

Definition: let F be a field.

$$\text{let } A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n = y_1 \quad \rightarrow \textcircled{1}$$

$$A_{21}x_1 + A_{22}x_2 + \dots + A_{2n}x_n = y_2 \quad \rightarrow \textcircled{2}$$

$$A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n = y_m \quad \rightarrow \textcircled{m}$$

be a system of m linear equations in n unknowns x_1, \dots, x_n

$A_{ij}, y_i \in F$ are known.

Any n tuples (x_1, \dots, x_n) which satisfies each equation in the system is called a solution of the system.

We write the system of equation in matrix form as

$$AX = Y.$$

where A is called coefficient matrix

X is an $n \times 1$ matrix

Y is an $m \times 1$ matrix

$$A = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{bmatrix}; X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$

If $y_1 = y_2 = \dots = y_m = 0$, then the system is called homogeneous.

Linear combination: let $c_1, c_2, \dots, c_m \in F$

$$\textcircled{1} \times c_1 + \textcircled{2} \times c_2 + \dots + \textcircled{m} \times c_m = \Rightarrow$$

$$[c_1 A_{11} + c_2 A_{12} + \dots + c_m A_{m1}] x_1 + [c_2 A_{21} + c_2 A_{22} + \dots + c_m A_{m2}] x_2 + \dots$$

$$+ [c_1 A_{1n} + c_2 A_{2n} + \dots + c_m A_{mn}] x_n = 0$$

is called a linear combination of m equations in the system.

Clearly any solution of the system of equation is also a solution of above linear combination.

Equivalent of the linear equation:-

If each equation in one system is a linear combination of equation in the II system, then the systems are said to be equivalent

Theorem:-

Prove the equivalent system of linear combination equations have exactly the same solution.

Proof:

$$A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n = y_1$$

$$A_{21}x_1 + A_{22}x_2 + \dots + A_{2n}x_n = y_2$$

$$A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n = y_m$$

Linear equations.

$$\text{Let } B_{11}x_1 + B_{12}x_2 + \dots + B_{1n}x_n = y_1$$

$$B_{21}x_1 + B_{22}x_2 + \dots + B_{2n}x_n = y_2$$

⋮

$$B_{k1}x_1 + B_{k2}x_2 + \dots + B_{kn}x_n = y_k$$

be a solution of k linear equations in which each equation is a linear combination of the equations in II system then every solution of II system is a solution of I system.

It may happen that some solution of II system may not be a solution of I system.

This will not happen if each equation in II system is a linear combination of equation in I system.

(ii) if the 2 system are equivalent they have the same solution.

Definition of an $m \times n$ matrix:-

An $m \times n$ matrix over the field F is a function (Rule) from the set of pairs of integers (i, j) ; $1 \leq i \leq m$ into the field F .

Definition:- An ERO [Elementary Row operation] is a function (or) a rule e which associates with each $m \times n$ matrix A an $m \times n$ matrix

$e(A)$

They are three ERO

- (i) Multiplying any Row by a non-zero scalar
- (ii) replacement of r th row by row r plus c times row s ($r \neq s$)
- (iii) Interchange of two rows of A .

Definition:- Row Equivalent

If A and B are $m \times n$ matrix over the field F , then B is said to be row equivalent to A . If B can be obtained from A by a finite sequence of EROs.

Definition:- Row Reduced matrix

An $m \times n$ matrix R is called a row reduced matrix if

1. First non-zero entry in each non-zero row is 1

2. each column of R which has the leading non-zero entry of

some row has all its other entries zero.

Example

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

is an Row-reduced matrix

Definition:- Row-reduced Echelon matrix

An $m \times n$ matrix R is said to be a row-reduced echelon matrix

- if
- R is called row-reduced matrix
 - every zero row of R occurs below every row which has a non-zero entry.
 - If rows $1, 2, \dots, r$ are all the non-zero rows of R and if the leading non-zero entry of row i occurs in column k_i , $i=1, 2, \dots, r$ then $k_1 < k_2 < \dots < k_r$.

Example:-

$$\begin{bmatrix} 0 & 1 & -3 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Theorem:-

If A is an $m \times n$ matrix and $m < n$, then the homogeneous system of equations $Ax = 0$ has a non-trivial solution.

Proof:-

Let A be an $m \times n$ matrix

$$\text{let } m < n \quad \rightarrow \textcircled{1}$$

Let R be the row-reduced echelon matrix which is row equivalent to A .

Then $Ax = 0$ and $Rx = 0$ have the same solutions $\rightarrow \textcircled{2}$

Let r be the number of non-zero rows

$$\text{Then } r \leq m$$

$$\text{But by } \textcircled{1} \quad m < n$$

$$\therefore r < n$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

∴ $Rx = 0$ has a non-trivial solution
 by ② ∴ $Ax = 0$ has a non-trivial solution.

Theorem: If A, B, C are matrices over F and the products BC and $A(BC)$ are defined then so are AB and $(AB)C$ and that $A(BC) = (AB)C$

Proof: Suppose B is an $m \times p$ matrix

BC is defined

∴ C has p rows and BC has m rows

$A(BC)$ is defined

A must have m columns

∴ we take A as an $m \times n$ matrix

A is of type $m \times n$

B is of type $n \times p$

∴ (AB) is defined and is of type $m \times p$

C has p rows

∴ $(AB)C$ is defined

- next we prove:-

$A(BC) = (AB)C$ for that it's enough if we prove

$$[A(BC)]_{ij} = [(AB)C]_{ij}$$

$$\therefore [A(BC)]_{ij} = \sum_y A_{iy} (BC)_{yj}$$

$$= \sum_y A_{iy} (BC)_{ij}$$

$$= \sum_y A_{iy} \sum_s B_{ys} C_{sj}$$

$$= \sum_r \sum_s A_{ir} B_{rs} C_{sj} = \sum_s \left[\sum_r A_{ir} B_{rs} \right] C_{sj}$$

$$= \sum_s (AB)_{is} C_{sj} = (AB)C_{ij}$$

$$\therefore A(BC) = (AB)C$$

Definition: Elementary matrix

An $n \times n$ matrix A is said to be elementary matrix if it can be obtained from a $n \times n$ identity matrix by a single elementary row operation.

Definition: Invertible matrix

Let A be an $n \times n$ matrix. If \exists an $n \times n$ matrix B such that $BA = I$ then B is called a left inverse of A . If \exists an $n \times n$ matrix B , such that $AB = I$, then B is called a right inverse of A .

If $AB = BA = I$, then B is called a 2-side inverse or inverse of A .

Lemma: If A has a left inverse of B and a right inverse C , then

Prove that $B = C$

Proof:-

B is a left of A

$$\therefore BA = I \rightarrow (1)$$

C is a right inverse of A

$$\therefore AC = I \rightarrow (2)$$

To Prove: $B = C$

$$\text{Consider } B = BI$$

$$= B(AC) = (BA)C$$

$$= IC$$

$$\therefore B = C$$

vector addition :-

 $(V, +)$

1. $a + b \in V$ closure Property
2. $a + (b + c) = (a + b) + c$ Associative.
3. $a + 0 = 0 + a = a$
4. $a + (-a) = (-a) + a = 0$
5. $a + b = b + a$

scalar multiplication :-

1. $1 \cdot a = a \quad 1 \in F$
2. $c_1(c_2 a) = (c_1 c_2) a$
3. $c_1(a + b) = c_1 a + c_1 b$
4. $(c_1 + c_2) a = c_1 a + c_2 a$

Definition :- Dimensional

The number of elements in a basis of a vector space V is called

the dimension of V .

if the dimension is finite, V is said to be finite dimensional.

Theorem :-

If V is finite dimension vector space, then prove that any two basis of V have the same number of elements.

Proof :-

V is finite dimensional

\therefore It has a finite basis

Let $\{b_1, b_2, \dots, b_m\}$ be a basis of V .

