LINEAR ALGEBRA NOTES MP274 1991

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Contents

1	Linear Transformations			
	1.1	Rank + Nullity Theorems (for Linear Maps)	3	
	1.2	Matrix of a Linear Transformation	6	
	1.3	Isomorphisms	12	
	1.4	Change of Basis Theorem for T_A	18	
2	Poly	ynomials over a field	20	
	2.1	Lagrange Interpolation Polynomials	21	
	2.2	Division of polynomials	24	
		2.2.1 Euclid's Division Theorem	24	
		2.2.2 Euclid's Division Algorithm	25	
	2.3	Irreducible Polynomials	26	
	2.4	Minimum Polynomial of a (Square) Matrix	32	
	2.5	Construction of a field of p^n elements	38	
	2.6	Characteristic and Minimum Polynomial of a Transformation	41	
		2.6.1 $M_{n \times n}(F[x])$ —Ring of Polynomial Matrices	42	
		2.6.2 $M_{n \times n}(F)[y]$ —Ring of Matrix Polynomials	43	
3	Invariant subspaces			
	3.1	T-cyclic subspaces	54	
		3.1.1 A nice proof of the Cayley-Hamilton theorem	57	
	3.2	An Algorithm for Finding m_T	58	
	3.3	Primary Decomposition Theorem	61	
4	The	Jordan Canonical Form	65	
	4.1	The Matthews' dot diagram	66	
	4.2	Two Jordan Canonical Form Examples	71	
		4.2.1 Example (a):	71	
		4.2.2 Example (b):	73	
	4.3	Uniqueness of the Jordan form	75	
	4.4	Non–derogatory matrices and transformations	78	
	4.5	Calculating A^{m} , where $A \in M_{n \times n}(\mathbb{C})$	79	
	4.6	Calculating e^A , where $A \in M_{n \times n}(\mathbb{C})$	81	
	4.7	Properties of the exponential of a complex matrix	82	
	4.8	Systems of differential equations	87	
	4.9	Markov matrices	89	
	4.10	The Real Jordan Form	94	
		4.10.1 Motivation	94	

	4.10.2 Determining the real Jordan form		
	4.10.3 A real algorithm for finding	g the real Jordan form $\ . \ . \ 100$	
5	The Rational Canonical Form 105		
	5.2 Deductions from the Rational Canonical Form		
	-		
	5.3.2 Invariant Factors		
6	The Smith Canonical Form 12		
	6.1 Equivalence of Polynomial Matric	es	
	6.2 Smith Canonical Form		
		Canonical Form 125	
	6.3 Invariant factors of a polynomial		
7	Various Applications of Rational Canonical Forms 13		
	7.1 An Application to commuting tra		
	7.2 Tensor products and the Byrnes-		
	· · ·	0	
	7.2.1 Properties of the tensor pr	oduct of matrices 136	
8	8 Further directions in linear algebra	ra 143	

Linear Transformations 1

We will study mainly finite-dimensional vector spaces over an arbitrary field F—i.e. vector spaces with a basis. (Recall that the dimension of a vector space V (dim V) is the number of elements in a basis of V.)

DEFINITION 1.1

(Linear transformation)

Given vector spaces U and V, $T: U \mapsto V$ is a linear transformation (LT) if

$$T(\lambda u + \mu v) = \lambda T(u) + \mu T(v)$$

for all $\lambda, \mu \in F$, and $u, v \in U$. Then T(u+v) = T(u) + T(v), $T(\lambda u) = \lambda T(u)$ and

$$T\left(\sum_{k=1}^{n} \lambda_k u_k\right) = \sum_{k=1}^{n} \lambda_k T(u_k).$$

EXAMPLES 1.1

Consider the linear transformation

$$T = T_A : V_n(F) \mapsto V_m(F)$$

where $A = [a_{ij}]$ is $m \times n$, defined by $T_A(X) = AX$.

Note that $V_n(F)$ = the set of all *n*-dimensional column vectors $\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ of

F—sometimes written F^n .

Note that if $T: V_n(F) \mapsto V_m(F)$ is a linear transformation, then $T = T_A$, where $A = [T(E_1)| \cdots |T(E_n)]$ and

$$E_1 = \begin{bmatrix} 1\\0\\\vdots\\0 \end{bmatrix}, \dots, E_n = \begin{bmatrix} 0\\\vdots\\0\\1 \end{bmatrix}$$

Note:

$$v \in V_n(F),$$
 $v = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = x_1 E_1 + \dots + x_n E_n$

If V is a vector space of all infinitely differentiable functions on \mathbb{R} , then

$$T(f) = a_0 \mathbf{D}^n f + a_1 \mathbf{D}^{n-1} f + \dots + a_{n-1} \mathbf{D} f + a_n f$$

defines a linear transformation $T: V \mapsto V$.

The set of f such that T(f) = 0 (i.e. the kernel of T) is important.

Let $T:U\mapsto V$ be a linear transformation. Then we have the following definition:

DEFINITIONS 1.1

(Kernel of a linear transformation)

$$\operatorname{Ker} T = \{ u \in U \mid T(u) = 0 \}$$

(Image of T)

$$\operatorname{Im} T = \{ v \in V \mid \exists u \in U \text{ such that } T(u) = v \}$$

Note: Ker T is a subspace of U. Recall that W is a subspace of U if

1. $0 \in W$,

- 2. W is closed under addition, and
- 3. W is closed under scalar multiplication.

PROOF. that $\operatorname{Ker} T$ is a subspace of U:

- 1. T(0) + 0 = T(0) = T(0 + 0) = T(0) + T(0). Thus T(0) = 0, so $0 \in \text{Ker } T$.
- 2. Let $u, v \in \text{Ker } T$; then T(u) = 0 and T(v) = 0. So T(u + v) = T(u) + T(v) = 0 + 0 = 0 and $u + v \in \text{Ker } T$.
- 3. Let $u \in \text{Ker } T$ and $\lambda \in F$. Then $T(\lambda u) = \lambda T(u) = \lambda 0 = 0$. So $\lambda u \in \text{Ker } T$.

EXAMPLE 1.1

$$\operatorname{Ker} T_A = N(A), \text{ the null space of } A$$
$$= \{X \in V_n(F) \mid AX = 0\}$$
and
$$\operatorname{Im} T_A = C(A), \text{ the column space of } A$$
$$= \langle A_{*1}, \dots, A_{*n} \rangle$$

Generally, if $U = \langle u_1, \ldots, u_n \rangle$, then $\operatorname{Im} T = \langle T(u_1), \ldots, T(u_n) \rangle$.

Note: Even if u_1, \ldots, u_n form a basis for $U, T(u_1), \ldots, T(u_n)$ may not form a basis for Im T. I.e. it may happen that $T(u_1), \ldots, T(u_n)$ are linearly dependent.

1.1 Rank + Nullity Theorems (for Linear Maps)

THEOREM 1.1 (General rank + nullity theorem)

If $T: U \mapsto V$ is a linear transformation then

$$\operatorname{rank} T + \operatorname{nullity} T = \dim U.$$

PROOF.

1. Ker $T = \{0\}$.

Then nullity T = 0. We first show that the vectors $T(u_1), \ldots, T(u_n)$, where u_1, \ldots, u_n are a basis for U, are LI (linearly independent): Suppose $x_1T(u_1) + \cdots + x_nT(u_n) = 0$ where $x_1, \ldots, x_n \in F$. Then

$$T(x_1u_1 + \dots + x_nu_n) = 0 \quad \text{(by linearity)}$$

$$x_1u_1 + \dots + x_nu_n = 0 \quad \text{(since Ker } T = \{0\})$$

$$x_1 = 0, \dots, x_n = 0 \quad \text{(since } u_i \text{ are LI})$$

Hence Im $T = \langle T(u_1), \ldots, T(u_n) \rangle$ so

$$\operatorname{rank} T + \operatorname{nullity} T = \dim \operatorname{Im} T + 0 = n = \dim V$$

2. Ker T = U. So nullity $T = \dim U$. Hence Im $T = \{0\} \Rightarrow \operatorname{rank} T = 0$

$$\Rightarrow \operatorname{rank} T + \operatorname{nullity} T = 0 + \dim U$$
$$= \dim U.$$

3. $0 < \text{nullity } T < \dim U$. Let u_1, \ldots, u_r be a basis for Ker T and $n = \dim U$, so r = nullity Tand r < n. Extend the basis u_1, \ldots, u_r to form a basis $u_1, \ldots, u_r, u_{r+1}, \ldots, u_n$ of U (refer to last year's notes to show that this can be done). Then $T(u_{r+1}), \ldots, T(u_n)$ span Im T. For

$$\operatorname{Im} T = \langle T(u_1), \dots, T(u_r), T(u_{r+1}), \dots, T(u_n) \rangle$$
$$= \langle 0, \dots, 0, T(u_{r+1}), \dots, T(u_n) \rangle$$
$$= \langle T(u_{r+1}), \dots, T(u_n) \rangle$$

So assume

$$\begin{aligned} x_1 T(u_{r+1}) + \dots + x_{n-r} T(u_n) &= 0 \\ \Rightarrow & T(x_1 u_{r+1} + \dots + x_{n-r} u_n) = 0 \\ \Rightarrow & x_1 u_{r+1} + \dots + x_{n-r} u_n \in \operatorname{Ker} T \\ \Rightarrow & x_1 u_{r+1} + \dots + x_{n-r} u_n = y_1 u_1 + \dots + y_r u_r \\ & \text{for some } y_1, \dots, y_r \\ \Rightarrow & (-y_1) u_1 + \dots + (-y_r) u_r + x_1 u_{r+1} + \dots + x_{n-r} u_n = 0 \end{aligned}$$

and since u_1, \ldots, u_n is a basis for U, all coefficients vanish. Thus

rank
$$T$$
 + nullity T = $(n - r) + r$
= n
= dim U .

We now apply this theorem to prove the following result:

THEOREM 1.2 (Dimension theorem for subspaces)

 $\dim(U \cap V) + \dim(U + V) = \dim U + \dim V$

where U and V are subspaces of a vector space W.

(Recall that $U + V = \{u + v \mid u \in U, v \in V\}$.) For the proof we need the following definition:

DEFINITION 1.2

If U and V are any two vector spaces, then the direct sum is

$$U \oplus V = \{(u, v) \mid u \in U, v \in V\}$$

(i.e. the cartesian product of U and V) made into a vector space by the component-wise definitions:

- 1. $(u_1, v_1) + (u_2, v_2) = (u_1 + u_2, v_1 + v_2),$
- 2. $\lambda(u, v) = (\lambda u, \lambda v)$, and
- 3. (0,0) is an identity for $U \oplus V$ and (-u, -v) is an additive inverse for (u, v).

We need the following result:

THEOREM 1.3

 $\dim(U \oplus V) = \dim U + \dim V$

PROOF.

Case 1: $U = \{0\}$

Case 2: $V = \{0\}$

Proof of cases 1 and 2 are left as an exercise.

Case 3: $U \neq \{0\}$ and $V \neq \{0\}$

Let u_1, \ldots, u_m be a basis for U, and v_1, \ldots, v_n be a basis for V.

We assert that $(u_1, 0), \ldots, (u_m, 0), (0, v_1), \ldots, (0, v_n)$ form a basis for $U \oplus V$. Firstly, spanning:

Let $(u, v) \in U \oplus V$, say $u = x_1u_1 + \cdots + x_mu_m$ and $v = y_1v_1 + \cdots + y_nv_n$. Then

$$\begin{array}{lll} (u,v) &=& (u,0) + (0,v) \\ &=& (x_1u_1 + \dots + x_mu_m, 0) + (0,y_1v_1 + \dots + y_nv_n) \\ &=& x_1(u_1,0) + \dots + x_m(u_m,0) + y_1(0,v_1) + \dots + y_n(0,v_n) \\ \\ \text{So} \quad U \oplus V &=& \langle (u_1,0), \dots, (u_m,0), (0,v_1), \dots, (0,v_n) \rangle \end{array}$$

Secondly, independence: assume $x_1(u_1, 0) + \cdots + x_m(u_m, 0) + y_1(0, v_1) + \cdots + y_n(0, v_n) = (0, 0)$. Then

$$(x_1u_1 + \dots + x_mu_m, y_1v_1 + \dots + y_nv_n) = 0$$

$$\Rightarrow \quad x_1u_1 + \dots + x_mu_m = 0$$

and $y_1v_1 + \dots + y_nv_n = 0$

$$\Rightarrow \quad x_i = 0, \forall i$$

and $y_i = 0, \forall i$

Hence the assertion is true and the result follows.

PROOF.

Let $T: U \oplus V \mapsto U + V$ where U and V are subspaces of some W, such that T(u, v) = u + v.

Thus $\operatorname{Im} T = U + V$, and

 $\begin{array}{lll} \mathrm{Ker}\, T &=& \{(u,v) \mid u \in U, v \in V, \mbox{ and } u+v=0\} \\ &=& \{(t,-t) \mid t \in U \cap V\} \end{array}$

Clearly then, dim $\operatorname{Ker} T = \dim (U \cap V)^1$ and so

$$\operatorname{rank} T + \operatorname{nullity} T = \dim(U \oplus V)$$
$$\Rightarrow \quad \dim(U+V) + \dim(U \cap V) = \dim U + \dim V.$$

1.2 Matrix of a Linear Transformation

DEFINITION 1.3

Let $T: U \mapsto V$ be a LT with bases $\beta: u_1, \ldots, u_n$ and $\gamma: v_1, \ldots, v_m$ for U and V respectively.

Then

$$T(u_j) = \begin{array}{ccc} a_{1j}v_1 \\ + \\ a_{2j}v_2 \\ + \\ \vdots \\ a_{mj}v_m \end{array} \quad a_{1j} \\ \vdots \\ a_{mj} \\ c_{mj} \\ c_{$$

The $m \times n$ matrix

$$A = [a_{ij}]$$

is called the matrix of T relative to the bases β and γ and is also written

$$A = [T]^{\gamma}_{\beta}$$

Note: The *j*-th column of A is the co-ordinate vector of $T(u_j)$, where u_j is the *j*-th vector of the basis β .

Also if
$$u = x_1u_1 + \dots + x_nu_n$$
, the co-ordinate vector $\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ is denoted by

 $[u]_{\beta}.$

¹True if $U \cap V = \{0\}$; if not, let S = Ker T and u_1, \ldots, u_r be a basis for $U \cap V$. Then $(u_1, -u_1), \ldots, (u_r, -u_r)$ form a basis for S and hence dim Ker $T = \dim S$.

EXAMPLE 1.2 Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in M_{2 \times 2}(F)$ and let $T : M_{2 \times 2}(F) \mapsto M_{2 \times 2}(F)$ be defined by

$$T(X) = AX - XA.$$

Then T is linear², and Ker T consists of all 2×2 matrices A where AX =XA.

Take β to be the basis E_{11} , E_{12} , E_{21} , and E_{22} , defined by

$$E_{11} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, E_{12} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, E_{21} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, E_{22} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

(so we can define a matrix for the transformation, consider these henceforth to be column vectors of four elements).

Calculate $[T]^{\beta}_{\beta} = B$:

$$T(E_{11}) = AE_{11} - E_{11}A$$

$$= \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$= \begin{bmatrix} 0 & -b \\ c & 0 \end{bmatrix}$$

$$= 0E_{11} - bE_{12} + cE_{21} + 0E_{22}$$

and similar calculations for the image of other basis vectors show that

$$B = \begin{bmatrix} 0 & -c & b & 0 \\ -b & a - d & 0 & b \\ c & 0 & d - a & -c \\ 0 & c & b & 0 \end{bmatrix}$$

EXERCISE: Prove that rank B = 2 if A is not a scalar matrix (i.e. if $A \neq tI_n$).

Later, we will show that $\operatorname{rank} B = \operatorname{rank} T$. Hence

nullity
$$T = 4 - 2 = 2$$

$$T(\lambda X + \mu Y) = A(\lambda X + \mu Y) - (\lambda X + \mu Y)A$$

= $\lambda(AX - XA) + \mu(AY - YA)$
= $\lambda T(X) + \mu T(Y)$

Note: $I_2, A \in \text{Ker } T$ which has dimension 2. Hence if A is not a scalar matrix, since I_2 and A are LI they form a basis for Ker T. Hence

 $AX = XA \quad \Rightarrow \quad X = \alpha I_2 + \beta A.$

DEFINITIONS 1.2

Let T_1 and T_2 be LT's mapping U to V. Then $T_1 + T_2 : U \mapsto V$ is defined by

$$(T_1 + T_2)(x) = T_1(x) + T_2(x) \quad ; \forall x \in U$$

For T a LT and $\lambda \in F$, define $\lambda T : U \mapsto V$ by

$$(\lambda T)(x) = \lambda T(x) \; \forall x \in U$$

Now . . .

$$\begin{split} [T_1+T_2]^{\gamma}_{\beta} &= & [T_1]^{\gamma}_{\beta} + [T_2]^{\gamma}_{\beta} \\ & [\lambda T]^{\gamma}_{\beta} &= & \lambda [T]^{\gamma}_{\beta} \end{split}$$

DEFINITION 1.4

$$\operatorname{Hom}\left(U,V\right) = \{T|T: U \mapsto V \text{ is a } LT\}.$$

Hom (U, V) is sometimes written L(U, V).

The zero transformation $0: U \mapsto V$ is such that $0(x) = 0, \forall x$. If $T \in \text{Hom}(U, V)$, then $(-T) \in \text{Hom}(U, V)$ is defined by

$$(-T)(x) = -(T(x)) \quad \forall x \in U.$$

Clearly, $\operatorname{Hom}(U, V)$ is a vector space. Also

$$\begin{array}{rcl} [0]_{\beta}^{\gamma} & = & 0 \\ \text{and} & [-T]_{\beta}^{\gamma} & = & -[T]_{\beta}^{\gamma} \end{array} \end{array}$$

The following result reduces the computation of T(u) to matrix multiplication:

THEOREM 1.4

$$[T(u)]_{\gamma} = [T]_{\beta}^{\gamma}[u]_{\beta}$$

PROOF.

Let $A = [T]^{\gamma}_{\beta}$, where β is the basis u_1, \ldots, u_n, γ is the basis v_1, \ldots, v_m , and

$$T(u_j) = \sum_{i=1}^m a_{ij} v_i.$$

Also let $[u]_{\beta} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$. Then $u = \sum_{j=1}^n x_j u_j$, so

$$T(u) = \sum_{j=1}^{n} x_j T(u_j)$$

=
$$\sum_{j=1}^{n} x_j \sum_{i=1}^{m} a_{ij} v_i$$

=
$$\sum_{i=1}^{m} \left(\sum_{j=1}^{n} a_{ij} x_j \right) v_i$$

$$\Rightarrow [T(u)]_{\gamma} = \begin{bmatrix} a_{11}x_1 + \dots + a_{1n}x_n \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n \end{bmatrix}$$

=
$$A[u]_{\beta}$$

DEFINITION 1.5

(Composition of LTs)

If $T_1: U \mapsto V$ and $T_2: V \mapsto W$ are LTs, then $T_2T_1: U \mapsto W$ defined by

$$(T_2T_1)(x) = T_2(T_1(x)) \qquad \forall x \in U$$

is a LT.

THEOREM 1.5

If β , γ and δ are bases for U, V and W, then

$$[T_2 T_1]^{\delta}_{\beta} = [T_2]^{\delta}_{\gamma} [T_1]^{\gamma}_{\beta}$$

PROOF. Let $u \in U$. Then

$$[T_2T_1(u)]_{\delta} = [T_2T_1]_{\beta}^{\delta}[u]_{\beta}$$

and
$$= [T_2(T_1(u))]_{\delta}$$
$$= [T_2]_{\gamma}^{\delta}[T_1(u)]_{\gamma}$$

Hence

$$[T_2 T_1]^{\delta}_{\beta} [u]_{\beta} = [T_2]^{\delta}_{\gamma} [T_1]^{\gamma}_{\beta} [u]_{\beta}$$
(1)

(note that we can't just "cancel off" the $[u]_{\beta}$ to obtain the desired result!) Finally, if β is u_1, \ldots, u_n , note that $[u_j]_{\beta} = E_j$ (since $u_j = 0u_1 + \cdots +$ $0u_{j-1} + 1u_j + 0u_{j+1} + \dots + 0u_n$) then for an appropriately sized matrix B,

 $BE_j = B_{*j}$, the *j*th column of *B*.

Then (1) shows that the matrices

$$[T_2T_1]^{\delta}_{\beta}$$
 and $[T_2]^{\delta}_{\gamma}[T_1]^{\gamma}_{\beta}$

have their first, second, \ldots , *n*th columns respectively equal.

~

EXAMPLE 1.3

If A is $m \times n$ and B is $n \times p$, then

$$T_A T_B = T_{AB}.$$

DEFINITION 1.6

(the identity transformation)

Let U be a vector space. Then the identity transformation $I_U: U \mapsto U$ defined by

$$I_U(x) = x \quad \forall x \in U$$

is a linear transformation, and

$$[I_U]^{\beta}_{\beta} = I_n \quad \text{if } n = \dim U.$$

Also note that $I_{V_n(F)} = T_{I_n}$.

THEOREM 1.6

Let $T: U \mapsto V$ be a LT. Then

$$I_V T = T I_U = T.$$

Then

$$T_{I_m}T_A = T_{I_mA} = T_A = T_A T_{AI_n} = T_{AI_n}$$

and consequently we have the familiar result

$$I_m A = A = A I_n.$$

DEFINITION 1.7

(Invertible LTs) Let $T : U \mapsto V$ be a LT. If $\exists S : V \mapsto U$ such that S is linear and satisfies

$$ST = I_U$$
 and $TS = I_V$

then we say that T is **invertible** and that S is an **inverse** of T. Such inverses are unique and we thus denote S by T^{-1} . Explicitly,

$$S(T(x)) = x \ \forall x \in U \quad and \quad T(S(y)) = y \ \forall y \in V$$

There is a corresponding definition of an **invertible matrix**: $A \in M_{m \times n}(F)$ is called invertible if $\exists B \in M_{n \times m}(F)$ such that

$$AB = I_m$$
 and $BA = I_n$

Evidently

THEOREM 1.7

 T_A is invertible iff A is invertible (i.e. if A^{-1} exists). Then,

$$(T_A)^{-1} = T_{A^{-1}}$$

THEOREM 1.8

If u_1, \ldots, u_n is a basis for U and v_1, \ldots, v_n are vectors in V, then there is one and only one linear transformation $T: U \to V$ satisfying

$$T(u_1) = v_1, \ldots, T(u_n) = v_n,$$

namely $T(x_1u_1 + \dots + x_nu_n) = x_1v_1 + \dots + x_nv_n$.

(In words, a linear transformation is determined by its action on a basis.)

1.3 Isomorphisms

DEFINITION 1.8

A linear map $T: U \mapsto V$ is called an **isomorphism** if T is 1-1 and onto, *i.e.*

- 1. $T(x) = T(y) \Rightarrow x = y \ \forall x, y \in U$, and
- 2. Im T = V, that is, if $v \in V$, $\exists u \in U$ such that T(u) = v.

Lemma: A linear map T is 1-1 iff Ker $T = \{0\}$. PROOF:

1. (\Rightarrow) Suppose T is 1-1 and let $x \in \text{Ker } T$.

We have T(x) = 0 = T(0), and so x = 0.

2. (\Leftarrow) Assume Ker $T = \{0\}$ and T(x) = T(y) for some $x, y \in U$. Then

$$T(x - y) = T(x) - T(y) = 0$$

$$\Rightarrow x - y \in \text{Ker } T$$

$$\Rightarrow x - y = 0 \Rightarrow x = y$$

THEOREM 1.9

Let $A \in M_{m \times n}(F)$. Then $T_A : V_n(F) \to V_m(F)$ is

- (a) onto: $\Leftrightarrow \dim C(A) = m \Leftrightarrow$ the rows of A are LI;
- (b) $1-1: \Leftrightarrow \dim N(A) = 0 \Leftrightarrow \operatorname{rank} A = n \Leftrightarrow \text{ the columns of } A \text{ are } LI.$

EXAMPLE 1.4

Let $T_A : V_n(F) \mapsto V_n(F)$ with A invertible; so $T_A(X) = AX$. We will show this to be an isomorphism.

1. Let $X \in \text{Ker} T_A$, i.e. AX = 0. Then

$$A^{-1}(AX) = A^{-1}0$$

$$\Rightarrow I_n X = 0$$

$$\Rightarrow X = 0$$

$$\Rightarrow \text{Ker } T = \{0\}$$

$$\Leftrightarrow T \text{ is } 1\text{-}1.$$

2. Let $Y \in V_n(F)$: then,

$$T(A^{-1}Y) = A(A^{-1}Y)$$
$$= I_n Y = Y$$
so Im $T_A = V_n(F)$

THEOREM 1.10

If T is an isomorphism between U and V, then

 $\dim U = \dim V$

PROOF.

Let u_1, \ldots, u_n be a basis for U. Then

$$T(u_1),\ldots,T(u_n)$$

is a basis for V (i.e. $\langle u_i \rangle = U$ and $\langle T(u_i) \rangle = V$, with u_i , v_i independent families), so

 $\dim U = n = \dim V$

THEOREM 1.11

 Φ : Hom $(U, V) \mapsto M_{m \times n}(F)$ defined by $\Phi(T) = [T]_{\beta}^{\gamma}$

is an isomorphism.

Here dim U = n, dim V = m, and β and γ are bases for U and V, respectively.

THEOREM 1.12

 $T: U \mapsto V \text{ is invertible}$ $\Leftrightarrow \quad T \text{ is an isomorphism between } U \text{ and } V.$

PROOF.

 \Rightarrow Assume T is invertible. Then

$$T^{-1}T = I_U$$

and $TT^{-1} = I_V$
 $\Rightarrow T^{-1}(T(x)) = x \ \forall x \in U$
and $T(T^{-1}(y)) = y \ \forall y \in V$

1. We prove $\operatorname{Ker} T = \{0\}$. Let T(x) = 0. Then

$$T^{-1}(T(x)) = T^{-1}(0) = 0 = x$$

So T is 1-1.

2. We show $\operatorname{Im} T = V$. Let $y \in V$. Now $T(T^{-1}(y)) = y$, so taking $x = T^{-1}(y)$ gives

$$T(x) = y.$$

Hence $\operatorname{Im} T = V$.

 \Leftarrow Assume T is an isomorphism, and let S be the inverse map of T

 $S: V \mapsto U$

Then $ST = I_U$ and $TS = I_V$. It remains to show that S is linear. We note that

$$x = S(y) \Leftrightarrow y = T(x)$$

And thus, using linearity of T only, for any $y_1, y_2 \in V$, $x_1 = S(y_1)$, and $x_2 = S(y_2)$ we obtain

$$S(\lambda y_1 + \mu y_2) = S(\lambda T(x_1) + \mu T(x_2))$$

= $S(T(\lambda x_1 + \mu x_2))$
= $\lambda x_1 + \mu x_2$
= $\lambda S(y_1) + \mu S(y_2)$

COROLLARY 1.1

If $A \in M_{m \times n}(F)$ is invertible, then m = n.

PROOF.

Suppose A is invertible. Then T_A is invertible and thus an isomorphism between $V_n(F)$ and $V_m(F)$.

Hence dim $V_n(F)$ = dim $V_m(F)$ and hence m = n.

THEOREM 1.13

If dim $U = \dim V$ and $T : U \mapsto V$ is a LT, then

T is 1-1 (injective) \Leftrightarrow T is onto (surjective) $(\Leftrightarrow T \text{ is an isomorphism })$

PROOF.

⇒ Suppose T is 1-1. Then Ker $T = \{0\}$ and we have to show that Im T = V.

$$\begin{aligned} \operatorname{rank} T + \operatorname{nullity} T &= \dim U \\ \Rightarrow \operatorname{rank} T + 0 &= \dim V \\ \text{i.e. } \dim(\operatorname{Im} T) &= \dim V \\ \Rightarrow \operatorname{Im} T &= V \quad \text{as } T \subseteq V. \end{aligned}$$

 \Leftarrow Suppose T is onto.

Then $\operatorname{Im} T = V$ and we must show that $\operatorname{Ker} T = \{0\}$. The above argument is reversible:

$$Im T = V$$

$$rank T = \dim V$$

$$= \dim U$$

$$= rank T + nullity T$$

$$\Rightarrow nullity T = 0$$
or Ker T = {0}

COROLLARY 1.2

Let $A, B \in M_{n \times n}(F)$. Then

$$AB = I_n \Rightarrow BA = I_n.$$

PROOF Suppose $AB = I_n$. Then Ker $T_B = \{0\}$. For

$$BX = 0 \Rightarrow A(BX) = A0 = 0$$

$$\Rightarrow I_n X = 0 \Rightarrow X = 0.$$

But dim $U = \dim V = n$, so T_B is an isomorphism and hence invertible. Thus $\exists C \in M_{n \times n}(F)$ such that

$$T_B T_C = I_{V_n(F)} = T_C T_B$$

$$\Rightarrow BC = I_n = CB,$$

noting that $I_{V_n(F)} = T_{I_n}$.

Now, knowing $AB = I_n$,

$$\Rightarrow A(BC) = A$$

$$(AB)C = A$$

$$I_nC = A$$

$$\Rightarrow C = A$$

$$\Rightarrow BA = I_n$$

DEFINITION 1.9

Another standard isomorphism: Let dim V = m, with basis $\gamma = v_1, \ldots, v_m$. Then $\phi_{\gamma} : V \mapsto V_m(F)$ is the isomorphism defined by

$$\phi_{\gamma}(v) = [v]_{\gamma}$$

THEOREM 1.14

$$\operatorname{rank} T = \operatorname{rank} [T]_{\beta}^{\gamma}$$

PROOF

$$\begin{array}{cccc} U & \xrightarrow{T} & V \\ \phi_{\beta} \downarrow & & \downarrow & \phi_{\gamma} \\ V_n(F) & \overrightarrow{T_A} & V_m(F) \end{array}$$

With

$$\begin{array}{c} \beta: u_1, \dots, u_n\\ \gamma: v_1, \dots, v_m \end{array} \text{ a basis for } \begin{array}{c} U\\ V, \end{array}$$

let $A = [T]^{\gamma}_{\beta}$. Then the commutative diagram is an abbreviation for the equation

$$\phi_{\gamma}T = T_A \phi_{\beta}. \tag{2}$$

Equivalently

$$\phi_{\gamma}T(u) = T_A\phi_{\beta}(u) \; \forall u \in U$$

or

$$[T(u)]_{\gamma} = A[u]_{\beta}$$

which we saw in Theorem 1.4.

But rank (ST) = rank T if S is invertible and rank (TR) = rank T if R is invertible. Hence, since ϕ_{β} and ϕ_{γ} are both invertible,

(2)
$$\Rightarrow \operatorname{rank} T = \operatorname{rank} T_A = \operatorname{rank} A$$

and the result is proven.

Note:

Observe that $\phi_{\gamma}(T(u_j)) = A_{*j}$, the *j*th column of *A*. So Im *T* is mapped under ϕ_{γ} into C(A). Also Ker *T* is mapped by ϕ_{β} into N(A). Consequently we get bases for Im *T* and Ker *T* from bases for C(A) and N(A), respectively.

$$\begin{aligned} (u \in \operatorname{Ker} T \Leftrightarrow T(u) = 0 & \Leftrightarrow & \phi_{\gamma}(T(u)) = 0 \\ & \Leftrightarrow & T_A \phi_{\beta}(u) = 0 \\ & \Leftrightarrow & \phi_{\beta}(u) \in N(A).) \end{aligned}$$

THEOREM 1.15

Let β and γ be bases for some vector space V. Then, with $n = \dim V$,

$$[I_V]^{\gamma}_{\beta}$$

is non-singular and its inverse

$$\left\{ \left[I_V \right]_{\beta}^{\gamma} \right\}^{-1} = \left[I_V \right]_{\gamma}^{\beta}.$$

PROOF

$$I_V I_V = I_V$$

$$\Rightarrow [I_V I_V]^{\beta}_{\beta} = [I_V]^{\beta}_{\beta} = I_n$$

$$= [I_V]^{\beta}_{\gamma} [I_V]^{\gamma}_{\beta}.$$

The matrix $P = [I_V]^{\gamma}_{\beta} = [p_{ij}]$ is called the *change of basis matrix*. For if $\beta : u_1, \ldots, u_n$ and $\gamma : v_1, \ldots, v_n$ then

$$u_j = I_V(u_j)$$

= $p_{1j}v_1 + \dots + p_{nj}v_n$ for $j = 1, \dots, n$.

It is also called the *change of co-ordinate matrix*, since

$$[v]_{\gamma} = [I_V(v)]^{\gamma}_{\beta}[v]_{\beta}$$

i.e. if

$$v = x_1u_1 + \dots + x_nu_n$$
$$= y_1v_1 + \dots + y_nv_n$$

then

$$\left[\begin{array}{c} y_1\\ \vdots\\ y_n \end{array}\right] = P \left[\begin{array}{c} x_1\\ \vdots\\ x_n \end{array}\right],$$

or, more explicitly,

$$y_1 = p_{11}x_1 + \dots + p_{1n}x_n$$

$$\vdots$$

$$y_n = p_{n1}x_1 + \dots + p_{1n}x_n$$

THEOREM 1.16 (Effect of changing basis on matrices of LTs)

Let $T: V \mapsto V$ be a LT with bases β and γ . Then

$$[T]^{\beta}_{\beta} = P^{-1}[T]^{\gamma}_{\gamma}P$$

where

$$P = [I_V]^{\gamma}_{\beta}$$

as above.

