \varphi_{\mathcal{V}}^{-1})([v]_{\mathcal{V}}) \\ &= {}_{\mathcal{W}}[T]_{\mathcal{V}} \cdot [v]_{\mathcal{V}}. \end{aligned}$$

where the last equality follows from [Proposition 2.2\(i\)](#).

*Proof of (ii):* This is obvious from the definitions of  ${}_{\mathcal{V}}[\text{id}_V]_{\mathcal{V}}$  and  ${}_{\mathcal{V}}[\square]_{\mathcal{V}}$ .

*Proof of (iii):* Notice that

$$\varphi_{\mathcal{U}} \circ (S \circ T) \circ \varphi_{\mathcal{V}}^{-1} = (\varphi_{\mathcal{U}} \circ S \circ \varphi_{\mathcal{W}}^{-1}) \circ (\varphi_{\mathcal{W}} \circ T \circ \varphi_{\mathcal{V}}^{-1})$$

The claim then follows from [Proposition 2.2\(iii\)](#).

*Proof of (iv):* This is an immediate consequence of either [\(iii\)](#) or of [Proposition 2.2\(iv\)](#).  $\square$

#### PROPOSITION 2.5

Let  $\mathcal{V} = (v_1, \dots, v_n)$  be an ordered basis for an  $\mathbb{F}$ -vector space  $V$ , and let  $T: V \rightarrow V$  be a linear isomorphism. Let  $\mathcal{W} = (w_1, \dots, w_n)$  where  $w_i = Tv_i$ . Then  $\mathcal{W}$  is an ordered basis for  $V$  and

$$\varphi_{\mathcal{W}, \mathcal{V}} = \varphi_{\mathcal{V}} \circ T^{-1} \circ \varphi_{\mathcal{V}}^{-1}, \quad \text{or} \quad {}_{\mathcal{W}}[\square]_{\mathcal{V}} = {}_{\mathcal{V}}[T^{-1}]_{\mathcal{V}}.$$

In particular, if  $V = \mathbb{F}^n$  and  $\mathcal{V}$  is the standard basis  $\mathcal{E}$ , then

$$\varphi_{\mathcal{W}, \mathcal{E}} = T^{-1}, \quad \text{or} \quad {}_{\mathcal{W}}[\square]_{\mathcal{E}} = \mathcal{M}(T^{-1}).$$

We think of this result as follows: If we change basis by applying an invertible linear transformation  $T$ , we obtain the coordinate vectors corresponding to the transformed basis by applying  $T^{-1}$  (in the old basis). This says that if we perform a *passive transformation*, i.e. a change of basis while keeping vectors themselves fixed, the coordinates change by the inverse of said transformation.

**PROOF.** Let  $v \in V$  and write  $v = \sum_{i=1}^n \alpha_i v_i$ . Then

$$Tv = \sum_{i=1}^n \alpha_i Tv_i = \sum_{i=1}^n \alpha_i w_i = \varphi_{\mathcal{W}}^{-1} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} = \varphi_{\mathcal{W}}^{-1} \circ \varphi_{\mathcal{V}}(v),$$

implying that

$$\varphi_{\mathcal{W}, \mathcal{V}} = \varphi_{\mathcal{W}} \circ \varphi_{\mathcal{V}}^{-1} = (T \circ \varphi_{\mathcal{V}}^{-1})^{-1} \circ \varphi_{\mathcal{V}}^{-1} = \varphi_{\mathcal{V}} \circ T^{-1} \circ \varphi_{\mathcal{V}}^{-1}$$

as claimed.  $\square$

[TODO] Recall that two matrices  $A, B \in \text{Mat}_n(\mathbb{F})$  are *similar* if there exists an invertible matrix  $P \in \text{Mat}_n(\mathbb{F})$  such that  $A = PBP^{-1}$ .

### 3 • Determinants

#### 3.1. Existence of determinants

If  $M_1, \dots, M_n, N$  are modules over a commutative ring  $R$ , a map

$$\varphi: M_1 \times \dots \times M_n \rightarrow N$$

is called *n-linear* if, for all  $i$ , the maps  $m_i \mapsto \varphi(m_1, \dots, m_n)$  are linear for all choices of  $m_j \in M_j$  where  $j \neq i$ . Since there is a natural isomorphism  $\text{Mat}_{m,n}(R) \cong (R^n)^m$ , a map  $\varphi: \text{Mat}_{m,n}(R) \rightarrow N$  that is linear in each row is also called *n-linear*.

In the case  $M_1 = \dots = M_n$ , we call  $\varphi$  *alternating* if  $\varphi(m_1, \dots, m_n) = 0$  whenever  $m_i = m_j$  for some  $i \neq j$ . Furthermore,  $\varphi$  is called *skew-symmetric* if

$$\begin{aligned} \varphi(m_1, \dots, m_{i-1}, m_i, m_{i+1}, \dots, m_{j-1}, m_j, m_{j+1}, \dots, m_n) \\ = -\varphi(m_1, \dots, m_{i-1}, m_j, m_{i+1}, \dots, m_{j-1}, m_i, m_{j+1}, \dots, m_n) \end{aligned}$$

for all  $i < j$ .

#### LEMMA 3.1

Let  $M$  and  $N$  be  $R$ -modules, and let  $\varphi: M^n \rightarrow N$  be an  $n$ -linear map.

- (i) If  $\varphi$  is alternating, then  $\varphi$  is skew-symmetric. If  $\text{char } R \neq 2$  then the converse also holds.
- (ii) If  $\varphi(m_1, \dots, m_n) = 0$  whenever  $m_i = m_{i+1}$  for some  $i = 1, \dots, n-1$ , then  $\varphi$  is alternating.

We shall not use the converse direction of [Lemma 3.1\(i\)](#) but we include it for completeness.

**PROOF. Proof of (i):** Consider  $m_1, \dots, m_n \in M$ , and let  $1 \leq i < j \leq n$ . Define a map  $\psi: M \times M \rightarrow N$  by

$$\psi(a, b) = \varphi(m_1, \dots, m_{i-1}, a, m_{i+1}, \dots, m_{j-1}, b, m_{j+1}, \dots, m_n),$$

and notice that it suffices to show that  $\psi(m_i, m_j) = -\psi(m_j, m_i)$ . But  $\psi$  is 2-linear and alternating, so for  $a, b \in M$  we have

$$\psi(a + b, a + b) = \psi(a, a) + \psi(a, b) + \psi(b, a) + \psi(b, b) = \psi(a, b) + \psi(b, a).$$

Thus  $\psi(m_i, m_j) = -\psi(m_j, m_i)$ , so  $\varphi$  is skew-symmetric as claimed.

Conversely, if  $\text{char } R \neq 2$  and  $\psi$  is skew-symmetric, then since  $\psi(a, b) = -\psi(b, a)$ , letting  $a = b$  we have  $2\psi(a, a) = 0$ , so  $\psi(a, a) = 0$ .

*Proof of (ii):* The argument above shows that, in particular, if  $A, B \in M^n$ , and  $B$  is obtained from  $A$  by interchanging two adjacent elements, then  $\varphi(B) = -\varphi(A)$ . Assuming now that  $B$  is obtained from  $A$  by interchanging the  $i$ th and  $j$ th elements in  $A$ , with  $i < j$ , we claim that we may obtain  $B$  by successively interchanging adjacent elements of  $A$ . Writing  $A = (m_1, \dots, m_n)$ , we first perform  $j - i$  such interchanges and arrive that the tuple

$$(m_1, \dots, m_{i-1}, m_{i+1}, \dots, m_{j-1}, m_j, m_i, m_{j+1}, \dots, m_n),$$

moving  $m_i$  to the right  $j - i$  places. Next we perform another  $j - i - 1$  interchanges, moving  $m_j$  to the left until we reach

$$B = (m_1, \dots, m_{i-1}, m_j, m_{i+1}, \dots, m_{j-1}, m_i, m_{j+1}, \dots, m_n).$$

Since each interchange results in a sign change, we have

$$\varphi(B) = (-1)^{2(j-i)-1} \varphi(A) = -\varphi(A).$$

If  $m_i = m_j$  for  $i < j$ , then we claim that  $\varphi(A) = 0$ . For let  $B$  be obtained from  $A$  by interchanging  $m_{i+1}$  and  $m_j$ . Then  $\varphi(B) = 0$ , so  $\varphi(A) = -\varphi(B) = 0$  by the above argument, and hence  $\varphi$  is alternating as claimed.  $\square$

#### DEFINITION 3.2: Determinant functions

If  $n$  be a positive integer, a *determinant function* is a map  $\varphi: \text{Mat}_n(R) \rightarrow R$  that is  $n$ -linear, alternating, and which satisfies  $\varphi(I_n) = 1$ .

If  $A \in \text{Mat}_n(R)$  with  $n > 1$  and  $1 \leq i, j \leq n$ , denote by  $M(A)_{i,j}$  the matrix in  $\text{Mat}_{n-1}(R)$  obtained by removing the  $i$ th row and the  $j$ th column of  $A$ . This is called the  $(i, j)$ -th *minor* of  $A$ . If  $\varphi: \text{Mat}_{n-1}(R) \rightarrow R$  is an  $(n-1)$ -linear function and  $A \in \text{Mat}_n(R)$ , then we write  $\varphi_{i,j}(A) = \varphi(M(A)_{i,j})$ . Then  $\varphi_{i,j}: \text{Mat}_n(R) \rightarrow R$  is clearly linear in all rows except row  $i$ , and is independent of row  $i$ .

#### THEOREM 3.3: Construction of determinants

Let  $n > 1$ , and let  $\varphi: \text{Mat}_{n-1}(R) \rightarrow R$  be alternating and  $(n-1)$ -linear. For  $j = 1, \dots, n$  define a map  $\psi_j: \text{Mat}_n(R) \rightarrow R$  by

$$\psi_j(A) = \sum_{i=1}^n (-1)^{i+j} a_{ij} \varphi_{i,j}(A),$$

for  $A = (a_{ij}) \in \text{Mat}_n(R)$ . Then  $\psi_j$  is alternating and  $n$ -linear. If  $\varphi$  is a determinant

function, then so is  $\psi_j$ .

**PROOF.** Let  $A = (a_{ij}) \in \text{Mat}_n(R)$ . Then  $A \mapsto a_{ij}$  is independent of all rows except row  $i$ , and  $\varphi_{i,j}$  is linear in all rows except row  $i$ . Thus  $A \mapsto a_{ij}\varphi_{i,j}(A)$  is linear in all rows except row  $i$ . Conversely,  $A \mapsto a_{ij}$  is linear in row  $i$ , and  $\varphi_{i,j}$  is independent of row  $i$ , so  $A \mapsto a_{ij}\varphi_{i,j}(A)$  is also linear in row  $i$ . Since  $\psi_j$  is a linear combination of  $n$ -linear maps, is it itself  $n$ -linear.