Let $\{a_1, a_2, \dots, a_n\}$ be another basis of V .

$\{b_1, b_2, \dots, b_m\}$ spans V and $\{a_1, a_2, \dots, a_n\}$ is linearly independent

Then $n \leq m \rightarrow \textcircled{1}$

By the same argument $m \leq n \rightarrow \textcircled{2}$

from $\textcircled{1}$ & $\textcircled{2}$

$$\therefore m = n$$

Theorem 2

If w_1, w_2 are finite dimension sub space of vector space V . Then

$w_1 + w_2$ is finite dimension and $\dim w_1 + \dim w_2 = \dim(w_1 \cap w_2) + \dim(w_1 + w_2)$.

Proof:

w_1, w_2 are sub space of V .

$\therefore w_1 \cap w_2$ is also sub space of V .

Let $S_1 = \{\alpha_1, \alpha_2, \dots, \alpha_k\}$ be a basis of $w_1 \cap w_2$

$$\therefore \dim(w_1 \cap w_2) = k$$

Thus S_1 is a part of a basis $S_2 = \{\alpha_1, \alpha_2, \dots, \alpha_k, \beta_1, \dots, \beta_m\}$ of w_1

Also S_1 is a part of a basis $S_3 = \{\alpha_1, \dots, \alpha_k, \gamma_1, \gamma_2, \dots, \gamma_n\}$ of w_2

The sub space $w_1 + w_2$ is spanned by

$$S = \{\alpha_1, \alpha_2, \dots, \alpha_k, \beta_1, \beta_2, \dots, \beta_m, \gamma_1, \gamma_2, \dots, \gamma_n\} \rightarrow \textcircled{1}$$

Next we prove S is linearly independent.

$$\text{we consider } \sum x_i \alpha_i + \sum y_j \beta_j + \sum z_r \gamma_r = 0$$

$$\Rightarrow \sum z_r \gamma_r = - \left[\sum x_i \alpha_i + \sum y_j \beta_j \right]$$

$$\Rightarrow \sum z_r \gamma_r \in w_2$$

$$\sum x_i \alpha_i + \sum y_j \beta_j \in w_1$$

These 2 are equal there are in $w_1 \cap w_2$

$\sum z_r \gamma_r$ is a linearly combination of $\alpha_1, \alpha_2, \dots, \alpha_k$

$$\text{(ii) } \sum z_r \gamma_r = \sum c_i \alpha_i$$

$$\Rightarrow \begin{cases} c_i = 0 \forall i \\ z_r = 0 \forall r \end{cases}$$

because S_3 is a basis of w_2

Put $x_i = 0, \forall i$ in (2)

$$\text{Then (2)} \Rightarrow \sum x_i \alpha_i + \sum y_j \beta_j = 0$$

$$\Rightarrow \begin{matrix} x_i = 0, \forall i \\ y_j = 0, \forall j \end{matrix} \text{ because } S_3 \text{ is a basis of } W_1$$

$$\therefore x_i = 0$$

$$y_j = 0$$

$$z_r = 0$$

$\Rightarrow S$ is a linearly independent set

$$\therefore \dim(W_1 + W_2) = k + m + n \rightarrow (3)$$

consider $\dim W_1 + \dim W_2 = k + m + k + n$
 $= k + (m + k + n)$
 $= \dim(W_1, W_2) + \dim(W_1 + W_2)$

$$\therefore \dim W_1 + \dim W_2 = \dim(W_1, W_2) + \dim(W_1 + W_2)$$

Ordered Basis :-

Let V be a finite dimension vector space. An ordered basis of V is a finite sequence of vector which is linearly independent and spans V .

Spanned :-

Let S be a set of vectors in V . Then the subspace spanned by S is defined to be the intersection W of all subspace of V which contains S .

Let $S = \{\alpha_1, \alpha_2, \dots, \alpha_n\}$ then W is called subspace spanned by S .

Basis and Dimension

Definition:- Let V be a vector space over F , let $S = \{\alpha_1, \alpha_2, \dots, \alpha_n\}$ be a subspace of V . Then S is said to be a linearly dependent set of

vectors if \exists scalars c_1, c_2, \dots, c_n in F not all zero such that

$$c_1 x_1 + c_2 x_2 + \dots + c_n x_n = 0 \rightarrow \textcircled{1}$$

If S is not linearly independent then it is said to be linearly dependent.

In that case

$$\textcircled{1} \Rightarrow c_1 = c_2 = c_3 = \dots = c_n = 0$$

Definition!:- Basis

Let V be a vector space over F and let S be a subset of V . If

1. S is linearly independent

2. S spans V then S is said to be a basis of V

V is said to be finite dimensional if it has a finite basis

Theorem:-

Prove that row equivalent matrices have the same row space.

Proof:-

Let A be $m \times n$ matrix over F .

Let $\alpha_1, \alpha_2, \dots, \alpha_m$ be the m row vectors of A

These m vectors span a subspace called row space of A .

If B is the row-equivalent matrix to A

Then \exists an $m \times m$ matrix P such that $B = PA \rightarrow \textcircled{1}$

Let $\beta_1, \beta_2, \dots, \beta_m$ be the m row vectors of B

Then from $\textcircled{1}$ we see that each β_i is a linear combination

of $\alpha_1, \alpha_2, \dots, \alpha_m$ and

$$\beta_i = p_{i1}\alpha_1 + p_{i2}\alpha_2 + \dots + p_{im}\alpha_m$$

$\Rightarrow \beta_i$ is in the subspace spanned by $\alpha_1, \alpha_2, \dots, \alpha_m$

\Rightarrow Row space of B is a subspace of row space of A 11
 \hookrightarrow (2)

$$B = PA$$

$$\Rightarrow A = P^{-1}B$$

\Rightarrow row space of A is a subspace of row space of B

By (2) & (3)

row space of $A =$ row space of B .

Linear transformation:

Let V & W be two vector spaces over the same field F . Then a function $T: V \rightarrow W$ is called a linear transformation

if $T(c\alpha + \beta) = cT\alpha + T\beta$, $\forall \alpha, \beta \in V$ and $c \in F$

Theorem: Let V be a vector space over the field F and $\beta = \{\alpha_1, \dots, \alpha_n\}$ be an ordered basis of V . Let W be a vector space over the same field F & let $\beta_1, \beta_2, \dots, \beta_n$ be any n vectors in W .

Then Prove that T is precisely one linear transformation from

V into W defined by $T\alpha_j = \beta_j$, $j = 1, 2, \dots, n$.

Proof:

Let $\alpha \in V$

Since $\beta = \{\alpha_1, \alpha_2, \dots, \alpha_n\}$ is an ordered basis of vectors. Then α be expressed as a unique linear combination of $\alpha_1, \alpha_2, \dots, \alpha_n$

Let $\alpha = x_1\alpha_1 + \dots + x_n\alpha_n$ \rightarrow (1)

Define $T: V \rightarrow W$ such that

$T\alpha = x_1\beta_1 + \dots + x_n\beta_n$ \rightarrow (2)

from (1) & (2), we see that

$T\alpha_j = \beta_j$, $j = 1, 2, \dots, n$.

To Prove: T is a linear transformation from V into W .

Let $\beta \in V$.

