# CHAPTER 1

# GTLA

## 1.1. Matrices and operations

Let  $\mathbb{F}$  be a field. Let  $m, n \in \mathbb{Z}_{>0}$ .

• An  $m \times n$  matrix with entries in  $\mathbb{F}$  is a table of elements of  $\mathbb{F}$  with m rows and n columns. More precisely, an  $m \times n$  matrix with entries in  $\mathbb{F}$  is a function

 $A: \{1, \dots, m\} \times \{1, \dots, n\} \longrightarrow \mathbb{F}.$ 

- A column vector of length n is an  $n \times 1$  matrix.
- A row vector of length n is an  $1 \times n$  matrix.
- The (i, j) entry of a matrix A is the element A(i, j) in row i and column j of A.

$$A = \begin{pmatrix} A(1,1) & A(1,2) & \cdots & A(1,m) \\ A(2,1) & A(2,2) & \cdots & A(2,m) \\ \vdots & & & \vdots \\ A(n,1) & A(n,2) & \cdots & A(n,m) \end{pmatrix}$$

Let  $M_{m \times n}(\mathbb{F})$  be the set of  $m \times n$  matrices with entries in  $\mathbb{F}$ .

• The sum of  $m \times n$  matrices A and B is the  $m \times n$  matrix A + B given by

(A+B)(i,j) = A(i,j) + B(i,j), for  $i \in \{1, \dots, m\}$  and  $j \in \{1, \dots, n\}.$ 

• The scalar multiplication of an element  $c \in \mathbb{F}$  with an  $m \times n$  matrix A is the  $m \times n$  matrix  $c \cdot A$  given by

$$(c \cdot A)(i, j) = c \cdot A(i, j),$$
 for  $i \in \{1, \dots, m\}$  and  $j \in \{1, \dots, n\}$ 

• The *product* of an  $m \times n$  matrix A and an  $n \times p$  matrix B is the  $m \times p$  matrix AB given by

$$(AB)(i,k) = \sum_{j=1}^{n} A(i,j)B(j,k)$$
  
=  $A(i,1)B(1,k) + A(i,2)B(2,k) + \dots + A(i,n)B(n,k),$ 

for  $i \in \{1, ..., m\}$  and  $k \in \{1, ..., p\}$ .

The zero matrix is the  $m \times n$  matrix  $0 \in M_{m \times n}(\mathbb{F})$  given by

$$0(i, j) = 0,$$
 for  $i \in \{1, \dots, m\}$  and  $j \in \{1, \dots, n\}$ .

The negative of a matrix  $A \in M_{m \times n}(\mathbb{F})$  is the matrix  $-A \in M_{m \times n}(\mathbb{F})$  given by

$$(-A)(i,j) = -A(i,j),$$
 for  $i \in \{1,\ldots,m\}$  and  $j \in \{1,\ldots,n\}.$ 

The following proposition says that  $M_{m \times n}(\mathbb{F})$  is an  $\mathbb{F}$ -vector space.

**Proposition 1.1.1.** — Let  $m, n \in \mathbb{Z}_{>0}$  and let  $M_{m \times n}(\mathbb{F})$  be the set of  $m \times n$  matrices with entries in  $\mathbb{F}$ .

(a) If  $A, B, C \in M_{m \times n}(\mathbb{F})$  then A + (B + C) = (A + B) + C.

- (b) If  $A, B \in M_{m \times n}(\mathbb{F})$  then A + B = B + A.
- (c) If  $A \in M_{m \times n}(\mathbb{F})$  then 0 + A = A and A + 0 = A.
- (d) If  $A \in M_{m \times n}(\mathbb{F})$  then (-A) + A = 0 and A + (-A) = 0.
- (e) If  $A \in M_{m \times n}(\mathbb{F})$  and  $c_1, c_2 \in \mathbb{F}$  then  $c_1 \cdot (c_2 \cdot A) = (c_1 c_2) \cdot A$ .
- (f) If  $A \in M_{m \times n}(\mathbb{F})$  and  $1 \in \mathbb{F}$  is the identity in  $\mathbb{F}$  then  $1 \cdot A = A$ .