Now assume that  $A$  has two equal adjacent rows, say  $a_k, a_{k+1} \in R^n$ . If  $i \neq k$  and  $i \neq k+1$ , then  $M(A)_{i,j}$  has two equal rows, so  $\varphi_{i,j}(A) = 0$ . Thus

$$\psi_j(A) = (-1)^{k+j} a_{kj} \varphi_{k,j}(A) + (-1)^{k+1+j} a_{(k+1)j} \varphi_{k+1,j}(A).$$

Since  $a_k = a_{k+1}$  we also have  $a_{kj} = a_{(k+1)j}$  and  $M(A)_{k,j} = M(A)_{k+1,j}$ . Thus  $\psi_j(A) = 0$ , so **Lemma 3.1(ii)** implies that  $\psi_j$  is alternating.

Finally suppose that  $\varphi$  is a determinant function. Then  $M(I_n)_{j,j} = I_{n-1}$  and we have

$$\psi_j(I_n) = (-1)^{j+j} \varphi_{j,j}(I_n) = \varphi(I_{n-1}) = 1,$$

so  $\psi_j$  is also a determinant function.  $\square$

#### COROLLARY 3.4: Existence of determinants

For every positive integer  $n$ , there exists a determinant function  $\text{Mat}_n(R) \rightarrow R$ .

**PROOF.** The identity map on  $\text{Mat}_1(R) \cong R$  is a determinant function for  $n = 1$ , and **Theorem 3.3** allows us to recursively construct a determinant for each  $n > 1$ .  $\square$

### 3.2. Uniqueness of determinants

#### THEOREM 3.5: Uniqueness of determinants

Let  $n$  be a positive integer. There is precisely one determinant function on  $\text{Mat}_n(R)$ , namely the function  $\det: \text{Mat}_n(R) \rightarrow R$  given by

$$\det A = \sum_{\sigma \in S_n} (\text{sgn } \sigma) a_{1\sigma(1)} \cdots a_{n\sigma(n)}$$

for  $A = (a_{ij}) \in \text{Mat}_n(R)$ . If  $\varphi: \text{Mat}_n(R) \rightarrow R$  is any alternating  $n$ -linear function, then

$$\varphi(A) = (\det A) \varphi(I_n).$$

We use the notation  $\det$  for the unique determinant on  $\text{Mat}_n(R)$  for all  $n$ .

**PROOF.** Let  $e_1, \dots, e_n$  denote the rows of  $I_n$ , and denote the rows of a matrix  $A = (a_{ij}) \in \text{Mat}_n(R)$  by  $a_1, \dots, a_n$ . Then  $a_i = \sum_{j=1}^n a_{ij}e_j$ , so

$$\varphi(A) = \sum_{k_1, \dots, k_n} a_{1k_1} \cdots a_{nk_n} \varphi(e_{k_1}, \dots, e_{k_n}),$$

where the sum is taken over all  $k_i = 1, \dots, n$ . Since  $\varphi$  is alternating we have  $\varphi(e_{k_1}, \dots, e_{k_n}) = 0$  if two of the indices  $k_1, \dots, k_n$  are equal. Thus it suffices to sum over those sequences  $(k_1, \dots, k_n)$  that are permutations of  $(1, \dots, n)$ , and so

$$\varphi(A) = \sum_{\sigma \in S_n} a_{1\sigma(1)} \cdots a_{n\sigma(n)} \varphi(e_{\sigma(1)}, \dots, e_{\sigma(n)}).$$

Next notice that, since  $\varphi$  is also skew-symmetric by [Lemma 3.1\(i\)](#), we have  $\varphi(e_{\sigma(1)}, \dots, e_{\sigma(n)}) = (-1)^m \varphi(e_1, \dots, e_n)$ , where  $m$  is the number of transpositions of  $(1, \dots, n)$  it takes to obtain the permutation  $(\sigma(1), \dots, \sigma(n))$ . But then  $(-1)^m$  is just the sign of  $\sigma$ , so

$$\varphi(A) = \sum_{\sigma \in S_n} (\text{sgn } \sigma) a_{1\sigma(1)} \cdots a_{n\sigma(n)} \varphi(I_n).$$

Finally, if  $\varphi$  is a determinant function, then  $\varphi(I_n) = 1$ , so we must have  $\varphi = \det$ . The rest of the theorem follows directly from this.  $\square$

### 3.3. Properties of determinants

#### THEOREM 3.6

Let  $A, B \in \text{Mat}_n(R)$ . Then

$$\det AB = (\det A)(\det B).$$

In particular,  $\det: \text{GL}_n(R) \rightarrow R^*$  is a group homomorphism.

**PROOF.** The map  $\varphi: \text{Mat}_n(R) \rightarrow R$  given by  $\varphi(A) = \det AB$  is clearly  $n$ -linear and alternating. Hence  $\varphi(A) = (\det A)\varphi(I)$ , and  $\varphi(I) = \det B$ .

Furthermore, if  $A$  is invertible, then  $1 = \det I = (\det A)(\det A^{-1})$ . Thus  $\det A \in R^*$ , so  $\det$  is a group homomorphism as claimed.  $\square$

#### COROLLARY 3.7

If  $A, B \in \text{Mat}_n(\mathbb{F})$  are similar matrices, then  $\det A = \det B$ .

**PROOF.** Let  $P \in \text{Mat}_n(\mathbb{F})$  be such that  $A = PBP^{-1}$ . [Theorem 3.6](#) then implies that

$$\det A = (\det P)(\det B)(\det P^{-1}) = (\det B)(\det PP^{-1}) = \det B. \quad \square$$

**Corollary 3.7** allows us to define the determinant of a general linear operator  $T: V \rightarrow V$  on a finite-dimensional  $\mathbb{F}$ -vector space. If  $\mathcal{V}$  and  $\mathcal{W}$  are bases for  $V$ , then the matrix representations  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$  and  ${}_{\mathcal{W}}[T]_{\mathcal{W}}$  are similar. This allows us to define the determinant  $\det T$  of  $T$  as the matrix representation  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$  for any basis  $\mathcal{V}$ .

**PROPOSITION 3.8**

Let  $A_{11}, \dots, A_{nn}$  be square matrices with entries in  $R$  and consider the block matrix

$$M = \begin{pmatrix} A_{11} & A_{12} & \cdots & A_{1n} \\ 0 & A_{22} & \cdots & A_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & A_{nn} \end{pmatrix},$$

where the remaining  $A_{ij}$  are matrices of appropriate dimensions. Then  $\det M = \prod_{i=1}^n \det A_{ii}$ .

**PROOF.** By induction it suffices to consider the case where  $M$  has the block form

$$M = \begin{pmatrix} A & C \\ 0 & B \end{pmatrix},$$

where  $A \in \text{Mat}_r(R)$ ,  $B \in \text{Mat}_s(R)$  and  $C \in \text{Mat}_{r,s}(R)$  for appropriate integers  $r, s$ . Notice that if we define the matrices

$$M_1 = \begin{pmatrix} I_r & 0 \\ 0 & B \end{pmatrix} \quad \text{and} \quad M_2 = \begin{pmatrix} A & C \\ 0 & I_s \end{pmatrix},$$

then  $M = M_1 M_2$ . But using **Theorem 3.3** we easily see that  $\det M_1 = \det B$  and  $\det M_2 = \det A$ , so it follows that

$$\det M = (\det M_1)(\det M_2) = (\det A)(\det B)$$

as desired. □

**PROPOSITION 3.9**

Let  $A \in \text{Mat}_n(R)$ . Then  $\det A = \det A^\top$ .

**PROOF.** Writing  $A = (a_{ij})$ , first notice that

$$\det A^\top = \sum_{\sigma \in S_n} (\text{sgn } \sigma^{-1}) a_{\sigma(1)1} \cdots a_{\sigma(n)n},$$

since  $\operatorname{sgn} \sigma = \operatorname{sgn} \sigma^{-1}$ . Next notice that, if  $j = \sigma(i)$ , then  $a_{\sigma(i)i} = a_{j\sigma^{-1}(j)}$ . Since  $R$  is commutative, it follows that

$$\det A^\top = \sum_{\sigma \in S_n} (\operatorname{sgn} \sigma^{-1}) a_{1\sigma^{-1}(1)} \cdots a_{n\sigma^{-1}(n)},$$

and since  $\sigma \mapsto \sigma^{-1}$  is a bijection on  $S_n$ , it follows that  $\det A^\top = \det A$  as desired.  $\square$

Let  $A \in \operatorname{Mat}_n(R)$ . For  $1 \leq i, j \leq n$ , the  $(i, j)$ -th cofactor of  $A$  is the number  $A_{i,j} = (-1)^{i+j} \det M(A)_{i,j}$ , where we recall that  $M(A)_{i,j}$  is the  $(i, j)$ -th minor of  $A$ . The cofactor matrix of  $A$  is the matrix  $\operatorname{cof} A \in \operatorname{Mat}_n(R)$  whose  $(i, j)$ -th entry is the cofactor  $A_{i,j}$ . Note that

$$(A^\top)_{i,j} = (-1)^{i+j} \det M(A^\top)_{i,j} = (-1)^{j+i} \det M(A)_{j,i} = A_{j,i},$$

so  $\operatorname{cof} A^\top = (\operatorname{cof} A)^\top$ . Of greater importance than the cofactor matrix is the *adjoint matrix* of  $A$ , written  $\operatorname{adj} A$ , which is just the transpose of  $\operatorname{cof} A$ . That is, the  $(i, j)$ -th entry of  $\operatorname{adj} A$  is the cofactor  $A_{j,i}$ . Similar to the cofactor matrix we have

$$\operatorname{adj} A^\top = (\operatorname{cof} A^\top)^\top = \operatorname{cof} A = (\operatorname{adj} A)^\top.$$

We have the following:

**PROPOSITION 3.10**

Let  $A \in \operatorname{Mat}_n(R)$ . Then

$$(\operatorname{adj} A)A = (\det A)I = A(\operatorname{adj} A).$$

**PROOF.** Writing  $A = (a_{ij})$  and fixing some  $j \in \{1, \dots, n\}$ , [Theorem 3.3](#) implies that

$$\det A = \sum_{i=1}^n (-1)^{i+j} a_{ij} \det M(A)_{i,j} = \sum_{i=1}^n a_{ij} A_{i,j},$$

which is just the  $(j, j)$ -th entry in the product  $(\operatorname{adj} A)A$ .

Next we claim that if  $k \neq j$ , then  $\sum_{i=1}^n a_{ik} A_{i,j} = 0$ . Let  $B = (b_{ij}) \in \operatorname{Mat}_n(R)$  be the matrix obtained from  $A$  by replacing the  $j$ th column of  $A$  by its  $k$ th column. Then  $B$  has two equal columns, so  $\det B = 0$ . Also,  $b_{ij} = a_{ik}$  and  $M(B)_{i,j} = M(A)_{i,j}$ , so it follows that

$$\begin{aligned} 0 = \det B &= \sum_{i=1}^n (-1)^{i+j} b_{ij} \det M(B)_{i,j} \\ &= \sum_{i=1}^n (-1)^{i+j} a_{ik} \det M(A)_{i,j} = \sum_{i=1}^n a_{ik} A_{i,j}. \end{aligned}$$

That is, the  $(j, k)$ -th entry of the product  $(\operatorname{adj} A)A$  is zero, so the off-diagonal entries of  $(\operatorname{adj} A)A$  are zero. In total we thus have  $(\operatorname{adj} A)A = (\det A)I$ .