Let $\beta = y_1\alpha_1 + y_2\alpha_2 + \dots + y_n\alpha_n$ \rightarrow (3)

Let $c \in F$

consider $c\alpha + \beta = c(x_1\alpha_1 + \dots + x_n\alpha_n) + (y_1\alpha_1 + \dots + y_n\alpha_n)$

$$= (cx_1 + y_1)\alpha_1 + \dots + (cx_n + y_n)\alpha_n \longrightarrow (4)$$

consider $cT\alpha + T\beta = c(x_1\beta_1 + \dots + x_n\beta_n) + (y_1\beta_1 + \dots + y_n\beta_n)$
 $= (cx_1 + y_1)\beta_1 + \dots + (cx_n + y_n)\beta_n \longrightarrow (5)$

By (4) & (5) T is a linear transformation

T is unique:

let $U: V \rightarrow W$ be another linear transformation such that

$$U\alpha_j = \beta_j, \quad j=1, 2, \dots, n$$

let $\alpha \in V$

$$\text{consider } U\alpha = U(x_1\alpha_1 + \dots + x_n\alpha_n)$$

$$= U\left(\sum_{i=1}^n x_i\alpha_i\right)$$

$$= \sum_{i=1}^n x_i U\alpha_i = \sum_{i=1}^n x_i \beta_i$$

$$\therefore U\alpha = T\alpha \Rightarrow U = T$$

$\therefore T$ is unique.

Null space & Range space

let V and W be two vector space over the same field F .

If T is a linear transformation from V into W , then the set

$N = \{\alpha \in V \mid T\alpha = 0\}$ is a subspace of V called the null space of T

let V and W be two vector space over the same field F .

If T is a linear transformation from V into W , then the set

$R_T = \{\beta \in W \mid \beta = T\alpha, \alpha \in V\}$ is a subspace of W called the

Range of T .

Theorem: Let V & W be two vector spaces over F . If T is a linear transformation from V into W , then $\text{rank}(T) + \text{nullity}(T) = \dim V$.

Proof:

Let $\dim V = n$

If $N = \{ \alpha \in V \mid T\alpha = 0 \}$. N is null space of T .

Let $R_T = \{ \beta \mid \beta = T\alpha, \alpha \in V \}$ be the range of T

Let $S_1 = \{ \alpha_1, \alpha_2, \dots, \alpha_k \}$ be a basis of N

Then nullity of $T = k$.

S_1 can be completed to a basis of V as

$$S_2 = \{ \alpha_1, \alpha_2, \dots, \alpha_k, \alpha_{k+1}, \dots, \alpha_n \}$$

Let $\alpha \in V$. Then $\alpha = c_1\alpha_1 + c_2\alpha_2 + \dots + c_k\alpha_k + c_{k+1}\alpha_{k+1} + \dots + c_n\alpha_n$.

$$T\alpha = c_1T\alpha_1 + c_2T\alpha_2 + \dots + c_kT\alpha_k + c_{k+1}T\alpha_{k+1} + \dots + c_nT\alpha_n$$

$$= 0 + 0 + \dots + 0 + c_{k+1}T\alpha_{k+1} + \dots + c_nT\alpha_n$$

$$\therefore T\alpha = c_{k+1}T\alpha_{k+1} + \dots + c_nT\alpha_n$$

$\Rightarrow S_3 = \{ T\alpha_{k+1}, T\alpha_{k+2}, \dots, T\alpha_n \}$ spans R_T

To Prove: S_3 is linearly independent

Consider $c_{k+1}T\alpha_{k+1} + \dots + c_nT\alpha_n = 0 \rightarrow \textcircled{1}$

$$\Rightarrow \sum_{i=k+1}^n c_i T\alpha_i = 0$$

$$\Rightarrow T \left(\sum_{i=k+1}^n c_i \alpha_i \right) = 0$$

$$\Rightarrow \sum_{i=k+1}^n c_i \alpha_i \in N$$

$S_1 = \{ \alpha_1, \alpha_2, \dots, \alpha_k \}$ is a basis of U .

Then $\sum_{i=k+1}^n c_i \alpha_i$ can be expressed as a linear combination of

$\alpha_1, \alpha_2, \dots, \alpha_k$

$$(ii) \sum_{i=k+1}^n c_i \alpha_i = d_1 \alpha_1 + d_2 \alpha_2 + \dots + d_k \alpha_k$$

$$= \sum_{j=1}^k d_j \alpha_j$$

$$\therefore \sum_{j=1}^k d_j \alpha_j - \sum_{i=k+1}^n c_i \alpha_i = 0$$

$$\Rightarrow \left. \begin{aligned} d_1 = d_2 = \dots = d_k = 0 \\ c_{k+1} = c_{k+2} = \dots = c_n = 0 \end{aligned} \right\} \text{ because}$$

$\{ \alpha_1, \alpha_2, \dots, \alpha_k, \alpha_{k+1}, \dots, \alpha_n \}$ is a basis of V .

In equation (i), $c_{k+1} = c_{k+2} = \dots = c_n = 0$

$\Rightarrow \{ T_{\alpha_{k+1}}, T_{\alpha_{k+2}}, \dots, T_{\alpha_n} \}$ is linearly independent.

Hence S_3 is a basis for R_T .

\therefore dimension of $R_T = n - k$.

$$(ii) \text{ Rank}(T) = n - k$$

$$\therefore \text{Rank } T + \text{nullity } T = (n - k) + k$$

$$= n$$

$$= \dim V$$

$$\therefore \text{Rank } T + \text{nullity } T = \dim V$$

Let $U \in \mathcal{L}(V, W)$ and $U = \sum_P \sum_Q A_{PQ} E^{PQ}$

Consider $U_{\alpha_j} = \sum_P \sum_Q A_{PQ} E^{PQ} (\alpha_j)$

$= \sum_P \sum_Q A_{PQ} \delta_{jQ} \beta_P$

$= \sum_P A_{Pj} \beta_P$

ie $U_{\alpha_j} = T_{\alpha_j} \Rightarrow U = T$

T is a linear combination of linear transformations in $\mathcal{L}(V, W)$
Hence linear transformations span $\mathcal{L}(V, W)$

To Prove: $m \times n$ linear transformations are linearly independent.

Consider $\sum_P \sum_Q A_{PQ} E^{PQ} = 0$

$\Rightarrow \sum_P \sum_Q A_{PQ} E^{PQ} (\alpha_j)$

$\Rightarrow A_{Pj} = 0, \forall P \in J$

since $\{\beta_1, \beta_2, \dots, \beta_m\}$ is a basis for W

Hence $m \times n$ linear transformations independent. \rightarrow (3)

from (2) & (3)

These $m \times n$ linear transformations form a basis for $\mathcal{L}(V, W)$

$\dim \mathcal{L}(V, W) = mn$

$\mathcal{L}(V, W)$ is finite dimensional & its dim is mn

Hence the proof.

Theorem: Let V be n -dimensional vector space over F & W be m -dimensional vector space over F and then $L(V, W)$ is finite dimensional and its dimension is $m \times n$.

Proof: $L(V, W)$ is the space of all linear transformations from V into W

Let $\beta = \{\alpha_1, \alpha_2, \dots, \alpha_n\}$ be a basis of V and

$\beta' = \{\beta_1, \beta_2, \dots, \beta_m\}$ be a basis of W

For each pair of integers (p, q) define a linear transformations

$$E^{p,q}(\alpha_i) = \begin{cases} 0 & \text{if } i \neq q \\ \beta_p & \text{if } i = q \end{cases}$$

$$= \text{diag } \beta_p \longrightarrow \textcircled{1}$$

Then there is a unique linear transformation

from V into W satisfying these conditions

we claim that these mn linear transformations form a basis in $L(V, W)$

First we prove mn linear transformations spans $L(V, W)$

for that we have show that if $T \in L(V, W)$.

$$\text{Then } T = \sum_{p=1}^m \sum_{q=1}^n A_{pq} E^{pq}$$

if $\alpha_j \in V$, then $T\alpha_j \in W$

$\{\beta_1, \beta_2, \dots, \beta_m\}$ is a basis of W .

Then $T\alpha_j$ is unique linear transformation of $\beta_1, \beta_2, \dots, \beta_m$

$$\text{as } T\alpha_j = \sum_{p=1}^m A_{pj} \beta_p \longrightarrow \textcircled{2}$$

Definition:- linear operator.