The Kronecker delta is given by

$$\delta_{ij} = \begin{cases} 1, & \text{if } i = j, \\ 0, & \text{otherwise} \end{cases}$$

The *identity matrix* is the  $n \times n$  matrix  $1 \in M_{n \times n}(\mathbb{F})$  given by

$$1(i, j) = \delta_{ij},$$
 for  $i \in \{1, ..., m\}$  and  $j \in \{1, ..., n\}$ 

The following proposition says that  $M_n(\mathbb{F})$  is a ring (usually noncommutative).

**Proposition 1.1.2.** — Let  $n \in \mathbb{Z}_{>0}$  and let  $M_n(\mathbb{F})$  be the set of  $n \times n$  matrices in  $\mathbb{F}$ . (a) If  $A, B, C \in M_n(\mathbb{F})$  then A + (B + C) = (A + B) + C.

- (b) If  $A, B \in M_n(\mathbb{F})$  then A + B = B + A.
- (c) If  $A \in M_n(\mathbb{F})$  then 0 + A = A and A + 0 = A.
- (d) If  $A \in M_n(\mathbb{F})$  then (-A) + A = 0 and A + (-A) = 0.
- (e) If  $A, B, C \in M_n(\mathbb{F})$  then A(BC) = (AB)C.
- (f) If  $A, B, C \in M_n(\mathbb{F})$  then (A+B)C = AC + BC and C(A+B) = CA + CB.
- (g) If  $A \in M_n(\mathbb{F})$  then 1A = A and A1 = A.

The transpose of an  $m \times n$  matrix A is the  $n \times m$  matrix  $A^t$  given by

$$A^{t}(i, j) = A(j, i),$$
 for  $i \in \{1, ..., n\}$  and  $j \in \{1, ..., m\}$ 

The following proposition says that transpose is an antiautomorphism of the ring  $M_n(\mathbb{F})$ .

**Proposition 1.1.3.** — Let  $m, n \in \mathbb{Z}_{>0}$ , let  $M_{m \times n}(\mathbb{F})$  be the set of  $m \times n$  matrices with entries in  $\mathbb{F}$ , and let  $M_n(\mathbb{F})$  be the set of  $n \times n$  matrices in  $\mathbb{F}$ .

- (a) If  $A, B \in M_{m \times n}(\mathbb{F})$  then  $(A + B)^t = A^t + B^t$ ,
- (b) If  $A \in M_{m \times n}(\mathbb{F})$  and  $c \in \mathbb{F}$  then  $(c \cdot A)^t = c \cdot A^t$ ,
- (c) If  $A, B \in M_n(\mathbb{F})$  then  $(AB)^t = B^t A^t$ .
- (d) If  $A \in M_n(\mathbb{F})$  then  $(A^t)^t = A$ .

#### 1.2. Vector spaces and linear transformations

Let  $\mathbb{F}$  be a field. A  $\mathbb{F}$ -vector space is a set V with functions

$$\begin{array}{ccccc} V \times V & \to & V \\ (v_1, v_2) & \mapsto & v_1 + v_2 \end{array} \quad \text{and} \quad \begin{array}{cccccc} \mathbb{K} \times V & \to & V \\ (c, v) & \mapsto & cv \end{array}$$

(addition and scalar multiplication) such that

- (a) If  $v_1, v_2, v_3 \in V$  then  $(v_1 + v_2) + v_3 = v_1 + (v_2 + v_3)$ ,
- (b) There exists  $0 \in V$  such that if  $v \in V$  then 0 + v = v and v + 0 = v,
- (c) If  $v \in V$  then there exists  $-v \in V$  such that v + (-v) = 0 and (-v) + v = 0,
- (d) If  $v_1, v_2 \in V$  then  $v_1 + v_2 = v_2 + v_1$ ,
- (e) If  $c \in \mathbb{F}$  and  $v_1, v_2 \in V$  then  $c(v_1 + v_2) = cv_1 + cv_2$ ,
- (f) If  $c_1, c_2 \in \mathbb{F}$  and  $v \in V$  then  $(c_1 + c_2)v = c_1v + c_2v$ ,
- (g) If  $c_1, c_2 \in \mathbb{F}$  and  $v \in V$  then  $c_1(c_2v) = (c_1c_2)v$ ,
- (h) If  $v \in V$  then 1v = v.