Finally we prove the equality  $A(\operatorname{adj} A) = (\det A)I$ . Applying the first equality to  $A^\top$  yields

$$(\operatorname{adj} A^\top)A^\top = (\det A^\top)I = (\det A)I,$$

and transposing we get

$$A(\operatorname{adj} A) = A(\operatorname{adj} A^\top)^\top = (\det A)I$$

as desired.  $\square$

#### COROLLARY 3.11

Let  $A \in \operatorname{Mat}_n(R)$ . The following are equivalent:

- (i)  $A$  is a (two-sided) unit in  $\operatorname{Mat}_n(R)$ .
- (ii)  $A$  is a left- or right-unit in  $\operatorname{Mat}_n(R)$ .
- (iii)  $\det A$  is a unit in  $R$ .

**PROOF.** If  $A$  is e.g. a left-unit, then [Theorem 3.6](#) implies that

$$1 = \det I_n = (\det A)(\det A^{-1}),$$

so  $\det A$  is a unit in  $R$ . Conversely, if  $\det A$  is a unit then [Proposition 3.10](#) implies that  $(\det A)^{-1}(\operatorname{adj} A)$  is a two-sided inverse of  $A$ .  $\square$

Notice that this gives us a second proof of the fact that a matrix is invertible just when it has either a left- or right-inverse. In fact, we see that this holds for matrices with entries in any commutative ring.

#### 3.4. Determinants and eigenvalues

Let  $V$  be a vector space of dimension  $n < \infty$ . If  $T \in \mathcal{L}(V)$ , then recall that an *eigenvalue* of  $T$  is an element  $\lambda \in \mathbb{F}$  such that there is a nonzero vector  $v \in V$  with  $Tv = \lambda v$ . The set of eigenvalues of  $T$  is called the *spectrum* of  $T$  and is denoted  $\operatorname{Spec} T$ . Clearly  $\lambda \in \operatorname{Spec} T$  if and only if  $\lambda I - T$  is not injective, i.e. if  $\det(\lambda I - T) = 0$ . This motivates the definition of the *characteristic polynomial*  $p_T(t) \in \mathbb{F}[t]$  of  $T$ , given by  $p_T(t) = \det(tI - T)$ . The eigenvalues of  $T$  are then precisely the roots of  $p_T(t)$ .

#### PROPOSITION 3.12

Let  $T \in \mathcal{L}(V)$ .

- (i)  $p_T(t)$  is a monic polynomial of degree  $n$ .



(ii) The constant term of  $p_T(t)$  equals  $(-1)^n \det T$ .

(iii) The coefficient of  $t^{n-1}$  in  $p_T(t)$  equals  $-\operatorname{tr} T$ .

Assume further that  $p_T(t)$  splits over  $\mathbb{F}$ . Then:

(iv)  $T$  has an eigenvalue.

(v)  $\det T$  is the product of the eigenvalues of  $T$ .

(vi)  $\operatorname{tr} T$  is the sum of the eigenvalues of  $T$ .

The condition that  $p_T(t)$  splits over  $\mathbb{F}$  means that  $p_T(t)$  decomposes into a product of linear factors on the form  $t - a \in \mathbb{F}[t]$  (up to multiplication by a constant). This is in particular the case if  $\mathbb{F}$  is algebraically closed.

**PROOF.** (i): Let  $A = (a_{ij}) \in \operatorname{Mat}_n(\mathbb{F})$  be a matrix representation of  $T$ . The  $(i, j)$ -th entry of  $tI - A$  is then  $t\delta_{ij} - a_{ij}$ , so

$$\det(tI - T) = \sum_{\sigma \in S_n} (\operatorname{sgn} \sigma) (t\delta_{1\sigma(1)} - a_{1\sigma(1)}) \cdots (t\delta_{n\sigma(n)} - a_{n\sigma(n)}) \quad (3.1)$$

by Theorem 3.5. Thus  $p_T(t)$  is a polynomial in  $t$ . Furthermore, the only entries in  $tI - A$  containing  $t$  are the diagonal entries, and the largest number of such entries occurring in a single term of (3.1) is  $n$ , so  $\deg p_T(t) \leq n$ . But notice that there is only one term in which  $t$  appears  $n$  times, namely the term corresponding to the identity permutation in  $S_n$ , giving the product of the diagonal entries in  $tI - A$ . This term equals

$$(t - a_{11})(t - a_{22}) \cdots (t - a_{nn}), \quad (3.2)$$

and multiplying out we see that the only resulting term containing  $t^n$  is  $t^n$  itself. Hence  $p_T(t)$  is monic and of degree  $n$ . Thus we may write  $p_T(t) = \sum_{i=0}^n c_i t^i$  for appropriate  $c_0, \dots, c_n \in \mathbb{F}$ .

(ii): Simply notice that

$$(-1)^n \det T = \det(-T) = p_T(0) = c_0$$

by  $n$ -linearity of  $\det$  and the definition of  $p_T(t)$ .

(iii): The only way for one of the terms in (3.1) to contain the factor  $t^{n-1}$  is for at least  $n-1$  of the  $b_{ij}$  to be a diagonal element. But in choosing  $n-1$  elements along the diagonal we are forced to also choose the final diagonal element, since otherwise  $\sigma$  would not be a permutation. Hence the factor  $t^n$  can only appear in the product (3.2). It is then clear that

$$c_{n-1} = -(a_{11} + \cdots + a_{nn}) = -\operatorname{tr} T$$

as claimed.

(iv): Now assume that  $p_T(t)$  splits over  $\mathbb{F}$ . Then some linear factor  $t - \lambda \in \mathbb{F}[t]$  divides  $p_T(t)$ , which implies that  $\lambda \in \mathbb{F}$  is an eigenvalue of  $T$ .

(v): Since  $p_T(t)$  is monic we have

$$p_T(t) = (t - \lambda_1)(t - \lambda_2) \cdots (t - \lambda_n)$$

for appropriate  $\lambda_1, \dots, \lambda_n \in \mathbb{F}$ . These are then the (not necessarily distinct) eigenvalues of  $T$ . Thus  $p_T(0) = (-1)^n \lambda_1 \cdots \lambda_n$ , and the claim follows from (ii).

(vi): We similarly find that  $c_{n-1} = -(\lambda_1 + \cdots + \lambda_n)$ , so the final claim follows from (iii).  $\square$

### 3.5. Proofs without determinants

#### Existence of eigenvalues

Assume that  $\mathbb{F}$  is algebraically closed, and consider  $T \in \mathcal{L}(V)$ . For  $d \in \mathbb{N}$ , let  $\mathbb{F}[t]_d$  denote the vector space of polynomials in  $\mathbb{F}[t]$  with degree strictly less than  $d$ , such that  $\dim \mathbb{F}[t]_d = d$ . Consider the map  $\text{ev}_T: \mathbb{F}[t]_{n^2+1} \rightarrow \mathcal{L}(V)$  given by  $\text{ev}_T(p) = p(T)$ . This cannot be injective, so there is some nonzero  $p(t) \in \mathbb{F}[t]_{n^2+1}$  such that  $p(T) = 0$ . Note that  $p(t)$  cannot be constant.

Since  $\mathbb{F}$  is algebraically closed, there exist  $c, \lambda_1, \dots, \lambda_m \in \mathbb{F}$  such that  $p(t) = c \prod_{i=1}^m (t - \lambda_i)$ . But then

$$0 = p(T) = c \prod_{i=1}^m (T - \lambda_i I),$$

so at least one  $T - \lambda_i I$  is not injective. Hence  $\lambda_i$  is an eigenvalue of  $T$ .

#### Trace is sum of eigenvalues

##### COROLLARY 3.13

Let  $\mathbb{F}$  be algebraically closed, and let  $T \in \mathcal{L}(V)$ . Then the sum of the eigenvalues of  $T$  is  $\text{tr } T$ .

**PROOF.** Let  $A \in \text{Mat}_n(\mathbb{F})$  be an upper triangular matrix [TODO reference to later, perhaps move things around.] for  $T$ . The diagonal elements of  $A$  are the eigenvalues, and the trace of  $T$  is just the sum of these elements.  $\square$

### 3.6. Cross products

**DEFINITION 3.14:** *Cross products*

Let  $v = (\alpha_1, \alpha_2, \alpha_3)$  and  $w = (\beta_1, \beta_2, \beta_3)$  be vectors in  $\mathbb{R}^3$ . The *cross product* of  $v$  and  $w$  is the vector

$$v \times w = \begin{pmatrix} \alpha_2\beta_3 - \alpha_3\beta_2 \\ \alpha_3\beta_1 - \alpha_1\beta_3 \\ \alpha_1\beta_2 - \alpha_2\beta_1 \end{pmatrix}.$$

Denote the standard basis on  $\mathbb{R}^3$  by  $\mathcal{E} = (e_1, e_2, e_3)$ . We easily see that  $e_i \times e_j = e_k$  when  $(i, j, k)$  is a cyclic permutation of  $(1, 2, 3)$ .

**LEMMA 3.15**

Let  $v, w, u \in \mathbb{R}^3$ . Then

$$\langle u, v \times w \rangle = \det(u, v, w).$$

**PROOF.** By multilinearity of the inner product and of determinants, it suffices to prove the lemma when  $u$  is a basis vector. But it is clear that

$$\langle e_i, v \times w \rangle = \det(e_i, v, w),$$

as desired. □

The product  $\langle u, v \times w \rangle$  is called the (*scalar*) *triple product* of  $u$ ,  $v$  and  $w$ , and is denoted  $[u, v, w]$ . We call it the *scalar* triple product to distinguish it from the *vector* triple product  $u \times (v \times w)$ , whose properties we will examine in [Corollary 3.18](#). The scalar triple product has some very nice properties summarised in the following proposition:

**PROPOSITION 3.16**

Let  $u, v, w \in \mathbb{R}^3$ .

- (i) The cross product map  $(v, w) \mapsto v \times w$  is bilinear.
- (ii)  $v \times w = -w \times v$ .
- (iii) The triple product  $[u, v, w]$  is invariant under cyclic permutations, i.e.

$$[u, v, w] = [v, w, u] = [w, u, v]$$

and invariant under interchange of inner product and cross product, i.e.

$$\langle u, v \times w \rangle = [u, v, w] = \langle u \times v, w \rangle.$$

- (iv)  $v \times w = 0$  if and only if  $v$  and  $w$  are linearly dependent.

(v)  $v \times w$  is orthogonal to both  $v$  and  $w$ .

**PROOF.** The first three claims follow from [Lemma 3.15](#) since the determinant is multilinear and alternating (hence skew-symmetric).

For the fourth claim, if  $v$  and  $w$  are linearly dependent then  $\det(u, v, w) = 0$  for all  $u \in \mathbb{R}^3$ , so  $v \times w = 0$ . Conversely, if  $v$  and  $w$  are linearly independent, then extending to a basis  $(u, v, w)$  for  $\mathbb{R}^3$  we have  $\det(u, v, w) \neq 0$ , implying that  $v \times w \neq 0$ .