PROOF

$$I_V T = T = T I_V$$

$$\Rightarrow [I_V T]^{\gamma}_{\beta} = [T I_V]^{\gamma}_{\beta}$$

$$\Rightarrow [I_V]^{\gamma}_{\beta} [T]^{\beta}_{\beta} = [T]^{\gamma}_{\gamma} [I_V]^{\gamma}_{\beta}$$

DEFINITION 1.10

(Similar matrices)

If A and B are two matrices in $M_{m \times n}(F)$, then if there exists a nonsingular matrix P such that

$$B = P^{-1}AP$$

we say that A and B are similar over F.

1.4 Change of Basis Theorem for T_A

In the MP274 course we are often proving results about linear transformations $T: V \mapsto V$ which state that a basis β can be found for V so that $[T]_{\beta}^{\beta} = B$, where B has some special property. If we apply the result to the linear transformation $T_A: V_n(F) \mapsto V_n(F)$, the change of basis theorem applied to T_A tells us that A is similar to B. More explicitly, we have the following:

THEOREM 1.17

Let $A \in M_{n \times n}(F)$ and suppose that $v_1, \ldots, v_n \in V_n(F)$ form a basis β for $V_n(F)$. Then if $P = [v_1| \cdots |v_n]$ we have

$$P^{-1}AP = [T_A]^{\beta}_{\beta}.$$

PROOF. Let γ be the standard basis for $V_n(F)$ consisting of the unit vectors E_1, \ldots, E_n and let $\beta : v_1, \ldots, v_n$ be a basis for $V_n(F)$. Then the change of basis theorem applied to $T = T_A$ gives

$$[T_A]^\beta_\beta = P^{-1}[T_A]^\gamma_\gamma P,$$

where $P = [I_V]^{\gamma}_{\beta}$ is the change of coordinate matrix.

Now the definition of P gives

$$v_1 = I_V(v_1) = p_{11}E_1 + \dots + p_{n1}E_n$$

:
 $v_n = I_V(v_n) = p_{1n}E_1 + \dots + p_{nn}E_n,$

or, more explicitly,

$$v_1 = \begin{bmatrix} p_{11} \\ \vdots \\ p_{n1} \end{bmatrix}, \quad \dots, \quad \begin{bmatrix} p_{1n} \\ \vdots \\ p_{nn} \end{bmatrix}.$$

In other words, $P = [v_1| \cdots |v_n]$, the matrix whose columns are v_1, \ldots, v_n respectively.

Finally, we observe that $[T_A]^{\gamma}_{\gamma} = A$.

2 Polynomials over a field

A polynomial over a field F is a sequence

 $(a_0, a_1, a_2, \ldots, a_n, \ldots)$ where $a_i \in F \forall i$

with $a_i = 0$ from some point on. a_i is called the *i*-th coefficient of f.

We define three special polynomials...

$$\begin{array}{rcl} 0 & = & (0,0,0,\ldots) \\ 1 & = & (1,0,0,\ldots) \\ x & = & (0,1,0,\ldots). \end{array}$$

The polynomial (a_0, \ldots) is called a constant and is written simply as a_0 . Let F[x] denote the set of all polynomials in x.

If $f \neq 0$, then the **degree** of f, written deg f, is the greatest n such that $a_n \neq 0$. Note that the polynomial 0 has <u>no</u> degree. a_n is called the 'leading coefficient' of f.

F[x] forms a vector space over F if we define

$$\lambda(a_0, a_1, \ldots) = (\lambda a_0, \lambda a_1, \ldots), \ \lambda \in F.$$

DEFINITION 2.1

(Multiplication of polynomials)

Let $f = (a_0, a_1, ...)$ and $g = (b_0, b_1, ...)$. Then $fg = (c_0, c_1, ...)$ where

$$c_n = a_0b_n + a_1b_{n-1} + \dots + a_nb_0$$

=
$$\sum_{\substack{i=0\\i=j=n}}^n a_ib_{n-i}$$

=
$$\sum_{\substack{0 \le i, 0 \le j\\i+j=n}} a_ib_j.$$

EXAMPLE 2.1

$$x^{2} = (0, 0, 1, 0, \ldots), x^{3} = (0, 0, 0, 1, 0, \ldots).$$

More generally, an induction shows that $x^n = (a_0, \ldots)$, where $a_n = 1$ and all other a_i are zero.

If deg f = n, we have $f = a_0 1 + a_1 x + \dots + a_n x^n$.

THEOREM 2.1 (Associative Law)

f(gh) = (fg)h

PROOF Take f, g as above and $h = (c_0, c_1, \ldots)$. Then $f(gh) = (d_0, d_1, \ldots)$, where

$$d_n = \sum_{i+j=n} (fg)_i h_j$$

=
$$\sum_{i+j=n} \left(\sum_{u+v=i} f_u g_v \right) h_j$$

=
$$\sum_{u+v+j=n} f_u g_v h_j.$$

Likewise $(fg)h = (e_0, e_1, \ldots)$, where

$$e_n = \sum_{u+v+j=n} f_u g_v h_j$$

Some properties of polynomial arithmetic:

$$\begin{array}{rcl} fg &=& gf\\ 0f &=& 0\\ 1f &=& f\\ f(g+h) &=& fg+fh\\ f\neq 0 \mbox{ and } g\neq 0 \ \Rightarrow & fg\neq 0\\ && \mbox{ and } \deg(fg) = \deg f + \deg g. \end{array}$$

The last statement is equivalent to

$$fg = 0 \Rightarrow f = 0 \text{ or } g = 0.$$

The we deduce that

$$fh = fg$$
 and $f \neq 0 \Rightarrow h = g$.

2.1 Lagrange Interpolation Polynomials

Let $P_n[F]$ denote the set of polynomials $a_0 + a_1x + \cdots + a_nx^n$, where $a_0, \ldots, a_n \in F$. Then $a_0 + a_1x + \cdots + a_nx^n = 0$ implies that $a_0 = 0, \ldots, a_n = 0$

 $a_0, \ldots, a_n \in F$. Then $a_0 + a_1x + \cdots + a_nx^n = 0$ implies that $a_0 = 0, \ldots, a_n = 0$. $P_n[F]$ is a subspace of F[x] and $1, x, x^2, \ldots, x^n$ form the 'standard' basis for $P_n[F]$. If $f \in P_n[F]$ and $c \in F$, we write

$$f(c) = a_0 + a_1c + \dots + a_nc^n.$$

This is the "value of f at c". This symbol has the following properties:

$$(f+g)(c) = f(c) + g(c)$$

$$(\lambda f)(c) = \lambda(f(c))$$

$$(fg)(c) = f(c)g(c)$$

DEFINITION 2.2

Let c_1, \ldots, c_{n+1} be distinct members of F. Then the Lagrange interpolation polynomials p_1, \ldots, p_{n+1} are polynomials of degree n defined by

$$p_{i} = \prod_{\substack{j=1\\j \neq i}}^{n+1} \left(\frac{x - c_{j}}{c_{i} - c_{j}} \right), \ 1 \le i \le n+1.$$

EXAMPLE 2.2

$$p_1 = \begin{pmatrix} \frac{x-c_2}{c_1-c_2} \end{pmatrix} \begin{pmatrix} \frac{x-c_3}{c_1-c_3} \end{pmatrix} \cdots \begin{pmatrix} \frac{x-c_{n+1}}{c_1-c_{n+1}} \end{pmatrix}$$

$$p_2 = \begin{pmatrix} \frac{x-c_1}{c_2-c_1} \end{pmatrix} \times \begin{pmatrix} \frac{x-c_3}{c_2-c_3} \end{pmatrix} \cdots \begin{pmatrix} \frac{x-c_{n+1}}{c_2-c_{n+1}} \end{pmatrix}$$

$$etc...$$

We now show that the Lagrange polynomials also form a basis for $P_n[F]$. PROOF Noting that there are n + 1 elements in the 'standard' basis, above, we see that dim $P_n[F] = n + 1$ and so it suffices to show that p_1, \ldots, p_{n+1} are LI.

We use the following property of the polynomials p_i :

$$p_i(c_j) = \delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j. \end{cases}$$

Assume that

$$a_1 p_1 + \dots + a_{n+1} p_{n+1} = 0$$

where $a_i \in F$, $1 \le i \le n+1$. Evaluating both sides at c_1, \ldots, c_{n+1} gives

$$a_1p_1(c_1) + \dots + a_{n+1}p_{n+1}(c_1) = 0$$

 \vdots
 $a_1p_1(c_{n+1}) + \dots + a_{n+1}p_{n+1}(c_{n+1}) = 0$

 \Rightarrow

$$a_1 \times 1 + a_2 \times 0 + \dots + a_{n+1} \times 0 = 0$$

 $a_1 \times 0 + a_2 \times 1 + \dots + a_{n+1} \times 0 = 0$
 \vdots
 $a_1 \times 0 + a_2 \times 0 + \dots + a_{n+1} \times 1 = 0$

Hence $a_i = 0 \ \forall i$ as required.

COROLLARY 2.1

If $f \in P_n[F]$ then

$$f = f(c_1)p_1 + \dots + f(c_{n+1})p_{n+1}$$

PROOF: We know that

$$f = \lambda_1 p_1 + \dots + \lambda_{n+1} p_{n+1}$$
 for some $\lambda_i \in F$.

Evaluating both sides at c_1, \ldots, c_{n+1} then, gives

$$f(c_1) = \lambda_1,$$

$$\vdots$$

$$f(c_{n+1}) = \lambda_{n+1}$$

as required.

COROLLARY 2.2

If $f \in P_n[F]$ and $f(c_1) = 0, \ldots, f(c_{n+1}) = 0$ where c_1, \ldots, c_{n+1} are distinct, then f = 0. (I.e. a non-zero polynomial of degree n can have at most n roots.)

COROLLARY 2.3

If b_1, \ldots, b_{n+1} are any scalars in F, and c_1, \ldots, c_{n+1} are again distinct, then there exists a unique polynomial $f \in P_n[F]$ such that

$$f(c_1) = b_1, \ldots, f(c_{n+1}) = b_{n+1};$$

namely

$$f = b_1 p_1 + \dots + b_{n+1} p_{n+1}$$

EXAMPLE 2.3

Find the quadratic polynomial

$$f = a_0 + a_1 x + a_2 x^2 \in P_2[\mathbb{R}]$$

such that

$$f(1) = 8, f(2) = 5, f(3) = 4.$$

Solution: $f = 8p_1 + 5p_2 + 4p_3$ where

$$p_1 = \frac{(x-2)(x-3)}{(1-2)(1-3)}$$

$$p_2 = \frac{(x-1)(x-3)}{(2-1)(2-3)}$$

$$p_3 = \frac{(x-1)(x-2)}{(3-1)(3-2)}$$

2.2 Division of polynomials

DEFINITION 2.3

If $f, g \in F[x]$, we say f divides g if $\exists h \in F[x]$ such that

$$g = fh$$

For this we write " $f \mid g$ ", and " $f \nmid g$ " denotes the negation "f does not divide g".

Some properties:

$$f \mid g \text{ and } g \neq 0 \Rightarrow \deg f \leq \deg g$$

and thus of course

$$f \mid 1 \Rightarrow \deg f = 0.$$

2.2.1 Euclid's Division Theorem

Let $f, g \in F[x]$ and $g \neq 0$. Then $\exists q, r \in F[x]$ such that

$$f = qg + r, \tag{3}$$

where r = 0 or deg $r < \deg g$. Moreover q and r are unique. Outline of Proof: If f = 0 or deg $f < \deg g$, (3) is trivially true (taking q = 0 and r = f). So assume deg $f \ge \deg g$, where

$$f = a_m x^m + a_{m-1} x^{m-1} + \dots + a_0,$$

$$g = b_n x^n + \dots + b_0$$

and we have a long division process, viz:

$$b_{n}x^{n} + \dots + b_{0} \boxed{\begin{array}{cccc} a_{m}b_{n}^{-1}x^{m-n} & + & \dots \\ a_{m}x^{m} & + & a_{m-1}x^{m-1} & + \dots + & a_{0} \\ a_{m}x^{m} & & \\ \end{array}}_{\text{etc...}}$$

(See S. Perlis, Theory of Matrices, p.111.)

2.2.2 Euclid's Division Algorithm

$$f = q_1 g + r_1 \quad \text{with} \quad \deg r_1 < \deg g$$

$$g = q_2 r_1 + r_2 \quad \text{with} \quad \deg r_2 < \deg r_1$$

$$r_1 = q_3 r_2 + r_3 \quad \text{with} \quad \deg r_3 < \deg r_2$$

$$\vdots \quad \dots \quad \dots \quad \dots$$

$$r_{n-2} = q_n r_{n-1} + r_n \quad \text{with} \quad \deg r_n < \deg r_{n-1}$$

$$r_{n-1} = q_{n+1} r_n$$

Then $r_n = \gcd(f, g)$, the **greatest common divisor** of f and g—i.e. r_n is a polynomial d with the property that

1. $d \mid f$ and $d \mid g$, and

2.
$$\forall e \in F[x], e \mid f \text{ and } e \mid g \Rightarrow e \mid d.$$

(This defines gcd(f,g) uniquely up to a constant multiple.)

We select the *monic* (i.e. leading coefficient = 1) gcd as "the" gcd. Also, $\exists u, v \in F[x]$ such that

$$r_n = \gcd(f,g)$$

= $uf + vg$

—find u and v by 'forward substitution' in Euclid's algorithm; viz.

$$r_1 = f + (-q_1)g$$

 $r_2 = g + (-q_2)r_1$

$$= g + (-q_2)(f + (-q_1)g) = g + (-q_2)f + (q_1q_2)g = (-q_2)f + (1 + q_1q_2)g \vdots r_n = (\dots) f + (\dots) g.$$

In general, $r_k = s_k f + t_k g$ for $-1 \le k \le n$, where

$$r_{-1}=f,\ r_0=g,\ s_{-1}=1,\ s_0=0,\ t_{-1}=0,\ t_0=1$$

and

$$s_k = -q_k s_{k-1} + s_{k-2}, \ t_k = -q_k t_{k-1} + t_{k-2}$$

for $1 \le k \le n$. (Proof by induction.)

The special case gcd(f,g) = 1 (i.e. f and g are **relatively prime**) is of great importance: here $\exists u, v \in F[x]$ such that

$$uf + vg = 1.$$

EXERCISE 2.1

Find $gcd(3x^2 + 2x + 4, 2x^4 + 5x + 1)$ in $\mathbb{Q}[x]$ and express it as uf + vg for two polynomials u and v.

2.3 Irreducible Polynomials

DEFINITION 2.4

Let f be a non-constant polynomial. Then, if

$$g \mid f \Rightarrow \begin{array}{c} g \text{ is a constant} \\ or \quad g = \text{constant} \times f \end{array}$$

we call f an irreducible polynomial.

Note: (**Remainder theorem**)

f = (x - a)q + f(a) where $a \in F$. So f(a) = 0 iff $(x - a) \mid f$.

EXAMPLE 2.4

 $f(x) = x^2 + x + 1 \in \mathbb{Z}_2[x]$ is irreducible, for $f(0) = f(1) = 1 \neq 0$, and hence there are no polynomials of degree 1 which divide f.

THEOREM 2.2

Let f be irreducible. Then if $f \not \mid g, \ \gcd(f,g) = 1$ and $\exists u, v \in F[x]$ such that

$$uf + vg = 1.$$

PROOF Suppose f is irreducible and f/g. Let $d = \gcd(f, g)$ so

$$d \mid f$$
 and $d \mid g$.

Then either d = cf for some constant c, or d = 1. But if d = cf then

 $\begin{array}{ll} f \mid d & \text{and} & d \mid g \\ \Rightarrow & f \mid g & -\!\!\!-\!\!\!\text{a contradiction.} \end{array}$

So d = 1 as required.

COROLLARY 2.4

If f is irreducible and $f \mid gh$, then $f \mid g$ or $f \mid h$. PROOF: Suppose f is irreducible and $f \mid gh$, $f \nmid g$. We show that $f \mid h$. By the above theorem, $\exists u, v$ such that

$$uf + vg = 1$$

$$\Rightarrow ufh + vgh = h$$

$$\Rightarrow f \mid h$$

THEOREM 2.3

Any non-constant polynomial is expressible as a product of irreducible polynomials where representation is unique up to the order of the irreducible factors.

Some examples:

$$(x+1)^2 = x^2 + 2x + 1$$

= $x^2 + 1$ in $\mathbb{Z}_2[x]$
 $(x^2 + x + 1)^2 = x^4 + x^2 + 1$ in $\mathbb{Z}_2[x]$
 $(2x^2 + x + 1)(2x + 1) = x^3 + x^2 + 1$ in $\mathbb{Z}_3[x]$
= $(x^2 + 2x + 2)(x + 2)$ in $\mathbb{Z}_3[x]$.

PROOF

Existence of factorization: If $f \in F[x]$ is not a constant polynomial, then f being irreducible implies the result.

Otherwise, $f = f_1 F_1$, with $0 < \deg f_1, \deg F_1 < \deg f$. If f_1 and F_1 are irreducible, stop. Otherwise, keep going.

Eventually we end with a decomposition of f into irreducible polynomials.

Uniqueness: Let

$$cf_1f_2\cdots f_m = dg_1g_2\cdots g_n$$

be two decompositions into products of constants (c and d) and monic irreducibles (f_i, g_j). Now

$$f_1 \mid f_1 f_2 \cdots f_m \Longrightarrow f_1 \mid g_1 g_2 \cdots g_n$$

and since f_i, g_i are irreducible we can cancel f_1 and some g_j .

Repeating this for f_2, \ldots, f_m , we eventually obtain m = n and c = d in other words, each expression is simply a rearrangement of the factors of the other, as required.

THEOREM 2.4

Let F_q be a field with q elements. Then if $n \in \mathbb{N}$, there exists an irreducible polynomial of degree n in F[x].

PROOF First we introduce the idea of the **Riemann zeta function**:

$$\zeta(s) = \sum_{n=1}^{\infty} \frac{1}{n^s} = \prod_{p \text{ prime}} \frac{1}{1 - \frac{1}{p^s}}$$

To see the equality of the latter expressions note that

$$\frac{1}{1-x} = \sum_{i=0}^{\infty} x^i = 1 + x + x^2 + \cdots$$

and so

R.H.S. =
$$\prod_{p \text{ prime}} \left(\sum_{i=0}^{\infty} \frac{1}{p^{is}} \right)$$

= $\left(1 + \frac{1}{2^s} + \frac{1}{2^{2s}} + \cdots \right) \left(1 + \frac{1}{3^s} + \frac{1}{3^{2s}} + \cdots \right)$
= $1 + \frac{1}{2^s} + \frac{1}{3^s} + \frac{1}{4^s} + \cdots$

—note for the last step that terms will be of form

$$\left(\frac{1}{p_1^{a_1}\cdots p_R^{a_R}}\right)^s$$

up to some prime p_R , with $a_i \ge 0 \ \forall i = 1, \ldots, R$. and as $R \to \infty$, the prime factorizations

$$p_1^{a_1}\cdots p_R^{a_R}$$

map onto the natural numbers, \mathbb{N} .

We let N_m denote the number of monic irreducibles of degree m in $F_q[x]$. For example, $N_1 = q$ since $x + a, a \in F_q$ are the irreducible polynomials of degree 1.

Now let $|f| = q^{\deg f}$, and |0| = 0. Then we have

$$|fg| = |f||g|$$
 since deg $fg = \deg f + \deg g$

and, because of the uniqueness of factorization theorem,

$$\sum_{\substack{f \text{ monic}}} \frac{1}{|f|^s} = \prod_{\substack{f \text{ monic and} \\ \text{irreducible}}} \frac{1}{1 - \frac{1}{|f|^s}}$$

Now the left hand side is

=

$$\sum_{n=0}^{\infty} \sum_{\substack{f \text{ monic and} \\ \deg f = n}} \frac{1}{|f|^s}$$

$$\sum_{n=0}^{\infty} \frac{q^n}{q^{ns}}$$

(there are q^n monic polynomials of degree n)

$$= \sum_{n=0}^{\infty} \frac{1}{q^{n(s-1)}}$$

= $\frac{1}{1 - \frac{1}{q^{s-1}}}$
and R.H.S. = $\prod_{n=1}^{\infty} \frac{1}{\left(1 - \frac{1}{q^{ns}}\right)^{N_n}}$.

а

Equating the two, we have

$$\frac{1}{1 - \frac{1}{q^{s-1}}} = \prod_{n=1}^{\infty} \frac{1}{\left(1 - \frac{1}{q^{ns}}\right)^{N_n}}.$$
(4)

We now take logs of both sides, and then use the fact that

$$\log\left(\frac{1}{1-x}\right) = \sum_{n=1}^{\infty} \frac{x^n}{n} \quad \text{if } |x| < 1;$$

so (4) becomes

$$\log \frac{1}{1 - q^{-(s-1)}} = \prod_{n=1}^{\infty} \frac{1}{\left(1 - \frac{1}{q^{ns}}\right)^{N_n}}$$
$$\Rightarrow \sum_{k=1}^{\infty} \frac{1}{kq^{(s-1)k}} = -\sum_{n=1}^{\infty} N_n \log\left(1 - \frac{1}{q^{ns}}\right)$$
$$= \sum_{n=1}^{\infty} N_n \sum_{m=1}^{\infty} \frac{1}{mq^{mns}}$$
so
$$\sum_{k=1}^{\infty} \frac{q^k}{kq^{sk}} = \sum_{n=1}^{\infty} N_n \sum_{m=1}^{\infty} \frac{n}{mnq^{mns}}$$
$$= \sum_{k=1}^{\infty} \frac{\sum_{n=1}^{\infty} nN_n}{kq^{ks}}.$$

Putting $x = q^s$, we have

$$\sum_{k=1}^{\infty} \frac{q^k x^k}{k} = \sum_{k=1}^{\infty} x^k \times \sum_{mn=k} nN_n,$$

and since both sides are power series, we may equate coefficients of \boldsymbol{x}^k to obtain

$$q^k = \sum_{mn=k} nN_n = \sum_{n|k} nN_n.$$
(5)

We can deduce from this that $N_n > 0$ as $n \to \infty$ (see Berlekamp's "Algebraic Coding Theory").

Now note that $N_1 = q$, so if k is a prime—say k = p, (5) gives

$$q^p = N_1 + pN_p = q + pN_p$$

 $\Rightarrow N_p = \frac{q^p - q}{p} > 0$ as $q > 1$ and $p \ge 2$.

This proves the theorem for n = p, a prime.

But what if k is not prime? Equation (5) also tells us that

$$q^k \ge kN_k.$$

Now let $k \geq 2$. Then

$$q^{k} = kN_{k} + \sum_{\substack{n|k\\n\neq k}} nN_{n}$$

$$\leq kN_{k} + \sum_{\substack{n|k\\n\neq k}} q^{n} \quad (\text{as } nN_{n} \leq q^{n})$$

$$\leq kN_{k} + \sum_{\substack{n=1\\n\neq k}}^{\lfloor k/2 \rfloor} q^{n}$$

$$< kN_{k} + \sum_{\substack{n=0\\n=0}}^{\lfloor k/2 \rfloor + 1} q^{n} \quad (\text{adding } 1)$$

$$= kN_{k} + \frac{q^{\lfloor k/2 \rfloor + 1} - 1}{q - 1} \quad (\text{sum of geometric series}).$$

But

$$\frac{q^{t+1}-1}{q-1} < q^{t+1} \qquad \text{if } q \ge 2,$$

 \mathbf{SO}

$$q^{k} < kN_{k} + q^{\lfloor k/2 \rfloor + 1}$$

$$\Rightarrow N_{k} > \frac{q^{k} - q^{\lfloor k/2 \rfloor + 1}}{k}$$

$$\geq 0 \quad \text{if } q^{k} \geq q^{\lfloor k/2 \rfloor + 1}.$$

Since q > 1 (we cannot have a field with a single element, since the additive and multiplicative identities cannot be equal by one of the axioms), the latter condition is equivalent to

$$k \ge \lfloor k/2 \rfloor + 1$$

which is true and the theorem is proven.

2.4 Minimum Polynomial of a (Square) Matrix

Let $A \in M_{n \times n}(F)$, and $g = ch_A$. Then g(A) = 0 by the Cayley–Hamilton theorem.

DEFINITION 2.5

Any non-zero polynomial g of minimum degree and satisfying g(A) = 0 is called a **minimum polynomial** of A.

Note: If f is a minimum polynomial of A, then f cannot be a constant polynomial. For if f = c, a constant, then $0 = f(A) = cI_n$ implies c = 0.

THEOREM 2.5

If f is a minimum polynomial of A and g(A) = 0, then $f \mid g$. (In particular, $f \mid ch_A$.)

PROOF Let g(A) = 0 and f be a minimum polynomial. Then

$$g = qf + r,$$

where r = 0 or deg r < deg f. Hence

$$g(A) = q(A) \times 0 + r(A)$$

$$0 = r(A).$$

So if $r \neq 0$, the inequality deg $r < \deg f$ would give a contradict the definition of f. Consequently r = 0 and $f \mid g$.

Note: It follows that if f and g are minimum polynomials of A, then f|g and g|f and consequently f = cg, where c is a scalar. Hence there is a unique monic minimum polynomial and we denote it by m_A .

EXAMPLES (of minimum polynomials):

- 1. $A = 0 \Leftrightarrow m_A = x$
- 2. $A = I_n \Leftrightarrow m_A = x 1$
- 3. $A = cI_n \Leftrightarrow m_A = x c$

4. $A^2 = A$ and $A \neq 0$ and $A \neq I_n \Leftrightarrow m_A = x^2 - x$.

EXAMPLE 2.5 $F = \mathbb{Q}$ and

$$A = \begin{bmatrix} 5 & -6 & -6 \\ -1 & 4 & 2 \\ 3 & -6 & -4 \end{bmatrix}.$$

Now

$$A \neq c_0 I_3, c_0 \in \mathbb{Q}, \text{ so } m_A \neq x - c_0,$$

$$A^2 = 3A - 2I_3$$

$$\Rightarrow m_A = x^2 - 3x + 2$$

This is an special case of a general algorithm:

(Minimum polynomial algorithm) Let $A \in M_{n \times n}(F)$. Then we find the least positive integer r such that A^r is expressible as a linear combination of the matrices

$$I_n, A, \ldots, A^{r-1},$$

say

•

$$A^{r} = c_0 + c_1 A + \dots + c_{r-1} A^{r-1}$$

(Such an integer must exist as I_n , A, \ldots, A^{n^2} form a linearly dependent family in the vector space $M_{n \times n}(F)$ and this latter space has dimension equal to n^2 .)

Then $m_A = x^r - c_{r-1}x^{r-1} - \dots - c_1x - c_0$.

THEOREM 2.6

If $f = x^n + a_{n-1}x^{n-1} + \dots + a_1x + a_0 \in F[x]$, then $m_{C(f)} = f$, where

$$C(f) = \begin{bmatrix} 0 & 0 & 0 & -a_0 \\ 1 & 0 & \cdots & 0 & -a_1 \\ 0 & 1 & 0 & -a_2 \\ \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 & -a_{n-1} \end{bmatrix}$$

PROOF For brevity denote C(f) by A. Then post-multiplying A by the respective unit column vectors E_1, \ldots, E_n gives

$$AE_{1} = E_{2}$$

$$AE_{2} = E_{3} \Rightarrow A^{2}E_{1} = E_{3}$$

$$\vdots$$

$$AE_{n-1} = E_{n} \Rightarrow A^{n-1}E_{1} = E_{n}$$

$$AE_{n} = -a_{0}E_{1} - a_{2}E_{2} - \dots - a_{n-1}E_{n}$$

$$= -a_{0}E_{1} - a_{2}AE_{1} - \dots - a_{n-1}A^{n-1}E_{1} = A^{n}E_{1},$$

$$\Rightarrow f(A)E_1 = 0 \Rightarrow$$
 first column of $f(A)$ zero

Now although matrix multiplication is not commutative, multiplication of two matrices, each of which is a polynomial in a given square matrix A, is commutative. Hence f(A)g(A) = g(A)f(A) if $f, g \in F[x]$. Taking g = x gives

$$f(A)A = Af(A).$$

Thus

$$f(A)E_2 = f(A)AE_1 = Af(A)E_1 = 0$$

and so the second column of A is zero. Repeating this for E_3, \ldots, E_n , we see that

f(A) = 0

and thus $m_A|f$.

To show $m_A = f$, we assume deg $m_A = t < n$; say

$$m_A = x^t + b_{t-1}x^{t-1} + \dots + b_0$$

Now

$$m_A(A) = 0$$

$$\Rightarrow A^t + b_{t-1}A^{t-1} + \dots + b_0I_n = 0$$

$$\Rightarrow (A^t + b_{t-1}A^{t-1} + \dots + b_0I_n)E_1 = 0,$$

and recalling that $AE_1 = E_2$ etc., and t < n, we have

$$E_{t+1} + b_{t-1}E_t + \dots + b_1E_2 + b_0E_1 = 0$$

which is a contradiction—since the E_i are independent, the coefficient of E_{t+1} cannot be 1.

Hence $m_A = f$.

Note: It follows that $ch_A = f$. Because both ch_A and m_A have degree n and moreover m_A divides ch_A .

EXERCISE 2.2

If $A = J_n(a)$ for $a \in F$, an elementary Jordan matrix of size n, show

 \mathbf{SO}

that $m_A = (x-a)^n$ where

$$A = J_n(a) = \begin{bmatrix} a & 0 & & 0 \\ 1 & a & \cdots & & \\ 0 & 1 & & & \\ \vdots & \ddots & & \vdots \\ 0 & 0 & \cdots & a & 0 \\ 0 & 0 & & 1 & a \end{bmatrix}$$

(i.e. A is an $n \times n$ matrix with a's on the diagonal and 1's on the subdiagonal).

Note: Again, the minimum polynomial happens to equal the characteristic polynomial here.

DEFINITION 2.6

(Direct Sum of Matrices)

Let A_1, \ldots, A_t be matrices over F. Then the direct sum of these matrices is defined as follows:

$$A_1 \oplus A_2 \oplus \cdots \oplus A_t = \begin{bmatrix} A_1 & 0 & \cdots & \\ 0 & A_2 & & \\ \vdots & & \ddots & \vdots \\ & & \cdots & 0 & A_t \end{bmatrix}.$$

Properties:

1.

$$(A_1 \oplus \cdots \oplus A_t) + (B_1 \oplus \cdots \oplus B_t) = (A_1 + B_1) \oplus \cdots \oplus (A_t + B_t)$$

2. If $\lambda \in F$,

$$\lambda(A_1 \oplus \cdots \oplus A_t) = (\lambda A_1) \oplus \cdots \oplus (\lambda A_t)$$

3.

$$(A_1 \oplus \cdots \oplus A_t)(B_1 \oplus \cdots \oplus B_t) = (A_1B_1) \oplus \cdots \oplus (A_tB_t)$$

4. If $f \in F[x]$ and A_1, \ldots, A_t are square,

$$f(A_1 \oplus \cdots \oplus A_t) = f(A_1) \oplus \cdots \oplus f(A_t)$$

DEFINITION 2.7

If $f_1, \ldots, f_t \in F[x]$, we call $f \in F[x]$ a least common multiple (lcm) of f_1, \ldots, f_t if

1. $f_1 \mid f, \dots f_t \mid f$, and 2. $f_1 \mid e, \dots f_t \mid e \Rightarrow f \mid e$.

This uniquely defines the lcm up to a constant multiple and so we set "the" lcm to be the monic lcm .

EXAMPLES 2.1

If $fg \neq 0$, $\operatorname{lcm}(f,g) \mid fg$. (Recursive property)

$$lcm(f_1, ..., f_{t+1}) = lcm(lcm(f_1, ..., f_t), f_{t+1}).$$

THEOREM 2.7

$$m_{A_1 \oplus \cdots \oplus A_t} = \operatorname{lcm}(m_{A_1}, \ldots, m_{A_t}),$$

Also

$$\operatorname{ch}_{A_1\oplus\cdots\oplus A_t} = \prod_{i=1}^t \operatorname{ch}_{A_i}.$$

PROOF Let f = L.H.S. and g = R.H.S. Then

$$f(A_1 \oplus \dots \oplus A_t) = 0$$

$$\Rightarrow \quad f(A_1) \oplus \dots \oplus f(A_t) = 0 \oplus \dots \oplus 0$$

$$\Rightarrow \quad f(A_1) = 0, \dots, f(A_t) = 0$$

$$\Rightarrow \quad m_{A_1} \mid f, \dots, m_{A_t} \mid f$$

$$\Rightarrow \quad g \mid f.$$

Conversely,

$$m_{A_1} \mid g, \dots, m_{A_t} \mid g$$

$$\Rightarrow \quad g(A_1) = 0, \dots, g(A_t) = 0$$

$$\Rightarrow \quad g(A_1) \oplus \dots \oplus g(A_t) = 0 \oplus \dots \oplus 0$$

$$\Rightarrow \quad g(A_1 \oplus \dots \oplus A_t) = 0$$

$$\Rightarrow \quad f = m_{A_1 \oplus \dots \oplus A_t} \mid g.$$

Thus f = g.