Let V be a vector space over the field F . Then T is called a linear operator on V if T is a linear transformation from V into V .

Definition: Invertible Transformation

A function $T: V \rightarrow W$ is said to be invertible if $\exists U: W \rightarrow V$ such that UT is the identity function on V & TU is the identity function on W .

(ii) if $TT^{-1} = T^{-1}T = I$, then T is said to be invertible.

Theorem: If T is invertible, then (i) T is 1-1. (ii) T is onto

Proof: (i) let $\alpha, \beta \in V$.

consider $T\alpha = T\beta$

$$T\alpha - T\beta = 0$$

$$T(\alpha - \beta) = 0$$

$$T^{-1}T(\alpha - \beta) = T^{-1}(0)$$

$$I(\alpha - \beta) = 0 = 0 = 0$$

$$\alpha = \beta$$

Hence T is one to one.

(ii) let $W = \{ \alpha, \beta \mid T\alpha = 0 \}$ and $R_T = \{ \beta \in W \mid \beta = T\alpha, \alpha \in V \}$

let $\beta \in W$, then

$$\beta = c_1 T\alpha_1 + c_2 T\alpha_2 + \dots + c_n T\alpha_n.$$

where $\{ \alpha_1, \alpha_2, \dots, \alpha_n \}$ is a basis for V

$$\det A \in \mathbb{W}$$

$$\Rightarrow \tau \alpha = 0$$

$$\Rightarrow \alpha = 0$$

we know that Rank τ + nullity τ = dim V .

$$\text{Rank } \tau + 0 = \dim V$$

$$R(\tau) = W$$

Theorem: Let $\tau: V \rightarrow W$ be a linear transformation. Then τ is non-singular iff τ carries each linearly independent subset of V into a linearly independent subset of W .

Proof: Let τ be non-singular, let $\{\alpha_1, \alpha_2, \dots, \alpha_k\}$ be a linearly independent subset of V .

To Prove: $\{\tau \alpha_1, \tau \alpha_2, \dots, \tau \alpha_k\}$ is linearly independent in W .

$$\text{Consider } c_1 \tau \alpha_1 + \dots + c_k \tau \alpha_k = 0 \quad \rightarrow (1)$$

$$\Rightarrow \tau (c_1 \alpha_1 + \dots + c_k \alpha_k) = 0$$

$$\Rightarrow c_1 \alpha_1 + \dots + c_k \alpha_k = 0$$

$$\Rightarrow c_1 = c_2 = \dots = c_k = 0$$

[$\because \tau$ is invertible]

$\therefore \{\tau \alpha_1, \tau \alpha_2, \dots, \tau \alpha_k\}$ is linearly independent in W .

Conversely,

let τ carry linearly independent set into linearly independent set

independent set

let $\alpha \neq 0 \in V$.

Then $\{\alpha\}$ is linearly independent.

$\therefore \{T\alpha\}$ is linearly independent

$$\therefore T\alpha \neq 0 = (a_1, \dots, a_n)$$

\therefore null space of $T = \{0\}$

$\Rightarrow T$ is non-singular.

Isomorphism

Definition: Let $T: V \rightarrow W$ be a linear transformation from a vector space V into W over the same field F . If T is 1-1 & onto, then T is said to be an isomorphism of V into W .

Theorem: Every n -dimension vector space is isomorphic to the space F^n .

Proof: Let V be a vector space F & let $\dim V = n$.

Let $\{\alpha_1, \alpha_2, \dots, \alpha_n\}$ be an ordered basis of V .

Let $\alpha \in V$, then α is linear combination of $\alpha_1, \alpha_2, \dots, \alpha_n$ as

$$\alpha = x_1\alpha_1 + \dots + x_n\alpha_n, \quad x_i \in F$$

Define $T: V \rightarrow F^n$ such that $T\alpha = (x_1, \dots, x_n)$

To Prove: T is linear transformation, T is 1-1 & T is onto

$$\text{Let } \alpha, \beta \in V \text{ & } c \in F$$

$$\text{Let } \alpha = x_1\alpha_1 + \dots + x_n\alpha_n, \quad \beta = y_1\alpha_1 + \dots + y_n\alpha_n$$

$$\text{Consider } c\alpha + \beta = (cx_1 + y_1)\alpha_1 + \dots + (cx_n + y_n)\alpha_n$$

$$\therefore T(c\alpha + \beta) = cT\alpha + T\beta$$

$\therefore T$ is a linear transformation.

ab
 T is 1-1:

consider $T\alpha = T\beta$

$$\Rightarrow (x_1, x_2, \dots, x_n) = (y_1, y_2, \dots, y_n)$$

$$\Rightarrow \alpha = \beta$$

$\therefore T$ is 1-1.

T is onto: if $(x_1, x_2, \dots, x_n) \in F^n$, then $\exists \alpha \in V$ such that

$$T\alpha = (x_1, x_2, \dots, x_n)$$

T is onto

Then V is isomorphic to F^n .

Matrix of a linear operator's

Definition: let $T: V \rightarrow V$ be a linear operator on V , let B be an ordered basis of V . Then the matrix of T relative to the basis B is an $n \times n$ matrix A whose entries A_{ij} are given by

$$T d_j = \sum_{i=1}^n A_{ij} d_i, \quad j=1, 2, \dots, n$$

We denote by $[T]_B$

linear function

Definition: let V be a vector space over the field F . If f is a function from V into F such that $f(c\alpha + \beta) = cf(\alpha) + f(\beta)$, $\forall \alpha, \beta \in V$ & $c \in F$, then f is called a linear functional on V .

Dual Basis

Definition: let V be a vector space over F , & let $\dim V = n$
let V^* be a dual space of V .

If $B = \{ \alpha_1, \alpha_2, \dots, \alpha_n \}$ is a basis of V , then \exists a unique linear functional f_i on V such that $f_i(\alpha_j) = \delta_{ij}$

Then n linear functions $\{ f_1, f_2, \dots, f_n \}$ form a basis of V^* called dual basis of B .

Definition: Annihilator

If V is a vector space over F and S is a subset of V , the annihilator of S is the set S^0 of linear functionals f on V such that $f(\alpha) = 0$ for every α in S .

Theorem: Let V be a finite dimension vector space over F and let W be a sub space of V then $\dim W + \dim W^0 = \dim V$.

Proof: Let $\dim W = k$ and let $\{ \alpha_1, \alpha_2, \dots, \alpha_k \}$ be a basis of W . Choose vector $\alpha_{k+1}, \dots, \alpha_n$ such that $S_1 = \{ \alpha_1, \dots, \alpha_n \}$ is a basis for V .

Let $\{ f_1, f_2, \dots, f_n \}$ be the dual basis of S_1 in V^* .

To Prove: $\{ f_{k+1}, \dots, f_n \}$ is a basis for the annihilator W^0 .

Then $f_i(\alpha_j) = \delta_{ij}$ and $\delta_{ij} = 0$ for $i \geq k+1$ & $j \leq k \rightarrow$ (1)

To Prove: $f_{k+1}, \dots, f_n \in W^0$

Let $\alpha \in W$, then α is a linear combination of $\alpha_1, \alpha_2, \dots, \alpha_k$

$$\Rightarrow \alpha = c_1 \alpha_1 + \dots + c_k \alpha_k$$

$$\Rightarrow f_i(\alpha) = c_1 f_i(\alpha_1) + \dots + c_k f_i(\alpha_k)$$

$$= 0 \text{ for } i \geq k+1$$

$$\therefore f_i(\alpha) = 0, \forall i \geq k+1 \text{ & } \alpha \in W$$

$\therefore S_2 = \{ f_{k+1}, \dots, f_n \} \in W^0$ and it is linearly independent set.

