

Notes on linear algebra

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1 • Linear equations and matrices

1.1. Linear equations

Throughout we let 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 $a_1, \dots, a_n, b \in F$. A *solution* to l is an element $v = (v_1, \dots, v_n) \in F^n$ 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 $v \in F^n$ 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 $A \in \mathcal{M}_{m,n}(F)$, $x \in F^n$ and $b \in F^m$. 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 $A, B \in \mathcal{M}_{m,n}(F)$ 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 $A, B \in \mathcal{M}_{m,n}(F)$ 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 $H \in \mathcal{M}_{m,n}(F)$ 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 $E = e(I) \in \mathcal{M}_m(F)$, then $e(A) = EA$ for $A \in \mathcal{M}_{m,n}(F)$. If $B \in \mathcal{M}_{m,n}(F)$, then A and B are row-equivalent if and only if $A = PB$, where $P \in \mathcal{M}_m(F)$ is a product of elementary matrices.

PROPOSITION 1.2

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

PROOF. The usual Gauss–Jordan elimination algorithm proves existence. If $H, K \in \mathcal{M}_{m,n}(R)$ 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-equivalent¹ 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 $h, k \in F^r$ and $h', k' \in F^{m-r}$ 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 n th 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 $A \in \mathcal{M}_n(F)$, 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 $H \in \mathcal{M}_n(F)$ be a row-reduced echelon matrix that is row-equivalent to A . Then $R = PA$, where $P \in \mathcal{M}_n(F)$ 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.

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

(i) \Leftrightarrow (iii): Clearly (iii) implies (i), and the above shows that (i) implies that $A = P^{-1}$.

(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 $H \in \mathcal{M}_{m,n}(F)$ 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 $A \in \mathcal{M}_n(F)$. 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 $B \in \mathcal{M}_n(F)$, 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 • Determinants

2.1. Existence of determinants

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

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

is called *n-linear* if the maps $m_i \mapsto \varphi(m_1, \dots, m_n)$ are linear for all $m_i \in M_i$. Since $\mathcal{M}_{m,n}(R) \cong (R^m)^n$, a map $\varphi: \mathcal{M}_{m,n}(R) \rightarrow N$ that is linear in each row is also called *n-linear*.

In the case $M_1 = \cdots = M_n$, we call φ *alternating* if $\varphi(m_1, \dots, m_n) = 0$ whenever $m_i = m_j$ for some $i \neq j$. Furthermore, φ 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 2.1

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

- (i) If φ is alternating, then φ is skew-symmetric.
- (ii) If $\varphi(m_1, \dots, m_n) = 0$ whenever $m_i = m_{i+1}$ for some $i = 1, \dots, n-1$, then φ is alternating.

PROOF. (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 ψ 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 φ is skew-symmetric as claimed.

(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 at 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 φ is alternating as claimed. \square

DEFINITION 2.2

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

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

THEOREM 2.3

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

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

for $A = (a_{ij}) \in \mathcal{M}_n(R)$. Then ψ_j is alternating and n -linear. If φ is a determinant function, then so is ψ_j .

PROOF. Let $A = (a_{ij}) \in \mathcal{M}_n(R)$. 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 ψ_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 2.1(ii)** implies that ψ_j is alternating.

Finally suppose that φ 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 ψ_j is also a determinant function. □

COROLLARY 2.4

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

PROOF. The identity map on $\mathcal{M}_1(R) \cong R$ is a determinant function for $n = 1$, and **Theorem 2.3** allows us to recursively construct a determinant for each $n > 1$. □

2.2. Uniqueness of determinants

THEOREM 2.5

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

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

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

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

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

PROOF. Let e_1, \dots, e_n denote the rows of I_n , and denote the rows of a matrix $A = (a_{ij}) \in \mathcal{M}_n(R)$ 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 φ 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 φ is also skew-symmetric by [Lemma 2.1\(i\)](#), 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 σ , so

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

Finally, if φ 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

2.3. Properties of determinants

THEOREM 2.6

Let $A, B \in \mathcal{M}_n(R)$. Then

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

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

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

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

PROPOSITION 2.7

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

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

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

PROOF. By induction it suffices to consider the case where A has the block form

$$A = \begin{pmatrix} B & C \\ 0 & D \end{pmatrix}.$$

Say that $B \in \mathcal{M}_r(R)$ and $D \in \mathcal{M}_s(R)$, and put $\varphi(B, C, D) = \det A$. Then $D \mapsto \varphi(B, C, D)$ is clearly s -linear and alternating, so [Theorem 2.3](#) implies that

$$\varphi(B, C, D) = (\det D)\varphi(B, C, I_s).$$

By subtracting multiples of the rows of I_s from C we obtain $\varphi(B, C, I_s) = \varphi(B, 0, I_s)$. Next, $B \mapsto \varphi(B, 0, I_s)$ is also r -linear and alternating, so

$$\varphi(B, 0, I_s) = (\det B)\varphi(I_r, 0, I_s).$$

But $\varphi(I_r, 0, I_s) = 1$, so summarising we have

$$\varphi(B, C, D) = (\det D)\varphi(B, C, I_s) = (\det D)\varphi(B, 0, I_s) = (\det D)(\det B),$$

as desired. \square

PROPOSITION 2.8

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

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

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

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 $A \in \mathcal{M}_n(R)$. 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 adjoint matrix of A is the matrix $\operatorname{adj} A \in \mathcal{M}_n(R)$ whose (i, j) -th entry is the cofactor $A_{j,i}$. 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 $\operatorname{adj} A^\top = (\operatorname{adj} A)^\top$. We have the following:

PROPOSITION 2.9

Let $A \in \mathcal{M}_n(R)$. 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 2.3 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 $B = (b_{ij}) \in \mathcal{M}_n(R)$ 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 2.10

Let $A \in \mathcal{M}_n(R)$. Then A is a unit in $\mathcal{M}_n(R)$ if and only if $\det A$ is a unit in R .