Linear transformations are for comparing vector spaces.

Let  $\mathbb{F}$  be a field and let V and W be  $\mathbb{F}$ -vector spaces. A linear transformation from V to W is a function  $f: V \to W$  such that

- (a) If  $v_1, v_2 \in V$  then  $f(v_1 + v_2) = f(v_1) + f(v_2)$ ,
- (b) If  $c \in \mathbb{F}$  and  $v \in V$  then f(cv) = cf(v).

One vector space can be a subspace of another.

Let V be an  $\mathbb{F}$ -vector space. A subspace of V is a subset  $W \subseteq V$  such that

- (a) If  $w_1, w_2 \in W$  then  $w_1 + w_2 \in W$ ,
- (b)  $0 \in W$ ,
- (c) If  $w \in W$  then  $-w \in W$ ,
- (d) If  $w \in W$  and  $c \in \mathbb{F}$  then  $cw \in W$ .

The tiniest vector space is the zero space.

The zero space, (0), is the set containing only 0 with the operations 0 + 0 = 0 and  $c \cdot 0$ , for  $c \in \mathbb{F}$ .

### 1.3. Kernels and images

The kernel, or null space, of a linear transformation  $f: V \to W$  is the set

$$\ker(f) = \{ v \in V \mid f(v) = 0 \}.$$

The *image* of a linear transformation  $f: V \to W$  is the set

$$\operatorname{im}(f) = \{ f(v) \mid v \in V \}.$$

**Proposition 1.3.1**. — Let  $f: V \to W$  be a linear transformation. Then

- (a) ker f is a subspace of V, and
- (b) im f is a subspace of W.

Let S and T be sets and let  $f: S \to T$  be a function. The function  $f: S \to T$  is *injective* if f satisfies:

if  $s_1, s_2 \in S$  and  $f(s_1) = f(s_2)$  then  $s_1 = s_2$ .

The function  $f: S \to T$  is surjective if f satisfies:

if  $t \in T$  then there exists  $s \in S$  such that f(s) = t.

**Proposition 1.3.2**. — Let  $f: V \to W$  be a linear transformation. Then

(a) ker  $f = \{0\}$  if and only if f is injective, and

(b) im f = W if and only if f is surjective.

The rank of a linear transformation  $f: V \to W$  is the dimension of the image of f and the *nullity* of a linear transformation f is the dimension of the kernel of f,

 $\operatorname{rank}(f) = \operatorname{dim}(\operatorname{im}(f))$  and  $\operatorname{nullity}(f) = \operatorname{dim}(\operatorname{ker}(f)).$ 

#### 1.4. Bases and dimension

Let  $\mathbb{F}$  be a field and let V be a vector space over  $\mathbb{F}$ . Let  $\{v_1, v_2, \ldots, v_k\}$  be a subset of V.

• The span of the set  $\{v_1, \ldots, v_k\}$  is

$$\operatorname{span}\{v_1, \dots, v_k\} = \{c_1v_1 + c_2v_2 + \dots + c_kv_k \mid c_1, c_2, \dots, c_k \in \mathbb{F}\}.$$

- A linear combination of  $v_1, v_2, \ldots, v_k$  is an element of span $\{v_1, \ldots, v_k\}$ .
- The set  $\{v_1, \ldots, v_k\}$  is *linearly independent* if it satisfies:

if  $c_1, \ldots, c_k \in \mathbb{F}$  and  $c_1v_1 + \cdots + c_kv_k = 0$  then  $c_1 = 0, c_2 = 0, \ldots, c_k = 0$ .

- A basis of V is a subset  $B \subseteq V$  such that
  - (a)  $\operatorname{span}(B) = V$ ,
  - (b) B is linearly independent.
- The dimension of V is the cardinality (number of elements) of a basis of V.

**Theorem 1.4.1.** — (Characterization of a basis) Let V be a vector space and let B be a subset of V. The following are equivalent:

(a) B is a basis of V;

(b) B is a minimal element of  $\{S \subseteq V \mid \text{span}(S) = V\}$ , ordered by inclusion;

(c) B is a maximal element of  $\{L \subseteq V \mid L \text{ is linearly independent}\}$ , ordered by inclusion.