Next we prove S_2 spans W^0

Let $f \in V^*$, then $f = \sum_{i=1}^n f(\alpha_i) f_i \rightarrow \textcircled{2}$

If $f \in W^0$, then $f(\alpha_i) = 0$ for $i \leq k$.

$\therefore \textcircled{2}$ becomes $f = \sum_{i=k+1}^n f(\alpha_i) f_i$

S_2 spans W^0

$\dim W^0 = n - k$

$\therefore \dim W + \dim W^0 = k + n - k = n$

$\therefore \dim W + \dim W^0 = \dim V$

The transpose of a linear transformation

Definition: Let V & W be two vector spaces over F . For each linear transformation T from V into W , there is a unique linear transformation

T^t from W^* into V^* such that $(T^t g)(\alpha) = g(T\alpha) \forall g \in W^*, \alpha \in V$.

$\textcircled{1} \leftarrow$

$$\begin{aligned} & \alpha = c_1 \alpha_1 + \dots + c_k \alpha_k \\ & \beta(\alpha) = c_1 \beta(\alpha_1) + \dots + c_k \beta(\alpha_k) = 0 \end{aligned}$$

$$\begin{aligned} & \beta(\alpha) = 0 \quad \forall \alpha \in W^0 \\ & \therefore \beta(\alpha_i) = 0 \quad \forall i = k+1, \dots, n \end{aligned}$$

Polynomial

Definition: Linear Algebra

Let F be a field. A linear algebra over the field F is a vector space over F with an additional operation called multiplication of vectors which associates with each pair of vectors α, β in a vector space V a vector $\alpha\beta$ in V called the product of α and β in such a way that

(i) Multiplication is associative

$$\alpha(\beta\gamma) = (\alpha\beta)\gamma$$

(ii) Multiplication is distributive w.r.t addition

$$\alpha(\beta + \gamma) = \alpha\beta + \alpha\gamma \text{ and}$$

$$(\alpha + \beta)\gamma = \alpha\gamma + \beta\gamma$$

(iii) For each scalar $c \in F$, $c(\alpha\beta) = \alpha(c\beta) = (c\alpha)\beta$

If there is an element $1 \in A$ such that $1\alpha = \alpha 1 = \alpha, \forall \alpha \in A$

then A is called linear algebra with identity over F .

The algebra is called commutative if $\alpha\beta = \beta\alpha, \alpha, \beta \in A$.

Theorem: Prove that F^∞ is a linear algebra with identity.

Proof: A vector in F^∞ is an infinite sequence of scalars in F

Let $f, g \in F^\infty, a, b \in F$

$$\text{let } f = (f_0, f_1, \dots) \text{ \& } g = (g_0, g_1, \dots)$$

$$\text{Then } af + bg = (af_0 + bg_0, af_1 + bg_1, \dots)$$

which define addition in F^∞

Product is defined by $(fg)_p = \sum_{i=0}^n f_i g_{n-i}$, $n=0,1,2,\dots$

$$\Rightarrow fg = (f_0g_0 + f_1g_1, f_0g_2 + 2f_1g_1 + f_2g_0, \dots)$$

consider $(gfn) = \sum_{i=1}^n g_i f_n = \sum f_i g_{n-i} = (fg)_n$

$$\therefore fg = gf$$

\therefore Multiplication is commutative

let $h = [h_0, h_1, \dots] \in F^\infty$

consider $[(fg)h]_n = \sum_{i=0}^n (fg)_i h_{n-i}$

$$= \sum_{i=0}^n \left(\sum_{j=0}^i f_j g_{i-j} \right) h_{n-i}$$

$$= \sum_{j=0}^n f_j [gh]_{n-j} = [f(gh)]_n, n=0,1,\dots$$

$$\Rightarrow (fg)h = f(gh)$$

\therefore Multiplication is associative

Multiplication is distributive:

$$[f(g+h)]_n = \sum_{i=0}^n f_i (g+h)_{n-i}$$

$$= (fg)_n + (fh)_n, n=0,1,2,\dots$$

$$\Rightarrow f(g+h) = fg + fh$$

let $\alpha, \beta \in F$ & $f \in F^\infty$

Taking $a = \alpha, b = \beta$ and $g = f$ in ①

We see that $\alpha(\alpha+\beta)f \in F^\infty$

$$\text{and } (\alpha+\beta)f = \alpha f + \beta f$$

$$\text{clearly } \alpha(fg) = (\alpha f)g = f(\alpha g)$$

$i = (1, 0, \dots) \in F^\infty$ is the identity element of F^∞

$\therefore F^\infty$ is a commutative linear Algebra with identity over F

Definition: Algebra of formal Power series

Let F be a field and S be the set of non-negative integers. Then the set of all functions from S into F is a vector space over F denoted by F^∞ .

F^∞ is a linear algebra with identity and it is called the algebra of formal Power series over F .

Polynomial.

Definition: Let $F[x]$ be the subspace of F^∞ spanned by the vectors $1, x, x^2, \dots$. An element of $F[x]$ is called a Polynomial over F .

$F[x]$ consists of all finite linear combination of x and its

Power.

Definition: Degree of Polynomial.

A non-zero vector $f \in F^\infty$ is a Polynomial if and only if \exists an integer $n \geq 0$ such that

$$i. f_n \neq 0$$

$$ii. f_k = 0, \forall \text{ integer } k > n$$

The integer n is unique and is called the degree of the Polynomial. It is denoted by $\text{degree of Polynomial}$.

Definition: Let L be a linear algebra over F . Let $f = \sum_{i=0}^n f_i x^i \in L$.

$$\text{If } \alpha \in L, \text{ then } f(\alpha) = \sum_{i=0}^n f_i \alpha^i.$$

Theorem: Let F be a field and \mathcal{L} be a linear algebra with identity over F . Suppose f & g are polynomials over F .

Let $\alpha \in \mathcal{L}$ and $c \in F$. Then prove that

i. $(cf + g)(\alpha) = cf(\alpha) + g(\alpha)$

ii. $(fg)(\alpha) = f(\alpha)g(\alpha)$

Proof:

$$\text{Let } f = \sum_{i=0}^m f_i x^i \quad \& \quad g = \sum_{j=0}^n g_j x^j$$

$$\begin{aligned} \text{(i)} \quad (cf + g) &= \sum_i (cf_i + g_i) x^i \\ &= c \sum_i f_i x^i + \sum_i g_i x^i \\ &= cf(\alpha) + g(\alpha) \end{aligned}$$

$$\text{(ii)} \quad fg = \sum_{ij} f_i g_j x^{i+j}$$

$$\begin{aligned} \text{Then } (fg)(\alpha) &= \sum_{ij} f_i g_j \alpha^{i+j} = \left(\sum_{i=0}^m f_i \alpha^i \right) \left(\sum_{j=0}^n g_j \alpha^j \right) \\ &= f(\alpha)g(\alpha). \end{aligned}$$

Definition: Isomorphism

Let F be a field and \mathcal{L} and \mathcal{L}^{\sim} be a linear algebra over F .

\mathcal{L} and \mathcal{L}^{\sim} are said to be isomorphic

if \exists a 1-1 mapping $\alpha \rightarrow \alpha^{\sim}$ from \mathcal{L} onto \mathcal{L}^{\sim} such that

i. $c(\alpha + d\beta)^{\sim} = c\alpha^{\sim} + d\beta^{\sim}$

ii. $(\alpha\beta)^{\sim} = \alpha^{\sim}\beta^{\sim}, \forall \alpha, \beta \in \mathcal{L} \quad \& \quad c, d \in F$

The mapping is called an isomorphism of \mathcal{L} onto \mathcal{L}^{\sim}

Polynomial Ideals

Lemma: Suppose f and d are non-zero polynomials over a field F such that $\deg d \leq \deg f$. Then prove that \exists a polynomial $g \in F[x]$ such that either $f - \deg g = 0$ or $\deg(f - dg) < \deg f$.