To prove the final claim, notice that

$$\langle v, v \times w \rangle = \det(v, v, w) = 0,$$

and similarly for  $w$ . □

#### PROPOSITION 3.17

Let  $a, b, v, w \in \mathbb{R}^3$ . Then

$$\langle a \times b, v \times w \rangle = \det \begin{pmatrix} \langle a, v \rangle & \langle b, v \rangle \\ \langle a, w \rangle & \langle b, w \rangle \end{pmatrix}.$$

In particular,

$$\|v \times w\|^2 = \det \begin{pmatrix} \|v\|^2 & \langle v, w \rangle \\ \langle v, w \rangle & \|w\|^2 \end{pmatrix}.$$

The latter identity is just Lagrange's identity in three dimensions. If  $\theta$  is the angle between  $v$  and  $w$ , then  $\langle v, w \rangle = \|v\|\|w\|\cos \theta$ , so

$$\|v \times w\|^2 = \|v\|^2\|w\|^2 - \langle v, w \rangle^2 = \|v\|^2\|w\|^2(1 - \cos^2 \theta) = \|v\|^2\|w\|^2 \sin^2 \theta.$$

Hence  $\|v \times w\| = \|v\|\|w\||\sin \theta|$ , which is the area of the parallelogram spanned by  $v$  and  $w$ . If  $u \in \mathbb{R}^3$  is another vector and  $\varphi$  is the angle between  $u$  and the normal of the plane spanned by  $v$  and  $w$  (e.g.  $v \times w$ ), then

$$|[u, v, w]| = |\langle u, v \times w \rangle| = \|u\|\|v \times w\||\cos \varphi| = \|u\|\|v\|\|w\||\sin \theta \cos \varphi|.$$

But this is the volume of the parallelepiped spanned by  $u, v$  and  $w$ . This gives a geometric interpretation (or 'proof') of the invariance of the scalar triple product.

**PROOF.** By linearity it suffices to prove the identity when the four vectors are basis vectors. If  $a = b$  or  $v = w$  then both sides are zero, so we may assume that  $a = e_i, b = e_j, v = e_k$  and  $w = e_l$  with  $i \neq j$  and  $k \neq l$ . By potentially swapping  $a$  and  $b$  and/or  $v$  and  $w$  we may assume that  $e_i \times e_j = e_p$  and  $e_k \times e_l = e_q$  for some  $p, q \in \{1, 2, 3\}$ .

If  $p = q$  then  $i = k$  and  $j = l$ , so both sides equal 1. If instead  $p \neq q$ , then the two cross products on the left-hand side are orthogonal, so the inner product is zero. Furthermore, either  $k$  or  $l$  equals  $p$ , so one of the rows in the right-hand side matrix is zero, and hence the determinant is zero.  $\square$

#### COROLLARY 3.18

Let  $u, v, w \in \mathbb{R}^3$ . Then

$$u \times (v \times w) = v\langle u, w \rangle - w\langle u, v \rangle. \quad (3.3)$$

In particular, the cross product satisfies the Jacobi identity

$$u \times (v \times w) + v \times (w \times u) + w \times (u \times v) = 0. \quad (3.4)$$

The identity (3.3) is sometimes called the ‘bac-cab rule’, a name that would have been self-explanatory had we used the names  $a, b$  and  $c$  instead of  $u, v$  and  $w$ . Note that to conform to this rule we need to write the vectors before the scalars.

**PROOF.** For  $x \in \mathbb{R}^3$  we have

$$\begin{aligned} \langle x, u \times (v \times w) \rangle &= [x, u, v \times w] \\ &= \langle x \times u, v \times w \rangle \\ &= \det \begin{pmatrix} \langle x, v \rangle & \langle u, v \rangle \\ \langle x, w \rangle & \langle u, w \rangle \end{pmatrix} \\ &= \langle x, v \rangle \langle u, w \rangle - \langle u, v \rangle \langle x, w \rangle \\ &= \langle x, v \langle u, w \rangle - w \langle u, v \rangle \rangle. \end{aligned}$$

The claim then follows since  $x$  was arbitrary.  $\square$

#### LEMMA 3.19

Let  $A \in \text{Mat}_d(\mathbb{R})$ . Every neighbourhood of  $A$  contains an invertible matrix different from  $A$ . In particular, there exists a sequence  $(A_n)_{n \in \mathbb{N}}$  of invertible matrices different from  $A$  such that  $A_n \rightarrow A$  for  $n \rightarrow \infty$ .

Since  $\text{Mat}_d(\mathbb{R})$  is a finite-dimensional vector space, it has a unique vector space topology. More concretely, all norms on  $\text{Mat}_d(\mathbb{R})$  are Lipschitz equivalent, so we may choose whatever norm we wish. We choose the Euclidean norm, identifying  $\text{Mat}_d(\mathbb{R})$  with  $\mathbb{R}^{d^2}$ .

**PROOF.** Let  $t \in \mathbb{R} \setminus \{0\}$ . Then  $A - tI$  is invertible if and only if  $\det(A - tI) \neq 0$ , but  $\det(A - tI)$  is a polynomial in  $t$ , so it has finitely many roots. Hence the nonzero roots of  $\det(A - tI)$  are bounded away from zero, so since  $A - tI \rightarrow A$  as  $t \rightarrow 0$ , the claim follows.  $\square$

**PROPOSITION 3.20: Transformation of cross products**

Let  $u, v, w \in \mathbb{R}^3$ , and let  $A \in \text{Mat}_3(\mathbb{R})$ . Then we have the following:

- (i)  $[Au, Av, Aw] = (\det A)[u, v, w]$ .
- (ii)  $Av \times Aw = (\text{cof } A)(v \times w) = (\text{adj } A)^\top(v \times w)$ .
- (iii) If  $A$  is orthogonal, then  $A(v \times w) = (\det A)(Av \times Aw)$ .

This gives a geometric interpretation of the determinant. If  $[u, v, w]$  is the signed volume of the parallelepiped spanned by  $u, v$  and  $w$ , and  $[Au, Av, Aw]$  is the signed volume of the parallelepiped spanned by  $Au, Av$  and  $Aw$ , then  $\det A$  is the factor by which this volume is increasing when applying  $A$  to each of  $u, v$  and  $w$ . In particular, this explains why the determinant of  $A$  is zero if and only if  $A$  is singular: This means that  $A$  sends a basis of  $\mathbb{R}^3$  to a linearly dependent set, and the parallelepiped spanned by such a set has zero volume.

**PROOF. Proof of (i):** Simply notice that

$$[Au, Av, Aw] = \det(Au, Av, Aw) = (\det A) \det(u, v, w) = (\det A) \langle u, v \times w \rangle,$$

where the second equality follows since  $\det(Au, Av, Aw)$  is also the determinant of the matrix

$$(Au \mid Av \mid Aw) = A(u \mid v \mid w),$$

and the determinant is multiplicative.

**Proof of (ii):** First assume that  $A$  is invertible. Then replacing  $u$  with  $A^{-1}u$  in (i) we obtain

$$\begin{aligned} \langle u, Av \times Aw \rangle &= (\det A) \langle A^{-1}u, v \times w \rangle \\ &= (\det A) \langle u, (A^{-1})^\top(v \times w) \rangle \\ &= \langle u, (\text{cof } A)(v \times w) \rangle, \end{aligned}$$

where the last equality follows from [Proposition 3.10](#). Hence we obtain the desired identity when  $A$  is invertible. Finally notice that both the maps  $A \mapsto \text{cof } A$  and  $A \mapsto Av \times Aw$  are continuous. Hence the claim for general  $A$  follows from [Lemma 3.19](#).

**Proof of (iii):** Notice that  $A^{-1} = A^\top$ , so this follows immediately from (ii).  $\square$



If  $A$  is a proper rotation, i.e. if  $A$  is orthogonal and  $\det A = 1$ , then [Proposition 3.20\(iii\)](#) implies that  $A(v \times w) = Av \times Aw$ . This allows us to define a cross product on any three-dimensional inner product space, when this is equipped with an orientation.

First, if  $\mathcal{V}$  and  $\mathcal{W}$  are ordered bases for any finite-dimensional real vector space  $V$ , then we say that  $\mathcal{V}$  and  $\mathcal{W}$  have the *same orientation* if the change of basis operator  $\varphi_{\mathcal{W}, \mathcal{V}}$  has positive determinant. It follows that orientation partitions the set of ordered bases for  $V$  into two *orientation classes*, each called an *orientation* of  $V$ . If  $V$  is equipped with an orientation  $\mathcal{O}$ , then we call this class the *positive orientation* of  $V$ , and the other class the *negative orientation* of  $V$ . An ordered basis for  $V$  is called *positive* if it lies in  $\mathcal{O}$  and *negative* if it does not.

Returning to the case where  $V$  is three-dimensional and equipped with an orientation, let  $\mathcal{V}$  and  $\mathcal{W}$  be positive ordered orthonormal bases for  $V$ . For vectors  $v, w \in V$  we can then consider the cross products of their coordinate vectors, i.e.

$$[v]_{\mathcal{V}} \times [w]_{\mathcal{V}} \quad \text{and} \quad [v]_{\mathcal{W}} \times [w]_{\mathcal{W}}.$$

Since  ${}_{\mathcal{W}}[\square]_{\mathcal{V}}$  is orthogonal with determinant 1, we have

$${}_{\mathcal{W}}[\square]_{\mathcal{V}}([v]_{\mathcal{V}} \times [w]_{\mathcal{V}}) = {}_{\mathcal{W}}[\square]_{\mathcal{V}} \cdot [v]_{\mathcal{V}} \times {}_{\mathcal{W}}[\square]_{\mathcal{V}} \cdot [w]_{\mathcal{V}} = [v]_{\mathcal{W}} \times [w]_{\mathcal{W}}.$$

Hence we have

$$\varphi_{\mathcal{V}}^{-1}([v]_{\mathcal{V}} \times [w]_{\mathcal{V}}) = \varphi_{\mathcal{W}}^{-1}([v]_{\mathcal{W}} \times [w]_{\mathcal{W}}),$$

so we may define the cross product of  $v$  and  $w$  as  $v \times w = \varphi_{\mathcal{V}}^{-1}([v]_{\mathcal{V}} \times [w]_{\mathcal{V}})$  where  $\mathcal{V}$  is any positive ordered orthonormal basis for  $V$ . Notice that this means that  $[v \times w]_{\mathcal{V}} = [v]_{\mathcal{V}} \times [w]_{\mathcal{V}}$ .

This allows us to generalise most of the above results to general vector spaces. For instance, using that the coordinate map  $\varphi_{\mathcal{V}}$  is an isometry, the scalar triple product of  $u, v, w \in V$  is given by

$$[u, v, w] = \langle u, v \times w \rangle = \langle [u]_{\mathcal{V}}, [v \times w]_{\mathcal{V}} \rangle = \langle [u]_{\mathcal{V}}, [v]_{\mathcal{V}} \times [w]_{\mathcal{V}} \rangle = \left[ [u]_{\mathcal{V}}, [v]_{\mathcal{V}}, [w]_{\mathcal{V}} \right],$$

and hence it has all the properties of the scalar triple product on  $\mathbb{R}^3$ , such as invariance under cyclic permutations. Notice also that it is indeed a *scalar* quantity, in the sense that it is invariant under a change of basis. Similarly, the ‘bac-cab rule’ [\(3.3\)](#) becomes

$$\begin{aligned} [u \times (v \times w)]_{\mathcal{V}} &= [u]_{\mathcal{V}} \times [v \times w]_{\mathcal{V}} \\ &= [u]_{\mathcal{V}} \times ([v]_{\mathcal{V}} \times [w]_{\mathcal{V}}) \\ &= [v]_{\mathcal{V}} \langle [u]_{\mathcal{V}}, [w]_{\mathcal{V}} \rangle - [w]_{\mathcal{V}} \langle [u]_{\mathcal{V}}, [v]_{\mathcal{V}} \rangle \\ &= [v]_{\mathcal{V}} \langle u, w \rangle - [w]_{\mathcal{V}} \langle u, v \rangle \\ &= [v \langle u, w \rangle - w \langle u, v \rangle]_{\mathcal{V}}. \end{aligned}$$

Hence  $u \times (v \times w) = v\langle u, w \rangle - w\langle u, v \rangle$  since  $\varphi_V$  is an isomorphism. In particular, the cross product on  $V$  also satisfies the Jacobi identity (3.4), so  $V$  becomes a Lie algebra whose Lie bracket is given by the cross product, i.e.  $[v, w] = v \times w$ .