EXAMPLE 2.6

Let A = C(f) and B = C(g). Then $m_{A \oplus B} = \operatorname{lcm}(f, g)$. Note: If

$$\begin{array}{rcl} f & = & cp_1^{a_1} \dots p_t^{a_t} \\ g & = & dp_1^{b_1} \dots p_t^{b_t} \end{array}$$

where $c, d \neq 0$ are in F and p_1, \ldots, p_t are distinct monic irreducibles, then

$$\gcd(f,g) = p_1^{\min(a_1,b_1)} \dots p_t^{\min(a_t,b_t)}, \\ \operatorname{lcm}(f,g) = p_1^{\max(a_1,b_1)} \dots p_t^{\max(a_t,b_t)}$$

Note

$$\min(a_i, b_i) + \max(a_i, b_i) = a_i + b_i.$$

 \mathbf{SO}

$$gcd(f, g) lcm(f, g) = fg.$$

EXAMPLE 2.7

If $A = \text{diag}(\lambda_1, \ldots, \lambda_n)$, then $m_A = (x - c_1) \cdots (x - c_t)$, where c_1, \ldots, c_t are the distinct members of the sequence $\lambda_1, \ldots, \lambda_n$.

PROOF. For A is the direct sum of the 1×1 matrices $\lambda_1, \ldots, \lambda_n$ having minimum polynomials $x - \lambda_1, \ldots, \lambda_n$. Hence

$$m_A = \operatorname{lcm} (x - \lambda_1, \dots, x - \lambda_n) = (x - c_1) \cdots (x - c_t).$$

We know that $m_A \mid ch_A$. Hence if

$$\mathrm{ch}_A = p_1^{a_1} \dots p_t^{a_t}$$

where $a_1 > 0, \ldots, a_t > 0$, and p_1, \ldots, p_t are distinct monic irreducibles, then

$$m_A = p_1^{b_1} \dots p_t^{b_t}$$

where $0 \leq b_i \leq a_i, \forall i = 1, \dots, t$.

We soon show that each $b_i > 0$, i.e. if $p \mid ch_A$ and p is irreducible then $p \mid m_A$.

2.5 Construction of a field of p^n elements

(where p is prime and $n \in \mathbb{N}$)

Let f be a monic irreducible polynomial of degree n in $\mathbb{Z}_p[x]$ —that is, $F_q = \mathbb{Z}_p$ here.

For instance,

$$n = 2, p = 2 \implies x^2 + x + 1 = f$$

 $n = 3, p = 2 \implies x^3 + x + 1 = f \text{ or } x^3 + x^2 + 1 = f.$

Let A = C(f), the companion matrix of f. Then we know f(A) = 0.

We assert that the set of all matrices of the form g(A), where $g \in \mathbb{Z}_p[x]$, forms a field consisting of precisely p^n elements. The typical element is

$$b_0I_n + b_1A + \dots + b_tA^t$$

where $b_0, \ldots, b_t \in \mathbb{Z}_p$.

We need only show existence of a multiplicative inverse for each element except 0 (the additive identity), as the remaining axioms clearly hold.

So let $g \in \mathbb{Z}_p[x]$ such that $g(A) \neq 0$. We have to find $h \in \mathbb{Z}_p[x]$ satisfying

 $g(A)h(A) = I_n.$

Note that $g(A) \neq 0 \Rightarrow f \not\mid g$, since

$$f \mid g \Rightarrow g = ff_1$$

and hence

$$g(A) = f(A)f_1(A) = 0f_1(A) = 0$$

Then since f is irreducible and f/g, there exist $u, v \in \mathbb{Z}_p[x]$ such that

$$uf + vg = 1.$$

Hence $u(A)f(A) + v(A)g(A) = I_n$ and $v(A)g(A) = I_n$, as required.

We now show that our new field is a \mathbb{Z}_p- vector space with basis consisting of the matrices

 $I_n, A, \ldots, A^{n-1}.$

Firstly the spanning property: By Euclid's division theorem,

$$g = fq + r$$

where $q, r \in \mathbb{Z}_p[x]$ and deg $r < \deg g$. So let

$$r = r_0 + r_1 x + \dots + r_{n-1} x^{n-1}$$

where $r_0, \ldots, r_{n-1} \in \mathbb{Z}_p$. Then

$$g(A) = f(A)q(A) + r(A)$$

= 0q(A) + r(A)
= r(A)
= r_0I_n + r_1A + \dots + r_{n-1}A^{n-1}

Secondly, linear independence over Z_p : Suppose that

$$r_0 I_n + r_1 A + \dots + r_{n-1} A^{n-1} = 0,$$

where $r_0, r_1, \ldots, r_{n-1} \in \mathbb{Z}_p$. Then r(A) = 0, where

$$r = r_0 + r_1 x + \dots + r_{n-1} x^{n-1}.$$

Hence $m_A = f$ divides r. Consequently r = 0, as deg f = n whereas deg r < n if $r \neq 0$.

Consequently, there are p^n such matrices g(A) in the field we have constructed.

Numerical Examples

EXAMPLE 2.8

Let
$$p = 2, n = 2, f = x^2 + x + 1 \in \mathbb{Z}_2[x]$$
, and $A = C(f)$. Then
$$A = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix},$$

and

$$F_4 = \{ a_0 I_2 + a_1 A \mid a_0, a_1 \in \mathbb{Z}_2 \} \\ = \{ 0, I_2, A, I_2 + A \}.$$

We construct addition and multiplication tables for this field, with $B = I_2 + A$ (as an exercise, check these):

В

0 0

 $A \mid B$

 $B I_2$

 $I_2 \mid A$

\oplus	0	I_2	A	В	\otimes	0	I_2	
0	0	I_2	A	В	0	0	0	
I_2	I_2	0	B	A	I_2	0	I_2	
A	A	B	0	I_2	A	0	A	
B	B	A	I_2	0	B	0	B	

EXAMPLE 2.9

Let $p = 2, n = 3, f = x^3 + x + 1 \in \mathbb{Z}_2[x]$. Then

$$A = C(f) = \begin{bmatrix} 0 & 0 & -1 \\ 1 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix},$$

and our eight-member field F_8 (usually denoted by GF(8) ["GF" corresponds to "Galois Field", in honour of Galois]) is

$$F_8 = \{ a_0 I_3 + a_1 A + a_2 A^2 \mid a_0, a_1, a_2 \in \mathbb{Z}_2 \}$$

= $\{ 0, I_3, A, A^2, I_3 + A, I_3 + A^2, A + A^2, I_3 + A + A^2 \}.$

Now find $(A^2 + A)^{-1}$.

Solution: use Euclid's algorithm.

$$x^{3} + x + 1 = (x + 1)(x^{2} + x) + 1.$$

Hence

$$x^{3} + x + 1 + (x + 1)(x^{2} + x) = 1$$

$$A^{3} + A + I_{3} + (A + I_{3})(A^{2} + A) = I_{3}$$

$$(A + I_{3})(A^{2} + A) = I_{3}$$

Hence $(A^2 + A)^{-1} = A + I_3$.

THEOREM 2.8

Every <u>finite</u> field has precisely p^n elements for some prime p—the least positive integer with the property that

$$\underbrace{1+1+1+\dots+1}_p = 0.$$

p is then called the **characteristic** of the field.

Also, if $x \in F$, a field of q elements, then it can be shown that if $x \neq 0$, then

$$x^{q-1} = 1$$

In the special case $F = \mathbb{Z}_p$, this reduces to **Fermat's Little Theorem**:

$$x^{p-1} \equiv 1 \pmod{p},$$

if p is prime not dividing x.

2.6 Characteristic and Minimum Polynomial of a Transformation

DEFINITION 2.8

(Characteristic polynomial of $T: V \mapsto V$) Let β be a basis for V and $A = [T]_{\beta}^{\beta}$. Then we define $ch_T = ch_A$. This polynomial is independent of the basis β :

PROOF (ch_T is independent of the basis.)

If γ is another basis for V and $B = [T]_{\gamma}^{\gamma}$, then we know $A = P^{-1}BP$ where P is the change of basis matrix $[I_V]_{\beta}^{\gamma}$.

Then

$$ch_A = ch_{P^{-1}BP}$$

$$= det(xI_n - P^{-1}BP) \text{ where } n = \dim V$$

$$= det(P^{-1}(xI_n)P - P^{-1}BP)$$

$$= det(P^{-1}(xI_n - B)P)$$

$$= det P^{-1} ch_B det P$$

$$= ch_B.$$

DEFINITION 2.9

If $f = a_0 + \cdots + a_t x^t$, where $a_0, \ldots, a_t \in F$, we define

$$f(T) = a_0 I_V + \dots + a_t T^t.$$

Then the usual properties hold:

$$f, g \in F[x] \Rightarrow (f+g)(T) = f(T)+g(T) \text{ and } (fg)(T) = f(T)g(T) = g(T)f(T).$$

LEMMA 2.1

$$f \in F[x] \Rightarrow [f(T)]_{\beta}^{\beta} = f\left([T]_{\beta}^{\beta}\right).$$

Note: The Cayley-Hamilton theorem for matrices says that $\operatorname{ch}_A(A) = 0$. Then if $A = [T]_{\beta}^{\beta}$, we have by the lemma

$$\left[\operatorname{ch}_{T}\left(T\right)\right]_{\beta}^{\beta} = \operatorname{ch}_{T}\left(A\right) = \operatorname{ch}_{A}\left(A\right) = 0,$$

so $\operatorname{ch}_{T}(T) = 0_{V}$.

DEFINITION 2.10

Let $T: V \to V$ be a linear transformation over F. Then any polynomial of least positive degree such that

$$f(T) = 0_V$$

is called a minimum polynomial of T.

We have corresponding results for polynomials in a transformation T to those for polynomials in a square matrix A:

$$g = qf + r \Rightarrow g(T) = q(T)f(T) + r(T).$$

Again, there is a unique monic minimum polynomial of T is denoted by m_T and called "the" minimum polynomial of T.

Also note that because of the lemma,

$$m_T = m_{[T]^\beta_\beta}.$$

For (with $A = [T]^{\beta}_{\beta}$)

- (a) $m_A(A) = 0$, so $m_A(T) = 0_V$. Hence $m_T | m_A$.
- (b) $m_T(T) = 0_V$, so $[m_T(T)]^{\beta}_{\beta} = 0$. Hence $m_T(A) = 0$ and so $m_A | m_T$.

EXAMPLES 2.2

 $T = 0_V \Leftrightarrow m_T = x.$ $T = I_V \Leftrightarrow m_T = x - 1.$ $T = cI_V \Leftrightarrow m_T = x - c.$ $T^2 = T \text{ and } T \neq 0_V \text{ and } T \neq I_V \Leftrightarrow m_T = x^2 - x.$

2.6.1 $M_{n \times n}(F[x])$ —Ring of Polynomial Matrices

EXAMPLE:

$$\begin{bmatrix} x^2 + 2 & x^5 + 5x + 1 \\ x + 3 & 1 \end{bmatrix} \in M_{2 \times 2}(\mathbb{Q}[x])$$

= $x^5 \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + x^2 \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + x \begin{bmatrix} 0 & 5 \\ 1 & 0 \end{bmatrix} + \begin{bmatrix} 2 & 1 \\ 3 & 1 \end{bmatrix}$

—we see that any element of $M_{n \times n}(F[x])$ is expressible as

$$x^{m}A_{m} + x^{m-1}A_{m-1} + \dots + A_{0}$$

where $A_i \in M_{n \times n}(F)$. We write the coefficient of x^i after x^i , to distinguish these entities from corresponding objects of the following ring.

2.6.2 $M_{n \times n}(F)[y]$ —Ring of Matrix Polynomials

This consists of all polynomials in y with coefficients in $M_{n \times n}(F)$. EXAMPLE:

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} y^5 + \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} y^2 + \begin{bmatrix} 0 & 5 \\ 1 & 0 \end{bmatrix} y + \begin{bmatrix} 2 & 1 \\ 3 & 1 \end{bmatrix} \in M_{2 \times 2}(F)[y].$$

THEOREM 2.9

The mapping

$$\Phi: M_{n \times n}(F)[y] \mapsto M_{n \times n}(F[x])$$

given by

$$\Phi(A_0 + A_1y + \dots + A_my^m) = A_0 + xA_1 + \dots + x^mA_m$$

where $A_i \in M_{n \times n}(F)$, is a 1–1 correspondence and has the following properties:

$$\Phi(X + Y) = \Phi(X) + \Phi(Y)$$

$$\Phi(XY) = \Phi(X)\Phi(Y)$$

$$\Phi(tX) = t\Phi(X) \quad \forall t \in F.$$

Also

$$\Phi(I_n y - A) = xI_n - A \quad \forall A \in M_{n \times n}(F).$$

THEOREM 2.10 ((Left) Remainder theorem for matrix polynomials)

Let
$$B_m y^m + \dots + B_0 \in M_{n \times n}(F)[y]$$
 and $A \in M_{n \times n}(F)$.
Then
 $B_m y^m + \dots + B_0 = (I_n y - A)Q + R$

where

$$R = A^m B_m + \dots + A B_1 + B_0$$

and $Q = C_{m-1} y^{m-1} + \dots + C_0$

where C_{m-1}, \ldots, C_0 are computed recursively:

$$B_{m} = C_{m-1}$$

$$B_{m-1} = -AC_{m-1} + C_{m-2}$$

$$\vdots$$

$$B_{1} = -AC_{1} + C_{0}.$$

PROOF. First we verify that $B_0 = -AC_0 + R$:

$$R = A^{m}B_{m} = A^{m}C_{m-1}$$

$$+A^{m-1}B_{m-1} -A^{m}C_{m-1} + A^{m-1}C_{m-2}$$

$$+ +$$

$$\vdots \qquad \vdots$$

$$+AB_{1} -A^{2}C_{1} + AC_{0}$$

$$+B_{0} = B_{0} + AC_{0}.$$

Then

$$(I_n y - A)Q + R = (I_n y)(C_{m-1}y^{m-1} + \dots + C_0) - A(C_{m-1}y^{m-1} + \dots + C_0) + A^m B_m + \dots + B_0 = C_{m-1}y^m + (C_{m-2} - AC_{m-1})y^{m-1} + \dots + (C_0 - AC_1)y + - AC_0 + R = B_m y^m + B_{m-1}y^{m-1} + \dots + B_1 y + B_0.$$

Remark. There is a similar "right" remainder theorem.

THEOREM 2.11

If p is an irreducible polynomial dividing ch_A , then $p \mid m_A$.

PROOF (From Burton Jones, "Linear Algebra"). Let $m_A = x^t + a_{t-1}x^{t-1} + \cdots + a_0$ and consider the matrix polynomial in y

$$\Phi^{-1}(m_A I_n) = I_n y^t + (a_{t-1} I_n) y^{t-1} + \dots + (a_0 I_n)$$

= $(I_n y - A) Q + A^t I_n + A^{t-1} (a_{t-1} I_n) + \dots + a_0 I_n$
= $(I_n y - A) Q + m_T(A)$
= $(I_n y - A) Q.$

Now take Φ of both sides to give

$$m_A I_n = (xI_n - A)\Phi(Q)$$

and taking determinants of both sides yields

$${m_A}^n = \operatorname{ch}_A \times \det \Phi(Q).$$

So letting p be an irreducible polynomial dividing ch_A , we have $p \mid \{m_A\}^n$ and hence $p \mid m_A$.

Alternative simpler proof (MacDuffee): $m_A(x) - m_A(y) = (x - y)k(x, y)$, where $k(x, y) \in F[x, y]$. Hence $m_A(x)I_n = m_A(xI_n) - m_A(A) = (xI_n - A)k(xI_n, A).$

Now take determinants to get

$$m_A(x)^n = ch_A(x) \det k(xI_n, A).$$

Exercise: If $\Delta(x)$ is the gcd of the elements of $adj(xI_n - A)$, use the equation $(xI_n - a)adj(xI_n - A) = ch_A(x)I_n$ and an above equation to deduce that $m_A(x) = ch_A(x)/\Delta(x)$.

EXAMPLES 2.3

With $A = 0 \in M_{n \times n}(F)$, we have $\operatorname{ch}_A = x^n$ and $m_A = x$. $A = \operatorname{diag}(1, 1, 2, 2, 2) \in M_{5 \times 5}(\mathbb{Q})$. Here

$$ch_A = (x-1)^2 (x-2)^3$$
 and $m_A = (x-1)(x-2)$.

DEFINITION 2.11

A matrix $A \in M_{n \times n}(F)$ is called diagonable over F if there exists a non-singular matrix $P \in M_{n \times n}(F)$ such that

$$P^{-1}AP = \operatorname{diag}(\lambda_1, \ldots, \lambda_n),$$

where $\lambda_1, \ldots, \lambda_n$ belong to F.

THEOREM 2.12

If A is diagonable, then m_A is a product of distinct linear factors.

PROOF

If $P^{-1}AP = \text{diag}(\lambda_1, \dots, \lambda_n)$ (with $\lambda_1, \dots, \lambda_n \in F$) then

$$m_A = m_{P^{-1}AP} = m_{\text{diag}}(\lambda_1, \dots, \lambda_n)$$
$$= (x - c_1)(x - c_2) \dots (x - c_t)$$

where c_1, \ldots, c_t are the distinct members of the sequence $\lambda_1, \ldots, \lambda_n$.

The converse is also true, and will (fairly) soon be proved.

EXAMPLE 2.10

$$A = J_n(a).$$

We saw earlier that $m_A = (x - a)^n$ so if $n \ge 2$ we see that A is not diagonable.

DEFINITION 2.12

(Diagonable LTs)

 $T: V \mapsto V$ is called **diagonable** over F if there exists a basis β for V such that $[T]^{\beta}_{\beta}$ is diagonal.

THEOREM 2.13

A is diagonable $\Leftrightarrow T_A$ is diagonable.

PROOF (Sketch)

 \Rightarrow Suppose $P^{-1}AP = \text{diag}(\lambda_1, \dots, \lambda_n)$. Now pre-multiplying by P and letting $P = [P_1| \cdots |P_n]$ we see that

$$T_A(P_1) = AP_1 = \lambda_1 P_1$$

$$\vdots$$

$$T_A(P_n) = AP_n = \lambda_n P_n$$

and we let β be the basis P_1, \ldots, P_n over $V_n(F)$. Then

$$[T_A]^{\beta}_{\beta} = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \ddots & \\ & & & & \lambda_n \end{bmatrix}.$$

 \Leftarrow Reverse the argument and use Theorem 1.17.

THEOREM 2.14

Let $A \in M_{n \times n}(F)$. Then if λ is an eigenvalue of A with multiplicity m, (that is $(x - \lambda)^m$ is the exact power of $x - \lambda$ which divides ch_A), we have

nullity
$$(A - \lambda I_n) \leq m$$
.

REMARKS. (1) If m = 1, we deduce that nullity $(A - \lambda I_n) = 1$. For the inequality

$$1 \leq \text{nullity} (A - \lambda I_n)$$

always holds.

(2) The integer nullity $(A - \lambda I_n)$ is called the *geometric multiplicity* of the eigenvalue λ , while *m* is referred to as the *algebraic multiplicity* of λ .

PROOF. Let v_1, \ldots, v_r be a basis for $N(A - \lambda I_n)$, where λ is an eigenvalue of A having multiplicity m. Extend this linearly independent family to a basis $v_1, \ldots, v_r, v_{r+1}, \ldots, v_n$ of $V_n(F)$. Then the following equations hold:

$$Av_{1} = \lambda v_{1}$$

$$\vdots$$

$$Av_{r} = \lambda v_{r}$$

$$Av_{r+1} = b_{11}v_{1} + \dots + b_{n1}v_{n}$$

$$\vdots$$

$$Av_{n} = b_{1n-r}v_{1} + \dots + b_{nn-r}v_{n}.$$

These equations can be combined into a single matrix equation:

$$\begin{aligned} A[v_1|\cdots|v_r|v_{r+1}|\cdots|v_n] &= [Av_1|\cdots|Av_r|Av_{r+1}|\cdots|Av_n] \\ &= [\lambda v_1|\cdots|\lambda v_r|b_{11}v_1+\cdots+b_{n1}v_n|\cdots|b_{1n-r}v_1+\cdots+b_{nn-r}v_n] \\ &= [v_1|\cdots|v_n] \left[\frac{\lambda I_r \mid B_1}{0 \mid B_2}\right]. \end{aligned}$$

Hence if $P = [v_1| \cdots |v_n]$, we have

$$P^{-1}AP = \begin{bmatrix} \lambda I_r & B_1 \\ 0 & B_2 \end{bmatrix}.$$

Then

$$\operatorname{ch}_A = \operatorname{ch}_{P^{-1}AP} = \operatorname{ch}_{\lambda I_r} \cdot \operatorname{ch}_{B_2} = (x - \lambda)^r \operatorname{ch}_{B_2}$$

and because $(x - \lambda)^m$ is the exact power of $x - \lambda$ dividing ch_A , it follows that

nullity
$$(A - \lambda I_n) = r \leq m$$
.

THEOREM 2.15

Suppose that $ch_T = (x - c_1)^{a_1} \cdots (x - c_t)^{a_t}$. Then T is diagonable if

nullity $(T - c_i I_v) = a_i$ for $1 \le i \le t$.

PROOF. We first prove that the subspaces $\text{Ker}(T - c_i I_V)$ are independent. (Subspaces V_1, \ldots, V_t are called *independent* if

 $v_1 + \dots + v_t = 0, v_i \in V_i, i = 1, \dots, t, \Rightarrow v_1 = 0, \dots, v_t = 0.$

Then dim $(V_1 + \cdots + V_t)$ = dim $(V_1) + \cdots +$ dim $V_t)$.)

Assume that

$$v_1 + \dots + v_t = 0,$$

where $v_i \in \text{Ker}(T - c_i I_v)$ for $1 \le i \le t$. Then

$$T(v_1 + \dots + v_t) = T(0)$$

 $c_1v_1 + \dots + c_tv_t = 0.$

Similarly we deduce that

$$c_1^2 v_1 + \dots + c_t^2 v_t = 0$$

$$\vdots$$

$$c_1^{t-1} v_1 + \dots + c_t^{t-1} v_t = 0.$$

We can combine these t equations into a single matrix equation

$$\begin{bmatrix} 1 & \cdots & 1 \\ c_1 & \cdots & c_t \\ \vdots \\ c_1^{t-1} & \cdots & c_t^{t-1} \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_t \end{bmatrix} = \begin{bmatrix} o \\ \vdots \\ 0 \end{bmatrix}.$$

However the coefficient matrix is the Vandermonde matrix, which is nonsingular as $c_i \neq c_j$ if $i \neq j$, so we deduce that $v_1 = 0, \dots, v_t = 0$. Hence with $V_i = \text{Ker} (T - c_i I_V)$, we have

$$\dim (V_1 + \dots + V_t) = \sum_{i=1}^t \dim V_i = \sum_{i=1}^t a_i = \dim V_i$$

Hence

$$V = V_1 + \dots + V_t$$

Then if β_i is a basis for V_i for $i \leq i \leq t$ and $\beta = \beta_1 \cup \cdots \cup \beta_t$, it follows that β is a basis for V. Moreover

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} (c_i I_{a_i})$$

and T is diagonable.

EXAMPLE. Let

$$A = \begin{bmatrix} 5 & 2 & -2 \\ 2 & 5 & -2 \\ -2 & -2 & 5 \end{bmatrix}.$$

(a) We find that $ch_A = (x-3)^2(x-9)$. Next we find bases for each of the eigenspaces $N(A - 9I_3)$ and $N(A - 3I_3)$:

First we solve $(A - 3I_3)X = 0$. We have

$$A - 3I_3 = \begin{bmatrix} 2 & 2 & -2 \\ 2 & 2 & -2 \\ -2 & -2 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

Hence the eigenspace consists of vectors $X = [x, y, z]^t$ satisfying x = -y+z, with y and z arbitrary. Hence

$$X = \begin{bmatrix} -y+z \\ y \\ z \end{bmatrix} = y \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} + z \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix},$$

so $X_{11} = [-1, 1, 0]^t$ and $X_{12} = [1, 0, 1]^t$ form a basis for the eigenspace corresponding to the eigenvalue 3.

Next we solve $(A - 9I_3)X = 0$. We have

$$A - 9I_3 = \begin{bmatrix} -4 & 2 & -2 \\ 2 & -4 & -2 \\ -2 & -2 & -4 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}.$$

Hence the eigenspace consists of vectors $X = [x, y, z]^t$ satisfying x = -zand y = -z, with z arbitrary. Hence

$$X = \begin{bmatrix} -z \\ -z \\ z \end{bmatrix} = z \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix}$$

and we can take $X_{21} = [-1, -1, 1]^t$ as a basis for the eigenspace corresponding to the eigenvalue 9.

Then $P = [X_{11}|X_{12}|X_{21}]$ is non-singular and

$$P^{-1}AP = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 9 \end{bmatrix}$$

THEOREM 2.16

If

$$m_T = (x - c_1) \dots (x - c_t)$$

for c_1, \ldots, c_t distinct in F, then T is diagonable and conversely. Moreover there exist unique linear transformations T_1, \ldots, T_t satisfying

$$I_V = T_1 + \dots + T_t,$$

$$T = c_1 T_1 + \dots + c_t T_t,$$

$$T_i T_j = 0_V \text{ if } i \neq j,$$

$$T_i^2 = T_i, \ 1 \le i \le t.$$

Also rank $T_i = a_i$, where $ch_T = (x - c_1)^{a_1} \cdots (x - c_t)^{a_t}$.

Remarks.

- 1. T_1, \ldots, T_t are called the *principal idempotents* of T.
- 2. If $g \in F[x]$, then $g(T) = g(c_1)T_1 + \dots + g(c_t)T_t$. For example $T^m = c_1^m T_1 + \dots + c_t^m T_t.$
- 3. If c_1, \ldots, c_t are non-zero (that is the eigenvalues of T are non-zero), the T^{-1} is given by

$$T^{-1} = c_1^{-1}T_1 + \dots + c_t^{-1}T_t.$$

Formulae 2 and 3 are useful in the corresponding matrix formulation. PROOF Suppose $m_T = (x - c_1) \cdots (x - c_t)$, where c_1, \ldots, c_t are distinct. Then $ch_T = (x - c_1)^{a_1} \cdots (x - c_t)^{a_t}$. To prove T is diagonable, we have to prove that nullity $(T - c_i I_V) = a_i$, $1 \le i \le t$

Let p_1, \ldots, p_t be the Lagrange interpolation polynomials based on c_1, \ldots, c_t , i.e.

$$p_i = \prod_{\substack{j=1\\j\neq i}}^t \left(\frac{x-c_j}{c_i-c_j}\right), \ 1 \le i \le t.$$

Then

$$g \in F[x] \Rightarrow g = g(c_1)p_1 + \dots + g(c_t)p_t.$$

In particular,

 $g = 1 \Rightarrow 1 = p_1 + \dots + p_t$

and

$$g = x \Rightarrow x = c_1 p_1 + \dots + c_t p_t.$$

Hence with $T_i = p_i(T)$,

$$I_V = T_1 + \dots + T_t$$

$$T = c_1 T_1 + \dots + c_t T_t$$

Next

$$m_T = (x - c_1) \dots (x - c_t) \mid p_i p_j \quad \text{if } i \neq j$$

$$\Rightarrow \quad (p_i p_j)(T) = 0_V \quad \text{if } i \neq j$$

$$\Rightarrow \quad p_i(T) p_j(T) = 0_V \quad \text{or } \quad T_i T_j = 0_V \quad \text{if } i \neq j$$

Then $T_i^2 = T_i(T_1 + \dots + T_t) = T_i I_V = T_i.$ Next

$$0_V = m_T(T) = (T - c_1 I_V) \cdots (T - c_t I_V).$$

Hence

dim
$$V$$
 = nullity $0_V \le \sum_{i=1}^t$ nullity $(T - c_i I_V) \le \sum_{i=1}^t a_i = \dim V.$

Consequently nullity $(T - c_i I_V) = a_i$, $1 \le i \le t$ and T is therefore diagonable.

Next we prove that rank $T_i = a_i$. From the definition of p_i , we have

nullity
$$p_i(T) \leq \sum_{\substack{j=1\\j\neq i}}^t$$
 nullity $(T - c_j I_V) = \sum_{\substack{j=1\\j\neq i}}^t a_j = \dim V - a_i.$

Also $p_i(T)(T - c_i I_V) = 0$, so $\operatorname{Im}(T - c_i I_V) \subseteq \operatorname{Ker} p_i(T)$. Hence

 $\dim V - a_i \le \operatorname{nullity} p_i(T)$

and consequently nullity $p_i(T) = \dim(V) - a_i$, so rank $p_i(T) = a_i$.

We next prove the uniqueness of T_1, \ldots, T_t . Suppose that S_1, \ldots, S_t also satisfy the same conditions as T_1, \ldots, T_t . Then

so $(c_j - c_i)T_iS_j = 0_V$ and $T_iS_j = 0_V$ if $i \neq j$. Hence

$$T_i = T_i I_V = T_i (\sum_{j=1}^t S_j) = T_i S_i$$
$$S_i = I_V S_i = (\sum_{j=1}^t T_j) S_i = T_i S_i$$

Hence $T_i = S_i$.

Conversely, suppose that T is diagonable and let β be a basis of V such that

$$A = [T]_{\beta}^{\beta} = \operatorname{diag}(\lambda_1, \dots, \lambda_n).$$

Then $m_T = m_A = (x - c_1) \cdots (x - c_t)$, where c_1, \ldots, c_t are the distinct members of the sequence $\lambda_1, \ldots, \lambda_n$.

COROLLARY 2.5

If

$$ch_T = (x - c_1) \dots (x - c_t)$$

with c_i distinct members of F, then T is diagonable. PROOF: Here $m_T = ch_T$ and we use theorem 3.3.

EXAMPLE 2.11

Let

$$A = \left[\begin{array}{cc} 0 & a \\ b & 0 \end{array} \right] \qquad a,b \in F, \ ab \neq 0, \ 1+1 \neq 0.$$

Then A is diagonable if and only if $ab = y^2$ for some $y \in F$. For $ch_A = x^2 - ab$, so if $ab = y^2$,

$$ch_A = x^2 - y^2 = (x+y)(x-y)$$

which is a product of distinct linear factors, as $y \neq -y$ here.

Conversely suppose that A is diagonable. Then as A is not a scalar matrix, it follows that m_A is not linear and hence

$$m_A = (x - c_1)(x - c_2),$$

where $c_1 \neq c_2$. Also $ch_A = m_A$, so $ch_A(c_1) = 0$. Hence

$$c_1^2 - ab = 0$$
, or $ab = c_1^2$.

For example, take $F = \mathbb{Z}_7$ and let a = 1 and b = 3. Then $ab \neq y^2$ and consequently A is not diagonable.

3 Invariant subspaces

DEFINITIONS 3.1

Subspaces V_1, \ldots, V_t of V are called **independent** if

$$v_1 + \dots + v_t = 0 \Rightarrow v_1 = 0, \dots, v_t = 0$$

$$\forall v_1 \in V_1, \dots, v_t \in V_t.$$

We say that V is the (internal) direct sum of the subspaces V_1, \ldots, V_t if

- (a) V_1, \ldots, V_t are independent and
- (b) $V = V_1 + \dots + V_t$.

I.e. every element $v \in V$ is uniquely expressible as

$$v = v_1 + \dots + v_t$$

with $v_i \in V_i$.

Then V is isomorphic to the (external) direct sum $V_1 \oplus \cdots \oplus V_t$ under the isomorphism $v \mapsto (v_1, \ldots, v_t)$ and we write $V = V_1 \oplus \cdots \oplus V_t$.

THEOREM 3.1

If $V = V_1 \oplus \cdots \oplus V_t$ (an internal direct sum) and β_1, \ldots, β_t are bases for V_1, \ldots, V_t respectively, then

$$\beta = \beta_1 \cup \cdots \cup \beta_t,$$

the sequence formed by juxtaposing the separate bases, is a basis for V. Also

$$\dim V = \dim V_1 + \dots + \dim V_t.$$

Proof: Left as an exercise.

DEFINITION. Let $T: V \mapsto V$ be a LT and W a subspace of V. Then if

$$w \in W \Rightarrow T(w) \in W,$$

we say W is a *T***-invariant subspace** of V. We can then consider the linear transformation $T_W: W \to W$ defined by

$$T_W(w) = T(w) \qquad \forall w \in W.$$

If β' is a basis for W, $\{0\} \subset W \subset V$, and β is an extension to a basis of V, then

$$[T]^{\beta}_{\beta} = \left[\begin{array}{c|c} [T_W]^{\beta'}_{\beta'} & B_1 \\ \hline 0 & B_2 \end{array} \right].$$

A situation of great interest is when we have *T*-invariant subspaces W_1, \ldots, W_t and $V = W_1 \oplus \cdots \oplus W_t$. For if $\beta = \beta_1 \cup \cdots \cup \beta_t$, where β_i is a basis for W_i , we see that

$$[T]^{\beta}_{\beta} = [T_{W_1}]^{\beta_1}_{\beta_1} \oplus \cdots \oplus [T_{W_t}]^{\beta_t}_{\beta_t}.$$

There are two important examples of T-invariant subspaces that arise in our study of Jordan and rational canonical forms - Ker $p^t(T)$ and T-cyclic subspaces.