Proof:- Let $f = a_0 + a_1x + \dots + a_{m-1}x^{m-1} + a_mx^m$; $a_m \neq 0$

and $d = b_0 + b_1x + \dots + b_{n-1}x^{n-1} + b_nx^n$; $b_n \neq 0$

$$(1) \quad f = a_mx^m + \sum_{i=0}^{m-1} a_ix^i \quad \rightarrow (1)$$

$$d = b_nx^n + \sum_{i=0}^{n-1} b_ix^i \quad \rightarrow (2)$$

Let $\deg d \leq \deg f$
 $m \geq n$

Multiplying (2) by $\left(\frac{a_m}{b_n}\right)x^{m-n}$

$$\text{we get } d \left(\frac{a_m}{b_n}\right)x^{m-n} = a_mx^m + \left(\frac{a_m}{b_n}\right)x^{m-n} \sum_{i=0}^{n-1} b_ix^i \quad \rightarrow$$

from (1) - (3) we see that either

$$f - d \left(\frac{a_m}{b_n}\right)x^{m-n} = 0 \quad \text{or}$$

$$\deg \left[f - d \left(\frac{a_m}{b_n}\right)x^{m-n} \right] < \deg f$$

Then \exists a polynomial $g(x) = \frac{a_m}{b_n}x^{m-n} \in F[x]$

satisfying the given condition.

Definition:- Let d be a non-zero polynomial over the field F . If $f \in F[x]$ then \exists at most one polynomial $q \in F[x]$ such that $f = dq$

we say that d divides f and f is divisible by d . It is

denoted by $q = f/d$

q is called the quotient of f and d .

Corollary:

Let $f \in F[x]$ and $c \in F$. $f(x)$ is divisible by $x-c$

if and only if $f(c) = 0$

Definition:- Let F be a field and $f \in F[x]$. If $f(c) = 0$, then

c in F is called a root or a zero of polynomial f .

Corollary:- Prove that a polynomial f of degree n over a field F has at most n roots in F .

Proof: The result is true for polynomial of degree 0 and 1

Suppose that it is true for polynomials of degree $(n-1)$

If a is a root of f , then

$f = (x-a)q$ where q is of degree $(n-1)$

Since $f(b) = 0$ if $a = b$ or $q(b) = 0$

q has at most $(n-1)$ roots

\therefore By induction f has at most n roots.

Definition:- If $f = c_0 + c_1x + c_2x^2 + \dots + c_nx^n \in F(x)$, then the

derivative of this polynomial is defined by

$$Df = f' = c_1 + 2c_2x + \dots + nc_nx^{n-1}$$

$$f'' = D^2f$$

$$f''' = D^3f \text{ and so on.}$$

Theorem : Taylor's formula

Let F be a field of characteristic p, let c ∈ F and n be a positive integer, let f be a polynomial over F and let deg ≤ n, then

$$f = \sum_{k=0}^n \frac{(D^k f)(c)}{k!} (x-c)^k$$

Proof : From binomial theorem, we have

$$(a+b)^m = \sum_{k=0}^m \binom{m}{k} a^{m-k} b^k \text{ where}$$

$$\binom{m}{k} = \frac{m!}{k! (m-k)!} = \frac{m(m-1) \dots (m-k+1)}{1 \cdot 2 \dots k}$$

By the binomial theorem

$$\begin{aligned} x^n &= [c + (x-c)]^n \\ &= \sum_{k=0}^n \binom{n}{k} c^{n-k} (x-c)^k \\ &= c^n + n c^{n-1} (x-c) + \dots + (x-c)^n \end{aligned}$$

If we take f = x^m, then f(c) = c^m.

$$\text{then } (D^0 f)(c) = D^0 c^m = c^m$$

$$(D^1 f)(c) = D c^m = m c^{m-1}$$

$$(D^m f)(c) = D^m c^m = m(m-1) \dots 1 = m!$$

$$\Rightarrow x^m = D^0 c^m + \frac{D c^m}{1!} (x-c) + \dots + \frac{D^m c^m}{m!} (x-c)^m$$

$$= \sum_{k=0}^m \frac{(D^k f)(c)}{k!} (x-c)^k \rightarrow \textcircled{2}$$

If $f = \sum_{m=0}^n a_m x^m \rightarrow (3)$

Then $D^k f(x) = D^k \left(\sum_{m=1}^n a_m x^m \right) (c)$

$= \sum a_m (D^k x^m) (c)$

$\therefore \sum_{k=1}^n \frac{D^k f(c)}{k!} (x-c)^k = \sum_k \sum_m a_m \left(\frac{D^k x^m}{k!} \right) (c) (x-c)^k$

$= \sum_{m=x}^m a_m x^m \quad [\text{using } (3)]$

Definition:-

Let c be a root of the polynomial f then the largest +ve integer r such that $(x-c)^r$ divides f is called the multiplicity of c .

Definition: Ideal

Let F be a field let m be a subspace of $F[x]$, m is said to be an ideal of $F[x]$ if $f \in F[x]$ and $g \in m \Rightarrow fg \in m$.

Definition: Principal ideal.

Let m be an ideal of $F[x]$. If every element in M is a said to be an ideal of $F[x]$. if $f \in F[x]$ and $g \in m \Rightarrow fg \in m$

Definition:-

If p_1, \dots, p_n are polynomial over a field F , not all of which are 0, the monic generator d of the ideal $p_1 F[x] + \dots + p_n F[x]$ is called the greatest common divisor (gcd) of p_1, p_2, \dots, p_n .

Relative Prime :-

The polynomials are said to be relatively prime if their gcd is 1 if ideal generated by primes $F[x]$.

The Prime Factorization of a Polynomial

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Definition: Let F be a field and f be a polynomial in $F[x]$, f is said to be reducible over F if \exists polynomial $g, h \in F[x]$ such that $f = gh$

Theorem: If F is a field a non-scalar monic polynomial in $F[x]$ can be factored as a product of monic primes in $F[x]$ in one and only one way for order only one way?

Proof:- Suppose that f is a non-scalar monic polynomial over F

Polynomial of degree one are irreducible

\therefore if $\deg f = 1$, then there is nothing to prove

$\therefore \deg f = n$ and $n > 1$ by induction assume that theorem is true

for non-scalar monic polynomial of degree less than n .

If f is irreducible, there is nothing to prove

Let $f = gh$ where g and h are non-scalar monic polynomials of degree less than n .

Then by assumption g and h can be factored as a product of monic primes in $F[x]$

Suppose that $f = p_1 \dots p_m = q_1 \dots q_n \longrightarrow \textcircled{1}$

where p_1, p_2, \dots, p_m and q_1, q_2, \dots, q_n are monic primes in $F[x]$

consider $p_1 \dots p_m = q_1 \dots q_n$

clearly p_m divides LHS

$\therefore p_m$ divides $q_1 q_2 \dots q_n$

$\Rightarrow P_m$ divides q_i

But P_m and q_i are monic primes

$$\therefore q_i = P_m \rightarrow \textcircled{2}$$

if $m=1$ or $n=1$, then $f = p_1 = q_1$ and there is nothing to prove

Suppose that $m > 1$ and $n > 1$.

By rearranging q 's, we can assume $P_m = q_n$

$$\therefore \textcircled{1} \text{ becomes } P_1 \dots P_{m-1}, P_m = q_1, \dots, q_{n-1}, q_n$$

$$\Rightarrow P_1 \dots P_{m-1} = q_1 q_2 \dots q_{n-1}$$

Now the polynomial $P_1 \dots P_{m-1}$ has degree less than n

\therefore By induction hypothesis we see that $q_1 q_2 \dots q_{n-1}$ is at most a rearrangement of $P_1 \dots P_{m-1}$

Hence the theorem.