PROOF. This follows directly from Proposition 2.9. \square

2.4. 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 $\lambda \in F$ 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* $p_T(t) \in F[t]$ of T , given by $p_T(t) = \det(tI - T)$. The eigenvalues of T are then precisely the roots of $p_T(t)$.

PROPOSITION 2.11

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

PROOF. (i): Let $A = (a_{ij}) \in \mathcal{M}_n(F)$ be a matrix representation of T and write $tI - A = (b_{ij})$, recall that

$$\det(tI - T) = \sum_{\sigma \in S_n} (\operatorname{sgn} \sigma) b_{1\sigma(1)} \cdots b_{n\sigma(n)} \quad (2.1)$$

by Theorem 2.5. Hence $p_T(t)$ is a linear combination of products of elements on the form $-a_{ij}$ or $t - a_{ij}$. 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 (2.1) 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

$$(t - a_{11})(t - a_{22}) \cdots (t - a_{nn}), \quad (2.2)$$

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 $c_0, \dots, c_n \in F$.

(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 (2.1) 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 σ would not be a permutation. Hence the factor t^n can only appear in the product (2.2). 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 F . Then some linear factor $t - \lambda \in F[t]$ divides $p_T(t)$, which implies that $\lambda \in F$ 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 $\lambda_1, \dots, \lambda_n \in F$. 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

2.5. Proofs without determinants

Existence of eigenvalues

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

Since F is algebraically closed, there exist $c, \lambda_1, \dots, \lambda_m \in F$ 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 λ_i is an eigenvalue of T .

Trace is sum of eigenvalues

LEMMA 2.12

Let $A \in \mathcal{M}_n(F)$ 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 denote the standard basis of F^n . First assume that the diagonal elements are nonzero. Then notice that $e_1 \in R(A)$, and that

$$Ae_i = a_1 e_1 + \cdots + a_{i-1} e_{i-1} + \lambda_i e_i$$

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

Conversely, assume that some diagonal element λ_i is zero. If $i = 1$, then $Ae_1 = 0$ so A is singular. If $i > 0$, then A maps $\text{span}(e_1, \dots, e_i)$ into $\text{span}(e_1, \dots, e_{i-1})$, so again A is singular. \square

LEMMA 2.13

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

PROOF. Let $\lambda \in F$, and denote the diagonal elements of A by $\lambda_1, \dots, \lambda_n$. By [lemma], 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 2.14

Let F be algebraically closed, and let $T \in \mathcal{L}(V)$. Then V has a basis with respect to which the matrix of 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 F -vector spaces of dimension $n-1$. Let $v_1 \in V$ be an eigenvector for T , and let $U = \text{span}(v_1)$. Since U is invariant under T , we may define a linear operator² $\tilde{T} \in \mathcal{L}(V/U)$ 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 .

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 $a_2, \dots, a_i \in F$ 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

COROLLARY 2.15

Let 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 $A \in \mathcal{M}_n(F)$ be an upper triangular matrix for T . The diagonal elements of A are the eigenvalues, and the trace of T is just the sum of these elements. \square

3 • Inner product spaces

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

² 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 $\tilde{T}: V/U \rightarrow V/U$.

LEMMA 3.1

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

THEOREM 3.2: The spectral theorem

Let F be either the real or the complex numbers, let V be a finite-dimensional inner product space over F , and consider $T \in \mathcal{L}(V)$. Then T is orthogonally diagonalisable if and only if

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

PROOF. Assume that either $F = \mathbb{R}$ and T is self-adjoint, or that $F = \mathbb{C}$ 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 $0 < d < n$.

Clearly $E_T(\lambda)$ is invariant under T , and we claim that it is also invariant under T^* . If T is self-adjoint this is obvious, and if T is normal then for all $w \in E_T(\lambda)$,

$$TT^*w = T^*Tw = T^*(\lambda w) = \lambda T^*w,$$

so we also have $T^*(w) \in E_T(\lambda)$. It follows from Lemma 3.1 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 these operators 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.

Conversely, assume that \mathcal{V} is an orthonormal basis of V consisting of eigenvectors for T , and let $A \in \mathcal{M}_n(F)$ be the matrix of T with respect to \mathcal{V} . Then A is diagonal with the eigenvalues of T on its diagonal. If $F = \mathbb{R}$ then the eigenvalues of T are real, so A is a real symmetric matrix, and hence T

is self-adjoint. If instead $F = \mathbb{C}$, then since A is diagonal we have $A^*A = AA^*$, which implies that T is normal. \square

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