3.1 *T*-cyclic subspaces

DEFINITION 3.1

The unique monic polynomial f in F[x] of least degree satisfying

$$f(T)(v) = 0$$

is called the **minimum polynomial of the vector** $v \in V$ relative to the transformation $T: V \mapsto V$ and is denoted $m_{T,v}$.

Then $f(T)(v) = 0 \Rightarrow m_{T,v} \mid f$, so $m_{T,v} \mid m_T$. Also $m_{T,v} = 1 \Leftrightarrow v = 0$ and so if $v \neq 0$, deg $m_{T,v} \ge 1$.

EXAMPLE. Let
$$T = T_A$$
, where $A = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$. Also let
 $v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

Then $Av_1 = \begin{bmatrix} 0\\1 \end{bmatrix} \neq c_0v_1$, so $m_{T,v_1} \neq x - c_0$. Next $A^2v_1 = \begin{bmatrix} 0\\0 \end{bmatrix}$, so $m_{T,v_1} = x^2$. Also $Av_2 = \begin{bmatrix} 0\\0 \end{bmatrix}$, so $m_{T,v_2} = x$.

DEFINITION 3.2

(T-cyclic subspace generated by v.)

If $v \in V$, the set of all vectors of the form f(T)(v), $f \in F[x]$, forms a subspace of V called the T-cyclic subspace generated by v. It is denoted by $C_{T,v}$.

PROOF. Exercise.

Also, $C_{T,v}$ is a T-invariant subspace of V. For

$$w \in C_{T,v} \Rightarrow w = f(T)(v)$$

$$\Rightarrow T(w) = T(f(T)(v)) = (Tf(T))(v) = ((xf)(T))(v) \in C_{T,v}.$$

We see that v = 0 if and only if $C_{T,v} = \{0\}$.

THEOREM 3.2

Let $v \neq 0, v \in V$. Then $C_{T,v}$ has the basis β

$$v, T(v), T^2(v), \dots, T^{k-1}(v)$$

where $k = \deg m_{T,v}$. (β is called the *T*-cyclic basis generated by v.) Note that dim $C_{T,v} = \deg m_{T,v}$.

Finally,

$$\left[T_{C_{T,v}}\right]_{\beta}^{\beta} = C(m_{T,v}),$$

the companion matrix of the minimum polynomial of v.

PROOF.

1. The *T*-cyclic basis is a basis for $C_{T,v}$:

Spanning:

Let
$$w \in \langle v, T(v), \dots, T^{k-1}(v) \rangle$$
, so
 $w = w_0 v + w_1 T(v) + \dots + w_{k-1} T^{k-1}(v)$
 $= (w_0 I_V + \dots + w_{k-1} T^{k-1})(v)$
 $= g(T)(v),$

where $g = w_0 + \cdots + w_{k-1} x^{k-1}$, so $w \in C_{T,v}$. Hence

$$\langle v, T(v), \ldots, T^{k-1}(v) \rangle \subseteq C_{T,v}.$$

Conversely, suppose that $w \in C_{T,v}$ so

$$w = f(T)(v)$$

and

$$f = qm_{T,v} + r$$

where $r = a_0 + a_1 x + \dots + a_{k-1} x^{k-1}$ and $a_0, \dots, a_{k-1} \in F$. So
 $f(T)(v) = q(T)m_{T,v}(T)(v) + r(T)(v)$

$$= q(T)m_{T,v}(T)(v) + a_0v + a_1T(v) + \dots + a_{k-1}T^{k-1}(v) = a_0v + a_1T(v) + \dots + a_{k-1}T^{k-1}(v) \in \langle v, T(v), \dots, T^{k-1}(v) \rangle.$$

Independence:

Assume

$$a_0v + a_1T(v) + \dots + a_{k-1}T^{k-1}(v) = 0,$$

where $a_0, \ldots, a_{k-1} \in F$; that is, f(T)(v) = 0 where

$$f = a_0 + a_1 x + \dots + a_{k-1} x^{k-1}.$$

Hence $m_{T,v} \mid f$ and since

$$\deg f = k - 1 < k = \deg m_{T,v},$$

we have f = 0 and thus $a_i = 0 \ \forall i$.

2. $[T_{C_{T,v}}]^{\beta}_{\beta} = C(m_{T,v})$:

Let $L = T_{C_{T,v}}$, the restriction of T to $C_{T,v}$.

We want to find $[L]^{\beta}_{\beta}$. So

$$L(v) = T(v) = 0v + 1T(v) + 0T^{2}(v) + \dots + 0T^{k-1}(v)$$

$$L(T(v)) = T^{2}(v) = 0v + 0T(v) + 1T^{2}(v) + \dots + 0T^{k-1}(v)$$

$$\vdots$$

$$L(T^{k-2}(v)) = T^{k-1}(v) = 0v + 0T(v) + 0T^{2}(v) + \dots + 1T^{k-1}(v)$$

Finally, to calculate $L(T^{k-1}(v)) = T^k(v)$, we let

$$m_{T,v} = a_0 + a_1 x + \dots + a_{k-1} x^{k-1} + x^k.$$

Then $m_{T,v}(T)(v) = 0$ and hence

$$L(T^{k-1}(v)) = T^{k}(v) = -a_0v - a_1T(v) - \dots - a_{k-1}T^{k-1}(v).$$

Hence

$$[L]_{\beta}^{\beta} = \begin{bmatrix} 0 & 0 & \dots & 0 & -a_{0} \\ 1 & 0 & & & -a_{1} \\ \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & -a_{k-1} \end{bmatrix} = C(m_{T,v}),$$

as required.

THEOREM 3.3

Suppose that $m_{T,v} = (x-c)^k$. Then the vectors

$$v, (T - cI_V)(v), \dots, (T - cI_V)^{k-1}(v)$$

form a basis β for $W = C_{T,v}$ which we call the elementary Jordan basis. Also

$$[T_W]^\beta_\beta = J_k(c).$$

More generally suppose $m_{T,v} = p^k$, where p is a monic irreducible polynomial in F[x], with $n = \deg p$. Then the vectors

$$v, T(v), \dots, T^{n-1}(v) p(T)(v), Tp(T)(v), \dots, T^{n-1}p(T)(v) \vdots \vdots \vdots \vdots \vdots p^{k-1}(T)(v), Tp^{k-1}(T)(v), \dots, T^{n-1}p^{k-1}(T)(v),$$

form a basis for $W = C_{T,v}$, which reduces to the elementary Jordan basis when p = x - c. Also

$$[T_W]^\beta_\beta = H(p^k),$$

where $H(p^k)$ is a hypercompanion matrix, which reduces to the elementary Jordan matrix $J_k(c)$ when p = x - c:

$$H(p^{k}) = \begin{bmatrix} C(p) & 0 & \cdots & 0 \\ N & C(p) & \cdots & 0 \\ 0 & N & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & \cdots & N & C(p) \end{bmatrix},$$

where there are k blocks on the diagonal and N is a square matrix of same size as C(p) which is everywhere zero, except in the top right-hand corner, where there is a 1. The overall effect is an unbroken subdiagonal of 1's.

3.1.1 A nice proof of the Cayley-Hamilton theorem

(From Insel, Friedberg and Spence.)

Let $f = ch_T$, for some $T: V \mapsto V$. We must show that $f(T) = 0_V$ —i.e. that $f(T)(v) = 0 \ \forall v \in V$.

This is immediate if v = 0, so assume $v \neq 0$ and let $W = C_{T,v}$. Let β' be a basis of W and β be an extension of β' to a basis of V. Then

$$[T]^{\beta}_{\beta} = \left[\begin{array}{c|c} [T_W]^{\beta'}_{\beta'} & B_1 \\ \hline 0 & B_2 \end{array} \right]$$

and $ch_T = ch_{T_W} \cdot ch_{B_2}$. So $ch_{T_W} \mid ch_T$ and since we know that

$$ch_{T_W} = m_{T,v},$$

we have $m_{T,v} \mid \operatorname{ch}_T$.

Hence $ch_T = gm_{T,v}$ and

$$ch_T(T)(v) = (g(T)m_{T,v}(T))(v) = g(T)(m_{T,v}(T)(v)) = g(T)(0) = 0.$$

3.2 An Algorithm for Finding m_T

We use the factorization of ch_T into monic irreducibles.

THEOREM 3.4

Suppose $T: V \mapsto V$,

$$m_T = p_1^{b_1} \dots p_t^{b_t}$$

where $b_1, \ldots, b_t \ge 1$, and p_1, \ldots, p_t are distinct monic irreducibles. Then for $i = 1, \ldots, t$ we have

(a)

$$V \supset \operatorname{Im} p_i(T) \supset \cdots \supset \operatorname{Im} p_i^{b_i-1}(T) \supset \operatorname{Im} p_i^{b_i}(T) = \cdots$$

(b)

$$\{0\} \subset \operatorname{Ker} p_i(T) \subset \cdots \subset \operatorname{Ker} p_i^{b_i-1}(T) \subset \operatorname{Ker} p_i^{b_i}(T) = \cdots$$

Note: In terms of nullities, conclusion (b) says that

$$0 < \nu(p_i(T)) < \dots < \nu(p_i^{b_i-1}(T)) < \nu(p_i^{b_i}(T)) = \dots$$

so this gives us a method of calculating b_i .

Presently we'll show that if $ch_T = p_1^{a_1} \dots p_t^{a_t}$, then

nullity
$$(p_i^{b_i}(T)) = a_i \deg p_i.$$

Hence b_i is also characterised as the smallest integer h such that

nullity
$$(p_i^h(T)) = a_i \deg p_i$$
.

Also note that (a) and (b) are equivalent, and it is the latter that we prove.

A notational simplification—the left F[x]-module notation. If $f \in F[x]$ and $v \in V$, we define

$$fv = f(T)(v).$$

It is easy to verify that

1.
$$(f+g)v = fv + gv$$
 $\forall f,g \in F[x], v \in V;$ 2. $f(v+w) = fv + fw$ $\forall f \in F[x], v, w \in V;$ 3. $(fg)v = f(gv)$ $\forall f,g \in F[x], v \in V;$ 4.

$$1v = v \qquad \forall v \in V.$$

These axioms, together with the four axioms for addition on V, turn V into what is called a "left F[x]-module". (So there are deeper considerations lurking in the background—ideas of greater generality which make the algorithm we unravel for the rational canonical form also apply to other things such as the theorem that any finite abelian group is a direct product of cyclic prime power subgroups.)

(i) We first prove that $\{0\} \subset \operatorname{Ker} p_i(T)$. We write $p = p_i$, $b = b_i$ for brevity; no confusion should arise since *i* is fixed.

PROOF. $m_T = pf$, $f \in F[x]$ and $f(T) \neq 0_V$. Hence $\exists v \in V$ such that $fv \neq 0$. Then

$$p(fv) = (pf)v = m_T v = 0,$$

so $fv \in \operatorname{Ker} p(T)$.

(ii) We next prove that

$$\operatorname{Ker} p^b(T) = \operatorname{Ker} p^{b+1}(T).$$

The containment

$$\operatorname{Ker} p^b(T) \subseteq \operatorname{Ker} p^{b+1}(T)$$

is obvious, so we need only show that

$$\operatorname{Ker} p^b(T) \supseteq \operatorname{Ker} p^{b+1}(T).$$

Let $w \in \text{Ker } p^{b+1}(T)$, i.e. $p^{b+1}w = 0$. Now if $m_T = p^b q$, then $gcd(p^b, q) = 1$. So $\exists u, v \in F[x]$ such that $1 = up^b + vq$. Hence

$$p^b = up^{2b} + vm_T.$$

Hence

$$p^{b}(T) = (up^{2b})(T) + v(T)m_{T}(T)$$

= $(up^{2b})(T)$

and thus

$$p^{b}w = (p^{b-1})(p^{b+1}w) = p^{b-1}0 = 0$$

and $w \in \operatorname{Ker} p^b$, as required.

(iii)

$$\operatorname{Ker} p^{h}(T) = \operatorname{Ker} p^{h+1}(T) \Rightarrow \operatorname{Ker} p^{h+1}(T) = \operatorname{Ker} p^{h+2}(T),$$

i.e.
$$\operatorname{Ker} p^{h}(T) \supseteq \operatorname{Ker} p^{h+1}(T) \Rightarrow \operatorname{Ker} p^{h+1}(T) \supseteq \operatorname{Ker} p^{h+2}(T).$$

PROOF. Suppose that $\operatorname{Ker} p^h(T) \supseteq \operatorname{Ker} p^{h+1}(T)$. Then

$$\begin{aligned} v \in \operatorname{Ker} p^{h+2}(T) &\Rightarrow p^{h+2}v = 0 \\ &\Rightarrow p^{h+1}(pv) = 0 \Rightarrow pv \in \operatorname{Ker} p^{h+1}(T) \\ &\Rightarrow pv \in \operatorname{Ker} p^{h}(T) \Rightarrow p^{h}(pv) = 0 \\ &\Rightarrow p^{h+1}v = 0 \Rightarrow v \in \operatorname{Ker} p^{h+1}(T). \end{aligned}$$

So it follows by induction from (ii) that

$$\operatorname{Ker} p_i^{b_i}(T) = \operatorname{Ker} p_i^{b_{i+1}}(T) = \cdots.$$

(iv)

$$\operatorname{Ker} p^{b-1}(T) \subset \operatorname{Ker} p^b(T)$$

and this forces a chain of proper inclusions:

$$\{0\} \subset \operatorname{Ker} p(T) \subset \cdots \subset \operatorname{Ker} p^{b-1}(T) \subset \operatorname{Ker} p^b(T) = \cdots$$

which is the desired result. For

 $p^{b-1}q(T) \neq 0_V$, so $\exists v$ such that $p^{b-1}qv \neq 0$. Then

$$qv \notin \operatorname{Ker} p^{b-1}(T),$$

but $qv \in \operatorname{Ker} p^b(T)$ as

$$p^b q v = m_T v = 0.$$

3.3 Primary Decomposition Theorem

THEOREM 3.5 (Primary Decomposition)

If $T: V \mapsto V$ is a LT with $m_T = p_1^{b_1} \dots p_t^{b_t}$, where p_1, \dots, p_t are monic irreducibles, then

$$V = \operatorname{Ker} p_1^{b_1}(T) \oplus \cdots \oplus \operatorname{Ker} p_t^{b_t}(T),$$

a direct sum of T-invariant subspaces. Moreover for $1 \le i \le t$,

$$\nu(p_i(T)^{b_i}) = a_i \deg p_i,$$

where $ch_T = p_1^{a_1} \cdots p_t^{a_t}$.

REMARK. The same proof gives a slightly more general result: If $p = p_1^{b_1} \cdots p_t^{b_t}$, then

$$\operatorname{Ker} p(T) = \operatorname{Ker} p_1^{b_1}(T) \oplus \cdots \operatorname{Ker} p_t^{b_t}(T).$$

We subsequently give an application of the decomposition theorem in this form to the solution of the *n*-th order linear differential equations with constant coefficients. (See Hoffman and Kunze, pages 184–185.) PROOF. Let $m_T = p_i^{b_i} q_i \ \forall i = 1, \dots, t$. Then

$$(q_i q_j)(T) = 0_V$$
 if $i \neq j$ as $m_T \mid q_i q_j$ if $i \neq j$.

Now $gcd(q_1, \ldots, q_t) = 1$, so $\exists f_1, \ldots, f_t \in F[x]$ such that

$$1 = f_1 q_1 + \dots + f_t q_t$$

and with $T_i = (f_i q_i)(T)$ we have

$$I_V = T_1 + \dots + T_t. \tag{6}$$

Also

$$T_i T_j = (f_i q_i)(T)(f_j q_j)(T)$$

= $(f_i f_j)(T)(q_i q_j)(T)$
= 0_V if $i \neq j$,

Then

$$V = \bigoplus_{i=1}^{t} \operatorname{Im} T_{i}.$$

For $T_i^2 = T_i(T_1 + \dots + T_t) = T_i I_V = T_i$. Next, $V = \operatorname{Im} T_1 + \dots + \operatorname{Im} T_t$. For $v \in V \Rightarrow v = I_V(v) = T_1(v) + \dots + T_t(v) \in \operatorname{Im} T_1 + \dots + \operatorname{Im} T_t$.

Next assume $v_1 + \cdots + v_t = 0$, $v_i \in \operatorname{Im} T_i$, $1 \le i \le t$. Then $v_i = T_i(u_i)$ and

$$T_1(u_1) + \dots + T_t(u_t) = 0$$

$$T_i(T_1(u_1) + \dots + T_t(u_t)) = T(0) = 0$$

$$T_iT_i(u_i) = 0$$

$$v_i = T_i(u_i) = 0.$$

We now show that

$$\operatorname{Im} T_i = \operatorname{Ker} p_i^{b_i}(T).$$

" \subseteq " Let $v \in \operatorname{Im} T_i$. Then

$$v = f_i q_i w$$

$$\Rightarrow p_i^{b_i} v = p_i^{b_i} f_i q_i w$$

$$= f_i (p_i^{b_i} q_i) w$$

$$= 0.$$

"⊇" Suppose $p_i^{b_i}v = 0$.

Now if $j \neq i$, we have $p_i^{b_i} \mid f_j q_j$, so

$$T_j(v) = f_j q_j v = 0.$$

 So

$$v = I_V(v) = T_1(v) + \dots + T_t(v) = T_i(v)$$

$$\in \operatorname{Im} T_i,$$

as required.

Finally, let $V_i = \operatorname{Ker} p_i^{b_i}(T)$ and $L_i = T_{V_i}$. Then because V_1, \ldots, V_t are *T*-invariant subspaces of *V*, we have

$$ch_T = ch_{L_1} \cdots ch_{L_t}.$$

Now $p_i^{b_i}(T)(v) = 0$ if $v \in V_i$, so $p_i^{b_i}(L_i) = 0_{V_i}$. Hence m_{L_i} has the form $m_{L_i} = p_i^{e_i}$. Hence ch_{L_i} has the form $ch_{L_i} = p_i^{d_i}$. Hence

$$ch_T = p_1^{a_1} \cdots p_t^{a_t} = p_1^{d_1} \cdots p_t^{d_t}$$

and consequently $d_i = a_i$.

Finally,

$$\lim V_i = \deg ch_{L_i} = \deg p_i^{a_i} = a_i \deg p_i.$$

(Incidentally, we mention that $m_T = \operatorname{lcm}(m_{L_1}, \ldots, m_{L_t})$). Hence

$$m_T = p_1^{b_1} \cdots p_t^{b_t} = p_1^{e_1} \cdots p_t^{e_t}$$

and consequently $e_i = b_i$. Hence $m_{L_i} = p_i^{b_i}$.)

0

THEOREM 3.6 (Commuting diagonable linear transformations)

If $T_1, \ldots, T_n : V \to V$ are commuting diagonable linear transformations, then there exists a basis β for V such that each of $[T_1]^{\beta}_{\beta}, \ldots, [T_m]^{\beta}_{\beta}$ are each diagonal.

(Matrix version) If A_1, \ldots, A_m are commuting diagonable matrices of the same size, then there exists a non-singular matrix P such that the matrices

$$P^{-1}A_1P,\ldots P^{-1}A_mP$$

are simultaneously diagonal.

PROOF. (From Samelson page 158). We prove the result when m = 2, the general case follows by an easy iteration. Suppose T_1 and T_2 are commuting diagonable linear transformations on V. Because m_{T_1} splits as a product of distinct linear factors, the primary decomposition theorem gives a direct sum decomposition as a sum of the T_1 -eigenspaces:

$$V = U_1 \oplus \cdots \oplus U_t.$$

It turns out that not only are the subspaces $U_i T_1$ -invariant, they are T_2 -invariant. For if $U_i = \text{Ker}(T_1 - cI_V)$, then

$$v \in U_i \implies T_1(v) = cv$$

$$\implies T_2(T_1(v)) = cT_2(v)$$

$$\implies T_1(T_2(v)) = cT_2(v)$$

$$\implies T_2(v) \in U_i.$$

Now because T_2 is diagonable, V has a basis consisting of T_2 -eigenvectors and it is an easy exercise to show that in a direct sum of T_2 -invariant subspaces, each non-zero "component" of a T_2 -eigenvector is itself a T_2 eigenvector; moreover each non-zero component is a T_1 - eigenvector. Hence V is spanned by a family of vectors which are simultaneously T_1 -eigenvectors and T_2 -eigenvectors. If β is a subfamily which forms a basis for V, then $[T_1]^{\beta}_{\beta}$ and $[T_2]^{\beta}_{\beta}$ are diagonal.

THEOREM 3.7 (Fitting's lemma)

Suppose $T: V \to V$ is a linear transformation over T and

 $\operatorname{Ker} \subset \operatorname{Ker} T^2 \subset \cdots \operatorname{Ker} T^n = \operatorname{Ker} T^{n+1} = \cdots$

Then $V = \operatorname{Im} T^n \oplus \operatorname{Ker} T^n$.

COROLLARY 3.1

If $T: V \to V$ is an indecomposable linear transformation (that is the only *T*-invariant subspaces of *V* are $\{0\}$ and *V*), then *T* is either nilpotent (that is $T^n = 0_V$ for some $n \ge 1$) or *T* is an isomorphism.

4 The Jordan Canonical Form

The following subspaces are central for our treatment of the Jordan and rational canonical forms of a linear transformation $T: V \to V$.

DEFINITION 4.1

With $m_T = p_1^{b_1} \dots p_t^{b_t}$ as before and $p = p_i$, $b = b_i$ for brevity, we define

$$N_{h,p} = \operatorname{Im} p^{h-1}(T) \cap \operatorname{Ker} p(T)$$

REMARK. In numerical examples, we will need to find a spanning family for $N_{h,p}$. This is provided by Problem Sheet 1, Question 11(a): we saw that if $T : U \to V$ and $S : V \to W$ are linear transformations, then If $\operatorname{Ker} p^h(T) = \langle u_1, \ldots, u_n \rangle$, then

$$N_{h,p} = \langle p^{h-1}u_1, \dots, p^{h-1}u_n \rangle,$$

where we have taken U = V = W and replaced S and T by p(T) and $p^{h-1}(T)$ respectively, so that $ST = p^h(T)$. Also

$$\dim(\operatorname{Im} T \cap \operatorname{Ker} S) = \nu(ST) - \nu(T).$$

Hence

$$\nu_{h,p} = \dim N_{h,p}$$

= dim(Im $p^{h-1}(T) \cap \operatorname{Ker} p(T)$)
= $\nu(p^h(T)) - \nu(p^{h-1}(T)).$

THEOREM 4.1

$$N_{1,p} \supseteq N_{2,p} \supseteq \cdots \supseteq N_{b,p} \neq \{0\} = N_{b+1,p} = \cdots$$

PROOF. Successive containment follows from

$$\operatorname{Im} L^{h-1} \supseteq \operatorname{Im} L^h$$

with L = p(T).

The fact that $N_{b,p} \neq \{0\}$ and that $N_{b+1,p} = \{0\}$ follows directly from the formula

dim
$$N_{h,p} = \nu(p^h(T)) - \nu(p^{h-1}(T)).$$

For simplicity, assume that p is linear, that is that p = x - c. The general story (when deg p > 1) is similar, but more complicated; it is delayed until the next section.

Telescopic cancellation then gives

THEOREM 4.2

$$\nu_{1,p} + \nu_{2,p} + \dots + \nu_{b,p} = \nu(p^b(T)) = a,$$

where p^a is the exact power of p dividing ch_T .

Consequently we have the decreasing sequence

$$\nu_{1,p} \ge \nu_{2,p} \ge \cdots \ge \nu_{b,p} \ge 1.$$

EXAMPLE 4.1

Suppose $T: V \mapsto V$ is a LT such that $p^4 || m_T, p = x - c$ and

$$\begin{split} \nu(p(T)) &= 3, \qquad \nu(p^2(T)) = 6, \\ \nu(p^3(T)) &= 8, \qquad \nu(p^4(T)) = 10. \end{split}$$

So

$$\operatorname{Ker} p(T) \subset \operatorname{Ker} p^{2}(T) \subset \operatorname{Ker} p^{3}(T) \subset \operatorname{Ker} p^{4}(T) = \operatorname{Ker} p^{5}(T) = \cdots$$

Then

$$\nu_{1,p} = 3, \qquad
\nu_{2,p} = 6 - 3 = 3,
\nu_{3,p} = 8 - 6 = 2, \quad
\nu_{4,p} = 10 - 8 = 2$$

SO

$$N_{1,p} = N_{2,p} \supset N_{3,p} = N_{4,p} \neq \{0\}.$$

4.1 The Matthews' dot diagram

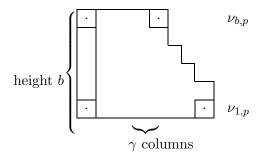
We would represent the previous example as follows:

•	•		$\nu_{4,p}$	Dots represent dimension:
•	·		$\nu_{3,p}$	
•	•	•	$\nu_{2,p}$	3 + 3 + 2 + 2 = 10
•	•	•	$\nu_{1,p}$	$\Rightarrow 10 = 4 + 4 + 2$

The conjugate partition of 10 is 4+4+2 (sum of column heights of diagram), and this will soon tell us that there is a corresponding contribution to the **Jordan canonical form** of this transformation, namely

$$J_4(c) \oplus J_4(c) \oplus J_2(c).$$

In general,



and we label the conjugate partition by

 $e_1 \ge e_2 \ge \cdots \ge e_{\gamma}.$

Finally, note that the total number of dots in the dot diagram is $\nu(p^b(T))$, by Theorem 4.2.

THEOREM 4.3

 $\exists v_1, \ldots, v_\gamma \in V$ such that

$$p^{e_1-1}v_1, p^{e_2-1}v_2, \dots, p^{e_{\gamma}-1}v_{\gamma}$$

form a basis for $\operatorname{Ker} p(T)$.

PROOF. Special case, but the construction is quite general.



choose a basis p^3v_1, p^3v_2 for $N_{4,p}$ $\begin{array}{c|c} \cdot & \cdot \\ \hline \end{array}$ extend to a basis p^3v_1, p^3v_2, pv_3 for $N_{2,p}$

Then p^3v_1 , p^3v_2 , pv_3 is a basis for $N_{1,p} = \operatorname{Ker} p(T)$.

THEOREM 4.4 (Secondary decomposition)

$$m_{T,v_i} = p^{e_i}$$

(ii)

$$\operatorname{Ker} p^{b}(T) = C_{T,v_{1}} \oplus \cdots \oplus C_{T,v_{\gamma}}$$

PROOF.

(i) We have $p^{e_i-1}v_i \in \operatorname{Ker} p(T)$, so $p^{e_i}v_i = 0$ and hence $m_{T,v_i} \mid p^{e_i}$. Hence $m_{T,v_i} = p^f$, where $0 \leq f \leq e_i$. But $r^{e_i-1}v_i \neq 0$ as it is part of a basis. Hence f > a and f

But $p^{e_i-1}v_i \neq 0$, as it is part of a basis. Hence $f \geq e_i$ and $f = e_i$ as required.

$$C_{T,v_i} \subseteq \operatorname{Ker} p^b(T).$$

For $p^{e_i}v_i = 0$ and so $p^{e_i}(fv_i) = 0$ $\forall f \in F[x]$. Hence as $e_i \leq b$, we have

$$p^{b}(fv_{i}) = p^{b-e_{i}}(p^{e_{i}}fv_{i}) = p^{b-e_{i}}0 = 0$$

and $fv_i \in \operatorname{Ker} p^b(T)$. Consequently $C_{T,v_i} \subseteq \operatorname{Ker} p^b(T)$ and hence

$$C_{T,v_1} + \dots + C_{T,v_{\gamma}} \subseteq \operatorname{Ker} p^b(T).$$

(b) We presently show that the subspaces C_{T,v_j} , $j = 1, \ldots, \gamma$ are independent, so

$$\dim(C_{T,v_1} + \dots + C_{T,v_{\gamma}}) = \sum_{j=1}^{\gamma} \dim C_{T,v_j}$$
$$= \sum_{j=1}^{\gamma} \deg m_{T,v_j} = \sum_{j=1}^{\gamma} e_j$$
$$= \nu(p^b(T))$$
$$= \dim \operatorname{Ker} p^b(T).$$

Hence

$$\operatorname{Ker} p^{b}(T) = C_{T,v_{1}} + \dots + C_{T,v_{\gamma}}$$
$$= C_{T,v_{1}} \oplus \dots \oplus C_{T,v_{\gamma}}.$$

The independence of the C_{T,v_i} is stated as a lemma: **Lemma**: Let $v_1, \ldots, v_{\gamma} \in V$, $e_1 \geq \cdots \geq e_{\gamma} \geq 1$; $m_{T,v_j} = p^{e_j} \quad 1 \leq j \leq \gamma$; p = x - c; Also $p^{e_1-1}v_1, \ldots, p^{e_{\gamma}-1}v_{\gamma}$ are LI. Then

$$f_1 v_1 + \dots + f_{\gamma} v_{\gamma} = 0 ; \quad f_1, \dots, f_{\gamma} \in F[x]$$

$$\Rightarrow \quad p^{e_j} \mid f_j \quad 1 \le j \le \gamma.$$

PROOF: (induction on e_1)

Firstly, consider $e_1 = 1$. Then

$$e_1 = e_2 = \dots = e_\gamma = 1.$$

Now $m_{T,v_j} = p^{e_j}$ and

$$p^{e_1-1}v_1,\ldots,p^{e_\gamma-1}v_\gamma$$

are LI, so v_1, \ldots, v_{γ} are LI. So assume

$$f_1v_1 + \dots + f_{\gamma}v_{\gamma} = 0 \qquad f_1, \dots, f_{\gamma} \in F[x].$$
(7)

and by the remainder theorem

$$f_j = (x - c)q_j + f_j(c).$$
 (8)

Thus

$$f_j v_j = q_j (x - c) v_j + f_j (c) v_j$$

= $f_j (c) v_j$.

So (7) implies

$$f_1(c)v_1 + \dots + f_{\gamma}(c)v_{\gamma}(c) = 0$$

$$\Rightarrow f_j(c) = 0 \qquad \forall j = 1, \dots, \gamma$$

and (8) implies

$$(x-c) \mid f_j \qquad \forall j$$

which is the result.

Now let $e_1 > 1$ and assume the lemma is true for $e_1 - 1$. If

$$m_{T,v_j} = p^{e_j};$$

$$p^{e_1-1}v_1, \dots, p^{e_\gamma-1}v_\gamma \text{ are LI},$$
and
$$f_1v_1 + \dots + f_\gamma v_\gamma = 0$$
(9)

as before, we have

$$f_1(pv_1) + \dots + f_\gamma(pv_\gamma) = 0 \tag{10}$$

where $m_{T,pv_j} = p^{e_j - 1}$.

Now let δ be the greatest positive integer such that $e_{\delta} > 1$; i.e. $e_{\delta+1} = 1$, but $e_{\delta} > 1$. Applying the induction hypothesis to (10), in the form

$$f_1(pv_1) + \dots + f_{\delta}(pv_{\delta}) = 0$$

we obtain

$$p^{e_j-1} \mid f_j \qquad \forall j = 1, \dots, \delta,$$

so we may write

$$f_j = p^{e_j - 1} g_j,$$
(where if $g_j = f_j$ if $j > \delta$). Now substituting in (9),
 $g_1 p^{e_1 - 1} v_1 + \dots + g_\gamma p^{e_\gamma - 1} v_\gamma = 0.$ (11)

 But

$$m_{T,p^{e_j-1}v_j} = p$$

so (11) and the case $e_1 = 1$ give

$$p \mid g_j \qquad \forall j,$$

as required.

A summary:

If $m_T = (x - c_1)^{b_1} \dots (x - c_t)^{b_t} = p_1^{b_1} \dots p_t^{b_t}$, then there exist vectors v_{ij} and positive integers e_{ij} $(1 \le i \le t, 1 \le j \le \gamma_i)$, where $\gamma_i = \nu(T - c_i I_V)$, satisfying

$$b_i = e_{i1} \ge \dots \ge e_{i\gamma_i}, \quad m_{T,v_{ij}} = p_i^{e_{ij}}$$

and

$$V = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C_{T, v_{ij}}$$

We choose the elementary Jordan bases

$$\beta_{ij}: v_{ij}, (T - c_i I_V)(v_{ij}), \dots, (T - c_i I_V)^{e_{ij}-1}(v_{ij})$$

for $C_{T,v_{ij}}$. Then if

$$\beta = \bigcup_{i=1}^{t} \bigcup_{j=1}^{\gamma_i} \beta_{ij},$$

 β is a basis for V and we have

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i) = J.$$

A direct sum of elementary Jordan matrices such as J is called a Jordan canonical form of T.

If $T = T_A$ and $P = [v_{11}| \dots |v_{t\gamma_t}]$, then

$$P^{-1}AP = J$$

and J is called a Jordan canonical form of A.