Determinants

Definition of a n -linear

Let K be a commutative ring with identity 1. Let r

be a ve integer.

Let $D: A \rightarrow D(A)$ be a function.

which gives to each $n \times n$ matrix A a scalar $D(A)$

D is said to be n -linear if for each i if $(1 \leq i \leq n)$

D is a linear function of the i th row when the other $(n-1)$

rows are held fixed.

Alternating function :-

Let D be a n -linear function. D is said to be alternating

- i. $D(A) = 0$ when 2 rows of A are equal
- ii. $D(A') = -D(A)$ where A' obtained from A by interchanging any two rows of A .

Determinant function :-

Let K be a commutative ring with identity 1. Let n be

a +ve integer. Let $D: K^{n \times n} \rightarrow K$ be a function from $n \times n$ matrix into K

1. D is n -linear
2. D is alternating
3. $D(I) = 1$, I - identity matrix.

Lemma :- Let D be a n -linear function on $n \times n$ matrices over K . Suppose D has the property that $D(A) = 0$ where two adjacent rows are equal

Prove that D is alternating.

Proof :-

we have to prove

(i) $D(A) = 0$ when any two rows are equal

(ii) $D(A') = -D(A)$ where A' is obtained from A by interchange two rows

when A' is obtained from A by interchanging two adjacent rows of A

then $D(A') = -D(A)$

Let B be the matrix obtained from A by interchanging rows i & j where $i < j$

First we interchange row i with row $(i+1)$ and continue till we get the rows in the order -

$d_1, \dots, d_{i-1}, d_{i+1}, \dots, d_j, d_i, d_{j+1}, \dots, d_n$

The number of interchange in these case = $j-1$

Next we move d_j to the i th position.

The number of interchange of rows = $j-1$

Thus B is obtained from A by $(j-1) + [(j-1) - 1]$ interchanges

$d_j - d_i - 2$ interchanges

$$D(B) = (-1)^{2j-2i-1} D(A)$$

$$= -D(A)$$

Condition ② is true.

Next we prove $D(A) = 0$ when any 2 rows are equal

Given $D(A) = 0$ when 2 adjacent rows are equal

If A, two adjacent rows are equal, then $D(A) = 0$

if not, we can make two any equal rows adjacent by row interchanges

as we did before.

\therefore In that case also $D(A) = 0$.

Elementary Canonical Forms

Characteristic values

Let V be a vector space over the field F and let T be a linear operator on V . A characteristic value of T is a scalar c in F such that there is a non-zero vector α in V with $T\alpha = c\alpha$. If c is a characteristic value of T , then

- any α such that $T\alpha = c\alpha$ is called a characteristic vector of T associated with the characteristic value c .
- the collection of all α such that $T\alpha = c\alpha$ is called the characteristic space associated with c .

Result: Let T be a linear operator on a finite-dimensional space V and let c be a scalar. The following are equivalent.

- c is a characteristic value of T
- The operator $(T - cI)$ is singular (not invertible)
- $\det(T - cI) = 0$

Definition: If A is an $n \times n$ matrix over the field F , a characteristic value of A in F is a scalar c in F such that the matrix $(A - cI)$ is singular.

Lemma: Similar matrices have the same characteristic polynomial.

Proof: If $B = P^{-1}AP$, then

$$\begin{aligned} \det(xI - B) &= \det(xI - P^{-1}AP) \\ &= \det(P^{-1}(xI - A)P) \\ &= \det P^{-1} \det(xI - A) \det P \\ &= \det(xI - A) \end{aligned}$$

Definition:- Let T be a linear operator on the finite dimensional space V .

We say that T is diagonalizable if there is a basis for V each vector of which is a characteristic vector of T .

Lemma:-

Suppose that $T\alpha = c\alpha$. If f is any polynomial, then $f(T)\alpha = f(c)\alpha$.

Proof:-

Given $T\alpha = c\alpha$

(i) c is a characteristic value of T and α is the characteristic vector such that $\alpha \neq 0$.

Now

$$T\alpha = c\alpha \Rightarrow T\alpha - c\alpha = 0$$

$$\Rightarrow (T - cI)\alpha = 0$$

$$\Rightarrow f(T - cI)\alpha = f(0)$$

$$\Rightarrow f(T)\alpha - f(c)\alpha = 0$$

$$\Rightarrow f(T)\alpha = f(c)\alpha$$

Annihilating Polynomial

Definition:- Minimal Polynomial

Let T be a linear operator on a finite dimensional vector space V over the field F . The minimal polynomial for T is the unique monic generator of the ideal of polynomials over F which annihilate T .

Theorem:- Let T be a linear operator on an n -dimensional vector space V . The characteristic and minimal polynomial for T have same roots except for multiplicities.

Proof:-

Let P be the minimal polynomial for T . Let c be a scalar.

To Prove: $P(c) = 0$ iff c is a characteristic value for T .

Suppose $P(c) = 0$, then $P = (x-c)q$, where q is a polynomial.

Since $\deg q < \deg P$.

Then by definition of minimal polynomial P tells us that $q(T) \neq 0$.

Choose a vector β such that

$$q(T)\beta \neq 0$$

Let $\alpha = q(T)\beta$, then $0 = P(T)\beta$

$$0 = (x-c)q(T)\beta$$

$$0 = (x-c)\alpha$$

Thus c is a characteristic value of T .

Conversely, suppose c is a characteristic value of T .

Then $T\alpha = c\alpha$, with $\alpha \neq 0$.

$$P(T)\alpha = P(c)\alpha$$

Since $P(T) = 0$ and $\alpha \neq 0$

$$\Rightarrow P(c) = 0$$

Invariant subspace.

Definition:-

Let V be a vector space and T be a linear operator on V .

If W is a subspace of V we say that W is invariant under T if for each vector α in W the vector $T\alpha$ is in W (ie) if $T(W)$ is contained in W .

$$(ie) T(W) \subseteq W$$

Lemma:-

Let W be an invariant subspace for T . The characteristic Polynomial for the restriction operator $T|_W$ divides the characteristic Polynomial for T . The minimal Polynomial for $T|_W$ divides the minimal Polynomial for T .

Proof:- Let the characteristic Polynomial for $T|_W$ divides the characteristic Polynomial for T .

$$A = \begin{bmatrix} B & C \\ 0 & D \end{bmatrix} \quad \text{where } A = [T]_B \quad \& \quad B = [T]_B$$

$$\det(xI - A) = \det \begin{bmatrix} x - A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & x - A_{22} & \dots & A_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \dots & x - A_{nn} \end{bmatrix}$$

Since $A_{ij} = 0$ if $j \leq r, i > r$.

$$\det(xI - A) = \det[xI - B] \det(xI - D)$$

$\det(xI - A)$ is characteristic Polynomial for T

$\det(xI - A)$ is characteristic Polynomial for $T|_W$.

$$\therefore A^k = \begin{bmatrix} B^k & C^k \\ 0 & D^k \end{bmatrix} \quad \text{where } C^k \text{ is some } r \times (n-r) \text{ matrix}$$

\therefore Any Polynomial which annihilates A

$$f(A) = 0$$

$$f(A) = \det(xI - A)$$

$$0 = \det(xI - B) \det(xI - D)$$

A also annihilates B

(ie) Any Polynomial which annihilates A also annihilates B .

pt 202 dnt. The minimal Polynomial for B divides the minimal Polynomial for A.

Definition :-

Let W be an invariant subspace for T and let α be a vector in V . The T -conductor of α into W is the set $\mathcal{C}_T(\alpha, W)$ which consists of all polynomials g such that $g(T)\alpha$ is in W .

notes :-

If the special case $W = \{0\}$ the conductor is called the T -conductor of α .