## 4 • Complexification

If  $W$  is a complex vector space, then we may restrict the scalar multiplication  $\mathbb{C} \times W \rightarrow W$  to a map  $\mathbb{R} \times W \rightarrow W$ . When we equip  $W$  with this restricted scalar multiplication instead of the original one, we call the resulting space the *real version* of  $W$  and denote it by  $W_{\mathbb{R}}$ .

Conversely, if  $V$  is a real vector space then we define the *complexification* of  $V$  as the vector space  $V^{\mathbb{C}}$  whose underlying set is  $V \times V$ , and which is equipped with componentwise addition and the complex scalar multiplication

$$(a + ib)(v, u) = (av - bu, au + bv),$$

for  $a, b \in \mathbb{R}$  and  $v, u \in V$ . We denote the vector  $(v, u)$  by  $v + iu$ .

If  $T: V \rightarrow W$  is a linear map between real vector spaces, then we define the complexification of  $T$  by

$$\begin{aligned} T^{\mathbb{C}}: V^{\mathbb{C}} &\rightarrow W^{\mathbb{C}}, \\ v + iu &\mapsto Tv + iTu. \end{aligned}$$

That is,  $T^{\mathbb{C}}$  is just the product map  $T \times T$ . This is easily seen to be complex-linear.

If  $V$  is a real inner product space, then we define an inner product by

$$\langle v + iu, x + iy \rangle = \langle v, x \rangle + \langle u, y \rangle + i(\langle u, x \rangle - \langle v, y \rangle).$$

Notice that this identity holds in any *complex* inner product space, where the notation  $v + iu$  instead means the sum of  $v$  and the scalar product of  $i$  and  $u$  (in justifying this claim, the reader will recall that the inner product on a complex space is *sesquilinear*).

## 5 • Operator adjoints

### DEFINITION 5.1: Operator adjoints

Let  $V$  and  $W$  be  $\mathbb{F}$ -vector spaces, and let  $T: V \rightarrow W$  be a linear map. The



(operator) adjoint of  $T$  is the pullback

$$\begin{aligned} T^*: W^* &\rightarrow V^*, \\ \varphi &\mapsto \varphi \circ T. \end{aligned}$$

Note that this is just the action of the dual functor on maps in the category of  $\mathbb{F}$ -vector spaces. Hence it already satisfies  $\text{id}_V^* = \text{id}_{V^*}$  and  $(ST)^* = T^*S^*$ , so that in particular  $(T^{-1})^* = (T^*)^{-1}$  when  $T$  is invertible. Furthermore, it is easy to show that the map  $T \mapsto T^*$  is linear. It is also injective, since if  $Tv \neq Sv$  then there is a  $\varphi \in W^*$  such that  $\varphi(Tv) \neq \varphi(Sv)$ . If  $V$  and  $W$  are finite-dimensional, it is therefore a linear isomorphism.

#### PROPOSITION 5.2

Let  $T \in \mathcal{L}(V, W)$ .

- (i)  $\ker T^* = (\text{im } T)^0$ .
- (ii)  $\text{im } T^* = (\ker T)^0$ .

**PROOF.** Roman (2008, Theorem 3.19). □

#### COROLLARY 5.3

If  $T \in \mathcal{L}(V, W)$  with  $V$  and  $W$  finite-dimensional, then  $\text{rank } T^* = \text{rank } T$ .

**PROOF.** Recall that the dimension of  $(\ker T)^0$  equals the codimension of  $\ker T$ , which is just  $\dim V - \dim \ker T$  when  $V$  is finite-dimensional (cf. Roman 2008, Theorem 3.15). We then have

$$\text{rank } T^* = \dim \text{im } T^* = \dim (\ker T)^0 = \dim V - \dim \ker T = \dim \text{im } T = \text{rank } T,$$

as desired. □

Note that if  $\mathcal{V} = (v_1, \dots, v_n)$  is an ordered basis for  $V$ ,  $\mathcal{V}^*$  the corresponding dual basis, and  $\mathcal{V}^{**}$  the double dual basis, then for  $\varphi = \varphi_1 v_1^* + \dots + \varphi_n v_n^*$  we have

$$v_i^{**}(\varphi) = \varphi_i = \varphi(v_i),$$

since both  $v_i^*(v_j) = \delta_{ij}$  and  $v_i^{**}(v_j^*) = \delta_{ij}$ , by definition of the dual basis.

#### PROPOSITION 5.4

If  $T \in \mathcal{L}(V, W)$  is a linear map between finite-dimensional vector spaces, and  $\mathcal{V}$

and  $\mathcal{W}$  are ordered bases for  $V$  and  $W$  respectively, then

$${}_{\mathcal{V}}[T^*]_{\mathcal{W}^*} = ({}_{\mathcal{W}}[T]_{\mathcal{V}})^{\top}.$$

**PROOF.** Write  $\mathcal{V} = (v_1, \dots, v_n)$  and  $\mathcal{W} = (w_1, \dots, w_m)$ . Then

$$({}_{\mathcal{W}}[T]_{\mathcal{V}})_{ij} = ([Tv_j]_{\mathcal{W}})_i = w_i^*(Tv_j),$$

and

$$({}_{\mathcal{V}}[T^*]_{\mathcal{W}^*})_{ij} = ([T^*w_j^*]_{\mathcal{V}^*})_i = v_i^*(T^*w_j^*) = T^*w_j^*(v_i) = w_j^*(Tv_i).$$

These expressions are the same, but with  $i$  and  $j$  switched.  $\square$

#### COROLLARY 5.5

The row rank and the column rank of a matrix  $A \in \text{Mat}_{m,n}(\mathbb{F})$  are equal.

**PROOF.** The matrix representation of the multiplication operator  $M_A$  with respect to the standard bases on  $\mathbb{F}^n$  and  $\mathbb{F}^m$  is just  $A$  itself, and [Proposition 5.4](#) then implies that the matrix representation of  $(M_A)^*$  with respect to the dual bases is  $A^{\top}$ . But the rank of an operator equals the rank of any matrix representation of that operator, so [Corollary 5.3](#) implies that  $A$  and  $A^{\top}$  have the same (column) rank. Finally, the column rank of  $A^{\top}$  is the row rank of  $A$ , proving the claim.  $\square$

If  $V$  is a finite-dimensional inner product space, for  $v \in V$  let  $\varphi_v$  denote the element in  $V^*$  given by  $\varphi_v(w) = \langle w, v \rangle$ . Further, let  $\Phi_V: V \rightarrow V^*$  denote the (conjugate-)linear isomorphism  $v \mapsto \varphi_v$ .

#### THEOREM 5.6

Let  $V$  and  $W$  be finite-dimensional inner product spaces, and let  $T \in \mathcal{L}(V, W)$ . Denoting the Hilbert space adjoint of  $T$  by  $T^{\dagger}: W \rightarrow V$  we have

$$T^* = \Phi_V \circ T^{\dagger} \circ \Phi_W^{-1},$$

i.e. the diagram

$$\begin{array}{ccc} V & \xrightleftharpoons[T^{\dagger}]{} & W \\ \Phi_V \downarrow & & \downarrow \Phi_W \\ V^* & \xleftarrow{T^*} & W^* \end{array}$$

commutes. [TODO also commutes when  $T$  is there?]

**PROOF.** Simply notice that, for  $v \in V$  and  $\varphi \in W^*$ , we have

$$T^*\varphi(v) = \varphi(Tv) = \langle Tv, \Phi_W^{-1}(\varphi) \rangle = \langle v, T^{\dagger}\Phi_W^{-1}(\varphi) \rangle = \Phi_V(T^{\dagger}\Phi_W^{-1}(\varphi))(v),$$

which implies the claim.  $\square$

## 6 • Triangularisation and diagonalisation

### 6.1. Triangularisation

Recall that a matrix  $A = (a_{ij}) \in \text{Mat}_n(R)$  is called *upper triangular* if  $a_{ij} = 0$  whenever  $i > j$ . If  $V$  is an  $n$ -dimensional  $\mathbb{F}$ -vector space and  $\mathcal{V}$  is an ordered basis for  $V$ , then we say that the operator  $T \in \mathcal{L}(V)$  is *upper triangular with respect to  $\mathcal{V}$*  if the matrix representation  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$  is upper triangular.

A subspace  $U$  of a vector space  $V$  is said to be *invariant under  $T \in \mathcal{L}(V)$*  if  $T(U) \subseteq U$ .

#### PROPOSITION 6.1

Let  $V$  be an  $\mathbb{F}$ -vector space with  $n = \dim V < \infty$ , and let  $\mathcal{V} = (v_1, \dots, v_n)$  be an ordered basis for  $V$ . An operator  $T \in \mathcal{L}(V)$  is upper triangular with respect to  $\mathcal{V}$  if and only if  $\text{span}(v_1, \dots, v_i)$  is invariant under  $T$  for all  $i \in \{1, \dots, n\}$ .

**PROOF.** This is obvious. □

#### LEMMA 6.2

Let  $V$  be an  $\mathbb{F}$ -vector space, and let  $T \in \mathcal{L}(V)$  be an isomorphism. If  $U$  is a finite-dimensional subspace of  $V$  that is invariant under  $T$ , then  $U$  is also invariant under  $T^{-1}$ .

**PROOF.** Since  $U$  is finite-dimensional and  $T|_U: U \rightarrow U$  is injective, applying the rank–nullity theorem implies that  $T|_U$  is also surjective. Hence if  $u \in U$ , then there exists a  $v \in U$  such that  $Tv = u$ . It follows that

$$T^{-1}u = T^{-1}Tv = v \in U,$$

so  $U$  is invariant under  $T^{-1}$ . □

#### PROPOSITION 6.3

Let  $V$  be a finite-dimensional  $\mathbb{F}$ -vector space, and let  $\mathcal{V}$  be an ordered basis for  $V$ . If  $T \in \mathcal{L}(V)$  is an isomorphism that is upper triangular with respect to  $\mathcal{V}$ , then  $T^{-1}$  is also upper triangular with respect to  $\mathcal{V}$ .

In particular, the subset of  $\text{GL}_n(\mathbb{F})$  consisting of upper triangular matrices is a subgroup.

**PROOF.** This is an obvious consequence of the above two results. □

**LEMMA 6.4**

Let  $A \in \text{Mat}_n(\mathbb{F})$  be upper triangular. Then  $A$  is invertible if and only if all its diagonal elements are nonzero.