4.2 Two Jordan Canonical Form Examples

4.2.1 Example (a):

Let
$$A = \begin{bmatrix} 4 & 0 & 1 & 0 \\ 2 & 2 & 3 & 0 \\ -1 & 0 & 2 & 0 \\ 4 & 0 & 1 & 2 \end{bmatrix} \in M_{4 \times 4}(\mathbb{Q}).$$

We find $ch_T = (x-2)^2(x-3)^2 = p_1^2 p_2^2$, where $p_1 = x - 2$, $p_2 = x - 3$.
CASE 1, $p_1 = x - 2$:

$$p_1(A) = A - 2I_4 = \begin{bmatrix} 2 & 0 & 1 & 0 \\ 2 & 0 & 3 & 0 \\ -1 & 0 & 0 & 0 \\ 4 & 0 & 1 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix},$$

so $\nu(p_1(A)) = \gamma_1 = 2$. Hence $b_1 = 1$ and the corresponding dot diagram has height 1, width 2, with associated Jordan blocks $J_1(2) \oplus J_1(2)$:

$$\begin{array}{c|c} \hline & & \\ \hline & & \\ \hline & \\ \end{array} \end{array} \overset{$ \hline $ \ \ $ \ \ $ \ \ $ \ \ $ \ \ $ \ \ $ \ \ $ \$$

Note that $C_{T_A, v_{11}}$ and $C_{T_A, v_{12}}$ have Jordan bases $\beta_{11} : v_{11}$ and $\beta_{12} : v_{12}$ respectively.

CASE 2,
$$p_2 = x - 3$$
:

$$p_2(A) = A - 3I_4 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 2 & -1 & 3 & 0 \\ -1 & 0 & -1 & 0 \\ 4 & 0 & 1 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 & -\frac{1}{3} \\ 0 & 1 & 0 & \frac{1}{3} \\ 0 & 0 & 1 & \frac{1}{3} \\ 0 & 0 & 0 & 0 \end{bmatrix},$$

so $\nu(p_2(A)) = 1 = \gamma_2$; also $\nu(p_2^2(A)) = 2$. Hence $b_2 = 2$ and we get a corresponding dot diagram consisting of two vertical dots, with associated Jordan block $J_2(3)$:

$$\begin{array}{c} \cdot \\ \cdot \\ \cdot \\ \end{array} \begin{array}{c} N_{2,x-3} \\ N_{1,x-3} \end{array}$$

We have to find a basis of the form $p_2(T_A)(v_{21}) = (A-3I_4)v_{21}$ for Ker $p_2(T_A) = N(A-3I_4)$.

To find v_{21} we first get a basis for $N(A - 3I_4)^2$. We have

$$p_2^2(A) = (A - 3I_4)^2 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ -3 & 1 & -4 & 0 \\ 0 & 0 & 0 & 0 \\ -1 & 0 & 2 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -2 & -1 \\ 0 & 1 & -10 & -3 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

and we find $X_1 = \begin{bmatrix} 2\\10\\1\\0 \end{bmatrix}$ and $X_2 = \begin{bmatrix} 1\\3\\0\\1 \end{bmatrix}$ is such a basis. Then we have

$$N_{2,p_2} = \langle p_2 X_1, p_2 X_2 \rangle$$

= $\langle p_2(A) X_1, p_2(A) X_2 \rangle = \langle (A - 3I_4) X_1, (A - 3I_4) X_2 \rangle$
= $\left\langle \begin{bmatrix} 3 \\ -3 \\ -3 \\ 9 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ -1 \\ 3 \end{bmatrix} \right\rangle = \left\langle \begin{bmatrix} 3 \\ -3 \\ -3 \\ 9 \end{bmatrix} \right\rangle.$

Hence we can take $v_{21} = X_1$. Then $m_{T_A, v_{21}} = (x-3)^2$. Also

Ker
$$p_2^{b_2}(T_A)$$
) = $N(p_2^2(A)) = N(A - 3I_4)^2 = C_{T_A, v_{21}}$.

Moreover $C_{T_A, v_{21}}$ has Jordan basis $\beta_{21} : v_{21}, (A - 3I_4)v_{21}$.

Finally we have $V_4(\mathbb{Q}) = C_{T_A, v_{11}} \oplus C_{T_A, v_{12}} \oplus C_{T_A, v_{21}}$ and $\beta = \beta_{11} \cup \beta_{12} \cup \beta_{21}$ is a basis for $V_4(\mathbb{Q})$. Then with

$$P = [v_{11}|v_{12}|v_{21}|(A-3I_4)v_{21}] = \begin{bmatrix} 0 & 0 & 2 & 3\\ 1 & 0 & 10 & -3\\ 0 & 0 & 1 & -3\\ 0 & 1 & 0 & 9 \end{bmatrix}$$

we have

$$P^{-1}AP = [T_A]^{\beta}_{\beta} = J_1(2) \oplus J_1(2) \oplus J_2(3) = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 1 & 3 \end{bmatrix}$$

4.2.2 Example (b):

Let $A \in M_{6 \times 6}(F)$ have the property that $ch_A = x^6$, $m_A = x^3$ and

$$\nu(A) = 3, \ \nu(A^2) = 5, \ (\nu(A^3) = 6).$$

Next, with $\nu_{h,x} = \dim_F N_{h,x}$ we have

$$\nu_{1,x} = \nu(A) = 3 = \gamma_1;$$

$$\nu_{2,x} = \nu(A^2) - \nu(A) = 5 - 3 = 2;$$

$$\nu_{3,x} = \nu(A^3) - \nu(A^2) = 6 - 5 = 1.$$

Hence the dot diagram corresponding to the (only) monic irreducible factor x of m_A is

$$\begin{array}{c|c} \cdot & N_{3,x} \\ \hline \cdot & \cdot & N_{2,x} \\ \hline \cdot & \cdot & \cdot & N_{1,x} \end{array}$$

Hence we read off that \exists a non-singular $P \in M_{6\times 6}(F)$ such that $P^{-1}AP = J_3(0) \oplus J_2(0) \oplus J_1(0)$. To find such a matrix P we proceed as follows:

(i) First find a basis for $N_{3,x}$. We do this by first finding a basis for $N(A^3)$: $X_1, X_2, X_3, X_4, X_5, X_6$. Then

$$N_{3,x} = \langle A^2 X_1, A^2 X_2, A^2 X_3, A^2 X_4, A^2 X_5, A^2 X_6 \rangle.$$

We now apply the LRA (left-to-right algorithm) to the above spanning family to get a basis A^2v_{11} for $N_{3,x}$, where A^2v_{11} is the first non-zero vector in the spanning family.

(ii) Now extend the linearly independent family A^2v_{11} to a basis for $N_{2,x}$. We do this by first finding a basis Y_1 , Y_2 , Y_3 , Y_4 , Y_5 for $N(A^2)$. Then

$$N_{2,x} = \langle AY_1, AY_2, AY_3, AY_4, AY_5 \rangle.$$

We now attach $A^2 v_{11}$ to the head of this spanning family:

$$N_{2,x} = \langle A^2 v_{11}, AY_1, AY_2, AY_3, AY_4, AY_5 \rangle$$

and apply the LRA to find a basis for $N_{2,x}$ which includes A^2X_1 . This will have the form A^2v_{11} , Av_{12} , where Av_{12} is the first vector in the list AY_1, \ldots, AY_5 which is not a linear combination of A^2v_{11} .

(iii) Now extend the linearly independent family A^2v_{11} , Av_{12} to a basis for $N_{1,x} = N(A)$. We do this by first finding a basis Z_1, Z_2, Z_3 for N(A). Then place the linearly independent family A^2v_{11} , Av_{12} at the head of this spanning family:

$$N_{1,x} = \langle A^2 v_{11}, A v_{12}, Z_1, Z_2, Z_3 \rangle.$$

The LRA is then applies to the above spanning family selects a basis of the form A^2v_{11} , Av_{12} , v_{13} , where v_{13} is the first vector among Z_1 , Z_2 , Z_3 which is not a linear combination of A^2v_{11} and Av_{12} . Then $m_{T_A,v_{11}} = x^3$, $m_{T_A,v_{12}} = x^2$, $m_{T_A,v_{13}} = x$. Also

$$\operatorname{Ker} p_1^{b_1}(T_A) = N(A^3) = C_{T_A, v_{11}} \oplus C_{T_A, v_{12}} \oplus C_{T_A, v_{13}}.$$

Finally, if we take Jordan bases

$$\begin{array}{rcl} \beta_{11} & : & v_{11}, \, Av_{11}, \, A^2v_{11}; \\ \beta_{12} & : & v_{12}, \, Av_{12}; \\ \beta_{13} & : & v_{13} \end{array}$$

for the three T–cyclic subspaces $C_{T_A, v_{11}}, C_{T_A, v_{12}}, C_{T_A, v_{13}}$, respectively, we then get the basis

$$\begin{array}{rcl} \beta & = & \beta_{11} \cup \beta_{12} \cup \beta_{13} \\ & = & v_{11}, \, Av_{11}, \, A^2 v_{11}; \, v_{12}, \, Av_{12}; \, v_{13} \end{array}$$

for $V_6(F)$. Then if

$$P = [v_{11}|Av_{11}|A^2v_{11}|v_{12}|Av_{12}|v_{13}]$$

we have

4.3 Uniqueness of the Jordan form

Let β be a basis for V for which $[T]_{\beta}^{\beta}$ is in Jordan canonical form

$$J = J_{e_1}(\lambda_1) \oplus \cdots J_{e_s}(\lambda_s).$$

If we change the order of the basis vectors in β , we produce a corresponding change in the order of the elementary Jordan matrices. It is customary to assume our Jordan forms arranged so as to group together into a block those elementary Jordan matrices having the same eigenvalue c_i :

$$J=J_1\oplus\cdots\oplus J_t,$$

where

$$J_i = \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i).$$

Moreover within this *i*-th block J_i , we assume the sizes $e_{i1}, \ldots, e_{i\gamma_i}$ of the elementary Jordan matrices decrease monotonically:

$$e_{i1} \geq \ldots \geq e_{i\gamma_i}.$$

We prove that with this convention, the above sequence is uniquely determined by T and the eigenvalue c_i .

We next observe that

$$ch_T = ch_J = \prod_{i=1}^t ch_{J_i} = \prod_{i=1}^t \prod_{j=1}^{\gamma_i} (x - c_i)^{e_{ij}} = \prod_{i=1}^t (x - c_i)^{e_{i1} + \dots + e_{i\gamma_i}}.$$

Hence c_1, \ldots, c_t are determined as the distinct eigenvalues of T.

DEFINITION 4.2

The numbers $e_{i1}, \ldots, e_{i\gamma_i}, 1 \leq i \leq t$, are called the Segre characteristic of T, while the numbers $\nu_{1,x-c_i}, \ldots, \nu_{b_i,x-c_i}, 1 \leq i \leq t$ are called the Weyr characteristic of T.

The polynomials $(x - c_i)^{e_{ij}}$ are called the elementary divisors of T.

LEMMA 4.1

Let

$$A = J_e(0) = \begin{bmatrix} 0 & 0 & & 0 \\ 1 & 0 & \cdots & & \\ 0 & 1 & & & \\ & \vdots & \ddots & & \vdots \\ 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & & 1 & 0 \end{bmatrix}.$$

Then

$$\nu(A^h) = \begin{cases} h & \text{if } 1 \le h \le e-1, \\ e & \text{if } e \le h. \end{cases}$$

Proof. A^h has 1 on the *h*-th sub-diagonal, 0 elsewhere, if $1 \le h \le e - 1$, whereas $A^h = 0$ if $h \ge e$.

Consequently

$$\nu(A^h) - \nu(A^{h-1}) = \begin{cases} 1 & \text{if } 1 \le h \le e, \\ 0 & \text{if } e < h. \end{cases}$$

We now can prove that the sequence $e_{i1} \ge \ldots e_{i\gamma_i}$ is determined uniquely by T and the eigenvalue c_i .

Let $p_k = x - c_k$ and

$$A = [T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i).$$

Then

$$\nu(p_k^h(T)) = \nu(p_k^h(A))$$

$$= \nu(\bigoplus_{i=1}^t \bigoplus_{j=1}^{\gamma_i} p_k^h(J_{e_{ij}}(c_i)))$$

$$= \nu(\bigoplus_{i=1}^t \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}^h(c_i - c_k))$$

$$= \sum_{i=1}^t \sum_{j=1}^{\gamma_i} \nu(J_{e_{ij}}^h(c_i - c_k)),$$

where we have used the fact that

$$p_k(J_{e_{ij}}(c_i)) = J_{e_{ij}}(c_i) - c_k I_n = J_{e_{ij}}(c_i - c_k).$$

However $J_{e_{ij}}(c_i - c_k)$ is a non-singular matrix if $i \neq k$, so

$$\nu(J_{e_{ij}}^h(c_i - c_k)) = 0$$

if $i \neq k$. Hence

$$\nu(p_k^h(T)) = \sum_{j=1}^{\gamma_k} \nu(J_{e_{kj}}^h(0)).$$

Hence

$$\nu_{h,x-c_k} = \nu(p_k^h(T)) - \nu(p_k^{h-1}(T)) = \sum_{j=1}^{\gamma_k} \left(\nu(J_{e_{kj}}^h(0)) - \nu(J_{e_{kj}}^{h-1}(0)) \right)$$
$$= \sum_{\substack{j=1\\h \le e_{kj}}}^{\gamma_k} 1.$$

Consequently $\nu_{h,x-c_k} - \nu_{h+1,x-c_k}$ is the number of e_{kj} which are equal to h. Hence by taking $h = 1, \ldots$, we see that the sequence $e_{k1}, \ldots, e_{k\gamma_k}$ is determined by T and c_k and is in fact the contribution of the eigenvalue c_k to the Segre characteristic of T.

REMARK. If A and B are similar matrices over F, then $B = P^{-1}AP$ say. Also A and B have the same characteristic polynomials. Then if c_k is an eigenvalue of A and B and $p_k = x - c_k$, we have

$$p_k^h(T_B) = p_k^h(B) = P^{-1}p_k^h(A)P = P^{-1}p_k^h(T_A)P$$

and hence

$$\nu(p_k^h(T_B)) = \nu(p_k^h(T_A))$$

for all $h \ge 1$.

Consequently the Weyr characteristics of T_A and T_B will be identical. Hence the corresponding dot diagrams and so the Segre characteristics will also be identical. Hence T_A and T_B have the same Jordan form.

EXAMPLE 4.2

Let
$$A = J_2(0) \oplus J_2(0)$$
 and $B = J_2(0) \oplus J_1(0) \oplus J_1(0)$. Then
 $ch_A = ch_B = x^4$ and $m_A = m_B = x^2$.

However A is not similar to B. For both matrices are in Jordan form and the Segre characteristics for T_A and T_B are 2, 2 and 2, 1, 1, respectively.

EXERCISE List all possible Jordan canonical forms of 2×2 and 3×3 matrices and deduce that if A and B have the same characteristic and same minimum polynomials, then A and B are similar if A and B are 2×2 or 3×3 .

REMARK. Of course if A and B have the same Jordan canonical form, then A and B are similar.

We now present some interesting applications of the Jordan canonical form.

4.4 Non-derogatory matrices and transformations

If $ch_A = m_A$, we say that the matrix A is **non-derogatory**.

THEOREM 4.5

Suppose that ch_T splits completely in F[x]. Then $ch_T = m_T \Leftrightarrow \exists$ a basis β for V such that

$$[T]^{\beta}_{\beta} = J_{b_1}(c_1) \oplus \ldots \oplus J_{b_t}(c_t),$$

where c_1, \ldots, c_t are distinct elements of F.

PROOF.

4

$$ch_T = \prod_{i=1}^t ch_{J_{b_i}(c_i)} = \prod_{i=1}^t (x - c_i)^{b_i},$$

$$m_T = lcm((x - c_1)^{b_1}, \dots, (x - c_t)^{b_t}) = (x - c_1)^{b_1} \dots (x - c_t)^{b_t} = ch_T.$$

 \Rightarrow Suppose that $ch_T = m_T = (x - c_1)^{a_1} \cdots (x - c_t)^{a_t}$.

We deduce that the dot diagram for each $p_i = (x - c_i)$ consists of a single column of b_i dots, where $p_i^{b_i} || m_T$; that is,

$$\dim_F N_{h,p_i} = 1$$
 for $h = 1, 2, \dots, b_i$.

Then, for each i = 1, 2, ..., t we have the following sequence of positive integers:

$$1 \le \nu(p_i(T)) < \nu(p_i^2(T)) < \dots < \nu(p_i^{b_i}(T)) = a_i.$$

But $a_i = b_i$ here, as we are assuming that $ch_T = m_T$. In particular, it follows that $\nu(p_i^h(T)) = h$ for $h = 1, 2, ..., b_i$ and h = 1 gives

$$\nu(p_i(T)) = 1 = \gamma_i.$$

So the bottom row of the *i*-th dot diagram has only one element; it looks like this:

$$b_i \begin{cases} \frac{\cdot}{\vdots} \\ \frac{\cdot}{\cdot} \end{cases}$$

and we get the secondary decomposition

$$\operatorname{Ker} p_i^{b_i}(T) = C_{T,v_{i1}}.$$

Further, if $\beta = \beta_{11} \cup \cdots \cup \beta_{t1}$, where β_{i1} is the elementary Jordan basis for $C_{T,v_{i1}}$, then

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i))$$
$$= \bigoplus_{i=1}^{t} J_{b_i}(c_i),$$

as required.

4.5 Calculating A^m , where $A \in M_{n \times n}(\mathbb{C})$.

THEOREM 4.6

Let $c \in F$.

(a)

$$J_n^m(c) = \begin{bmatrix} c^m & 0 & \cdots & \cdots & 0\\ \binom{m}{1}c^{m-1} & c^m & \cdots & \cdots & 0\\ \binom{m}{2}c^{m-2} & \binom{m}{1}c^{m-1} & \cdots & \cdots & 0\\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots\\ \binom{m}{m} & & \cdots & \cdots & 0\\ 0 & \binom{m}{m} & \cdots & \cdots & 0\\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots\\ 0 & \cdots & \binom{m}{m} & \cdots & \binom{m}{1}c^{m-1} & c^m \end{bmatrix}$$

if
$$1 \le m \le n-1$$
;

(b)

$$J_n^m(c) = \begin{bmatrix} c^m & 0 & \cdots & 0 & 0\\ \binom{m}{1}c^{m-1} & c^m & \cdots & 0 & 0\\ \binom{m}{2}c^{m-2} & \binom{m}{1}c^{m-1} & \cdots & 0 & 0\\ \vdots & \vdots & \vdots & \vdots & \vdots\\ \binom{m}{n-1}c^{m-n+1} & \binom{m}{n-2}c^{m-n+2} & \cdots & \binom{m}{1}c^{m-1} & c^m \end{bmatrix}$$

if $n-1 \leq m$, where $\binom{m}{k}$ is the binomial coefficient

$$\binom{m}{k} = \frac{m!}{k!(m-k)!} = \frac{m(m-1)\cdots(m-n+1)}{k!}$$

PROOF. $J_n(c) = cI_n + N$, where N has the special property that N^k has 1 on the k-th sub-diagonal and 0 elsewhere, for $0 \le k \le n-1$.

Then because cI_n and N commute, we can use the binomial theorem:

$$J_n^m(c) = (cI_n + N)^m$$

= $\sum_{k=0}^m {m \choose k} (cI_n)^{m-k} N^k$
= $\sum_{k=0}^m {m \choose k} c^{m-k} N^k.$

(a). Let $1 \le m \le n-1$. Then in the above summation, the variable k must satisfy $0 \le k \le n-1$. Hence $J_n^m(c)$ is an $n \times n$ matrix having $\binom{m}{k}c^{m-k}$ on the k-th sub-diagonal, $0 \le k \le m$ and 0 elsewhere. (b). Let $n-1 \le m$. Then

$$J_{n}^{m}(c) = \sum_{k=0}^{m} \binom{m}{k} c^{m-k} N^{k} = \sum_{k=0}^{n-1} \binom{m}{k} c^{m-k} N^{k},$$

as $N^k = 0$ if $n \le k$. Hence $J_n^m(c)$ is an $n \times n$ matrix having $\binom{m}{k}c^{m-k}$ on the k-th sub-diagonal, $0 \le k \le n-1$ and 0 elsewhere.

COROLLARY 4.1

Let $F = \mathbb{C}$. Then

$$\lim_{m \to \infty} J_n^m(c) = 0 \quad \text{if} \quad |c| < 1.$$

PROOF. Suppose that |c| < 1. Let $n - 1 \le m$. Then

$$J_{n}^{m}(c) = \sum_{k=0}^{n-1} \binom{m}{k} c^{m-k} N^{k}.$$

But for fixed k, $0 \le k \le n-1$, $c^{m-k} \to 0$ as $m \to \infty$. For

$$\binom{m}{k} = \frac{m(m-1)\cdots(m-k+1)}{k!}$$

is a polynomial in m of degree k and

$$|m^{j}c^{m-k}| = |m^{j}e^{(m-k)\log c}| = m^{j}e^{(m-k)\log |c|} \to 0 \text{ as } m \to \infty,$$

as $\log c = \log |c| + i \arg c$ and $\log |c| < 0$.

The last corollary gives a more general result:

COROLLARY 4.2

Let $A \in M_{n \times n}(\mathbb{C})$ and suppose that all the eigenvalues of A are less than 1 in absolute value. Then

$$\lim_{m \to \infty} A^m = 0.$$

PROOF. Suppose $ch_A = (x - c_1)^{a_1} \cdots (x - c_t)^{a_t}$, where c_1, \ldots, c_t are the distinct eigenvalues of A and $|c_1| < 1, \ldots, |c_t| < 1$.

Then if J is the Jordan canonical form of A, there exists a non-singular matrix $P \in M_{n \times n}(\mathbb{C})$, such that

$$P^{-1}AP = J = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i).$$

Hence

$$P^{-1}A^{m}P = (P^{-1}AP)^{m} = J^{m} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_{i}} J^{m}_{e_{ij}}(c_{i}).$$

Hence $P^{-1}A^mP \to 0$ as $m \to \infty$, because $J^m_{e_{ij}}(c_i) \to 0$.

4.6 Calculating e^A , where $A \in M_{n \times n}(\mathbb{C})$.

We first show that the matrix limit

$$\lim_{M \to \infty} \left(I_n + A + \frac{1}{2!}A^2 + \dots + \frac{1}{M!}A^M \right)$$

exists. We denote this limit by e^A and write

$$e^{A} = I_{n} + A + \frac{1}{2!}A^{2} + \dots + \frac{1}{m!}A^{m} + \dots = \sum_{m=0}^{\infty} \frac{1}{m!}A^{m}.$$

To justify this definition, we let $A^m = [a_{ij}^{(m)}]$. We have to show that

$$\left(I_n + A + \frac{1}{2!}A^2 + \dots + \frac{1}{M!}A^M\right)_{ij} = a_{ij}^{(0)} + \frac{1}{1!}a_{ij}^{(1)} + \dots + \frac{1}{M!}a_{ij}^{(M)}$$

tends to a limit as $M \to \infty$; in other words, we have to show that the series

$$\sum_{m=0}^{\infty} \frac{1}{m!} a_{ij}^{(m)}$$

converges. To do this, suppose that

$$|a_{ij}| \leq \rho, \quad \forall i, j.$$

Then it is an easy induction to prove that

$$|a_{ij}^{(m)}| \le n^{m-1} \rho^m \quad \text{if} \quad m \ge 1.$$

Then the above series converges by comparison with the series

$$\sum_{m=0}^{\infty} \frac{1}{m!} n^{m-1} \rho^m.$$

4.7 Properties of the exponential of a complex matrix THEOREM 4.7

(i)
$$e^0 = I_n;$$

(ii) $e^{\operatorname{diag}(\lambda_1,\dots,\lambda_n)} = \operatorname{diag}(e^{\lambda_1},\dots,e^{\lambda_n});$
(iii) $e^{P^{-1}AP} = P^{-1}e^AP;$

(iv)
$$e^{\bigoplus_{i=1}^{t} A_i} = \bigoplus_{i=1}^{t} e^{A_i};$$

(v) if A is diagonable and has principal idempotent (spectral) decomposition:

$$A = c_1 E_1 + \dots + c_t E_t,$$

then

$$e^A = e^{c_1} E_1 + \dots + e^{c_t} E_t;$$

(vi)

$$\frac{d}{dt}e^{tA} = Ae^{tA},$$

if A is a constant matrix;

(vii) $e^A = p(A)$, where $p \in \mathbb{C}[x]$;

(viii) e^A is non-singular and

$$(e^A)^{-1} = e^{-A};$$

(ix)
$$e^A e^B = e^{A+B}$$
 if $AB = BA$;

(x)

$$e^{J_n(c)} = \begin{bmatrix} e^c & 0 & 0 & \cdots & 0\\ e^c/1! & e^c & 0 & \cdots & 0\\ e^c/2! & e^c/1! & e^c & \cdots & 0\\ \vdots & \ddots & \ddots & \ddots & \vdots\\ e^c/(n-2)! & & \ddots & e^c/1! & e^c & 0\\ e^c/(n-1)! & e^c/(n-2)! & \cdots & e^c/2! & e^c/1! & e^c \end{bmatrix}.$$

(xi)

$$e^{tJ_n(c)} = \begin{bmatrix} e^{tc} & 0 & 0 & \cdots & 0\\ te^{tc}/1! & e^{tc} & 0 & \cdots & 0\\ t^2 e^{tc}/2! & e^{tc}/1! & e^{tc} & \cdots & 0\\ \vdots & \ddots & \ddots & \ddots & \vdots\\ t^{n-2} e^{tc}/(n-2)! & & \cdot te^{tc}/1! & e^{tc} & 0\\ t^{n-1} e^{tc}/(n-1)! & t^{n-2} e^{tc}/(n-2)! & \cdots & t^2 e^{tc}/2! & te^{tc}/1! & e^c \end{bmatrix}.$$

(xii) If

$$P^{-1}AP = J = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i).$$

then

$$P^{-1}e^{A}P = J = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} e^{J_{e_{ij}}(c_i)}.$$

PROOF.

(i)

$$e^0 = \sum_{m=0}^{\infty} \frac{1}{m!} 0^k = I_n;$$

(ii) Let $A = \operatorname{diag}(\lambda_1, \ldots, \lambda_n)$. Then

$$A^{m} = \operatorname{diag} \left(\lambda_{1}^{m}, \dots, \lambda_{n}^{m}\right)$$
$$\sum_{m=0}^{\infty} \frac{1}{m!} A^{m} = \operatorname{diag} \left(\sum_{m=0}^{\infty} \frac{\lambda_{1}^{m}}{m!}, \dots, \sum_{m=0}^{\infty} \frac{\lambda_{n}^{m}}{m!}\right)$$
$$= \operatorname{diag} \left(e^{\lambda_{1}}, \dots, e^{\lambda_{n}}\right).$$

(iii)

$$e^{P^{-1}AP} = \sum_{m=0}^{\infty} \frac{1}{m!} (P^{-1}AP)^m$$
$$= \sum_{m=0}^{\infty} \frac{1}{m!} (P^{-1}A^m P)$$
$$= P^{-1} \left(\sum_{m=0}^{\infty} \frac{1}{m!} A^m\right) P$$
$$= P^{-1}e^A P.$$

(iv) and (v) are left as exercises.

(vi) Using the earlier notation, $A^m = [a_{ij}^{(m)}]$, we have

$$e^{tA} = \sum_{m=0}^{\infty} \frac{1}{m!} (tA)^m$$
$$= \sum_{m=0}^{\infty} \frac{t^m}{m!} A^m$$
$$= \left[\sum_{m=0}^{\infty} \frac{t^m a_{ij}^{(m)}}{m!} \right]$$
$$\frac{d}{dt} e^{tA} = \left[\frac{d}{dt} \sum_{m=0}^{\infty} \frac{t^m a_{ij}^{(m)}}{m!} \right]$$
$$= \left[\sum_{m=1}^{\infty} \frac{t^m a_{ij}^{(m+1)}}{(m-1)!} \right]$$
$$= \left[\sum_{m=0}^{\infty} \frac{t^m a_{ij}^{(m+1)}}{(m)!} \right]$$

84

$$= \sum_{m=0}^{\infty} \frac{t^m}{m!} A^{m+1}$$
$$= A e^{tA}.$$

(vii) Let deg $m_A = r$. Then the matrices I_n, A, \ldots, A^{r-1} are linearly independent over \mathbb{C} , as if

$$m_A = x^r - a_{r-1}x^{r-1} - \dots - a_0,$$

then

$$m_A(A) = 0 \Rightarrow A^r = a_0 I_n + a_1 A + \dots + a_{r-1} A^{r-1}$$

Consequently for each $m \ge 1$, we can express A^m as a linear combination over \mathbb{C} of I_n, A, \ldots, A^{r-1} :

$$A^{m} = a_{0}^{(m)}I_{n} + a_{1}^{(m)}A + \dots + a_{r-1}^{(m)}A^{r-1}$$

and hence

$$\sum_{m=0}^{M} \frac{1}{m!} A^{m} = \sum_{m=0}^{M} \frac{a_{0}^{(m)}}{m!} I_{n} + \sum_{m=0}^{M} \frac{a_{1}^{(m)}}{m!} A + \dots + \sum_{m=0}^{M} \frac{a_{r-1}^{(m)}}{m!} A^{r-1},$$

or

$$[t_{ij}^{(M)}] = s_{0M}I_n + s_{1M}A + \dots + s_{r-1M}A^{r-1},$$

say.

Now $[t_{ij}^{(M)}] \to e^A$ as $M \to \infty$.

Also the above matrix equation can be regarded as n^2 equations in

$$s_{0M}, s_{1M}, \ldots, s_{r-1, M}$$

Also the linear independence of I_n, A, \ldots, A^{r-1} implies that this system has a unique solution. Consequently we can express $s_{0M}, s_{1M}, \ldots, s_{r-1,M}$ as linear combinations with coefficients independent of M of the sequences $t_{ij}^{(M)}$. Hence, because each of the latter sequences converges, it follows that each of the sequences $s_{0M}, s_{1M}, \ldots, s_{r-1,M}$ converges to $s_0, s_1, \ldots, s_{r-1}$, respectively. Consequently

$$\sum_{k=0}^{r-1} s_{kM} A^k \to \sum_{k=0}^{r-1} s_k A^k$$

and

$$e^A = s_0 I_n + s_1 A + \dots + s_{r-1} A^{r-1},$$

a polynomial in A.

(viii) – (ix) Suppose that AB = BA. Then e^{tB} is a polynomial in B and hence A commutes with e^{tB} . Similarly, A and B commute with e^{A+B} . Now let

$$C(t) = e^{t(A+B)}e^{-tB}e^{-tA}, t \in \mathbb{R}.$$

Then $C(0) = I_n$. Also

$$C'(t) = (A+B)e^{t(A+B)}e^{-tB}e^{-tA} + e^{t(A+B)}(-B)e^{-tB}e^{-tA} + e^{t(A+B)}e^{-tB}(-A)e^{-tA} = 0.$$

Hence C(t) is a constant matrix and C(0) = C(1). That is

$$I_n = e^{A+B} e^{-B} e^{-A}, (12)$$

for any matrices A and B which commute.

The special case B = -A then gives

$$I_n = e^0 e^A e^{-A} = e^A e^{-A},$$

thereby proving that e^A is non-singular and $(e^A)^{-1} = e^{-A}$.

Then multiplying both sides of equation (12) on the left by $e^A e^B$ gives the equation $e^A e^B = e^{A+B}$.

In $\S4.8$ we give an application to the solution of a system of differential equations.

(x) Let $J_n(c) = cI_n + N$, where $N = J_n(0)$. Then

$$e^{J_n(c)} = e^{cI_n + N} = e^{cI_n} e^N$$
$$= (e^c I_n) \sum_{m=0}^{\infty} \frac{1}{m!} N^m$$
$$= \sum_{m=0}^{n-1} \frac{e^c}{m!} N^m.$$

(xi) Similar to above.

4.8 Systems of differential equations

THEOREM 4.8

If X = X(t) satisfies the system of differential equations

$$\dot{X} = AX,$$

for $t \ge t_0$, where A is a constant matrix, then

$$X = e^{(t-t_0)A} X(t_0).$$

PROOF. Suppose $\dot{X} = AX$ for $t \ge t_0$. Then

$$\frac{d}{dt} (e^{-tA}X) = (-Ae^{-tA})X + e^{-tA}\dot{X}$$

$$= (-Ae^{-tA})X + e^{-tA}(AX)$$

$$= (-Ae^{-tA})X + (Ae^{-tA})X$$

$$= (-Ae^{-tA} + Ae^{-tA})X$$

$$= 0X = 0.$$

Hence the vector $e^{-tA}X$ is constant for $t \ge t_0$. Thus

$$e^{-tA}X = e^{-t_0A}X(t_0)$$

and

$$X = e^{tA} e^{-t_0 A} X(t_0) = e^{(t-t_0)A} X(t_0).$$

EXAMPLE 4.3

Solve $\dot{X} = AX$, where

$$A = \begin{bmatrix} 0 & 4 & -2 \\ -1 & -5 & 3 \\ -1 & -4 & 2 \end{bmatrix}.$$

Solution: $\exists P \text{ with}$

$$P^{-1}AP = J_2(-1) \oplus J_1(-1)$$
$$= \begin{bmatrix} -1 & 0 & 0 \\ 1 & -1 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

and

$$P^{-1}(tA)P = \begin{bmatrix} -t & 0 & 0\\ t & -t & 0\\ 0 & 0 & -t \end{bmatrix}.$$

Thus

$$P^{-1}e^{tA}P = e^{tJ_2(-1)} \oplus J_1(-1)$$

= $e^{tJ_2(-1)} \oplus e^{tJ_1(-1)}$
= $\begin{bmatrix} e^{-t} & 0 & 0\\ te^{-t} & e^{-t} & 0\\ \hline 0 & 0 & e^{-t} \end{bmatrix} = K(t), \text{ say.}$

So $e^{tA} = PK(t)P^{-1}$. Now

$$X = e^{tA}X_0 = e^{-t}P\begin{bmatrix} 1 & 0 & 0 \\ t & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$
$$= e^{-t}P\begin{bmatrix} a \\ at+b \\ c \end{bmatrix},$$

where for brevity we have set
$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} = P^{-1}X_0.$$

4.9 Markov matrices

DEFINITION 4.3

A real $n \times n$ matrix $A = [a_{ij}]$ is called a Markov matrix, or row-stochastic matrix if

(i) $a_{ij} \ge 0$ for $1 \le i, j \le n$;

(ii)
$$\sum_{j=1}^{n} a_{ij} = 1$$
 for $1 \le i \le n$.