**Theorem 1.4.2.** — (Existence of a basis) Let V be a vector space over a field  $\mathbb{F}$ . Then (a) V has a basis, and

(b) Any two bases of V have the same number of elements.

### 1.5. Addition, scalar multiplication and composition of linear transformations

The sum of two linear transformations  $f_1: V \to W$  and  $f_2: V \to W$  is the linear transformation  $(f_1 + f_2): V \to W$ .

$$(f_1 + f_2)(v) = f_1(v) + f_2(v), \quad \text{for } v \in V.$$

Let  $f: V \to W$  be a linear transformation and let  $c \in \mathbb{F}$ . The scalar multiplication of f by c is the linear transformation  $(cf): V \to W$  given by

$$(cf)(v) = c \cdot f(v), \quad \text{for } v \in V.$$

The composition of a linear transformation  $f_2: V \to W$  and a linear transformation  $f_1: W \to Z$  is the linear transformation  $(f_1 \circ f_2): V \to Z$  given by

$$(f_1 \circ f_2)(v) = f_1(f_2(v)), \quad \text{for } v \in V$$

#### 1.6. Matrices of linear transformations and change of basis matrices

Let V and W be  $\mathbb{F}$ -vector spaces. Let B be a basis of V and let C be a basis of W. Let  $f: V \to W$  be a linear transformation. The matrix of  $f: V \to W$  with respect to the bases B and C is the matrix

$$f_{CB} \in M_{C \times B}(\mathbb{F})$$
 given by  $f(b) = \sum_{c \in C} f_{CB}(c, b)c$  for  $b \in B$ 

(here we view matrices in  $M_{C\times B}(\mathbb{F})$  as functions  $A: C \times B \to \mathbb{F}$  so that the (c, b) entry of the matrix A is the value A(c, b)).

**Proposition 1.6.1**. — Let V and W and Z be  $\mathbb{F}$ -vector spaces with bases B, C and D, respectively. Let

$$f: V \to W, \quad g: V \to W, \quad h: W \to Z$$
 be linear transformations

and let  $c \in \mathbb{F}$ . Then

$$(cf)_{CB} = c \cdot f_{CB}, \qquad f_{CB} + g_{CB} = (f+g)_{CB} \qquad and \qquad (h \circ g)_{DB} = h_{DC}g_{CB}.$$

Let V be an  $\mathbb{F}$ -vector space and let B and C be bases of V. The change of basis matrix from B to C is the matrix  $P_{CB} \in M_{C \times B}(\mathbb{F})$  given by

(1.6.1) 
$$b = \sum_{c \in C} P_{CB}(c, b)c, \quad \text{for } b \in B$$

**Proposition 1.6.2.** — Let  $g: V \to W$  and  $f: V \to V$  be linear transformations. Let

 $B_1$  and  $B_2$  be bases of V, and let  $C_1$  and  $C_2$  be bases of W,

and let  $P_{B_1B_2}$  and  $P_{C_2C_1}$  be the change of basis matrices defined as in (1.6.1). Then

$$g_{C_2B_2} = P_{C_2C_1}g_{C_1B_1}P_{B_1B_2}$$
 and  $f_{B_2B_2} = P_{B_1B_2}^{-1}f_{B_1B_1}P_{B_1B_2}$ .

**Proposition 1.6.3.** — Let  $P \in M_n(\mathbb{F})$ . The matrix P is invertible if and only if the columns of P are linearly independent in  $\mathbb{F}^n$ .

**1.6.1.** Minimal and characteristic polynomials (annihilators of  $\mathbb{F}[x]$ -modules). — Let  $A \in M_n(\mathbb{F})$ . Let

 $\varphi_A \colon \mathbb{F}[x] \longrightarrow M_n(\mathbb{F})$  $c_0 + c_1 x + \dots + c_r x^r \mapsto c_0 + c_1 A + \dots + c_r A^r$ 

The kernel of  $\varphi_A$  is

$$\ker(\varphi_A) = \{ p(x) \in \mathbb{F}[x] \mid \varphi_A(p(x)) = 0. \}$$

**Proposition 1.6.4.** — There exists a unique monic polynomial  $m(x) \in \mathbb{F}[x]$  such that  $\ker(\varphi_A) = m(x)\mathbb{F}[x]$ .