**PROOF.** Denote the diagonal elements of  $A$  by  $\lambda_1, \dots, \lambda_n$ , and let  $(e_1, \dots, e_n)$  be the standard basis of  $\mathbb{F}^n$ . First assume that the diagonal elements are nonzero. Then notice that  $e_1 \in R(A)$ , and that

$$Ae_i = a_{1i}e_1 + \dots + a_{i-1,i}e_{i-1} + \lambda_i e_i$$

for appropriate  $a_1, \dots, a_{i-1} \in \mathbb{F}$ . By induction we then have  $e_i \in R(A)$ . Since  $(e_1, \dots, e_n)$  is a basis, this implies that  $R(A) = \mathbb{F}^n$ .

Conversely, assume that some diagonal element  $\lambda_i$  is zero. Then

$$A\text{span}(e_1, \dots, e_i) \subseteq \text{span}(e_1, \dots, e_{i-1}),$$

so the null-space of  $A$  is nontrivial, and hence  $A$  is singular.  $\square$

**LEMMA 6.5**

Let  $A \in \text{Mat}_n(\mathbb{F})$  be upper triangular. Then the eigenvalues of  $A$  are its diagonal elements.

**PROOF.** Let  $\lambda \in \mathbb{F}$ , and denote the diagonal elements of  $A$  by  $\lambda_1, \dots, \lambda_n$ . By Lemma 6.4, the matrix  $\lambda I - A$  is singular if and only if  $\lambda - \lambda_i = 0$  for some  $i$ , and hence  $\lambda_1, \dots, \lambda_n$  are the eigenvalues of  $A$ .  $\square$

**PROPOSITION 6.6**

Let  $\mathbb{F}$  be algebraically closed, and let  $V$  be a finite-dimensional  $\mathbb{F}$ -vector space. If  $T \in \mathcal{L}(V)$ , then  $V$  has an ordered basis with respect to which  $T$  is upper triangular.

**PROOF.** This is obvious if  $\dim V = 1$ , so assume that  $n = \dim V > 1$ , and assume that the claim is true for  $\mathbb{F}$ -vector spaces of dimension  $n - 1$ . Since  $\mathbb{F}$  is algebraically closed,  $T$  has an eigenvector  $v_1 \in V$ . Then  $U = \text{span}(v_1)$  is invariant under  $T$ , so we may define a linear operator<sup>2</sup>  $\tilde{T} \in \mathcal{L}(V/U)$  by  $\tilde{T}(v + U) = Tv + U$ . Since  $\dim V/U = n - 1$ , by induction there is a basis  $v_2 + U, \dots, v_n + U$  of  $V/U$  with respect to which the matrix of  $\tilde{T}$  is upper triangular. It is easy to show that the collection  $v_1, \dots, v_n$  is linearly independent, hence a basis for  $V$ .

<sup>2</sup> The operator  $\tilde{T}$  may arise as follows: Let  $\pi: V \rightarrow V/U$  be the quotient map. Then  $U \subseteq \ker(\pi \circ T)$  since  $U$  is invariant under  $T$ , so  $\pi \circ T$  descends to a linear map  $\tilde{T}: V/U \rightarrow V/U$ .

Now notice that

$$Tv_i + U = \tilde{T}(v_i + U) \in \text{span}(v_2 + U, \dots, v_i + U)$$

for  $i \in \{2, \dots, n\}$ . That is, there exist  $a_2, \dots, a_i \in \mathbb{F}$  such that

$$Tv_i + U = (a_2v_2 + \dots + a_iv_i) + U.$$

But then  $Tv_i \in \text{span}(v_1, \dots, v_i)$  for all  $i \in \{2, \dots, n\}$ , and since  $U$  is invariant under  $T$  this also holds for  $i = 1$ . Hence  $T$  is upper triangular with respect to the basis  $v_1, \dots, v_n$  of  $V$ .  $\square$

#### THEOREM 6.7: Schur's Theorem

Let  $V$  be a finite-dimensional complex inner product space. If  $T \in \mathcal{L}(V)$ , then  $V$  has an ordered orthonormal basis with respect to which  $T$  is upper triangular.

**PROOF.** By Proposition 6.6  $V$  has an ordered basis  $\mathcal{V} = (v_1, \dots, v_n)$  with respect to which  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$  is upper triangular. Now apply the Gram–Schmidt procedure to  $\mathcal{V}$  and obtain an orthonormal basis  $\mathcal{U} = (u_1, \dots, u_n)$  for  $V$  such that

$$\text{span}(u_1, \dots, u_i) = \text{span}(v_1, \dots, v_i)$$

for all  $i \in \{1, \dots, n\}$ . Then Proposition 6.1 shows that  ${}_{\mathcal{U}}[T]_{\mathcal{U}}$  is also upper triangular, proving the claim.  $\square$

## 6.2. Orthonormal diagonalisation

Let  $V$  and  $W$  be finite-dimensional inner product spaces, and let  $T \in \mathcal{L}(V, W)$ . Recall that the *adjoint* of  $T$  is the operator  $T^* \in \mathcal{L}(W, V)$  with the property that

$$\langle T^*w, v \rangle_V = \langle w, Tv \rangle_W,$$

or by complex conjugation equivalently

$$\langle Tv, w \rangle_W = \langle v, T^*w \rangle_V,$$

for all  $v \in V$  and  $w \in W$ . An operator with this property is unique if it exists, since if  $S \in \mathcal{L}(W, V)$  is another such operator, then  $\langle v, Sw \rangle_V = \langle v, T^*w \rangle_V$  for all  $v$  and  $w$ , so  $S = T^*$ .

For existence, for  $w \in W$  define  $\psi_w \in V^*$  by  $\psi_w(v) = \langle Tv, w \rangle$ , and let  $\Psi_W: W \rightarrow V^*$  be the map  $\Psi_W(w) = \psi_w$ . Then define  $T^* = \Phi_V^{-1} \circ \Psi_W$ . Both  $\Phi_V$  and  $\Psi_W$  are (conjugate-)linear, so  $T^*$  is linear. Furthermore we have

$$\langle v, T^*w \rangle_V = \langle v, \Phi_V^{-1} \circ \Psi_W(w) \rangle_V = \psi_w(v) = \langle Tv, w \rangle_W$$

as required.



An operator  $U: V \rightarrow W$  is an *isometry* if

$$\langle Uv, Uu \rangle_W = \langle v, u \rangle_V$$

for all  $v, u \in V$ . Clearly  $U$  is injective. If  $U$  is also surjective (i.e. if  $\dim V = \dim W < \infty$ ), then it is called *unitary*. Notice that if  $U$  is an isometry, then

$$\langle U^*Uv, u \rangle_V = \langle Uv, Uu \rangle_W = \langle v, u \rangle_V,$$

implying that  $U^*U = \text{id}_V$ , and the converse clearly also holds. If  $U$  is also surjective, then it is an isomorphism and so also  $UU^* = \text{id}_W$  (an operator with this property is called a *coisometry*). In this case  $U^* = U^{-1}$ .

In the case  $W = V$  we say that  $T$  is *normal* if  $TT^* = T^*T$ , and that  $T$  is *self-adjoint* if  $T^* = T$ . Clearly both self-adjoint and unitary operators (with  $V = W$ ) are normal.

#### LEMMA 6.8

Let  $V$  and  $W$  be finite-dimensional inner product spaces, and let  $\mathcal{V}$  and  $\mathcal{W}$  be ordered orthonormal bases for  $V$  and  $W$ .

(i) The coordinate map  $\varphi_{\mathcal{V}}$  is unitary, i.e.

$$\langle [v]_{\mathcal{V}}, [u]_{\mathcal{V}} \rangle = \langle v, u \rangle \quad (6.1)$$

for all  $v, u \in V$ .

Let further  $T: V \rightarrow W$  be a linear map, and let  $A \in \text{Mat}_{m,n}(\mathbb{K})$ .

(ii)  $(M_A)^* = M_{A^*}$ . In particular, if  $V = \mathbb{K}^n$  and  $W = \mathbb{K}^m$  then  $\mathcal{M}(T^*) = \mathcal{M}(T)^*$ .

(iii)  $({}_{\mathcal{W}}[T]_{\mathcal{V}})^* = {}_{\mathcal{V}}[T^*]_{\mathcal{W}}$ .

**PROOF.** (i): By bi- or sesquilinearity of the inner product it suffices to prove (6.1) for a basis for  $V$ . And writing  $\mathcal{V} = (v_1, \dots, v_n)$  we find that

$$\langle [v_i]_{\mathcal{V}}, [v_j]_{\mathcal{V}} \rangle = \langle e_i, e_j \rangle = \delta_{ij} = \langle v_i, v_j \rangle$$

for  $1 \leq i, j \leq n$ .

(ii): Notice that

$$\langle M_{A^*}w, v \rangle = \langle A^*w, v \rangle = v^*(A^*w) = (Av)^*w = \langle w, Av \rangle = \langle w, M_A v \rangle$$

for all  $v \in \mathbb{K}^n$  and  $w \in \mathbb{K}^m$ . By uniqueness of the adjoint operator, it follows that  $(M_A)^* = M_{A^*}$ . Furthermore, we have

$$M_{\mathcal{M}(T^*)} = T^* = (M_{\mathcal{M}(T)})^* = M_{\mathcal{M}(T)^*}.$$

It follows that  $\mathcal{M}(T^*) = \mathcal{M}(T)^*$ .

(iii): Notice that

$$(\varphi_W \circ T \circ \varphi_V^{-1})^* = (\varphi_W \circ T \circ \varphi_V^*)^* = \varphi_V \circ T^* \circ \varphi_W^* = \varphi_V \circ T^* \circ \varphi_W^{-1},$$

and taking standard matrix representations, it follows from (ii) that  $({}_W[T]_V)^* = {}_V[T^*]_W$ .  $\square$

#### PROPOSITION 6.9

Let  $V$  be a finite-dimensional inner product space, and let  $T \in \mathcal{L}(V)$  and  $\lambda \in \mathbb{K}$ . Then  $\lambda \text{id}_V - T$  is invertible if and only if  $\bar{\lambda} \text{id}_V - T^*$  is invertible. In other words,  $\lambda$  is an eigenvalue of  $T$  if and only if  $\bar{\lambda}$  is an eigenvalue of  $T^*$ .

**PROOF.** Since the map  $T \mapsto T^*$  is idempotent it suffices to prove one implication, so assume that  $\lambda \text{id}_V - T$  is invertible. Then there exists an  $S \in \mathcal{L}(V)$  such that

$$S(\lambda \text{id}_V - T) = (\lambda \text{id}_V - T)S = \text{id}_V,$$

and taking adjoints we find that

$$(\bar{\lambda} \text{id}_V - T^*)S^* = S^*(\bar{\lambda} \text{id}_V - T^*) = \text{id}_V.$$

That is,  $\bar{\lambda} \text{id}_V - T^*$  is invertible as claimed.  $\square$

**REMARK 6.10.** Note that this does *not* say that  $v \in V$  is an eigenvector of  $T^*$  if it is an eigenvector of  $T$ . A counterexample is given by the matrix

$$A = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix},$$

which has the eigenvector  $(1, 0)$  with eigenvalue 1. However, while 1 is also an eigenvalue of the transpose  $A^\top$  (with eigenvector  $(1, 1)$ ),  $(1, 0)$  is not an eigenvector of  $A^\top$ .