Remark: (ii) is equivalent to $AJ_n = J_n$, where $J_n = [1, ..., 1]^t$. So 1 is always an eigenvalue of a Markov matrix.

EXERCISE 4.1

If A and B are $n \times n$ Markov matrices, prove that AB is also a Markov matrix.

THEOREM 4.9

Every eigenvalue λ of a Markov matrix satisfies $|\lambda| \leq 1$.

PROOF Suppose $\lambda \in \mathbb{C}$ is an eigenvalue of A and $X \in V_n(\mathbb{C})$ is a corresponding eigenvector. Then

$$AX = \lambda X. \tag{13}$$

Let k be such that $|x_j| \leq |x_k|, \forall j, 1 \leq j \leq n$. Then equating the k-th component of each side of equation (13) gives

$$\sum_{j=1}^{n} a_{kj} x_j = \lambda x_k.$$
(14)

Hence

$$|\lambda x_k| = |\lambda| \cdot |x_k| = |\sum_{j=1}^n a_{kj} x_j| \le \sum_{j=1}^n a_{kj} |x_j|$$
(15)

$$\leq \sum_{j=1}^{n} a_{kj} |x_k| = |x_k|.$$
 (16)

Hence $|\lambda| \leq 1$.

DEFINITION 4.4

A positive Markov matrix is one with all positive elements (i.e. strictly greater than zero). For such a matrix A we may write "A > 0".

THEOREM 4.10

If A is a positive Markov matrix, then 1 is the only eigenvalue of modulus 1. Moreover nullity $(A - I_n) = 1$.

PROOF Suppose $|\lambda| = 1$, $AX = \lambda X$, $X \in V_n(\mathbb{C})$, $X \neq 0$. Then inequalities (15) and (16) reduce to

$$|x_k| = \left|\sum_{j=1}^n a_{kj} x_j\right| \le \sum_{j=1}^n a_{kj} |x_j| \le \sum_{j=1}^n a_{kj} |x_k| = |x_k|.$$
(17)

Then inequalities (17) and a sandwich principle, give

$$|x_j| = |x_k| \quad \text{for} \quad 1 \le j \le n. \tag{18}$$

Also, as equality holds in the triangle inequality section of inequalities (17), this forces all the complex numbers $a_{kj}x_j$ to lie in the same direction:

$$\begin{aligned} a_{kj}x_j &= t_j a_{kk}x_k, \ , t_j > 0, \ 1 \le j \le n, \\ x_j &= \tau_j x_k, \end{aligned}$$

where $\tau_j = (t_j a_{kk})/a_{kj} > 0$.

Then equation (18) implies $\tau_j = 1$ and hence $x_j = x_k$ for $1 \le j \le n$. Consequently $X = x_k J_n$, thereby proving that $N(A - I_n) = \langle J_n \rangle$. Finally, equation (14) implies

$$\sum_{j=1}^{n} a_{kj} x_j = \lambda x_k = \sum_{j=1}^{n} a_{kj} x_k = x_k,$$

so $\lambda = 1$.

COROLLARY 4.3

If A is a positive Markov matrix, then A^t has 1 as the only eigenvalue of modulus 1. Also nullity $(A^t - I_n) = 1$.

PROOF The eigenvalues of A^t are precisely the same as those of A, even up to multiplicities. For

$$\operatorname{ch}_{A^t} = \det \left(xI_n - A^t \right) = \det \left(xI_n - A \right)^t = \det \left(xI_n - A \right) = \operatorname{ch}_A.$$

Also $\nu(A^t - I_n) = \nu(A - I_n)^t = \nu(A - I_n) = 1.$

THEOREM 4.11

If A is a positive Markov matrix, then

(i) $(x-1)||m_A;$

(ii)
$$A^m \to B$$
, where $B = \begin{bmatrix} X^t \\ \vdots \\ \hline X^t \end{bmatrix}$ is a positive Markov matrix and where

X is uniquely defined as the (positive) vector satisfying $A^t X = X$ whose components sum to 1.

Remark: In view of part (i) and the equation $\nu(A - I_n) = 1$, it follows that $(x - 1)|| \operatorname{ch}_A$.

PROOF As $\nu(A - I_n) = 1$, the Jordan form of A has the form $J_b(1) \oplus K$, where $(x - 1)^b || m_A$. Here K is the direct sum of all Jordan blocks corresponding to all the eigenvalues of A other than 1 and hence $K^m \to 0$.

Now suppose that b > 1; then $J_b(1)$ has size b > 1. Then $\exists P$ such that

$$P^{-1}AP = J_b(1) \oplus K,$$

$$P^{-1}A^m P = J_b^m(1) \oplus K^m.$$

Hence the 2 × 1 element of $J_b^m(1)$ equals $\binom{m}{1} \to \infty$ as $m \to \infty$.

However the elements of A^m are ≤ 1 , as A^m is a Markov matrix. Consequently the elements of $P^{-1}A^mP$ are bounded as $m \to \infty$. This contradiction proves that b = 1.

Hence $P^{-1}A^mP \to I_1 \oplus 0$ and $A^m \to P(I_1 \oplus 0)P^{-1} = B$.

We see that rank $B = \operatorname{rank}(I_1 \oplus 0) = 1$.

Finally it is easy to prove that B is a Markov matrix. So

$$B = \begin{bmatrix} t_1 X^t \\ \vdots \\ t_n X^t \end{bmatrix}$$

for some non-negative column vector X and where t_1, \ldots, t_n are positive. We can assume that the entries of X sum to 1. It then follows that $t_1 = \cdots = t_n = 1$ and hence

$$B = \begin{bmatrix} \frac{X^t}{\vdots} \\ \hline \frac{X^t}{X^t} \end{bmatrix}.$$
 (19)

Now $A^m \to B$, so $A^{m+1} = A^m \cdot A \to BA$. Hence B = BA and

$$A^t B^t = B^t. (20)$$

Then equations (19) and (20) imply

$$A^t[X|\cdots|X] = [X|\cdots|X]$$

and hence $A^t X = X$.

However $X \ge 0$ and $A^t > 0$, so $X = A^t X > 0$.

DEFINITION 4.5

We have thus proved that there is a positive eigenvector X of A^t corresponding to the eigenvalue 1, where the components of X sum to 1. Then because we know that the eigenspace $N(A^t - I_n)$ is one-dimensional, it follows that this vector is unique.

This vector is called the **stationary vector** of the Markov matrix A.

EXAMPLE 4.4

Let

$$A = \begin{bmatrix} 1/2 & 1/4 & 1/4 \\ 1/6 & 1/6 & 2/3 \\ 1/3 & 1/3 & 1/3 \end{bmatrix}.$$

Then

$$A^{t} - I_{3} \text{ row-reduces to} \begin{bmatrix} 1 & 0 & -4/9 \\ 0 & 1 & -2/3 \\ 0 & 0 & 0 \end{bmatrix}.$$

Hence
$$N(A^t - I_3) = \left\langle \begin{bmatrix} 4/9\\2/3\\1 \end{bmatrix} \right\rangle = \left\langle \begin{bmatrix} 4/19\\6/19\\9/19 \end{bmatrix} \right\rangle$$
 and
$$\lim_{m \to \infty} A^m = \frac{1}{19} \begin{bmatrix} 4 & 6 & 9\\4 & 6 & 9\\4 & 6 & 9 \end{bmatrix}.$$

We remark that $ch_A = (x - 1)(x^2 - 1/24)$.

DEFINITION 4.6

A Markov Matrix is called **regular** or **primitive** if $\exists k \geq 1$ such that $A^k > 0$.

THEOREM 4.12

If A is a primitive Markov matrix, then A satisfies the same properties enunciated in the last two theorems for positive Markov matrices.

PROOF Suppose $A^k > 0$. Then $(x - 1) || \operatorname{ch}_{A^k}$ and hence $(x - 1) || \operatorname{ch}_A$, as

$$ch_A = (x - c_1)^{a_1} \cdots (x - c_t)^{a_t} \Rightarrow ch_{A^k} = (x - c_1^k)^{a_1} \cdots (x - c_t^k)^{a_t}.$$
 (21)

and consequently $(x-1)||m_A$.

Also as 1 is the only eigenvalue of A^k with modulus 1, it follows from equation (21) that 1 is the only eigenvalue of A with modulus 1.

The proof of the second theorem goes through, with the difference that to prove the positivity of X we observe that $A^t X = X$ implies $(A^k)^t X = X$.

EXAMPLE 4.5

The following Markov matrix is primitive (its fourth power is positive) and is related to the 5x + 1 problem:

$$\left[\begin{array}{rrrrr} 0 & 0 & 1 & 0 \\ 1/2 & 0 & 1/2 & 0 \\ 0 & 0 & 1/2 & 1/2 \\ 0 & 1/2 & 1/2 & 0 \end{array}\right].$$

Its stationary vector is $[\frac{1}{15}, \frac{2}{15}, \frac{8}{15}, \frac{4}{15}]^t$.

We remark that $ch_A = (x - 1)(x + 1/2)(x^2 + 1/4)$.

4.10 The Real Jordan Form

4.10.1 Motivation

If A is a real $n \times n$ matrix, the characteristic polynomial of A will in general have real roots and complex roots, the latter occurring in complex pairs. In this section we show how to derive a canonical form B for A which has real entries. It turns out that there is a simple formula for e^B and this is useful in solving $\dot{X} = AX$, as it allows one to directly express the complete solution of the system of differential equations in terms of real exponentials and sines and cosines.

We first introduce a real analogue of $J_n(a+ib)$. It's the matrix $K_n(a, b) \in M_{2n \times 2n}(\mathbb{R})$ defined as follows:

Let $D = \begin{bmatrix} a & b \\ -b & a \end{bmatrix} = aI_2 + bJ$ where $J^2 = -I_2$ (*J* is a matrix version of $i = \sqrt{-1}$, while *D* corresponds to the complex number a + ib) then

$$e^{D} = e^{aI_{2}+bJ}$$

$$= e^{aI_{2}}e^{bJ}$$

$$= e^{a}I_{2}\left[I_{2} + \frac{bJ}{1!} + \frac{(bJ)^{2}}{2!} + \cdots\right]$$

$$= e^{a}\left[\left\{I_{2} - \frac{b^{2}}{2!}I_{2} + \frac{b^{4}}{4!}I_{2} + \cdots\right\} + \left\{\frac{b}{1!}J - \frac{b^{3}}{3!}J + \cdots\right\}\right]$$

$$= e^{a}\left[(\cos b)I_{2} + (\sin b)J\right]$$

$$= e^{a}\left[\cos b \sin b - \sin b \cos b\right].$$

Replacing a and b by ta and tb, where $t \in \mathbb{R}$, gives

$$e^{tD} = e^{at} \left[\begin{array}{c} \cos bt & \sin bt \\ -\sin bt & \cos bt \end{array} \right].$$

DEFINITION 4.7

Let a and b be real numbers and $K_n(a, b) \in M_{2n \times 2n}(\mathbb{R})$ be defined by

$$K_n(a, b) = \begin{bmatrix} D & 0 & \cdots & \\ \hline I_2 & D & \\ 0 & I_2 & \\ & & \ddots & \\ & & & D \end{bmatrix}$$

where
$$D = \begin{bmatrix} a & b \\ -b & a \end{bmatrix}$$
. Then it is easy to prove that

$$e^{K_n(a,b)} = \begin{bmatrix} e^D & 0 & \cdots & \\ \hline e^D/1! & e^D & \cdots & \\ \hline e^D/2! & e^D/1! & \ddots & \\ \vdots & \ddots & \ddots & \\ e^D/(n-1) & \cdots & \cdots & e^D/1! & e^D \end{bmatrix}$$

EXAMPLE 4.6

$$K_2(0, 1) = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & -1 & 0 \end{bmatrix}$$

.

and

$$e^{tK_2(0,1)} = \begin{bmatrix} \cos t & \sin t & 0 & 0 \\ -\sin t & \cos t & 0 & 0 \\ t \cos t & t \sin t & \cos t & \sin t \\ -t \sin t & t \cos t & -\sin t & \cos t \end{bmatrix}$$

4.10.2 Determining the real Jordan form

If $A = [a_{ij}]$ is a complex matrix, let $\overline{A} = [\overline{a}_{ij}]$. Then 1.

$$\overline{A\pm B}=\overline{A}\pm\overline{B},\ \overline{cA}=\overline{c}\overline{A}\ c\in\mathbb{C},\ \overline{AB}=\overline{A}\cdot\overline{B}.$$

2. If $A \in M_{n \times n}(\mathbb{R})$ and $a_0, \ldots, a_r \in \mathbb{C}$, then

$$\overline{a_0I_n + \cdots + a_rA^r} = \overline{a}_0I_n + \cdots + \overline{a}_rA^r.$$

3. If W is a subspace of $V_n(\mathbb{C})$, then so is $\overline{W} = \{\overline{w} | w \in W\}$. Moreover if $W = \langle w_1, \ldots, w_r \rangle$, then

$$\overline{W} = \langle \overline{w}_1, \dots, \overline{w}_r \rangle.$$

4. If w_1, \ldots, w_r are linearly independent vectors in $V_n(\mathbb{C})$, then so are $\overline{w}_1, \ldots, \overline{w}_r$. Hence if w_1, \ldots, w_r form a basis for a subspace W, then $\overline{w}_1, \ldots, \overline{w}_r$ form a basis for \overline{W} .

5. Let A be a real $n \times n$ matrix and $c \in \mathbb{C}$. Then (a)

(b)

$$W = N((A - cI_n)^h) \Rightarrow \overline{W} = N((A - \overline{c}I_n)^h).$$
(b)

$$W = W_1 \oplus \dots \oplus W_r \Rightarrow \overline{W} = \overline{W}_1 \oplus \dots \oplus \overline{W}_r.$$
(c)

$$W = C_{T_A, v} \Rightarrow \overline{W} = C_{T_A, \overline{v}}.$$

(d)

(e)

$$W = \bigoplus_{i=1}^{r} C_{T_A, v_i} \Rightarrow \overline{W} = \bigoplus_{i=1}^{r} C_{T_A, \overline{v}_i}.$$

$$m_{T_A,v} = (x-c)^e \Rightarrow m_{T_A,\overline{v}} = (x-\overline{c})^e.$$

Let $A \in M_{n \times n}(\mathbb{R})$. Then $m_A \in \mathbb{R}[x]$ and so any complex roots will occur in conjugate pairs.

Suppose that c_1, \ldots, c_r are the distinct real eigenvalues and c_{r+1}, \ldots, c_{r+s} , $\bar{c}_{r+1}, \ldots, \bar{c}_{r+s}$ are the distinct non-real roots and

$$m_A = (x - c_1)^{b_1} \dots (x - c_r)^{b_r} (x - c_{r+1})^{b_{r+1}} \dots (x - c_{r+s})^{b_{r+s}} \times (x - \bar{c}_{r+1})^{b_{r+1}} \dots (x - \bar{c}_{r+s})^{b_{r+s}}.$$

For each complex eigenvalue c_i , $r+1 \le i \le r+s$, there exists a secondary decomposition

$$N(A - c_i I_n)^{b_i} = \bigoplus_{j=1}^{\gamma_i} C_{T_A, v_{ij}}, \quad m_{T_A, v_{ij}} = (x - c_i)^{e_{ij}}$$

Hence we have a corresponding secondary decomposition for the eigenvalue \bar{c}_i :

$$N(A - \bar{c}_i I_n)^{b_i} = \bigoplus_{j=1}^{\gamma_i} C_{T_A, \bar{v}_{ij}}, \quad m_{T_A, \bar{v}_{ij}} = (x - \bar{c}_i)^{e_{ij}}.$$

For brevity, let $c = c_i$, $v = v_{ij}$, $e = e_{ij}$. Let

$$P_1 = v, P_2 = (A - cI_n)P_1, \dots, P_e = (A - cI_n)P_{e-1}$$

and

$$P_1 = X_1 + iY_1, P_2 = X_2 + iY_2, \dots, P_e = X_e + iY_e; c = a + ib.$$

Then we have the following equations, posed in two different ways:

$$\begin{array}{rcrcrcrc} AP_{1} & = & cP_{1} + P_{2} \\ & & \\ AP_{e} & = & cP_{e} \end{array} \begin{array}{rcrc} AX_{1} & = & aX_{1} - bY_{1} + X_{2} \\ AY_{1} & = & bX_{1} + aY_{1} + Y_{2} \\ & & \\ \vdots \\ AX_{e} & = & aX_{e} - bY_{e} \\ AY_{e} & = & bX_{e} + aY_{e}. \end{array}$$

In matrix terms we have

$$A[X_1|Y_1|X_2|Y_2|\cdots|X_e|Y_e] = \begin{bmatrix} a & b & & & \\ -b & a & & & \\ \hline 1 & 0 & a & b & & \\ 0 & 1 & -b & a & & \\ \vdots & & & \ddots & \\ & & & & & \hline 1 & 0 & a & b & \\ 0 & 1 & -b & a & & \\ \vdots & & & \ddots & \\ & & & & & \hline 1 & 0 & a & b & \\ 0 & 0 & 1 & -b & a & \\ \vdots & & & \ddots & \\ & & & & & \hline 1 & 0 & a & b & \\ 0 & & & & & -b & a \end{bmatrix}.$$

The large "real jordan form" matrix is the $2e \times 2e$ matrix $K_e(a, b)$. Note: If e = 1, no I_2 block is present in this matrix.

The spaces $C_{T_A,v}$ and $C_{T_A,\bar{v}}$ are independent and have bases P_1, \ldots, P_e and $\bar{P}_1, \ldots, \bar{P}_e$, respectively.

Consequently the vectors

$$P_1,\ldots,P_e,\,\bar{P}_1,\ldots,\bar{P}_e$$

form a basis for $C_{T_A,v} + C_{T_A,\bar{v}}$. It is then an easy exercise to deduce that the real vectors $X_1, Y_1, \ldots, X_e, Y_e$ form a basis β for the *T*-invariant subspace

$$W = C_{T_A, v} + C_{T_A, \bar{v}}.$$

Writing $T = T_A$ for brevity, the above right hand batch of equations tells us that $[T_W]^{\beta}_{\beta} = K_e(a, b)$. There will be s such real bases corresponding to each of the complex eigenvalues $c_{r+1} \dots, c_{r+s}$. Joining together these bases with the real elementary Jordan bases arising from any real eigenvalues c_1, \ldots, c_r gives a basis β for $V_n(\mathbb{C})$ such that if P is the non-singular real matrix formed by these basis vectors, then

$$P^{-1}AP = [T_A]^\beta_\beta = J \oplus K,$$

where

$$J = \bigoplus_{i=1}^{r} \bigoplus_{j=1}^{\gamma_i} J_{e_{ij}}(c_i), \quad K = \bigoplus_{i=r+1}^{r+s} \bigoplus_{j=1}^{\gamma_i} K_{e_{ij}}(a_i, b_i),$$

where $c_i = a_i + ib_i$ for $r+1 \le i \le r+s$.

The matrix $J \oplus K$ is said to be in real Jordan canonical form.

EXAMPLE 4.7

$$A = \begin{bmatrix} 1 & 1 & 0 & 0 \\ -2 & 0 & 1 & 0 \\ 2 & 0 & 0 & 1 \\ -2 & -1 & -1 & -1 \end{bmatrix} \quad \text{so} \quad m_A = (x^2 + 1)^2 \\ = (x - i)^2 (x + i)^2.$$

Thus with $p_1 = x - i$, we have the dot diagram

$$\begin{array}{c} \cdot \\ \cdot \\ \cdot \\ \end{array} \begin{array}{c} N_{2,p_1} \\ N_{1,p_1} = N(A - iI_4). \end{array}$$

Thus we find an elementary Jordan basis for N_{1,p_1} :

$$X_{11} + iY_{11}, \ (A - iI_4)(X_{11} + iY_{11}) = X_{12} + iY_{12}$$

yielding

$$\begin{aligned} AX_{11} &= -Y_{11} + X_{12} \\ AY_{11} &= X_{11} + Y_{12}. \end{aligned}$$
(22)

Now we know

$$m_{T_A, X_{11}+iY_{11}} = (x-i)^2$$

$$\Rightarrow (A-iI_4)^2 (X_{11}+iY_{11}) = 0$$

$$\Rightarrow (A-iI_4) (X_{12}+iY_{12}) = 0$$

$$\Rightarrow AX_{12} = -Y_{12}$$

$$AY_{12} = X_{12}.$$
(23)

Writing the four real equations (22) and (23) in matrix form, with

$$P = [X_{11}|Y_{11}|X_{12}|Y_{12}],$$

then P is non-singular and

$$P^{-1}AP = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ \hline 1 & 0 & 0 & 1 \\ 0 & 1 & -1 & 0 \end{bmatrix}.$$

The numerical determination of P is left as a tutorial problem.

4.10.3 A real algorithm for finding the real Jordan form

Referring to the last example, if we write $Z = \begin{bmatrix} A & I_4 \\ -I_4 & A \end{bmatrix}$, then

$$Z\left[\frac{X_{11}}{Y_{11}}\right] = \left[\frac{X_{12}}{Y_{12}}\right],$$

$$Z\left[\frac{-Y_{11}}{X_{11}}\right] = \left[\frac{-Y_{12}}{X_{12}}\right],$$

$$Z\left[\frac{X_{12}}{Y_{12}}\right] = \left[\frac{0}{0}\right],$$

$$Z\left[\frac{-Y_{12}}{X_{12}}\right] = \left[\frac{0}{0}\right].$$

Then the vectors

$$\left[\frac{X_{11}}{Y_{11}}\right], \ \left[\frac{-Y_{11}}{X_{11}}\right], \ Z\left[\frac{X_{11}}{Y_{11}}\right], \ Z\left[\frac{-Y_{11}}{X_{11}}\right]$$

actually form an \mathbb{R} -basis for N(Z). This leads to a method for finding the real Jordan canonical form using real matrices. (I am indebted to Dr. B.D. Jones for introducing me to the Z matrix approach.)

More generally, we observe that a collection of equations of the form

$$AX_{ij1} = a_i X_{ij1} - b_i Y_{ij1} + X_{ij2}$$

$$AY_{ij1} = b_i X_{ij1} + a_i Y_{ij1} + Y_{ij2}$$

$$\vdots$$

$$AX_{ije_{ij}} = a_i X_{ije_{ij}} - b_i Y_{ije_{ij}}$$

$$AY_{ije_{ij}} = b_i X_{ije_{ij}} + a_i Y_{ije_{ij}}$$

can be written concisely in real matrix form, giving rise to an elementary Jordan basis corresponding to an elementary divisor $x^{e_{ij}}$ for the following real matrix: Let

$$Z_i = \left[\begin{array}{cc} A - a_i I_n & b_i I_n \\ -b_i I_n & A - a_i I_n \end{array} \right].$$

Then

$$Z_{i} \begin{bmatrix} X_{ij1} \\ Y_{ij1} \end{bmatrix} = \begin{bmatrix} X_{ij2} \\ Y_{ij2} \end{bmatrix}$$
$$\vdots$$
$$Z_{i} \begin{bmatrix} X_{ije_{ij}} \\ Y_{ije_{ij}} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

LEMMA 4.2

If V is a \mathbb{C} -vector space with basis v_1, \ldots, v_n , then V is also an \mathbb{R} -vector space with basis

$$v_1, iv_1, \ldots, v_n, iv_n$$

Hence

$$\dim_{\mathbb{R}} V = 2 \dim_{\mathbb{C}} V.$$

DEFINITION 4.8

Let $A \in M_{n \times n}(\mathbb{R})$ and c = a + ib be a complex eigenvalue of A with $b \neq 0$. Let $Z \in M_{2n \times 2n}(\mathbb{R})$ be defined by

$$Z = \begin{bmatrix} A - aI_n & bI_n \\ -bI_n & A - aI_n \end{bmatrix} = (A - aI_n) \otimes I_n - I_n \otimes (bJ).$$

Also let p = x - c.

LEMMA 4.3

Let $\Phi: V_{2n}(\mathbb{R}) \to V_n(\mathbb{C})$ be the mapping defined by

$$\Phi\left(\left[\frac{X}{Y}\right]\right) = X + iY, \quad X, Y \in V_n(\mathbb{R}).$$

Then

(i) Φ is an \mathbb{R} --isomorphism;

(ii)
$$\Phi\left(\left[\frac{-Y}{X}\right]\right) = i(X + iY);$$

(iii)
$$\Phi\left(Z^h\left[\frac{X}{Y}\right]\right) = p^h(A)(X+iY);$$

- $(iv) \ \Phi\left(Z^h\left[\frac{-Y}{X}\right]\right) = ip^h(A)(X+iY);$
- (v) Φ maps $N(Z^h)$ onto $N(p^h(A);$

COROLLARY 4.4

If

$$p^{e_1-1}(A)(X_1+iY_1),\ldots,p^{e_{\gamma}-1}(A)(X_{\gamma}+iY_{\gamma})$$

form a \mathbb{C} -basis for N(p(A)), then

$$Z^{e_1-1}\left[\frac{X_1}{Y_1}\right], Z^{e_1-1}\left[\frac{-Y_1}{X_1}\right], \dots, Z^{e_{\gamma}-1}\left[\frac{X_{\gamma}}{Y_{\gamma}}\right], Z^{e_{\gamma}-1}\left[\frac{-Y_{\gamma}}{X_{\gamma}}\right]$$

form an \mathbb{R} -basis for N(Z) and conversely.

Remark: Consequently the dot diagram for the eigenvalue 0 for the matrix Z has the same height as that for the eigenvalue c of A, with each row expanded to twice the length.

To find suitable vectors $X_1, Y_1, \ldots, X_{\gamma}, Y_{\gamma}$, we employ the usual algorithm for finding the Jordan blocks corresponding to the eigenvalue 0 of the matrix Z, with the extra proviso that we always ensure that the basis for $N_{h,x}$ is chosen to have the form

$$Z^{h-1}\left[\frac{X_1}{Y_1}\right], Z^{h-1}\left[\frac{-Y_1}{X_1}\right], \dots, Z^{h-1}\left[\frac{X_r}{Y_r}\right], Z^{h-1}\left[\frac{-Y_r}{X_r}\right],$$

where r = (nullity $Z^h -$ nullity $Z^{h-1})/2.$

This can be ensured by extending a spanning family for $N(Z^h)$:

$$\left[\frac{X_1}{Y_1}\right], \dots, \left[\frac{X_{\nu(Z^h)}}{Y_{\nu(Z^h)}}\right]$$

to the form

$$\left[\frac{X_1}{Y_1}\right], \left[\frac{-Y_1}{X_1}\right], \dots, \left[\frac{X_{\nu(Z^h)}}{Y_{\nu(Z^h)}}\right], \left[\frac{-Y_{\nu(Z^h)}}{X_{\nu(Z^h)}}\right].$$

EXAMPLE 4.8

$$A = \begin{bmatrix} 1 & 1 & 0 & 0 \\ -2 & 0 & 1 & 0 \\ 2 & 0 & 0 & 1 \\ -2 & -1 & -1 & -1 \end{bmatrix} \in M_{4 \times 4}(\mathbb{R}) \text{ has } m_A = (x^2 + 1)^2. \text{ Find a real}$$

non-singular matrix P such that $P^{-1}AP$ is in real Jordan form. Solution:

$$Z = \begin{bmatrix} 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 \\ -2 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 2 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ -2 & -1 & -1 & -1 & 0 & 0 & 0 & 1 \\ -1 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & -1 & 0 & 0 & -2 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 & 2 & 0 & 0 & 1 \\ 0 & 0 & 0 & -1 & -2 & -1 & -1 & -1 \end{bmatrix}$$

102

basis for
$$N(Z^2)$$
:
$$\begin{bmatrix} 1 & 1 & 1/2 & 1/2 \\ -2 & -1/2 & 0 & -1/2 \\ 2 & 1/2 & 1 & 3/2 \\ -2 & -3/2 & -2 & -3/2 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

blown–up basis for $N(Z^2)$:

ſ	-	1	-1	1	0	1/2	0	1/2	0]
	_	-2	0	-1/2	-1	0	0	-1/2	0
		2	0	1/2	0	1	-1	3/2	0
	_	-2	0	-3/2	0	-2	0	-3/2	-1
		1	1	0	1	0	1/2	0	1/2
		0	-2	1	-1/2	0	0	0	-1/2
		0	2	0	1/2	1	1	0	3/2
	-	0	-2	0	-3/2	0	-2	1	-3/2

$$\rightarrow \text{ left-to-right basis for } N(Z^2): \begin{bmatrix} 1 & -1 & 1 & 0 \\ -2 & 0 & -1/2 & -1 \\ 2 & 0 & 1/2 & 0 \\ -2 & 0 & -3/2 & 0 \\ 1 & 1 & 0 & 1 \\ 0 & -2 & 1 & -1/2 \\ 0 & 2 & 0 & 1/2 \\ 0 & -2 & 0 & -3/2 \end{bmatrix}$$

We then derive a spanning family for $N_{2,x}$:

$$Z \times \text{ basis matrix} = \begin{bmatrix} 0 & 0 & 1/2 & 0 \\ 0 & 0 & -1/2 & -1/2 \\ 0 & 0 & 1/2 & 1/2 \\ 0 & 0 & -1/2 & -1/2 \\ 0 & 0 & 0 & 1/2 \\ 0 & 0 & 1/2 & -1/2 \\ 0 & 0 & -1/2 & 1/2 \\ 0 & 0 & 1/2 & -1/2 \end{bmatrix} \rightarrow \text{ basis for } N_{2,x}:$$

1/2	0
-1/2	-1/2
1/2	1/2
-1/2	-1/2
0	1/2
1/2	-1/2
-1/2	1/2
1/2	-1/2

 $\begin{bmatrix} 1/2 & 0 \\ -1/2 & -1/2 \\ 1/2 & 1/2 \\ -1/2 & -1/2 \\ 0 & 1/2 \\ 1/2 & -1/2 \\ 1/2 & -1/2 \\ 1/2 & -1/2 \\ 1/2 & -1/2 \end{bmatrix}$ Consequently we read off that $Z\left[\frac{X_{11}}{Y_{11}}\right] = \left[\frac{X_{12}}{Y_{12}}\right]$ is a basis for $N_{2,x} = N_{1,x} = N(Z)$. where

$$P = [X_{11}|Y_{11}|X_{12}|Y_{12}] = \begin{bmatrix} 1 & 0 & 1/2 & 0 \\ -1/2 & 1 & -1/2 & 1/2 \\ 1/2 & 0 & 1/2 & -1/2 \\ -3/2 & 0 & -1/2 & 1/2 \end{bmatrix}.$$

Then

$$P^{-1}AP = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & -1 & 0 \end{bmatrix},$$

which is in real Jordan form.

5 The Rational Canonical Form

Here p is a monic irreducible factor of the minimum polynomial m_T and is not necessarily of degree one.

Let F_p denote the field constructed earlier in the course, consisting of all matrices of the form f(B), $f \in F[x]$, where B = C(p), the companion matrix of p. (We saw that if deg p = n, then

$$F_p = \{a_0 I_n + \dots + a_{n-1} B^{n-1} | a_0, \dots, a_{n-1} \in F\}.$$

Let $\overline{f} = f(B)$, where $f \in F[x]$. Then this new symbol has the following properties:

- (i) $\overline{f} + \overline{g} = \overline{f + g}; \ \overline{f} \ \overline{g} = \overline{fg};$
- (ii) $\overline{f} = \overline{0} \Leftrightarrow p \mid f;$
- (iii) $\overline{f} = \overline{g} \Leftrightarrow p \mid (f g);$
- (iv) \overline{f}^{-1} exists $\Leftrightarrow p$ does not divide f.

Note: If p = x - c, then $F_p = F$.

THEOREM 5.1

 $N_{h,p}$ becomes a vector space over F_p if we define

$$\overline{f}v = fv = f(T)(v).$$

First we must verify that the above definition is well-defined, that is, independent of the particular polynomial f used to define the field element \overline{f} . So suppose $\overline{f} = \overline{g}$. Then f = g + kp, $k \in F[x]$. Hence

$$fv = (g + kp)v = gv + k(pv) = gv + k0 = gv,$$

as $v \in \operatorname{Im} p^{h-1}(T) \cap \operatorname{Ker} p(T)$ and consequently pv = 0.