Let  $A \in M_n(\mathbb{F})$ .

• The minimal polynomial of A is the monic polynomial  $m(x) \in \mathbb{F}[x]$  such that

$$\ker \varphi_A = m(x)\mathbb{F}[x].$$

• The matrix  $x - A \in M_n(\mathbb{F}[x])$ . The characteristic polynomial of A is det(x - A).

**Proposition 1.6.5.** — (Cayley-Hamlton theorem) Let  $A \in M_n(\mathbb{F})$  and let m(x) be the minimal polynomial of A. Then

$$\det(x - A) \in m(x)\mathbb{F}[x].$$

**1.6.2.** Diagonalization (simple and semisimple  $\mathbb{F}[x]$ -modules). — Let  $\mathbb{F}$  be a field and let  $A \in M_n(\mathbb{F})$ .

• A subspace  $U \subseteq \mathbb{F}^n$  is A-invariant, or U is an A-submodule of  $\mathbb{F}^n$ , if U satisfies:

if 
$$u \in U$$
 then  $Au \in U$ .

• Let  $\lambda \in \mathbb{F}$ . An eigenvector of A of eigenvalue  $\lambda$  is  $p \in \mathbb{F}^n$  such that  $p \neq 0$  and

$$Ap = \lambda p.$$

• The matrix A is *diagonalizable* if there exist  $P \in GL_n(\mathbb{F})$  and  $\lambda_1, \ldots, \lambda_n \in \mathbb{F}$  such that

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

**HW**: Show that p is an eigenvector of A if and only if  $\mathbb{F}p$  is A-invariant.

**HW**: Show that p is an eigenvector of A if and only if  $p \in \ker(A - \lambda)$ .

**HW**: Show that if  $D = \text{diag}(\lambda_1, \ldots, \lambda_n)$  and  $P^{-1}AP = D$  then

$$det(A) = \lambda_1 \cdots \lambda_n$$
 and  $det(x - A) = (x - \lambda_1) \cdots (x - \lambda_n).$ 

**Proposition 1.6.6**. — Let  $\mathbb{F}$  be a field and let  $A \in M_n(\mathbb{F})$ .

(a) If  $p_1, \ldots, p_k$  are eigenvectors of A with eigenvalues  $\lambda_1, \ldots, \lambda_k$  and  $\lambda_1, \ldots, \lambda_k$  are all distinct then  $p_1, \ldots, p_k$  are linearly independent.

(b) Let  $A \in M_n(\mathbb{F})$ . Then A is diagonalizable if and only if there exist n linearly independent eigenvectors of A.

(c) If  $\mathbb{F}$  is algebraically closed then A has an eigenvector.

### 1.6.3. Some proofs. -

**Proposition 1.6.7.** — Let V and W and Z be  $\mathbb{F}$ -vector spaces with bases B, C and D, respectively. Let

 $f: V \to W, \quad g: V \to W, \quad h: W \to Z$  be linear transformations and let  $c \in \mathbb{F}$ . Then

$$(cf)_{CB} = c \cdot f_{CB}, \qquad f_{CB} + g_{CB} = (f+g)_{CB} \qquad and \qquad (h \circ g)_{DB} = h_{DC}g_{CB}$$

*Proof.* — Let  $b \in B$  and  $c' \in C$ . Taking the coefficient of c' on each side of

$$\sum_{c \in C} (\alpha f)_{CB}(c, b)c = (\alpha f)(b) = \alpha \cdot f(b) = \alpha \cdot \left(\sum_{c \in C} f_{CB}(c, b)c\right) = \sum_{c \in C} \alpha f_{CB}(c, b)c$$

gives  $(\alpha f)_{CB}(c', b) = \alpha \cdot f_{CB}(c', b)$ . So  $(\alpha f)_{CB} = \alpha \cdot f_{CB}$ .