While this does not hold in general, in [Proposition 6.14\(ii\)](#) we will see that it holds for *normal* operators.  $\lrcorner$

#### PROPOSITION 6.11

Let  $V$  and  $W$  be real inner product spaces, and let  $T \in \mathcal{L}(V, W)$ . Then we have

$$(T^\mathbb{C})^* = (T^*)^\mathbb{C},$$

i.e., the adjoint of the complexification of  $T$  is the complexification of the adjoint of  $T$ . In particular

- (i)  $T$  is normal if and only if  $T^\mathbb{C}$  is normal, and

(ii)  $T$  is self-adjoint if and only if  $T^{\mathbb{C}}$  is self-adjoint.

**PROOF.** For  $v, u, x, y \in V$  we have

$$\begin{aligned}
 \langle (T^*)^{\mathbb{C}}(x + iy), v + iu \rangle &= \langle T^*x + iT^*y, v + iu \rangle \\
 &= \langle T^*x, v \rangle + \langle T^*y, u \rangle + i(\langle T^*y, u \rangle - \langle T^*x, v \rangle) \\
 &= \langle x, Tv \rangle + \langle y, Tu \rangle + i(\langle y, Tu \rangle - \langle x, Tv \rangle) \\
 &= \langle x + iy, Tv + iTu \rangle \\
 &= \langle x + iy, T^{\mathbb{C}}(v + iu) \rangle.
 \end{aligned}$$

Uniqueness of adjoints thus yields the claim.

Assume that  $T$  is normal. Then

$$T^{\mathbb{C}}(T^{\mathbb{C}})^* = T^{\mathbb{C}}(T^*)^{\mathbb{C}} = (TT^*)^{\mathbb{C}} = (T^*T)^{\mathbb{C}} = (T^*)^{\mathbb{C}}T^{\mathbb{C}} = (T^{\mathbb{C}})^*T^{\mathbb{C}},$$

so  $T^{\mathbb{C}}$  is normal. The converse follows similarly. If  $T$  is self-adjoint, then

$$(T^{\mathbb{C}})^* = (T^*)^{\mathbb{C}} = T^{\mathbb{C}},$$

and similarly if  $T^{\mathbb{C}}$  is self-adjoint. □

#### LEMMA 6.12

Let  $V$  be a finite-dimensional vector space, let  $T \in \mathcal{L}(V)$ , and let  $\mathcal{V}$  be an ordered basis for  $V$ . Then  $v \in V$  is an eigenvector for  $T$  if and only if  $[v]_{\mathcal{V}}$  is an eigenvector for  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$  with the same eigenvalue.

**PROOF.** Let  $\lambda \in \mathbb{F}$  be the eigenvalue of  $v$ . Then

$${}_{\mathcal{V}}[T]_{\mathcal{V}} \cdot [v]_{\mathcal{V}} = [Tv]_{\mathcal{V}} = [\lambda v]_{\mathcal{V}} = \lambda[v]_{\mathcal{V}}.$$

For the converse, a similar calculation shows that  $[Tv]_{\mathcal{V}} = [\lambda v]_{\mathcal{V}}$ . Since  $\varphi_{\mathcal{V}}$  is an isomorphism, it follows that  $Tv = \lambda v$  as desired. □

#### LEMMA 6.13

Let  $V$  be a real vector space, and let  $T \in \mathcal{L}(V)$ . If  $\lambda \in \mathbb{R}$  is an eigenvalue of the complexification  $T^{\mathbb{C}}$  of  $T$ , then  $\lambda$  is also an eigenvalue of  $T$ .

**PROOF.** Let  $v + iu \in V^{\mathbb{C}}$  be an eigenvector of  $T^{\mathbb{C}}$  corresponding to  $\lambda$ . Then

$$Tv + iTu = T^{\mathbb{C}}(v + iu) = \lambda(v + iu) = \lambda v + i\lambda u.$$

It follows that  $Tv = \lambda v$  as desired. □



**PROPOSITION 6.14**

Let  $T \in \mathcal{L}(V)$  be a normal operator.

- (i)  $\|Tv\| = \|T^*v\|$  for all  $v \in V$ .
- (ii) If  $\lambda \in \mathbb{K}$  is an eigenvalue of  $T$ , then  $\bar{\lambda}$  is an eigenvalue of  $T^*$  with the same eigenvectors. In other words,  $E_T(\lambda) = E_{T^*}(\bar{\lambda})$ .
- (iii) If  $\mu \in \mathbb{K}$  is another eigenvalue of  $T$  distinct from  $\lambda$ , then  $E_T(\lambda)$  and  $E_T(\mu)$  are orthogonal.
- (iv) If  $T$  is self-adjoint, then it has an eigenvalue and all its eigenvalues are real.
- (v) If  $T$  is unitary, then all its eigenvalues lie on the unit circle  $S^1 \subseteq \mathbb{C}$ .

In [Corollary 6.18](#) we will prove the converses of (iv) and (v) under the assumption that  $T$  is normal, using the spectral theorem (cf. [Theorem 6.17](#)). We will use (iv) in the proof of the spectral theorem, and we have proved (v) already to make explicit that it does not depend on the spectral theorem.

**PROOF.** *Proof of (i):* Notice that

$$\|Tv\|^2 = \langle Tv, Tv \rangle = \langle T^*Tv, v \rangle = \langle TT^*v, v \rangle = \langle T^*v, T^*v \rangle = \|T^*v\|^2.$$

*Proof of (ii):* If  $T$  is normal then so is  $\lambda \text{id}_V - T$ , so (i) implies that

$$\|(\lambda \text{id}_V - T)v\| = \|(\bar{\lambda} \text{id}_V - T^*)v\|,$$

so  $v \in V$  is an eigenvector for  $T$  with eigenvalue  $\lambda$  if and only if  $v$  is an eigenvector for  $T^*$  with eigenvalue  $\bar{\lambda}$ .

*Proof of (iii):* Let  $v \in E_T(\lambda)$  and  $u \in E_T(\mu)$ . Since  $w$  is also an eigenvector for  $T^*$  with eigenvalue  $\bar{\mu}$ , we have

$$\lambda \langle v, u \rangle = \langle Tv, u \rangle = \langle v, T^*u \rangle = \mu \langle v, u \rangle.$$

Since  $\lambda \neq \mu$  we must have  $\langle v, u \rangle = 0$  as claimed.

*Proof of (iv):* If  $T$  is self-adjoint and  $v \in V$  is an eigenvector for  $T$  with  $\lambda \in \mathbb{K}$ , then

$$\lambda \langle v, v \rangle = \langle Tv, v \rangle = \langle v, Tv \rangle = \bar{\lambda} \langle v, v \rangle,$$

and since  $v \neq 0$  we must have  $\lambda = \bar{\lambda}$ . Hence  $\lambda$  is real.

If  $\mathbb{K} = \mathbb{C}$  then  $V$  has a complex eigenvalue, which is real by the above argument. Assume instead that  $\mathbb{K} = \mathbb{R}$  and consider the complexification  $T^{\mathbb{C}}$  of  $T$ . This is self-adjoint by [Proposition 6.11](#), so it has a real eigenvalue by the above. But then [Lemma 6.13](#) implies that this also is an eigenvalue of  $T$ .

*Proof of (v):* Let  $\lambda \in \mathbb{K}$  be an eigenvalue of  $T$  with eigenvector  $v$ . Then

$$\langle v, v \rangle = \langle Tv, Tv \rangle = \langle \lambda v, \lambda v \rangle = \lambda \bar{\lambda} \langle v, v \rangle = |\lambda|^2 \langle v, v \rangle,$$

so  $|\lambda| = 1$ .  $\square$

Let  $T: V \rightarrow V$  is an operator on an  $\mathbb{F}$ -vector space  $V$ , and let  $U$  be a subspace of  $V$  that is invariant under  $T$ . If  $W$  is a complement of  $U$ , i.e.  $V = U \oplus W$ , then  $W$  is not necessarily invariant under  $T$ . However, we have the following:

#### LEMMA 6.15

Let  $T \in \mathcal{L}(V)$  be an operator on a finite-dimensional inner product space  $V$ . If a subspace  $U$  of  $V$  is invariant under  $T$ , then  $U^\perp$  is invariant under  $T^*$ .

**PROOF.** Let  $v \in U^\perp$ . For  $u \in U$  we have  $Tu \in U$ , so

$$\langle T^*v, u \rangle = \langle v, Tu \rangle = 0.$$

Since this holds for all  $u \in U$ , it follows that  $T^*v \in U^\perp$  as desired.  $\square$

#### LEMMA 6.16

$V$  be a finite-dimensional inner product space over  $\mathbb{K}$ , and consider  $T \in \mathcal{L}(V)$ . If either

- (i)  $\mathbb{K} = \mathbb{R}$  and  $T$  is self-adjoint, or
- (ii)  $\mathbb{K} = \mathbb{C}$  and  $T$  is normal,

then  $T$  is orthogonally diagonalisable.

**PROOF.** Assume that either  $\mathbb{K} = \mathbb{R}$  and  $T$  is self-adjoint, or that  $\mathbb{K} = \mathbb{C}$  and  $T$  is normal. We prove by induction in  $n = \dim V$  that  $T$  is orthogonally diagonalisable. If  $n = 1$  then this follows since  $T$  has an eigenvalue, so assume that the claim is proved for operators on spaces of dimension strictly less than  $n$ .

Let  $\lambda \in \text{Spec } T$ , and consider the corresponding eigenspace  $E_T(\lambda)$ . If  $d := \dim E_T(\lambda) = n$ , then any orthonormal basis of  $E_T(\lambda)$  will suffice. Assume therefore that  $0 < d < n$ .