The four addition axioms hold as V is already a vector space over F; The remaining vector space axioms then follow from the left F[x]-module axioms:

- (i) $(\overline{f} + \overline{g}) = \overline{(f+g)}v = (f+g)v = fv + gv = \overline{f}v + \overline{g}v;$
- (ii) $\overline{f}(v+w) = f(v+w) = fv + fw = \overline{f}v + \overline{f}w;$

(iii)
$$\overline{f}(\overline{g}v) = \overline{f}(gv) = f(gv) = (fg)v = (\overline{fg})v;$$

(iv) $\overline{1}v = 1v = v.$

Remark: An F-basis for $N_{h,p}$ will be an F_p -spanning family for $N_{h,p}$, but will not, in general, be an F_p -basis for $N_{h,p}$. The precise connection between F-independence and F_p -independence is given by the following theorem:

THEOREM 5.2

Vectors v_1, \ldots, v_r form an F_p -basis for $N_{h,p}$ if and only if the vectors

form an *F*-basis for $N_{h,p}$.

COROLLARY 5.1

$$\nu_{h,p} = \dim_{F_p} N_{h,p} = \frac{1}{\deg p} \dim_F N_{h,p} = \frac{\nu(p^h(T)) - \nu(p^{h-1}(T))}{\deg p}.$$

The exposition for p = x - c now goes over to general p, with small changes. We again have the decreasing sequence of dimensions:

$$\nu_{1,p} \geq \cdots \geq \nu_{b,p} \geq 1,$$

where $\nu_{1,p} = \dim_{F_p} \operatorname{Ker} p(T) = \frac{\nu(p(T))}{\deg p}$. Also

$$\nu_{1,p} + \dots + \nu_{b,p} = \frac{\nu(p^b(T))}{\deg p},$$
(24)

where $p^b \parallel \mathbf{m}_T$.

There is a corresponding dot diagram where the number of dots in the h-th row from the bottom represents the integer $\nu_{h,p}$. We also have a similar theorem to an earlier one, in terms of the conjugate partition

$$e_1 \geq \cdots \geq e_\gamma \geq 1$$

of the partition (24) above, where $\gamma = \nu_{1,p} = \dim_{F_p} \operatorname{Ker} p(T) = \frac{\nu(p(T))}{\deg p}$.

THEOREM 5.3

Vectors $v_1, \ldots, v_{\gamma} \in V$ can be found with the property that

$$p^{e_1-1}v_1,\ldots,p^{e_\gamma-1}v_\gamma$$

form an F_p -basis for Kerp(T). Moreover

- (i) $m_{T,v_j} = p^{e_j};$
- (ii) Ker $p^b(T) = C_{T,v_1} \oplus \cdots \oplus C_{T,v_{\gamma}}$.

In conclusion, if $m_T = p_1^{b_1} \cdots p_t^{b_t}$, we now have the direct sum decomposition

$$V = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C_{T, v_{ij}},$$

where $m_{T,v_{ij}} = p_i^{e_{ij}}$ and

$$e_{i1} = b_i \ge \ldots \ge e_{i\gamma_i}$$

form the conjugate partition for the dot diagram corresponding to p_i . Here

$$\gamma_i = \frac{\nu(p_i(T))}{\deg p_i}.$$

Taking T-cyclic bases β_{ij} for $C_{T, v_{ij}}$, then gives a basis

$$\beta = \bigcup_{i=1}^t \bigcup_{j=1}^{\gamma_i} \beta_{ij}$$

for V. Moreover

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C(p_i^{e_{ij}}).$$

The matrix on the right is said to be in **rational canonical form**.

If instead, we take the following basis β_{ij}' for $C_{T,\,v_{ij}}$

$$\beta'_{ij} : \begin{cases} v_{ij}, & T(v_{ij}), & \dots, & T^{n-1}(v_{ij}) \\ p_i(T)(v_{ij}), & Tp_i(T)(v_{ij}), & \dots, & T^{n-1}p_i(T)(v_{ij}) \\ \vdots & \vdots & \vdots & \vdots \\ p_i^{e_{ij}-1}(T)(v_{ij}), & Tp_i^{e_{ij}-1}(T)(v_{ij}), & \dots, & T^{n-1}p_i^{e_{ij}-1}(T)(v_{ij}), \end{cases}$$

(with $n = \deg p_i$) which reduces to the Jordan basis when $p_i = x - c_i$, it is not difficult to verify that we get a corresponding matrix $H(p_i^{e_{ij}})$ called a **hypercompanion** matrix, which reduces to the elementary Jordan matrix $J_{e_{ij}}(c_i)$ when $p_i = x - c_i$:

$$H(p_i^{e_{ij}}) = \begin{bmatrix} C(p_i) & 0 & \cdots & 0 \\ N & C(p_i) & \cdots & 0 \\ 0 & N & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & \cdots & N & C(p_i) \end{bmatrix},$$

where there are e_{ij} blocks on the diagonal and N is a square matrix of same size as $C(p_i)$ which is everywhere zero, except in the top right-hand corner, where there is a 1. The overall effect is an unbroken subdiagonal of 1's.

We then get the corresponding rational canonical form:

$$[T]^{\beta'}_{\beta'} = \bigoplus_{i=1}^t \bigoplus_{j=1}^{\gamma_i} H(p_i^{e_{ij}}).$$

Computational Remark:

We can do our computations completely over F, without going into F_p , as follows. Suppose v_1, \ldots, v_r form an F-spanning family for $N_{h,p}$. Then we could, in principle, perform the LRA over F_p on this spanning family and find an F_p -basis v_{c_1}, \ldots, v_{c_R} . A little thought reveals that if we had instead applied the LRA algorithm over F to the expanded sequence:

$$v_1, T(v_1), \ldots, T^{n-1}(v_1); \ldots; v_r, T(v_r), \ldots, T^{n-1}(v_r)$$

we would have obtained the *F*-basis for $N_{h,p}$:

$$v_{c_1}, T(v_{c_1}), \ldots, T^{n-1}(v_{c_1}); \ldots; v_{c_R}, T(v_{c_R}), \ldots, T^{n-1}(v_{c_R})$$

from which we select the desired F_p -basis v_{c_1}, \ldots, v_{c_R} .

Let
$$A = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 2 \\ 1 & 0 & 0 & 0 & 2 & 1 \\ 0 & 1 & 0 & 0 & 2 & 2 \\ 2 & 0 & 1 & 0 & 1 & 2 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \in M_{6 \times 6}(\mathbb{Z}_3).$$

Here $m_A = p^2$, $p = x^2 + x + 2 \in F[x]$, $F = \mathbb{Z}_3$.

$$p(A) = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 2 & 1 & 0 \\ 0 & 1 & 2 & 2 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 & 2 \\ 0 & 0 & 1 & 2 & 2 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}, \quad \nu(p(A)) = 4, \ \nu_{1,p} = \frac{\nu(p(A))}{\deg p} = 2.$$

$$p^{2}(A) = 0, \ \nu(p^{2}(A)) = 6, \ \nu_{2,p} = \frac{\nu(p^{2}(A)) - \nu(p(A))}{\deg p} = \frac{6-4}{2} = 1.$$

Hence we have a corresponding ${\cal F}_p$ dot diagram:

$$\begin{array}{c|c} \cdot & & N_{2,p} \\ \hline \cdot & \cdot & & N_{1,p} \end{array}$$

We have to find an F_p -basis $p(A)v_{11}$ for $N_{2,p}$ and extend this to an F_p -basis $p(A)v_{11}, v_{12}$ for N(p(A)). An *F*-basis for $N(p^2(A))$ is E_1, \ldots, E_6 . Then

$$N_{2,p} = \langle p(A)E_1, \dots, p(A)E_6 \rangle$$

and the LRA give $p(A)E_2$ as an F_p -basis for $N_{2,p}$ so we can take $v_{11} = E_2$. We find the columns of the following matrix form an F-basis for N(p(A)):

We place $p(A)E_2$ in front and then pad the resulting matrix to get

[0	0	1	1	0	0	0	0	0	2 -	
$\begin{bmatrix} 0\\2 \end{bmatrix}$	0	0	1	2	0	1	2	0	1	
1	2	0	0	1	2	1	0	1	2	
1	1	0	2	1	1	0	2	0	0	•
0	1	0	0	0	1	1	1	0	1	
$\begin{bmatrix} 0\\ 0 \end{bmatrix}$	0	0	1	0	0	0	0	1	1	

The first four columns $p(A)E_2$, $Ap(A)E_2$, E_1 , AE_1 of this matrix form a LR F-basis for N(p(A)) and hence $p(A)E_2$, E_1 form an F_p -basis for N(p(A)). So we can take $v_{12} = E_1$.

Then $V_6(\mathbb{Z}_3) = N(p^2(A)) = C_{T_A, v_{11}} \oplus C_{T_A, v_{12}}$. Then joining hypercompanion bases for $C_{T_A, v_{11}}$ and $C_{T_A, v_{12}}$:

 $v_{11}, Av_{11}, p(A)v_{11}, Ap(A)v_{11}$ and v_{12}, Av_{11}

gives a basis v_{11} , Av_{11} , $p(A)v_{11}$, $Ap(A)v_{11}$; v_{12} , Av_{11} for $V_6(\mathbb{Z}_3)$. Finally if P is the non-singular matrix whose columns are these vectors, we transform A into direct sum of hypercompanion matrices:

$$P^{-1}AP = H(p^2) \oplus H(p) = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 2 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 2 \end{bmatrix}$$

Explicitly, we have

$$P = \begin{bmatrix} 0 & 0 & 0 & 0 & 1 & 1 \\ 1 & 0 & 2 & 0 & 0 & 1 \\ 0 & 1 & 1 & 2 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

5.1 Uniqueness of the Rational Canonical Form

Suppose that $T:V\to V$ is a linear transformation over F and that β is a basis for V such that

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C(p_i^{e_{ij}}).$$
(25)

where

$$e_{i1} \ge \ldots \ge e_{i\gamma_i} \ge 1 \tag{26}$$

and p_1, \ldots, p_t are distinct monic irreducible polynomials.

We show that the polynomials p_i and the sequences (26) are determined by the transformation T.

First, it is not difficult to show that

$$\beta = \bigcup_{i=1}^t \bigcup_{j=1}^{\gamma_i} \beta_{ij},$$

where

$$\beta_{ij}: v_{ij}, T(v_{ij}), \dots, T^{n_{ij}-1}(v_{ij})$$

and $n_{ij} = \deg p_i^{e_{ij}}$ and $m_{T,v_{ij}} = p_i^{e_{ij}}$. Then we have the direct sum decomposition

$$V = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C_{T, v_{ij}}.$$

Also if we write $b_i = e_{i1}$, we have

$$\operatorname{Ker} p_i^{b_i}(T) = \bigoplus_{j=1}^{\gamma_i} C_{T, v_{ij}}$$

and hence

$$V = \bigoplus_{i=1}^{t} \operatorname{Ker} p_i^{b_i}(T).$$

Then from equation (25) above, it follows that

$$\mathbf{m}_T = \operatorname{lcm} p_i^{e_{ij}} = p_1^{b_1} \cdots p_t^{b_t},$$

thereby determining p_1, \ldots, p_t up to order.

Then it can be shown that if $1 \leq h \leq b_i$, then N_{h,p_i} has F_{p_i} basis

$$p_i^{e_{i1}-1}v_{i1},\ldots,p_i^{e_{ij_h}-1}v_{ij_h},$$

where e_{i1}, \ldots, e_{ij_h} are the integers not less than h.

There are consequently $\dim_{F_{p_i}} N_{h, p_i} = \nu_{h, p_i}$ such integers and hence the number of integers $e_{i1}, \ldots, e_{i\gamma_i}$ equal to h is equal to $\nu_{h, p_i} - \nu_{h+1, p_i}$, which depends only on T. In other words, for each i, the sequence $e_{i1}, \ldots, e_{i\gamma_i}$ depends only on T.

5.2 Deductions from the Rational Canonical Form

THEOREM 5.4

$$\frac{\nu(p_i^{b_i}(T))}{\deg p_i} = a_i$$

where $p_i^{a_i} || \operatorname{ch}_T$, and $p_i^{b_i} || m_T$.

Note that this determines b_i —we may evaluate

$$\frac{\nu(p_i^h(T))}{\deg p_i}$$

for h = 1, 2, ... until we get a value of a_i . Then that $h = b_i$. PROOF \exists a basis for V such that

$$A = [T]^\beta_\beta = \bigoplus_{i=1}^t \bigoplus_{j=1}^{\gamma_i} C(p_i^{e_{ij}}).$$

 So

$$\mathbf{ch}_T = \prod_{i=1}^t \prod_{j=1}^{\gamma_i} \mathbf{ch}_{B_{i,j}}$$

where, for brevity, we write $B_{i,j} = C(p_i^{e_{ij}})$. Hence

$$ch_T = \prod_{i=1}^t \prod_{j=1}^{\gamma_i} p_i^{e_{ij}}$$
$$= \prod_{i=1}^t p_i^{j=1} e_{ij}$$
$$= \prod_{i=1}^t p_i^{\frac{\nu(p_i^{b_i}(T))}{\deg p_i}}$$

as required.

THEOREM 5.5

 $ch_T = m_T \Leftrightarrow \exists a \text{ basis } \beta \text{ for } V \text{ such that}$

$$[T]^{\beta}_{\beta} = C(p_1^{b_1}) \oplus \ldots \oplus C(p_t^{b_t})$$

where p_1, \ldots, p_t are distinct monic irreducibles and $b_1 \ge \ldots b_t \ge 1$.

Note that if $ch_A = m_A$ (i.e. $T = T_A$ in the above), we say that the matrix A is **non-derogatory**. PROOF \Leftarrow

 \Rightarrow Suppose that $ch_T = m_T$.

We deduce that the dot diagram for each p_i consists of a single column of b_i dots, where $p_i^{b_i} || m_T$; that is,

$$\dim_{F_p} N_{h,p_i} = 1$$
 for $h = 1, 2, \dots, b_i$.

Observe that

$$\frac{\nu(p_i^h(T))}{\deg p_i} \in \mathbb{N},$$

for it may be written

$$\sum_{j=1}^{h} \frac{\nu(p_i^j(T)) - \nu(p_i^{j-1}(T))}{\deg p_i}$$
$$= \sum_{j=1}^{h} \dim_{F_p} N_{j,p_i} \in \mathbb{N}.$$

Then, for each i = 1, 2, ..., t we have the following sequence of positive integers:

$$1 \leq \frac{\nu(p_i(T))}{\deg p_i} < \frac{\nu(p_i^2(T))}{\deg p_i} < \ldots < \frac{\nu(p_i^{b_i}(T))}{\deg p_i} = a_i.$$

But $a_i = b_i$ here, as we are assuming that $ch_T = m_T$. In particular, it follows that

$$\frac{\nu(p_i^h(T))}{\deg p_i} = h \qquad \text{for } h = 1, 2, \dots, b_i$$

and h = 1 gives

$$\frac{\nu(p_i(T))}{\deg p_i} = 1 = \gamma_i.$$

So the bottom row of the *i*-th dot diagram has only one element; it looks like this:



and we get the secondary decomposition

$$\operatorname{Ker} p_i^{o_i}(T) = C_{T,v_{i1}}.$$

Further, if $\beta = \beta_{11} \cup \cdots \cup \beta_{t1}$, where β_{i1} is the *T*-cyclic basis for $C_{T,v_{i1}}$, then

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C(p_i^{e_{ij}})$$
$$= \bigoplus_{i=1}^{t} C(p_i^{b_i})$$
$$= C(p_1^{b_1}) \oplus \ldots \oplus C(p_t^{b_t})$$

as required.

THEOREM 5.6

 $m_T = p_1 p_2 \dots p_t$, a product of distinct monic irreducibles, if and only if \exists a basis β for V such that

$$[T]_{\beta}^{\beta} = \underbrace{C(p_{1}) \oplus \ldots \oplus C(p_{1})}_{\gamma_{1} \text{ times}} \oplus \underbrace{C(p_{t}) \oplus \ldots \oplus C(p_{t})}_{\gamma_{t} \text{ times}}.$$
(27)

Note: This is a generalization of an earlier result, namely that a transformation is diagonable if and only if its minimum polynomial splits into a product of distinct linear factors. PROOF

 \Leftarrow Assume $\exists \beta$ such that (27) holds. Then

$$m_T = \lim_{\gamma_1} (\underbrace{p_1, \dots, p_1}_{\gamma_1}, \dots, \underbrace{p_t, \dots, p_t}_{\gamma_t})$$
$$= \lim_{\gamma_1} (p_1, \dots, p_t)$$
$$= p_1 p_2 \dots p_t.$$

⇒ Assume $m_T = p_1 \dots p_t$. Then $b_i = 1$ for $i = 1, \dots, t$ (i.e. the *i*-th dot diagram has height 1) and $\exists \beta$ such that

$$[T]^{\beta}_{\beta} = \bigoplus_{i=1}^{t} \bigoplus_{j=1}^{\gamma_i} C(p_i),$$

as
$$e_{ij} = 1 \ \forall i, j$$
.

5.3 Elementary divisors and invariant factors

5.3.1 Elementary Divisors

DEFINITION 5.1

The polynomials $p_i^{e_{ij}}$ occurring in the rational canonical form of T are called the **elementary divisors** of T. Similarly the elementary divisors of a matrix $A \in M_{n \times n}(F)$ are the polynomials $p_i^{e_{ij}}$ occurring in the rational canonical form of A.

THEOREM 5.7

Linear transformations $T_1, T_2 : V \to V$ have the same elementary divisors if and only if there exists an isomorphism $L : V \to V$ such that

$$T_2 = L^{-1}T_1L.$$

PROOF

"only if". Suppose that T_1 and T_2 have the same elementary divisors. Then \exists bases β , γ for V such that

$$[T_1]^\beta_\beta = [T_2]^\gamma_\gamma = A.$$

Then we have the equations

$$\begin{split} \phi_{\beta}T_1 &= T_A\phi_{\beta} \\ \phi_{\gamma}T_2 &= T_A\phi_{\gamma}. \end{split}$$

Hence

$$\phi_{\beta}T_1\phi_{\beta}^{-1} = T_A = \phi_{\gamma}T_2\phi_{\gamma}^{-1},$$

 \mathbf{SO}

$$\phi_{\gamma}^{-1}\phi_{\beta}T_1\phi_{\beta}^{-1}\phi_{\gamma} = T_2,$$

or

$$L^{-1}T_1L = T_2,$$

where $L = \phi_{\beta}^{-1} \phi_{\gamma}$ is an isomorphism.

"if". Suppose that $L^{-1}T_1L = T_2$. Then

$$m_{T_1} = m_{T_2} = p_1^{b_1} \cdots p_t^{b_t}, \quad \text{say};$$

also for all i and h, because

$$p_i^h(T_2) = p_i^h(L^{-1}T_1L) = L^{-1}p_i^h(T_1)L,$$

we have

$$\nu(p_i^h(T_2)) = \nu(p_i^h(T_1)).$$

Hence for each p_i , the corresponding dot diagrams for T_1 and T_2 are identical and consequently the elementary divisors for T_1 and T_2 are identical.

COROLLARY 5.2

Let $A, B \in M_{n \times n}(F)$. Then A is similar to B if and only if A and B have the same elementary divisors.

PROOF

$$\begin{array}{lll} A \text{ is similar to } B & \Leftrightarrow & \exists \ P \text{ non-singular, with } P^{-1}AP = B \\ & \Leftrightarrow & \exists \ P \text{ non-singular, with } T_P^{-1}T_AT_P = T_B \\ & \Leftrightarrow & \exists \ L \text{ an isomorphism, with } L^{-1}T_AL = T_B. \end{array}$$

5.3.2 Invariant Factors

THEOREM 5.8

Let $T: V \to V$ be a linear transformation over F. Then there exist non-constant monic polynomials $d_1, \ldots, d_s \in F[x]$, such that

- (i) d_k divides d_{k+1} for $1 \le k \le s-1$;
- (ii) vectors $v_1, \ldots, v_s \in V$ exist such that

$$V = \bigoplus_{k=1}^{s} C_{T, v_k},$$

where $m_{T,v_k} = d_k$.

REMARK: If β is the basis for V obtained by stringing together the T-cyclic bases for each C_{T, v_k} , we obtain the matrix direct sum

$$[T]^{\beta}_{\beta} = \bigoplus_{k=1}^{s} C(d_k).$$

This matrix is also said to be in rational canonical form. PROOF

Let $s = \max(\gamma_1, \ldots, \gamma_t)$ and if $1 \leq i \leq t$ and $\gamma_i < j \leq s$, define $e_{ij} = 0$ and $v_{ij} = 0$, the zero vector of V. Now arrange the polynomials $p_i^{e_{ij}}$, $1 \leq i \leq t$; $1 \leq j \leq s$ as a $t \times s$ rectangular array:

$p_1^{e_{1s}}$	• • •	$p_1^{e_{11}}$
÷	:	÷
$p_t^{e_{ts}}$	•••	$p_t^{e_{t1}}$

Let

$$d_1 = p_1^{e_{1s}} \cdots p_t^{e_{ts}}, \dots, d_s = p_1^{e_{11}} \cdots p_t^{e_{t1}}$$

be the products along columns of the array, from left to right. Then d_1, \ldots, d_s are monic non-constant polynomials and

$$d_1|d_2|\cdots|d_s.$$

Also $C_{T,v_{ij}} = \{0\}$ if $v_{ij} = 0$, so V is the direct sum of the following ts T-cyclic subspaces:

Then by Problem Sheet 5, Question 15(b), if we let

$$v_1 = v_{1s} + \cdots + v_{ts}, \dots, v_s = v_{11} + \cdots + v_{t1},$$

we have $m_{T,v_1} = d_1, ..., m_{T,v_s} = d_s$ and

$$C_{T,v_1} = C_{T,v_{1s}} \oplus \cdots \oplus C_{T,v_{ts}}$$

$$\vdots \qquad \vdots$$

$$C_{T,v_s} = C_{T,v_{11}} \oplus \cdots \oplus C_{T,v_{t1}}.$$

Consequently

$$V = C_{T,v_1} \oplus \cdots \oplus C_{T,v_s}$$

DEFINITION 5.2

Polynomials d_1, \ldots, d_s satisfying the conditions of the above theorem are called **invariant factors** of T.

There is a similar definition for matrices: if $A \in M_{n \times n}(F)$ is similar to a direct sum

$$\bigoplus_{k=1}^{s} C(d_k),$$

where d_1, \ldots, d_s are non-constant monic polynomials in F[x] such that d_k divides d_{k+1} for $1 \le k \le s-1$, then d_1, \ldots, d_s are called **invariant factors** of A. So the invariant factors of A are the invariant factors of T_A .

THEOREM 5.9

The invariant factors of a linear transformation $T: V \to V$ are uniquely defined by T.

PROOF

Reverse the construction in the proof of the above theorem using Question 15(a) of Problem Sheet 5, thereby recapturing the rectangular array of elementary divisors, which in turn is uniquely determined by T.

EXAMPLE 5.1

Suppose $T: V \to V$ has elementary divisors

$$p_1^2, p_1^3, p_1^3; p_2, p_2^2, p_2^2, p_2^4; p_3, p_3, p_3^4, p_3^5, p_3^5.$$

Form the rectangular array

1	1	p_{1}^{2}	p_{1}^{3}	p_1^3
1	p_2	p_{2}^{2}	p_{2}^{2}	p_{2}^{4}
p_3	p_3	p_{3}^{4}	p_{3}^{5}	p_{3}^{5}

Then the invariant factors of T are obtained by respectively multiplying along columns:

THEOREM 5.10

If d_1, \ldots, d_s are the invariant factors of $T: V \to V$, then

- (i) $m_T = d_s;$
- (ii) $\operatorname{ch}_T = d_1 \cdots d_s$.

PROOF

Suppose $B = [T]_{\beta}^{\beta} = \bigoplus_{k=1}^{s} C(d_k)$ is the canonical form corresponding to the invariant factors d_1, \ldots, d_s of T. Then

$$m_T = m_B = \operatorname{lcm}(m_{C(d_1)}, \dots, m_{C(d_s)})$$

= $\operatorname{lcm}(d_1, \dots, d_s) = d_s.$

Also

$$\operatorname{ch}_T = \operatorname{ch}_B = \prod_{k=1}^s \operatorname{ch}_{C(d_k)} = \prod_{k=1}^s d_k.$$

We shall soon see that the invariant factors of a linear transformation or matrix are of independent interest. For example the invariant factors allow us to calculate the dimension of the vector space $Z_{L,M}$ consisting of all linear transformations $N: U \to V$ which satisfy the equation MN = NL, where $L: U \to U$ and $M: V \to V$ are given linear transformations over F.

It turns out that there is a more direct way of finding the invariant factors of T. To introduce this algorithm, we need to discuss an interesting equivalence relation on $M_{m \times n}(F[x])$, which in turn leads to the so-called **Smith canonical form** of a matrix over F[x].

6 The Smith Canonical Form

6.1 Equivalence of Polynomial Matrices

DEFINITION 6.1

A matrix $P \in M_{n \times n}(F[x])$ is called a **unit** in $M_{n \times n}(F[x])$ if $\exists Q \in M_{n \times n}(F[x])$ such that

$$PQ = I_n.$$

Clearly if P and Q are units, so is PQ.

THEOREM 6.1

A matrix $P \in M_{n \times n}(F[x])$ is a unit in $M_{n \times n}(F[x])$ if and only if det P = c, where $c \in F$ and $c \neq 0$.

proof

"only if". Suppose P is a unit. Then $PQ = I_n$ and

$$\det PQ = \det P \, \det Q = \det I_n = 1.$$

However det P and det Q belong to F[x], so both are in fact non-zero elements of F.

"if". Suppose $P \in M_{n \times n}(F[x])$ satisfies det P = c, where $c \in F$ and $c \neq 0$. Then

$$P \operatorname{adj} P = (\det P)I_n = cI_n$$

Hence $PQ = I_n$, where $Q = c^{-1} \operatorname{adj} P \in M_{n \times n}(F[x])$. Hence P is a unit in $M_{n \times n}(F[x])$.

EXAMPLE 6.1

$$P = \begin{bmatrix} 1+x & -x \\ x & 1-x \end{bmatrix} \in M_{2\times 2}(F[x]) \text{ is a unit, as det } P = 1.$$

THEOREM 6.2

Elementary row matrices in $M_{n \times n}(F[x])$ are units:

- (i) E_{ij} : interchange rows *i* and *j* of I_n ;
- (ii) $E_i(t)$: multiply row *i* of I_n by $t \in F$, $t \neq 0$;
- (iii) $E_{ij}(f)$: add f times row j of I_n to row $i, f \in F[x]$.

In fact det $E_{ij} = -1$; det $E_i(t) = t$; det $E_{ij}(f) = 1$. Similarly for elementary column matrices in $M_{n \times n}(F[x])$:

$$F_{ij}, F_i(t), F_{ij}(f).$$

REMARK: It follows that a product of elementary matrices in $M_{n \times n}(F[x])$ is a unit. Later we will be able to prove that the converse is also true.

DEFINITION 6.2

Let $A, B \in M_{m \times n}(F[x])$. Then A is equivalent to B over F[x] if units $P \in M_{m \times m}(F[x])$ and $Q \in M_{n \times n}(F[x])$ exist such that

$$PAQ = B.$$

THEOREM 6.3

Equivalence of matrices over F[x] defines an equivalence relation on $M_{m \times n}(F[x])$.

6.1.1 Determinantal Divisors

DEFINITIONS 6.1

Let $A \in M_{m \times n}(F[x])$. Then for $1 \le k \le \min(m, n)$, let $d_k(A)$ denote the gcd of all $k \times k$ minors of A.

 $d_k(A)$ is sometimes called the k^{th} determinantal divisor of A. Note: $gcd(f_1, \ldots, f_n) \neq 0 \Leftrightarrow$ at least one of f_1, \ldots, f_n is non-zero.

 $\rho(A)$, the **determinantal rank** of A, is defined to be the largest integer r for which there exists a non-zero $r \times r$ minor of A.

THEOREM 6.4

For $1 \leq k \leq \rho(A)$, we have $d_k(A) \neq 0$. Also $d_k(A)$ divides $d_{k+1}(A)$ for $1 \leq k \leq \rho(A) - 1$.

proof

Let $r = \rho(A)$. Then there exists an $r \times r$ non-zero minor and hence $d_r(A) \neq 0$. Then because each $r \times r$ minor is a linear combination over F[x] of $(r-1) \times (r-1)$ minors of A, it follows that some $(r-1) \times (r-1)$ minor of A is also non-zero and hence $d_{r-1}(A) \neq 0$; also $d_{r-1}(A)$ divides each minor of size r - 1 and consequently divides each minor of size r; hence $d_{r-1}(A)$ divides $d_r(A)$, the gcd of all minors of size r. This argument can be repeated with r replaced by r - 1 and so on.

THEOREM 6.5

Let $A, B \in M_{m \times n}(F[x])$. Then if A is equivalent to B over F[x], we have

(i) $\rho(A) = \rho(B) = r;$

(ii) $d_k(A) = d_k(B)$ for $1 \le k \le r$.

proof

Suppose PAQ = B, where P and Q are units. First consider PA. The rows of PA are linear combinations over F[x] of the rows of A, so it follows that each $k \times k$ minor of PA is a linear combination of the $k \times k$ minors of A. Similarly each column of (PA)Q is a linear combinations over F[x] of the columns of PA, so it follows that each $k \times k$ minor of B = (PA)Q is a linear combination over F[x] of the k × k minor of F[x] of the $k \times k$ minor of A.

It follows that all minors of B with size $k > \rho(A)$ must be zero and hence $\rho(B) \le \rho(A)$. However B is equivalent to A, so we deduce that $\rho(A) \le \rho(B)$ and hence $\rho(A) = \rho(B)$.

Also $d_k(B)$ is a linear combination over F[x] of all $k \times k$ minors of Band hence of all $k \times k$ minors of A. Hence $d_k(A)|d_k(B)$ and by symmetry, $d_k(B)|d_k(A)$. Hence $d_k(A) = d_k(B)$ if $1 \le k \le r$.

6.2 Smith Canonical Form

THEOREM 6.6 (Smith canonical form)

Every non-zero matrix $A \in M_{m \times n}(F[x])$ with $r = \rho(A)$ is equivalent to a matrix of the form

	f_1	0	• • •	0		0 -]
	0	f_2	· · · · · · ·	0	• • •	0	
D	÷	÷	÷	÷	÷	÷	= PAQ
D =	0	0	:	f_r	• • •	0	= PAQ
	÷		÷	÷	·	÷	
	0	0		0		0	

where $f_1, \ldots, f_r \in F[x]$ are monic, $f_k | f_{k+1}$ for $1 \le k \le r-1$, P is a product of elementary row matrices, and Q is a product of elementary column matrices.

DEFINITION 6.3

The matrix D is said to be in **Smith canonical form**.

proof

This is presented in the form of an algorithm which is in fact used by CMAT to find unit matrices P and Q such that PAQ is in Smith canonical form.

Our account is based on that in the book "Rings, Modules and Linear Algebra," by B. Hartley and T.O. Hawkes.

We describe a sequence of elementary row and column operations over F[x], which when applied to a matrix A with $a_{11} \neq 0$ either yields a matrix C of the form

$$C = \begin{bmatrix} f_1 & 0 \cdots 0 \\ 0 & & \\ \vdots & C^* \\ 0 & & \end{bmatrix}$$

where f_1 is monic and divides every element of C^* , or else yields a matrix B in which $b_{11} \neq 0$ and

$$\deg b_{11} < \deg a_{11}. \tag{28}$$

Assuming this, we start with our non-zero matrix A. By performing suitable row and column interchanges, we can assume that $a_{11} \neq 0$. Now repeatedly perform the algorithm mentioned above. Eventually we must reach a matrix of type C, otherwise we would produce an infinite strictly decreasing sequence of non-negative integers by virtue of inequalities of type (28).

On reaching a matrix of type C, we stop if $C^* = 0$. Otherwise we perform the above argument on C^* and so on, leaving a trail of diagonal elements as we go.

Two points must be made:

- (i) Any elementary row or column operation on C^* corresponds to an elementary operation on C, which does not affect the first row or column of C.
- (ii) Any elementary operation on C^* gives a new C^* whose new entries are linear combinations over F[x] of the old ones; consequently these new entries will still be divisible by f_1 .

Hence in due course we will reach a matrix D which is in Smith canonical form.

We now detail the sequence of elementary operations mentioned above. Case 1. $\exists a_{1j}$ in row 1 with a_{11} not dividing a_{1j} . Then

$$a_{1j} = a_{11}q + b,$$

by Euclid's division theorem, where $b \neq 0$ and $\deg b < \deg a_{11}$. Subtract q times column 1 from column j and then interchange columns 1 and j. This yields a matrix of type B mentioned above.

Case 2. $\exists a_{i1}$ in column 1 with a_{11} not dividing a_{i1} . Proceed as in Case 1, operating on rows rather than columns, again reaching a matrix of type *B*. Case 3. Here a_{11} divides every element in the first row and first column. Then by subtracting suitable multiples of column 1 from the other columns, we can replace all the entries in the first row other than a_{11} by 0. Similarly for the first column. We then have a matrix of the form

$$E = \begin{bmatrix} e_{11} & 0 \cdots & 0 \\ 0 & & \\ \vdots & E^* \\ 0 & & \end{bmatrix}.$$

If e_{11} divides every element of E^* , we have reached a matrix of type C. Otherwise $\exists e_{ij}$ not divisible by e_{11} . We then add row i to row 1, thereby reaching Case 1.