Let  $b \in B$  and  $c' \in C$ . Taking the coefficient of c' on each side of

$$\sum_{c \in C} (f+g)_{CB}(c,b)c = (f+g)(b) = f(b) + g(b) = \sum_{c \in C} (f_{CB}(c,b)c + \sum_{c \in C} g_{CB}(c,b)c) = \sum_{c \in C} (f_{CB}(c,b)c + g_{CB}(c,b)c + g_{CB}(c,b)c) = \sum_{c \in C} (f_{CB}(c,b) + g_{CB}(c,b)$$

gives  $(f_{CB} + g_{CB})(c', b) = f_{CB}(c', b) + g_{CB}(c', b)$ . So  $f_{CB} + g_{CB} = (f + g)_{CB}$ .

Let  $b \in B$  and  $d' \in D$ . Taking the coefficient of d' on each side of

$$\sum_{d \in D} (h \circ g)_{DB}(d, b)d = (h \circ g)(b) = h(g(b)) = h\left(\sum_{c \in C} g_{CB}(c, b)c\right)$$
$$= \sum_{c \in C} g_{CB}(c, b)h(c) = \sum_{c \in C} \sum_{d \in D} g_{CB}(c, b)h_{DC}(d, c)d,$$

gives  $(h \circ g)_{DB}(d', b) = \sum_{c \in C} \sum_{d \in D} h_{DC}(d, c)g_{CB}(c, b) = (h_{DC}g_{CB})(d', b).$ So  $(h \circ g)_{DB} = (h_{DC}g_{CB}).$ 

**Proposition 1.6.8**. — Let  $g: V \to W$  and  $f: V \to V$  be linear transformations. Let

 $B_1$  and  $B_2$  be bases of V, and let  $C_1$  and  $C_2$  be bases of W,

and let  $P_{B_1B_2}$  and  $P_{C_2C_1}$  be the change of basis matrices defined as in (1.6.1). Then

$$g_{C_2B_2} = P_{C_2C_1}g_{C_1B_1}P_{B_1B_2}$$
 and  $f_{B_2B_2} = P_{B_1B_2}^{-1}f_{B_1B_1}P_{B_1B_2}$ .

*Proof.* — Let  $\beta, \beta' \in B_2$ . Comparing coefficients of  $\beta'$  on each side of

$$\beta = \sum_{b \in B_1} P_{B_1 B_2}(b, \beta) b = \sum_{b \in B_1} P_{B_1 B_2}(b, \beta) \sum_{\beta' \in B_2} P_{B_2 B_1}(\beta', b) \beta'$$
$$= \sum_{b \in B_1} \sum_{\beta' \in B_2} P_{B_2 B_1}(\beta', b) P_{B_1 B_2}(b, \beta) \beta' = \sum_{b \in B_1} \sum_{\beta' \in B_2} (P_{B_2 B_1} P_{B_1 B_2})(\beta', \beta) \beta'$$

gives

$$(P_{B_2B_1}P_{B_1B_2})(\beta',\beta) = \delta_{\beta'\beta}.$$

So  $P_{B_2B_1} = P_{B_1B_2}^{-1}$ . Let  $\beta \in B_1$  and  $c \in B_2$ . Taking the coefficient of b' on each side of

$$f(c) = \sum_{c' \in B_2} f_{B_2 B_2}(c', c)c' = \sum_{b' \in B_1} f_{B_2 B_2}(c', c)P_{B_1 B_2}(b', c')b'$$

and

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$$f(c) = f\left(\sum_{b \in B_1} P_{B_1 B_2}(b, c)b\right) = \sum_{b \in B_1} P_{B_1 B_2}(b, c)f(b) = \sum_{b \in B_1} P_{B_1 B_2}(b, c)\sum_{b' \in B_1} f_{B_1 B_1}(b', b)b'$$

gives

$$(P_{B_1B_2}f_{B_2B_2})(\beta,b) = (f_{B_1B_1}P_{B_1B_2})(\beta,b).$$

So

 $P_{B_1B_2}f_{B_2B_2} = f_{B_1B_1}P_{B_1B_2}$  and thus  $f_{B_2B_2} = P_{B_1B_2}^{-1}f_{B_1B_1}P_{B_1B_2}$ . Let  $\gamma' \in C_2$  and  $\beta \in B_2$ . Taking the coefficient of  $\gamma$  on each side of