The space  $E_T(\lambda) = E_{T^*}(\bar{\lambda})$  is clearly invariant under both  $T$  and  $T^*$ . It follows from Lemma 6.15 that  $E_T(\lambda)^\perp$  is also invariant under both  $T$  and  $T^*$ . We furthermore have  $\dim E_T(\lambda)^\perp = n - d$  and  $0 < n - d < n$ . Let  $T_\parallel \in \mathcal{L}(E_T(\lambda))$  and  $T_\perp \in \mathcal{L}(E_T(\lambda)^\perp)$  denote the restrictions of  $T$  to  $E_T(\lambda)$  and  $E_T(\lambda)^\perp$  respectively. Both  $T_\parallel$  and  $T_\perp$  are also self-adjoint or normal, depending on the

hypothesis, so the induction hypothesis furnishes orthonormal bases  $\mathcal{U}$  and  $\mathcal{W}$  for  $E_T(\lambda)$  and  $E_T(\lambda)^\perp$  consisting of eigenvectors of  $T$ . But then  $\mathcal{V} = \mathcal{U} \cup \mathcal{W}$  is an orthonormal basis for  $V$  as desired.  $\square$

**THEOREM 6.17: The spectral theorem**

Let  $V$  be a finite-dimensional inner product space over  $\mathbb{K}$ , and let  $T \in \mathcal{L}(V)$ . Then the following are equivalent:

- (i)  $\mathbb{K} = \mathbb{R}$  and  $T$  is self-adjoint, or  $\mathbb{K} = \mathbb{C}$  and  $T$  is normal.
- (ii)  $T$  is orthogonally diagonalisable.
- (iii)  $T$  has the orthogonal spectral resolution

$$T = \sum_{\lambda \in \text{Spec } T} \lambda P_\lambda,$$

where  $P_\lambda$  is the orthogonal projection onto the eigenspace  $E_T(\lambda)$ . In particular,  $V$  is an orthogonal direct sum of the eigenspaces of  $T$ , i.e.

$$V = \bigoplus_{\lambda \in \text{Spec } T} E_T(\lambda).$$

- (iv)  $T$  is unitarily (when  $\mathbb{K} = \mathbb{C}$ ) or orthogonally (when  $\mathbb{K} = \mathbb{R}$ ) equivalent to a multiplication operator  $M_A \in \mathcal{L}(\mathbb{K}^n)$  where  $A$  is a diagonal matrix, and the diagonal of  $A$  contains the eigenvalues of  $T$  with multiplicity. If  $\mathcal{V}$  is an ordered orthonormal basis for  $V$  consisting of eigenvectors for  $T$ , then we may choose  $A = {}_{\mathcal{V}}[T]_{\mathcal{V}}$  and

$$T = \varphi_{\mathcal{V}}^{-1} \circ M_A \circ \varphi_{\mathcal{V}},$$

with  $\varphi_{\mathcal{V}}$  unitary.

Note that the first part of property (iii) means that

$$\text{id}_V = \sum_{\lambda \in \text{Spec } T} P_\lambda$$

is a resolution of the identity, i.e. that  $P_\lambda P_\mu = 0$  for  $\lambda \neq \mu$ , and that this is composed of orthogonal projections.

**PROOF.** (i)  $\Rightarrow$  (ii): This is just [Lemma 6.16](#).

(i) & (ii)  $\Rightarrow$  (iii): The first claim says that distinct eigenspaces are orthogonal, which is just a restatement of [Proposition 6.14\(iii\)](#). To prove the second,

let  $\mathcal{V} = (v_1, \dots, v_n)$  be an orthonormal basis for  $V$  consisting of eigenvectors for  $T$ , and let  $\lambda_1, \dots, \lambda_n$  be the corresponding eigenvalues. Then for any  $v = \alpha_1 v_1 + \dots + \alpha_n v_n$  we have  $P_{\lambda_i} v = \alpha_i v_i$ , so

$$\left( \sum_{\lambda \in \text{Spec } T} P_\lambda \right) v = \sum_{\lambda \in \text{Spec } T} P_\lambda v = \sum_{i=1}^n \alpha_i v_i = v.$$

For the third claim, notice that

$$\left( \sum_{\lambda \in \text{Spec } T} \lambda P_\lambda \right) v = \sum_{\lambda \in \text{Spec } T} \lambda P_\lambda v = \sum_{i=1}^n \lambda_i \alpha_i v_i = \sum_{i=1}^n \alpha_i T v_i = T v.$$

The final claim follows from the first two.

(iii)  $\Rightarrow$  (ii): This follows from the decomposition of  $V$  into an orthogonal sum of eigenspaces, by constructing an orthonormal basis for each eigenspace.

(ii)  $\Rightarrow$  (iv): Let  $\mathcal{V} = (v_1, \dots, v_n)$  be an ordered orthonormal basis for  $\mathcal{V}$  consisting of eigenvectors for  $T$  with corresponding eigenvalues  $\lambda_1, \dots, \lambda_n$ , and consider the matrix representation  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$ . If  $(e_1, \dots, e_n)$  is the standard basis on  $\mathbb{K}^n$ , then Lemma 6.12 implies that the vectors  $[v_i]_{\mathcal{V}} = e_i$  are eigenvectors for  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$ . Hence  ${}_{\mathcal{V}}[T]_{\mathcal{V}}$  is diagonal, so the basis representation  $\varphi_{\mathcal{V}} \circ T \circ \varphi_{\mathcal{V}}^{-1}$  is multiplication by a diagonal matrix. Next notice that

$$T = \varphi_{\mathcal{V}}^{-1} \circ (\varphi_{\mathcal{V}} \circ T \circ \varphi_{\mathcal{V}}^{-1}) \circ \varphi_{\mathcal{V}},$$

so it suffices to show that  $\varphi_{\mathcal{V}}$  is unitary (orthogonal). But this follows by Lemma 6.8.

(iv)  $\Rightarrow$  (i): First assume that  $\mathbb{K} = \mathbb{C}$ . Since  $\varphi_{\mathcal{V}}$  is unitary we have  $\varphi_{\mathcal{V}}^{-1} = \varphi_{\mathcal{V}}^*$ , so

$$T^* = (\varphi_{\mathcal{V}}^* \circ M_A \circ \varphi_{\mathcal{V}})^* = \varphi_{\mathcal{V}}^* \circ M_A^* \circ \varphi_{\mathcal{V}} = \varphi_{\mathcal{V}}^{-1} \circ M_{A^*} \circ \varphi_{\mathcal{V}}.$$

Since  $A$  is diagonal,  $T$  clearly commutes with  $T^*$ , hence is normal.

If instead  $\mathbb{K} = \mathbb{R}$ , the same argument shows that  $T^* = \varphi_{\mathcal{V}}^{-1} \circ M_{A^{\top}} \circ \varphi_{\mathcal{V}}$ , but since  $A$  is diagonal this is just  $T$ , so  $T$  is self-adjoint.  $\square$

#### COROLLARY 6.18

Let  $T \in \mathcal{L}(V)$  be a normal operator on a complex vector space  $V$ .

(i)  $T$  is self-adjoint if and only if  $\text{Spec } T \subseteq \mathbb{R}$ .

(ii)  $T$  is unitary if and only if  $\text{Spec } T \subseteq S^1$ .

Note that this does not hold on a real vector space, since then a normal operator is not necessarily diagonalisable.

**PROOF.** *Proof of (i):* The ‘only if’ part follows from [Proposition 6.14\(iv\)](#), so assume that  $\text{Spec } T \subseteq \mathbb{R}$  and notice that

$$T^* = \left( \sum_{\lambda \in \text{Spec } T} \lambda P_\lambda \right)^* = \sum_{\lambda \in \text{Spec } T} \bar{\lambda} P_\lambda^* = \sum_{\lambda \in \text{Spec } T} \lambda P_\lambda,$$

since each  $\lambda \in \mathbb{R}$ , and each  $P_\lambda$  is an orthogonal projection, hence self-adjoint.

Alternatively, choose a diagonal matrix  $A \in \text{Mat}_n(\mathbb{K})$  in accordance with [Theorem 6.17\(iv\)](#). Since the diagonal of  $A$  contains the eigenvalues of  $T$ , we have  $A^* = A$ , and so it follows that  $T^* = T$ .

*Proof of (ii):* Similarly, the ‘only if’ part is just [Proposition 6.14\(v\)](#). Assume that  $\text{Spec } T \subseteq S^1$  and notice that

$$T^* = \sum_{\lambda \in \text{Spec } T} \bar{\lambda} P_\lambda.$$

Since the projections  $P_\lambda$  are pairwise orthogonal, we have

$$T^*T = \sum_{\lambda \in \text{Spec } T} \bar{\lambda} \lambda P_\lambda = \sum_{\lambda \in \text{Spec } T} |\lambda|^2 P_\lambda = \sum_{\lambda \in \text{Spec } T} P_\lambda = \text{id}_V,$$

so  $U$  is unitary.

Alternatively, let  $A$  be as above. Then all diagonal elements in  $A$  are nonzero, so  $A$  is invertible, and we clearly have  $A^*A = I_n$ . Hence also  $T^*T = \text{id}_V$ , so  $T$  is unitary.  $\square$

## 7 • Projections

Let  $V$  be an  $\mathbb{F}$ -vector space. A linear operator  $P: V \rightarrow V$  is called a *projection* if it is idempotent, i.e. if  $P^2 = P$ .

### PROPOSITION 7.1

*A linear map  $P: V \rightarrow V$  is a projection if and only if there exist subspaces  $U$  and  $W$  of  $V$  such that  $V = U \oplus W$  and  $P|_U = \iota_U$ . In this case  $U = \text{im } P$  and  $W = \ker P$ .*

We say that  $P$  is the projection onto  $U$  along  $W$ .

**PROOF.** Assume that  $P$  is a projection, and let  $v \in \text{im } P$ . Then  $v = Pu$  for some  $u \in V$ , and

$$Pv = P^2u = Pu = v.$$

If also  $v \in \ker P$ , then  $v = 0$ . Furthermore, for any  $v \in V$  we have  $v = Pv + (v - Pv) \in \text{im } P \oplus \ker P$ , so  $\text{im } P$  and  $\ker P$  are indeed complements in  $V$ .

The converse is obvious, and so is the characterisation of  $U$  and  $W$ .  $\square$

Now let  $V$  be a real finite-dimensional<sup>3</sup> inner product space. A projection  $P: V \rightarrow V$  is *orthogonal* if  $\text{im } P$  and  $\ker P$  are orthogonal subspaces of  $V$ .

#### PROPOSITION 7.2

*A projection  $P: V \rightarrow V$  is orthogonal if and only if  $P$  is self-adjoint.*

**PROOF.** Say that  $P$  is a projection onto  $U$  along  $W$ . Assume that  $P$  is orthogonal and let  $v, w \in V$ . Since then  $Pv \in U$  and  $v - Pv \in W$ , and similarly for  $w$ , we get

$$\langle v - Pv, Pw \rangle = 0 = \langle Pv, w - Pw \rangle.$$

This implies that

$$\langle v, Pw \rangle = \langle Pv, Pw \rangle = \langle Pv, w \rangle = \langle v, P^*w \rangle,$$

which shows that  $P = P^*$ .

Conversely assume that  $P$  is self-adjoint. For  $u \in U$  and  $w \in W$  we then have

$$\langle u, w \rangle = \langle Pu, w \rangle = \langle u, Pw \rangle = \langle u, 0 \rangle = 0,$$

so  $U$  and  $W$  are orthogonal. □

#### PROPOSITION 7.3

*Let  $T: V \rightarrow W$  be an injective linear operator between real inner product spaces  $V$  and  $W$ , and let  $P$  be the orthogonal projection onto  $\text{im } T$ . Then  $P = T(T^*T)^{-1}T^*$ .*

**PROOF.** First note that  $T^*T$  is indeed injective (hence invertible) since  $T$  is. This follows from the identity  $\ker T^* = (\text{im } T)^\perp$ .

Next notice that the rank of  $P$  is  $\dim \text{im } T$ . But  $T^*$  is surjective since  $T$  is injective, so the rank of  $T(T^*T)^{-1}T^*$  is also  $\dim \text{im } T$ . It thus suffices to show that  $P$  and  $T(T^*T)^{-1}T^*$  agree on  $\text{im } T$ , and writing  $w = Tv$  we have

$$T(T^*T)^{-1}T^*w = T(T^*T)^{-1}(T^*T)v = Tv = w,$$

as desired. □

## References

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<sup>3</sup> Since projection operators are clearly bounded, the discussion below readily generalises to infinite-dimensional inner product spaces.

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