EXAMPLE 6.2

(of the Smith Canonical Form)

$$A = \left[\begin{array}{cc} 1+x^2 & x \\ x & 1+x \end{array} \right]$$

We want D = PAQ in Smith canonical form. So we construct the augmented matrix

	work	t on rows	5	work on columns			
		\downarrow		\downarrow			
	1	0	$\begin{array}{c} 1+x^2\\ x\end{array}$	x	1 0		
	0	1	x	$x \\ 1+x$	0 1		
$R_1 \to R_1 - xR_2 \Rightarrow$	1	-x	1	$-x^{2}$	1 0		
	0	1	x	1+x	$0 \ 1$		
$C_2 \rightarrow C_2 + x^2 C_1 \Rightarrow$	1	-x	1	0	$1 \ x^2$		
	0	1	x	$1 + x + x^3$	$0 \ 1$		
$R_2 \rightarrow R_2 - xR_1 \Rightarrow$	1	-x	1	0	$1 x^2$		
	-x	$1 + x^2$	0	$1 + x + x^3$	0 1		
		↑		Ť	↑ (
		P		D	Q		

Invariants are $f_1 = 1$, $f_2 = 1 + x + x^3$. Note also

$$f_1 = d_1(A), \qquad f_2 = \frac{d_2(A)}{d_1(A)}.$$

6.2.1 Uniqueness of the Smith Canonical Form

THEOREM 6.7

Every matrix $A \in M_{m \times n}(F[x])$ is equivalent to precisely one matrix is Smith canonical form.

proof Suppose A is equivalent to a matrix B in Smith canonical form. That is,

$$B = \begin{bmatrix} f_1 & & \\ & \ddots & \\ & & f_r \\ \hline & & 0 & 0 \end{bmatrix} \text{ and } f_1 \mid f_2 \mid \cdots \mid f_r.$$

Then $r = \rho(A)$, the determinantal rank of A. But if $1 \le k \le r$,

$$d_k(A) = d_k(B) = f_1 f_2 \dots f_k$$

and so the f_i are uniquely determined by

$$f_1 = d_1(A)$$

$$f_2 = \frac{d_2(A)}{d_1(A)}$$

$$\vdots$$

$$f_r = \frac{d_r(A)}{d_{r-1}(A)}$$

6.3 Invariant factors of a polynomial matrix

DEFINITION 6.4

The polynomials f_1, \ldots, f_r in the Smith canonical form of A are called the **invariant factors** of A^{3}

Note: CMAT calls the invariant factors of xI - B, where $B \in M_{n \times n}(F)$, the "similarity invariants" of B.

We next find these similarity invariants. They are

$$\underbrace{1,1,\ldots,1}_{n-s}, d_1,\ldots,d_s$$

where d_1, \ldots, d_s are what earlier called the invariant factors of T_B .

 $^{{}^{3}}$ **NB.** This is a slightly different, though similar, form of "invariant factor" to that we met a short while ago.

LEMMA 6.1

The Smith canonical form of $xI_n - C(d)$ where d is a monic polynomial of degree n is

diag
$$(\underbrace{1,\ldots,1}_{n-1},d).$$

proof Let $d = x^n + a_{n-1}x^{n-1} + \dots + a_0 \in F[x]$, so

$$xI_n - C(d) = \begin{bmatrix} x & 0 & & a_0 \\ -1 & x & \cdots & & a_1 \\ 0 & -1 & & a_2 \\ \vdots & \ddots & & \vdots \\ & & & x & a_{n-2} \\ 0 & & \cdots & -1 & x + a_{n-1} \end{bmatrix}$$

Now use the row operation

$$R_1 \to R_1 + xR_2 + x^2R_3 + \dots + x^{n-1}R_n$$

to obtain

$$\begin{bmatrix} 0 & 0 & & d \\ -1 & x & \cdots & a_1 \\ 0 & -1 & & a_2 \\ \vdots & \ddots & \vdots \\ & & & x & a_{n-2} \\ 0 & & \cdots & -1 & x + a_{n-1} \end{bmatrix}$$

(think about it!) and then column operations

$$C_2 \to C_2 + xC_1, \dots, C_{n-1} \to C_{n-1} + xC_{n-2}$$

and then

$$C_n \to C_n + a_1 C_1 + a_2 C_2 + \dots + a_{n-2} C_{n-2} + (x + a_{n-1}) C_{n-1}$$

yielding

$$\begin{bmatrix} 0 & 0 & \dots & 0 & d \\ -1 & 0 & & & 0 \\ 0 & -1 & & & \\ & & \ddots & & \vdots \\ 0 & & \dots & -1 & 0 \end{bmatrix}.$$

126

Trivially, elementary operations now form the matrix

diag
$$(\underbrace{1,\ldots,1}_{n-1},d).$$

THEOREM 6.8

Let $B \in M_{n \times n}(F)$. Then if the invariant factors of B are d_1, \ldots, d_s , then the invariant factors of $xI_n - B$ are

$$\underbrace{1,\ldots,1}_{n-s}, d_1, d_2, \ldots, d_s.$$

proof There exists non-singular $P \in M_{n \times n}(F)$ such that

$$P^{-1}BP = \bigoplus_{k=1}^{s} C(d_k).$$

Then

$$P^{-1}(xI_n - B)P = xI_n - \bigoplus_{k=1}^s C(d_k)$$
$$= \bigoplus_{k=1}^s (xI_{m_k} - C(d_k)) \quad \text{where } m_k = \deg d_k.$$

But by the lemma, each $xI_{m_k} - C(d_k)$ is equivalent over F[x] to diag $(1, \ldots, 1, d_k)$ and hence $xI_n - B$ is equivalent to

$$\bigoplus_{k=1}^{s} \operatorname{diag}(1, \dots, 1, d_k) \sim \begin{bmatrix} 1 & & & & \\ & \ddots & & & \\ & & 1 & & \\ & & & d_1 & & \\ & & & & \ddots & \\ & & & & & d_s \end{bmatrix}.$$

EXAMPLE 6.3

Find the invariant factors of

$$B = \begin{bmatrix} 2 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ 0 & -1 & 0 & -1 \\ 1 & 1 & 1 & 2 \end{bmatrix} \in M_{4 \times 4}(\mathbb{Q})$$

by finding the Smith canonical form of $xI_4 - B$. Solution:

$$xI_4 - B = \begin{bmatrix} x - 2 & 0 & 0 & 0 \\ 1 & x - 1 & 0 & 0 \\ 0 & 1 & x & 1 \\ -1 & -1 & -1 & x - 2 \end{bmatrix}$$

We start off with the row operations

$$\begin{array}{rccc} R_1 & \rightarrow & R_1 - (x-2)R_2 \\ R_1 & \leftrightarrow & R_2 \\ R_4 & \rightarrow & R_4 + R_1 \end{array}$$

and get

$$(\text{column ops.}) \Rightarrow \begin{bmatrix} 1 & x-1 & 0 & 0 \\ 0 & -(x-1)(x-2) & 0 & 0 \\ 0 & 1 & x & 1 \\ 0 & x-2 & -1 & x-2 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -(x-1)(x-2) & 0 & 0 \\ 0 & 1 & x & 1 \\ 0 & -(x-1)(x-2) & 0 & 0 \\ 0 & x-2 & -1 & x-2 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & x & 1 \\ 0 & -(x-1)(x-2) & 0 & 0 \\ 0 & x-2 & -1 & x-2 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & x & 1 \\ 0 & 0 & x(x-1)(x-2) & (x-1)(x-2) \\ 0 & 0 & -1-x(x-2) & 0 \\ = -(x-1)^2 \} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & x(x-1)(x-2) & (x-1)(x-2) \\ 0 & 0 & -(x-1)^2 & 0 \end{bmatrix} .$$

Now, for brevity, we work just on the 2×2 block in the bottom right corner:

$$\Rightarrow \begin{bmatrix} (x-1)(x-2) & x(x-1)(x-2) \\ 0 & -(x-1)^2 \end{bmatrix}$$

$$C_{2} \to C_{2} - xC_{1} \Rightarrow \begin{bmatrix} (x-1)(x-2) & 0\\ 0 & -(x-1)^{2} \end{bmatrix}$$

$$R_{1} \to R_{1} + R_{2} \Rightarrow \begin{bmatrix} (x-1)(x-2) & (x-1)^{2}\\ 0 & -(x-1)^{2} \end{bmatrix}$$

$$C_{2} \to C_{2} - C_{1} \Rightarrow \begin{bmatrix} (x-1)(x-2) & x-1\\ 0 & -(x-1)^{2} \end{bmatrix}$$

$$C_{1} \leftrightarrow C_{2} \Rightarrow \begin{bmatrix} x-1 & (x-1)(x-2)\\ -(x-1)^{2} & 0 \end{bmatrix}$$

$$C_{2} \to C_{2} - (x-2)C_{1} \Rightarrow \begin{bmatrix} x-1 & 0\\ -(x-1)^{2} & (x-2)(x-1)^{2} \end{bmatrix}$$

$$R_{2} \to R_{2} + (x-1)R_{1} \Rightarrow \begin{bmatrix} x-1 & 0\\ 0 & (x-2)(x-1)^{2} \end{bmatrix}$$

and here we stop, as we have a matrix in Smith canonical form. Thus

$$xI_4 - B \sim \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & x-1 & \\ & & & (x-1)^2(x-2) \end{bmatrix}$$

so the invariant factors of B are the non-trivial ones of $xI_4 - B$, i.e.

(x-1) and $(x-1)^2(x-2)$.

Also, the elementary divisors of B are

$$(x-1)$$
, $(x-1)^2$ and $(x-2)$

so the Jordan canonical form of B is

$$J_2(1) \oplus J_1(1) \oplus J_1(2).$$

THEOREM 6.9

Let $A, B \in M_{n \times n}(F)$. Then A is similar to B

$$\Leftrightarrow xI_n - A \text{ is equivalent to } xI_n - B$$

$$\Leftrightarrow xI_n - A \text{ and } xI_n - B \text{ have the same}$$

Smith canonical form.

 proof

 \Rightarrow Obvious. If $P^{-1}AP = B, P \in M_{n \times n}(F)$ then

$$P^{-1}(xI_n - A)P = xI_n - P^{-1}AP$$
$$= xI_n - B.$$

 \Leftarrow If $xI_n - A$ and $xI_n - B$ are equivalent over F[x], then they have the same invariant factors and so have the same non-trivial invariant factors. That is, A and B have the same invariant factors and hence are similar.

Note: It is possible to start from $xI_n - A$ and find $P \in M_{n \times n}(F)$ such that

$$P^{-1}AP = \bigoplus_{k=1}^{s} C(d_k)$$

where

 $P_1(xI_n - B)Q_1 = \text{diag}(1, \dots, 1, d_1, \dots, d_s).$

(See Perlis, Theory of matrices, p. 144, Corollary 8–1 and p. 137, Theorem 7–9.)

THEOREM 6.10

Every unit in $M_{n \times n}(F[x])$ is a product of elementary row and column matrices.

PROOF: Problem sheet 7, Question 12.

7 Various Applications of Rational Canonical Forms

7.1 An Application to commuting transformations

THEOREM 7.1 (Cecioni 1908, Frobenius 1910)

Let $L: U \mapsto U$ and $M: V \mapsto V$ be given LTs. Then the vector space $Z_{L,M}$ of all LTs $N: U \mapsto V$ satisfying

$$MN = NL$$

has dimension

$$\sum_{k=1}^{s} \sum_{l=1}^{t} \deg \gcd(d_k, D_l),$$

where d_1, \ldots, d_s and D_1, \ldots, D_t are the invariant factors of L and M respectively.

COROLLARY 7.1

Now take U = V and L = M. Then $Z_{L,L}$ the vector space of LTs satisfying

$$NL = LN$$
,

has dimension

$$\sum_{k=1}^{s} (2s - 2k + 1) \deg d_k.$$

proof Omitted, but here's a hint:

$$gcd(d_k, d_l) = \begin{cases} d_k \text{ if } k \leq l \text{ ; i.e. if } d_k \mid d_l \\ d_l \text{ if } k > l \text{ ; i.e. if } d_l \mid d_k. \end{cases}$$

N.B. Let P_L be the vector space of all LTs of the form

$$f(L): U \mapsto U \qquad f \in F[x].$$

Then $P_L \subseteq Z_{L,L}$ and we have the following...

THEOREM 7.2

$$P_L = Z_{L,L} \Leftrightarrow m_L = \operatorname{ch}_L.$$

proof First note that $\dim P_L = \deg m_L$ as

$$I_V, L, \ldots, L^{\deg m_L - 1}$$

form a basis for P_L . So, since $P_L \subseteq Z_{L,L}$ we have

$$P_L = Z_{L,L} \iff \dim P_L = \dim Z_{L,L}$$
$$\Leftrightarrow \quad \deg m_L = \sum_{k=1}^s (2s - 2k + 1) \deg d_k$$
$$\Leftrightarrow \quad s = 1$$
$$\Leftrightarrow \quad \operatorname{ch}_L = m_L.$$

proof (a sketch) of Cecioni-Frobenius theorem. We start with the invariant factor decompositions

$$U = \bigoplus_{k=1}^{s} C_{L,u_k}$$
 and $V = \bigoplus_{l=1}^{t} C_{M,v_l}$

where $m_{L,u_k} = d_k$ for k = 1, ..., s, and $m_{M,v_l} = D_l$ for l = 1, ..., t. Let MN = NL ...

$$\Rightarrow \quad M^n N = NL^n \quad \forall n \ge 1 \Rightarrow \quad f(M)N = Nf(L) \quad \forall f \in F[x].$$

Define vectors $w_1, \ldots, w_s \in V$ by $w_k = N(u_k)$, and observe

$$d_k(M)(w_k) = d_k(M)(N(u_k)) = N(d_k(L)(u_k)) = N(0) = 0.$$

Then we have the

Definition: Let W be the set of all (w_1, \ldots, w_s) such that $w_1, \ldots, w_s \in V$ and

$$d_k(M)(w_k) = 0 \qquad \forall k = 1, \dots, s$$

We assert that W is a vector space and the mapping

$$N \mapsto (w_1, \ldots, w_s)$$

is an isomorphism between $Z_{L,M}$ and W; proof is left as an exercise.

Now let

$$w_k = \sum_{l=1}^{t} c_{kl}(M)(v_l)$$
 $k = 1, \dots, s \text{ and } c_{kl} \in F[x].$

N.B.

$$f(M)(v_l) = g(M)(v_l) \text{ say}$$

$$\Leftrightarrow \quad D_l \mid f - g.$$

So, if we restrict c_{kl} by the condition

$$\deg c_{kl} < \deg D_l \quad \text{if } c_{kl} \neq 0 \tag{29}$$

then the c_{kl} are uniquely defined for each k. EXERCISE: Now let

$$g_{kl} = \gcd(d_k, D_l).$$

Then from the condition $d_k(M)(w_k) = 0$, show that

$$\frac{D_l}{g_{kl}} \mid c_{kl} \tag{30}$$

i.e. that

$$c_{kl} = b_{kl} \frac{D_l}{g_{kl}} \qquad b_{kl} \in F[x].$$
(31)

Then the matrices $[c_{kl}]$, where c_{kl} satisfy (30), form a vector space (call it X) which is isomorphic to W.

Then in (31),

(29)
$$\iff \deg b_{kl} < \deg g_{kl} \qquad \text{if } b_{kl} \neq 0.$$

Clearly then,

$$\dim X = \dim Z_{L,M} = \sum_{k=1}^{s} \sum_{l=1}^{t} \deg g_{kl}$$

as required.

EXAMPLE 7.1

(of the vector space X, when s = t = 2) Say

$$\left[\deg g_{kl}\right] = \left[\begin{array}{cc} 2 & 0\\ 1 & 3 \end{array}\right].$$

Then X consists of all matrices of the form

$$\begin{bmatrix} c_{kl} \end{bmatrix} = \begin{bmatrix} (a_0 + a_1 x) \cdot \frac{D_1}{g_{11}} & 0 \cdot \frac{D_2}{g_{12}} \\ b_0 \cdot \frac{D_1}{g_{21}} & (c_0 + c_1 x + c_2 x^2) \cdot \frac{D_2}{g_{22}} \end{bmatrix}$$
$$= a_0 \begin{bmatrix} \frac{D_1}{g_{11}} & 0 \\ 0 & 0 \end{bmatrix} + a_1 \begin{bmatrix} \frac{xD_1}{g_{11}} & 0 \\ 0 & 0 \end{bmatrix} + b_0 \begin{bmatrix} 0 & 0 \\ \frac{D_1}{g_{21}} & 0 \end{bmatrix} + \cdots$$

 \dots and so on.

EXAMPLE 7.2

The most general 3×3 matrix which commutes with others.

Let $A \in M_{3\times 3}(\mathbb{Q})$ such that there exists non-singular $P \in M_{3\times 3}(\mathbb{Q})$ with

$$P^{-1}AP = C(x-1) \oplus C((x-1)^2)$$
$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 2 \end{bmatrix} = J, \text{ say},$$

where C(p) denotes the companion matrix of p, as usual.

Then $P = [u_1 \mid u_2 \mid T(u_2)]$ where $T = T_A$ and

$$m_{T,u_1} = x - 1,$$
 $m_{T,u_2} = (x - 1)^2.$

Also $V_3(\mathbb{Q}) = C_{T,u_1} \oplus C_{T,u_2}$. Note that the invariant factors of T are (x-a) and $(x-1)^2$. We find all 3×3 matrices B such that

$$BA = AB,$$

i.e. $T_BT_A = T_AT_B.$

Let $N = T_B$. Then N must satisfy

$$N(u_1) = Bu_1 = c_{11}u_1 + c_{12}u_2 \quad \text{and} \\ N(u_2) = Bu_2 = c_{21}u_1 + c_{22}u_2 \quad \text{where } c_{kl} \in \mathbb{Q}[x].$$
(32)

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Now

$$\left[\deg \gcd(d_k, d_l) \right] = \left[\begin{array}{cc} 1 & 1 \\ 1 & 2 \end{array} \right]$$

SO

$$[c_{kl}] = \left[\begin{array}{cc} a_0 & b_0(x-1) \\ c_0 & d_0 + d_1x \end{array}\right]$$

where a_0 etc. $\in \mathbb{Q}$, so (32) gives

$$Bu_{1} = a_{0}u_{1} + b_{0}(x - 1)u_{2}$$

= $a_{0}u_{1} - b_{0}u_{2} + b_{0}T(u_{2})$ (33)
$$Bu_{2} = c_{0}u_{1} + (d_{0} + d_{1}x)u_{2}$$

$$= c_0 u_1 + d_0 u_2 + d_1 T(u_2).$$
(34)

Noting that

$$m_{T,u_1} = x - 1 \Rightarrow T(u_1) = u_1$$

and
$$m_{T,u_2} = (x - 1)^2 = x^2 - 2x + 1 \Rightarrow T^2(u_2) = 2T(u_2) - u_2,$$

we have from (34) that

$$T(Bu_2) = c_0 T(u_1) + d_0 T(u_2) + d_1 T^2(u_2)$$

= $c_0 u_1 - d_1 u_2 + (d_0 + 2d_1) T(u_2).$

In terms of matrices,

$$B[u_1|u_2|T(u_2)] = [u_1|u_2|T(u_2)] \begin{bmatrix} a_0 & c_0 & c_0 \\ -b_0 & d_0 & -d_1 \\ b_0 & d_1 & d_0 + 2d_1 \end{bmatrix}$$

i.e. $BP = PK$, say
or $B = PKP^{-1}$.

This gives the most general matrix B such that

$$BA = AB.$$

Note: BA = AB becomes

$$PKP^{-1} PJP^{-1} = PJP^{-1} PKP^{-1}$$
$$\Leftrightarrow KJ = JK.$$

7.2 Tensor products and the Byrnes-Gauger theorem

We next apply the Cecioni-Frobenius theorem to derive a third criterion for deciding whether or not two matrices are similar.

DEFINITION 7.1

(Tensor or Kronecker product)

If $A \in M_{m_1 \times n_1}(F)$ and $B \in M_{m_2 \times n_2}(F)$ we define

$$A \otimes B = \begin{bmatrix} a_{11}B & a_{12}B & \cdots \\ \hline a_{21}B & a_{22}B & \cdots \\ \vdots & \vdots & \ddots \end{bmatrix} \in M_{m_1m_2 \times n_1n_2}(F).$$

In terms of elements,

$$(A \otimes B)_{(i,j),(k,l)} = a_{ij}b_{kl}$$

-the element at the intersection of the *i*-th row block, *k*-th row sub-block, and the *j*-th column block, *l*-th column sub-block.⁴

EXAMPLE 7.3

(Tensor-product-taking is obviously far from commutative!)

7.2.1Properties of the tensor product of matrices

- (i) $(tA) \otimes B = A \otimes (tB) = t(A \otimes B), t \in F;$
- (ii) $A \otimes B = 0 \Leftrightarrow A = 0$ or B = 0;
- (iii) $A \otimes (B \otimes C) = (A \otimes B) \otimes C;$
- (iv) $A \otimes (B + C) = (A \otimes B) + (A \otimes C);$

(v) $(B+C) \otimes D = (B \otimes D) + (C \otimes D);$ ⁴That is, the $((i-1)m_2 + k, (j-1)n_2 + l)$ -th element in the tensor product is $a_{ij}b_{kl}$.

- (vi) $(A \otimes B)(C \otimes D) = (AC) \otimes (BD);$
- (vii) $(B \oplus C) \otimes D = (B \otimes D) \oplus (C \otimes D);$
- (viii) $P(A \otimes (B \oplus C))P^{-1} = (A \otimes B) \oplus (A \otimes C)$ for a suitable row permutation matrix P;
- (ix) det $(A \otimes B) = (\det A)^n (\det B)^m$ if A is $m \times m$ and B is $n \times n$;
- (x) Let $f(x, y) = \sum_{i=0}^{m} \sum_{j=0}^{n} c_{ij} x^{i} y^{j} \in F[x, y]$ be a polynomial in x and y over F and define

$$f(A;B) = \sum_{i=0}^{m} \sum_{j=0}^{n} c_{ij} (A^i \otimes B^j)$$

Then if $\operatorname{ch}_A = \prod_{k=1}^s (x - \lambda_k)$ and $\operatorname{ch}_B = \prod_{l=1}^t (x - \mu_l)$, we have $\operatorname{ch}_{f(A;B)} = \prod_{k=1}^s \prod_{l=1}^t (x - f(\lambda_k, \mu_l));$

- (xi) Taking f(x, y) = xy gives

$$\operatorname{ch}_{A\otimes B} = \prod_{k=1}^{s} \prod_{l=1}^{t} (x - \lambda_k \mu_l);$$

(xii) Taking f(x, y) = x - y gives

$$\operatorname{ch}_{(A\otimes I_n - I_m \otimes B)} = \prod_{k=1}^s \prod_{l=1}^t (x - (\lambda_k - \mu_l));$$

Remark: (ix) can be proved using the uniqueness theorem for alternating m-linear functions met in MP174; (x) follows from the the equations

$$P^{-1}AP = J_1 \quad \text{and} \quad Q^{-1}BQ = J_2,$$

where J_1 and J_2 are the Jordan forms of A and B, respectively. Then J_1 and J_2 are lower triangular matrices with the eigenvalues λ_k , $1 \leq k \leq m$ and μ_l , $1 \leq l \leq n$ of A and B as diagonal elements.

Then

$$P^{-1}A^iP = J_1^i$$
 and $Q^{-1}B^jQ = J_2^j$

and more generally

$$(P \otimes Q)^{-1} \sum_{i=0}^{s} \sum_{j=0}^{t} c_{ij} (A^{i} \otimes B^{j}) (P \otimes Q) = \sum_{i=0}^{s} \sum_{j=0}^{t} c_{ij} (J_{1}^{i} \otimes J_{2}^{j}).$$

The matrix on the right–hand side is lower triangular and has diagonal elements

$$f(\lambda_k, \mu_l), 1 \le k \le m, 1 \le l \le n.$$

THEOREM 7.3

Let β be the standard basis for $M_{m \times n}(F)$ —i.e. the basis consisting of the matrices

$$E_{11},\ldots,E_{mn}$$

and γ be the standard basis for $M_{p \times n}(F)$. Let A be $p \times m$, and

$$T_1: M_{m \times n}(F) \mapsto M_{p \times n}(F)$$

be defined by $T_1(X) = AX$. Then

$$[T_1]^{\gamma}_{\beta} = A \otimes I_n.$$

Similarly if B is $n \times p$, and

$$T_2: M_{m \times n}(F) \mapsto M_{m \times p}(F)$$

is defined by $T_2(Y) = YB$, then

$$[T_2]^{\delta}_{\beta} = A \otimes I_n$$

(where δ is the standard basis for $M_{m \times p}(F)$).

proof Left for the intrepid reader. A hint:

$$E_{ij}E_{kl} = \begin{cases} 0 & \text{if } j \neq k, \\ E_{il} & \text{if } j = k \end{cases}$$

COROLLARY 7.2

Let A be $m \times m$, B be $n \times n$, X be $m \times n$, and

$$T: M_{m \times n}(F) \mapsto M_{m \times n}(F)$$

be defined by T(X) = AX - XB. Then

$$[T]^{\beta}_{\beta} = A \otimes I_n - I_m \otimes B^t,$$

where β is the standard basis for $M_{m \times n}(F)$.

DEFINITION 7.2

For brevity in the coming theorems, we define

$$\nu_{A,B} = \nu(A \otimes I_n - I_m \otimes B^t)$$

where A is $m \times m$ and B is $n \times n$.

THEOREM 7.4

$$\nu_{A,B} = \nu(A \otimes I_n - I_m \otimes B^t)$$
$$= \sum_{k=1}^s \sum_{l=1}^t \deg \gcd(d_k, D_l)$$

where

$$d_1 \mid d_2 \mid \dots \mid d_s$$
 and
 $D_1 \mid D_2 \mid \dots \mid D_t$

are the invariant factors of A and B respectively.

proof With the transformation T from corollary 7.2 above, we note that

$$\nu_{A,B} = \text{nullity } T$$

= dim{ X \in M_{m \times n}(F) | AX = XB }
= dim{ N \in Hom (V_n(F), V_m(F)) | T_AN = NT_B }

and the Cecioni-Frobenius theorem gives the result.

LEMMA 7.1 (Byrnes-Gauger)

(This is needed in the proof of the Byrnes-Gauger theorem following.) Suppose we have two monotonic increasing integer sequences:

$$\begin{cases} m_1 \leq m_2 \leq \cdots \leq m_s & \text{and} \\ n_1 \leq n_2 \leq \cdots \leq n_s \end{cases}$$

Then

$$\sum_{k=1}^{s} \sum_{l=1}^{s} \left\{ \min(m_k, m_l) + \min(n_k, n_l) - 2\min(m_k, n_l) \right\} \ge 0$$

Further, equality occurs iff the sequences are identical.

proof

Case 1: k = l.

The terms to consider here are of the form

$$m_k + n_k - 2\min(m_k, n_k)$$

which is obviously ≥ 0 . Also, the term is equal to zero iff $m_k = n + k$. Case 2: $k \neq l$; without loss of generality take k < l.

Here we pair the off-diagonal terms (k, l) and l, k.

$$\{\min(m_k, m_l) + \min(n_k, n_l) - 2\min(m_k, n_l)\} + \{\min(m_l, m_k) + \min(n_l, n_k) - 2\min(m_l, n_k)\} = \{m_k + n_l - 2\min(m_k, n_l)\} + \{m_l + n_k - 2\min(m_l, n_k)\} \ge 0, \quad \text{obviously.}$$

Since the sum of the diagonal terms and the sum of the pairs of sums of off-diagonal terms are non-negative, the sum is non-negative. Also, if the sum is zero, so must be the sum along the diagonal terms, making

$$m_k = n_k \qquad \forall k.$$

THEOREM 7.5 (Byrnes-Gauger)

If A is $m \times m$ and B is $n \times n$ then

$$\nu_{A,A} + \nu_{B,B} \ge 2\nu_{A,B}$$

with equality if and only if m = n and A and B are similar.

 proof

$$\nu_{A,A} + \nu_{B,B} - 2\nu_{A,B}$$

$$= \sum_{k_1=1}^{s} \sum_{k_2=1}^{s} \deg \gcd(d_{k_1}, d_{k_2}) + \sum_{l_1=1}^{t} \sum_{l_2=1}^{t} \deg \gcd(D_{l_1}, D_{l_2})$$

$$-2\sum_{k=1}^{s} \sum_{l=1}^{t} \deg \gcd(d_k, D_l).$$

We now extend the definitions of d_1, \ldots, d_s and D_1, \ldots, D_t by renaming them as follows, with $N = \max(s, t)$:

and
$$\underbrace{\underbrace{1,\ldots,1}_{N-s}}_{N-t} D_1,\ldots,D_t \mapsto F_1,\ldots,F_N$$
.

This is so we may rewrite the above sum of three sums as a single sum, viz:

$$\nu_{A,A} + \nu_{B,B} - 2\nu_{A,B} = \sum_{k=1}^{N} \sum_{l=1}^{N} \{\deg \gcd(f_k, f_l) + \deg \gcd(F_k, F_l) - 2 \deg \gcd(f_k, F_l)\}.$$
(35)

We now let p_1, \ldots, p_r be the distinct monic irreducibles in $m_A m_B$ and write

$$\begin{cases} f_k = p_1^{a_{k1}} & p_2^{a_{k2}} & \dots & p_r^{a_{kr}} \\ F_k = p_1^{b_{k1}} & p_2^{b_{k2}} & \dots & p_r^{b_{kr}} \end{cases}$$
 $1 \le k \le N$

where the sequences $\{a_{ki}\}_{i=1}^r$, $\{b_{ki}\}_{i=1}^r$ are monotonic increasing non-negative integers. Then

$$gcd(f_k, F_l) = \prod_{i=1}^r p_i^{\min(a_{ki}, b_{li})}$$

$$\Rightarrow \qquad \deg gcd(f_k, F_l) = \sum_{i=1}^r \deg p_i \min(a_{ki}, b_{li})$$
and
$$\deg gcd(f_k, f_l) = \sum_{i=1}^r \deg p_i \min(a_{ki}, a_{li})$$
and
$$\deg gcd(F_k, F_l) = \sum_{i=1}^r \deg p_i \min(b_{ki}, b_{li}).$$

Then equation (35) may be rewritten as

$$\nu_{A,A} + \nu_{B,B} - 2\nu_{A,B}$$

$$= \sum_{k=1}^{N} \sum_{l=1}^{N} \sum_{i=1}^{r} \deg p_{i} \{\min(a_{ki}, a_{li}) + \min(b_{ki}, b_{li}) -2\min(a_{ki}, b_{li})\}$$

$$= \sum_{i=1}^{r} \deg p_{i} \sum_{k=1}^{N} \sum_{l=1}^{N} \{\min(a_{ki}, a_{li}) + \min(b_{ki}, b_{li}) -2\min(a_{ki}, b_{li})\}.$$

The latter double sum is of the form in lemma 7.1 and so, since deg $p_i > 0$, we have

$$\nu_{A,A} + \nu_{B,B} - 2\nu_{A,B} \ge 0,$$

proving the first part of the theorem.

Next we show that equality to zero in the above is equivalent to similarity of the matrices:

$$\nu_{A,A} + \nu_{B,B} - 2\nu_{A,B} = 0$$

$$\Leftrightarrow \sum_{i=1}^{r} \deg p_i \sum_{k=1}^{N} \sum_{l=1}^{N} \{\min(a_{ki}, a_{li}) + \min(b_{ki}, b_{li}) - 2\min(a_{ki}, b_{li})\} = 0$$

- $\Leftrightarrow~$ sequences $\{a_{ki}\},\{b_{ki}\}$ identical (by lemma 7.1)
- \Leftrightarrow A and B have same invariant factors
- \Leftrightarrow A and B are similar ($\Rightarrow m = n$).

EXERCISE 7.1

Show if if

$$P^{-1}A_1P = A_2$$
 and $Q^{-1}B_1Q = B_2$

then

$$(P^{-1} \otimes Q^{-1})(A_1 \otimes I_m - I_n \otimes B_1^t)(P \otimes Q) = A_2 \otimes I_n - I_m \otimes B_2^t.$$

(This is another way of showing that if A and B are similar then

$$\nu_{A,A} + \nu_{B,B} - 2\nu_{A,B} = 0.$$

8 Further directions in linear algebra

- 1. Dual space of a vector space; Tensor products of vector spaces; exterior algebra of a vector space. See C.W. Curtis, *Linear Algebra, an introductory approach* and T.S. Blyth, *Module theory*.
- 2. Quadratic forms, positive definite matrices (see L. Mirsky, Introduction to linear Algebra), singular value decomposition (see G. Strang, Linear Algebra).
- 3. Iterative methods for finding inverses and solving linear systems. See D.R. Hill and C.B. Moler, *Experiments in Computational Matrix Algebra*.
- 4. Positive matrices and Markov matrices are important in economics and statistics. For further reading on the structure of Markov matrices and more generally, non–negative matrices, the following books are recommended:
 - N.J. Pullman. Matrix Theory and its Applications, 1976. Marcel Dekker Inc. New York.
 - [2] M. Pearl. Matrix Theory and Finite Mathematics, 1973. McGraw-Hill Book Company, New York.
 - [3] H. Minc. Nonnegative Matrices, 1988. John Wiley and Sons, New York.
- 5. There are at least two research journals devoted to linear and multilinear algebra in our Physical Sciences Library: *Linear and Multilinear Algebra* and *Linear Algebra and its applications*.