$$\sum_{\gamma \in C_2} g_{C_2 B_2}(\gamma, \beta) \gamma = g(\beta) = g(\sum_{b \in B_1} P_{B_1 B_2}(b, \beta)b) = \sum_{b \in B_1} P_{B_1 B_2}(b, \beta)g(b)$$

$$= \sum_{b \in B_1} P_{B_1 B_2}(b, \beta) \sum_{c \in C_1} g_{C_1 B_1}(c, b)c$$

$$= \sum_{b \in B_1} P_{B_1 B_2}(b, \beta) \sum_{c \in C_1} g_{C_1 B_1}(c, b) \sum_{\gamma \in C_2} P_{C_2 C_1}(\gamma, c)\gamma$$

$$= \sum_{b \in B_1, c \in C_1, \gamma \in C_2} P_{C_2 C_1}(\gamma, c)g_{C_1 B_1}(c, b)P_{B_1 B_2}(b, \beta)\gamma$$

$$= \sum_{\gamma \in C_2} (P_{C_2 C_1} g_{C_1 B_1} P_{B_1 B_2})(\gamma, \beta)\gamma$$

gives  $g_{C_2B_2}(\gamma',\beta) = (P_{C_2C_1}g_{C_1B_1}P_{B_1B_2})(\gamma',\beta)$ . So  $g_{C_2B_2} = P_{C_2C_2}g_{C_1B_1}P_{B_1B_2}$ .

**Proposition 1.6.9.** — Let  $P \in M_n(\mathbb{F})$ . The matrix P is invertible if and only if the columns of P are linearly independent in  $\mathbb{F}^n$ .

Proof. —

⇒: Assume P is invertible. Let  $p_1, \ldots, p_n$  be the columns of P. To show:  $\{p_1, \ldots, p_n\}$  is linearly independent. Assume  $c_1, \ldots, c_n \in \mathbb{F}$  and  $c_1p_1 + \cdots + c_np_n = 0$ . Let  $c = (c_1, \ldots, c_n)^t \in \mathbb{F}^n$ . Since  $c_1p_1 + \cdots + c_np_n = 0$  then Pc = 0. So  $c_1 = 0, \ldots, c_n = 0$ .  $\Leftarrow$ : Assume the columns of P are linearly independent. To show: There exists  $Q \in M_n(\mathbb{F})$  such that QP = 1. Let  $p_1, \ldots, p_n$  be the columns of P. Since  $B = \{p_1, \ldots, p_n\}$  is linearly independent and dim $(\mathbb{F}^n) = n$  then B is a maximal linearly independent set. Thus, by Theorem 1.4.1, B is a basis.

Let  $S = \{e_1, \ldots, e_n\}$  where  $e_i$  has 1 in the *i*th spot and 0 elsewhere. Then  $P = P_{BS}$ , the change of basis matrix from S to B.

Let  $Q = P_{SB}$ , the change of basis matrix from B to S. Then  $QP = P_{SB}P_{BS} = P_{SS} = 1$ . So P is invertible.

**Proposition 1.6.10**. — There exists a unique monic polynomial  $m(x) \in \mathbb{F}[x]$  such that  $\ker(\varphi_A) = m(x)\mathbb{F}[x]$ .

Let  $r = \min\{\deg(p) \mid p \in \ker(\varphi_A)\}$  and let  $p(x) \in \ker(\varphi_A)$  with  $\deg(p) = r$  and let

$$m(x) = \frac{1}{a_r} p(x),$$
 where  $p(x) = a_r x^r + \dots + a_1 x + a_0.$ 

To show:  $\ker(\varphi_A) = m(x)\mathbb{F}[x]$ . To show: (a)  $\ker(\varphi_A) \subseteq m(x)\mathbb{F}[x]$ . To show: (b)  $\ker(\varphi_A) \supseteq m(x)\mathbb{F}[x]$ .

(a) Assume 
$$f \in \ker(\varphi_A)$$
.  
Then there exist  $q(x), g(x) \in \mathbb{F}[x]$  with  $\deg(g(x)) < r$  such that  
 $f(x) = q(x)m(x) + g(x)$ .

Since  $f(x) \in \ker(\varphi_A)$  and  $q(x)m(x) \in \ker(\varphi_A)$  then  $g(x) \in \ker(\varphi_A)$ . Since  $\deg(g(x)) < r$  then g(x) = 0. So f(x) = q(x)m(x). So  $f(x) \in m(x)\mathbb{F}[x]$ . (b) Let  $f(x) \in m(x)\mathbb{F}[x]$ . To show:  $f(x) \in \ker(\varphi_A)$ . Since  $f(x) \in m(x)\mathbb{F}[x]$  there exists  $q(x) \in \mathbb{F}[x]$  such that f(x) = q(x)m(x). So  $f(A) = q(A)m(A) = q(A) \cdot 0 = 0$ . So  $f(A) \in \ker(\varphi_A)$ . So  $\ker(\varphi_A) = m(x)\mathbb{F}[x]$ .

So  $\operatorname{Ker}(\varphi_A) = \operatorname{In}(x) \operatorname{Ir}[x].$ 

**Proposition 1.6.11.** — (Cayley-Hamlton theorem) Let  $A \in M_n(\mathbb{F})$  and let m(x) be the minimal polynomial of A. Then

$$\det(x - A) \in m(x)\mathbb{F}[x].$$

*Proof.* — Let  $p = \det(x - A)$ . BY CRAMER'S RULE,

(x - A)adj $(x - A) = det(x - A)1_n$ , where  $1_n$  is the  $n \times n$  identity matrix.

Evaluating both sides at A gives that p(A) = 0. So  $p \in \ker(\varphi_A)$ .

**Proposition 1.6.12.** — Let  $\mathbb{F}$  be a field and let  $A \in M_n(\mathbb{F})$ . (a) If  $p_1, \ldots, p_k$  are eigenvectors of A with eigenvalues  $\lambda_1, \ldots, \lambda_k$  and  $\lambda_1, \ldots, \lambda_k$  are all distinct then  $p_1, \ldots, p_k$  are linearly independent. (b) Let  $A \in M_n(\mathbb{F})$ . Then A is diagonalizable if and only if there exist n linearly indepen-

(b) Let  $A \in M_n(\mathbb{F})$ . Then A is diagonalizable if and only if there exist n linearly independent eigenvectors of A.

(c) If  $\mathbb{F}$  is algebraically closed then A has an eigenvector.

Proof. (a) Assume  $c_1p_1 + \dots + c_np_n = 0$ . To show: If  $j \in \{1, \dots, n\}$  then  $c_j = 0$ . Assume  $j \in \{1, \dots, n\}$ . Then  $0 = (A - d_1) \cdots (A - d_{j-1})(A - d_{j+1}) \cdots (A - d_n)(c_1p_1 + \dots + c_np_n)$   $= \cdots$ 

$$= c_j (d_j - d_1) \cdots (d_j - d_{j-1}) (d_j - d_{j+1}) \cdots (d_j - d_n) p_j.$$

So  $c_j p_j = 0$ . So  $c_j = 0$ .

(b) Let  $p_1, \ldots, p_n$  be the columns of P. Then AP = PD gives that  $p_1, \ldots, p_n$  are eigenvectors of A.

Rewriting this equation as

$$AP = PD$$
, where  $D = \operatorname{diag}(\lambda_1, \dots, \lambda_n)$ ,

the eigenvectors of A are the columns of P. By Proposition 1.6.3, P being invertible is equivalent to its n columns being linearly independent.

(c) Since  $\mathbb{F}$  is algebraically closed m(x) factors: there exists  $a_1, \ldots, a_n \in \mathbb{F}$  such that

$$m(x) = (x - a_1) \cdots (x - a_n).$$

Since  $(A-a_2)\cdots(A-a_n) \neq 0$  there exists  $w \in V$  such that  $v = (A-a_2)\cdots(A-a_n)w \neq 0$ . Then  $(A-a_1)(v) = m(A)(w) = 0$ . So  $A(v) = a_1v$ .