Direct sum decomposition

Definition :-

Let W_1, W_2, \dots, W_k be subspaces of the vector V we say that W_1, \dots, W_k are independent if $\alpha_1 + \alpha_2 + \dots + \alpha_k = 0$, $\alpha_i \in W_i$ implies that each α_i is 0.

Definition :-

If V is a vector space, a Projection of V is a linear operator E on V such that $E^2 = E$.

Theorem :-

If $V = W_1 \oplus \dots \oplus W_k$, then there exists k linear operators

E_1, \dots, E_k on V such that

(i) each E_i is a Projection ($E_i^2 = E_i$)

(ii) $E_i E_j = 0$, if $i \neq j$

(iii) $I = E_1 + \dots + E_k$, (iv) the range of E_i is W_i

Conversely if E_1, \dots, E_k are k linear operators on V which satisfy conditions (i), (ii), and (iii) and if we let W_i be the range of E_i , then $V = W_1 \oplus W_2 \oplus \dots \oplus W_k$

Proof:

we have to proof only the converse statement

suppose E_1, \dots, E_k are linearly operators on V

which satisfy the first three conditions and let w_i be the range of E_i . Then $V = w_1 + w_2 + \dots + w_k$.

For by condition (iii)

$$\alpha = E_1 \alpha + \dots + E_k \alpha$$

for each $\alpha \in V$, and $E_i \alpha \in w_i$

This expression for α 's unique, because $\alpha = \alpha_1 + \dots + \alpha_k$

with $\alpha_j \in w_j$ say $\alpha_j = E_j \beta_j$

Then using (i) and (ii) we have

$$\begin{aligned} E_j \alpha &= \sum_{i=1}^k E_j E_i \alpha \\ &= \sum_{i=1}^k E_j E_i E_j \beta_j \\ &= E_j^2 \beta_j = E_j \beta_j = \alpha_j \end{aligned}$$

This show that V is direct sum of the w_i 's

Hence the Proof.

Unit - 5

Invariant Direct Sum

Definition:-

If α is a vector in V , then \exists unique vectors $\alpha_1, \alpha_2, \dots, \alpha_k$ with $\alpha_i \in W_i$ such that $\alpha = \alpha_1 + \alpha_2 + \dots + \alpha_k$ and then $T\alpha = T\alpha_1 + \dots + T\alpha_k$ we say that T is direct sum of the operators T_1, T_2, \dots, T_k .

Theorem:-

Let T be a linear operator on the space V and let W_1, \dots, W_k and E_1, \dots, E_k be on the subspace of V and Projection on V . Then a necessary and sufficient condition that each subspace W_i be invariant under T is that T commutative with each of the Projections E_i that is $TE_i = E_iT, i=1, 2, \dots, k$.

Proof:-

suppose that T commutes with each E_i

let α be in W_j

$$\begin{aligned} \text{Then } E_j\alpha &= \alpha \text{ and } T\alpha = T(E_j\alpha) \\ &= E_j(T\alpha) \end{aligned}$$

which show that $T\alpha$ is in the range of E_j

ies W_j is invariant under T

Conversely, now Assume that each W_i is invariant under T

To show that $TE_j = E_jT$

let α be any vector in V , Then $\alpha = \alpha_1 + \dots + \alpha_k$

$$\alpha = E_1\alpha + \dots + E_k\alpha$$

[$\because W_i$ is invariant under $T \therefore \alpha_j = E_j\alpha$]

$$T\alpha = TE_1\alpha + \dots + TE_k\alpha$$

Since each $E_i\alpha$ is in W_i which is invariant under T

we have $T(E_i\alpha) = E_i\beta_i$ for some vector β_i

$$\text{Then } E_j TE_i\alpha = E_j E_i \beta_i$$

$$= \begin{cases} 0 & \text{if } i \neq j \\ E_j \beta_j & \text{if } i = j \end{cases}$$

$$\text{Thus } E_j T\alpha = E_j TE_1\alpha + \dots + E_j TE_k\alpha$$

$$= E_j \beta_j$$

$$= TE_j\alpha$$

$$\therefore E_j T\alpha = TE_j\alpha$$

$$E_j T = TE_j$$

This holds for each $\alpha \in V$.

$$\text{Hence } E_j T = TE_j$$

T commutes with each of the projections $E_i, i=1, \dots, k$.

Primary Decomposition Theorem.

Theorem: Let T be a linear operator on the finite dimensional vector space V over the field F . Let P be the minimal polynomial for T ,

$P = P_1^{r_1} \dots P_k^{r_k}$ where the P_i are distinct irreducible monic polynomials

over F and the r_i are positive integers. Let W_i be the null space of

$P_i(T)^{r_i}, i=1, \dots, k$, then

- (i) $V = W_1 \oplus W_2 \oplus \dots \oplus W_k$
- (ii) each W_i is invariant under T
- (iii) if T_i is the operator for induced on W_i by T , then the minimal polynomial for T_i is $p_i(x)$

Proof:-

If the direct sum decomposition (i) is valid.

Our aim is to get a projections E_1, E_2, \dots, E_k associated with the decomposition.

The projection E_i will be the identity on W_i and zero on the other W_j .

\therefore we shall find a polynomial h_i such that $h_i(T)$ is the identity on W_i and zero on the other W_j .

$$\therefore h_1(T) + \dots + h_k(T) = I$$

For each i , let $f_i = \frac{P}{p_i \cdot \prod_{j \neq i} p_j} = \prod_{j \neq i} p_j \cdot \gamma_j$

Since p_1, \dots, p_k are distinct prime polynomials

The polynomials f_1, f_2, \dots, f_k are relatively prime.

$$\therefore (f_i, f_j) = 1$$

Thus there are polynomials g_1, g_2, \dots, g_k such that $\sum_{j=1}^k f_j g_j = 1$

Also note that if $i \neq j$ then $f_i f_j$ is divisible by the

polynomial p_i because $f_i f_j$ contains each p_m as a factor

we shall show that the polynomials $h_i = f_i g_i$

$$\text{let } E_i = h_i(T)$$

$$= f_i(T) g_i(T)$$

Since $h_1 + \dots + h_k = 1$ and P divides $f_i f_j$ for $i \neq j$

$$E_i E_j = 0 \text{ if } i \neq j$$

Thus the E_i are projections which correspond to some direct sum decomposition of the space V .

The range of E_i is exactly the subspace of W_i

\Rightarrow Each vector in the range of E_i is in W_i

If α is the range of E_i , then $\alpha = E_i \alpha$

$$P_i (T)^{r_i} \alpha = P_i (T)^{r_i} E_i \alpha$$

$$= P_i (T)^{r_i} (T)^{r_i} g_i (T)^{q_i} \alpha$$
$$= 0$$

because $P^{r_i} f_i g_i$ is divisible by the minimal polynomial P .

conversely

let g be any polynomial such that

$$g(T_i) = 0$$

$$\text{Then } g(T) f_i(T) = 0$$

Thus $g f_i$ is divisible by the minimal polynomial P of T

$$P_i^{r_i} f_i \text{ divides } g f_i$$

$$\Rightarrow P_i^{r_i} \text{ divides } g$$

Hence the minimal polynomial for T_i is $P_i^{r_i}$

Theorem:- let T be a linear operator on finite dimensional space V . If T is diagonalizable and if c_1, c_2, \dots, c_k are distinct characteristic values of T , then there exists linear operators

E_1, E_2, \dots, E_k on V such that $(E_1 + \dots + E_k)(v) = v$

(i) $T = c_1 E_1 + c_2 E_2 + \dots + c_k E_k$

(ii) $I = E_1 + E_2 + \dots + E_k$

(iii) $E_i E_j = 0 \quad i \neq j$

(iv) $E_i^2 = E_i$ (E_i is a Projection)

(v) The range of E_i is the characteristic space for T with c_i

Conversely, if there exists k distinct scalars c_1, \dots, c_k and k non zero linear operators E_1, \dots, E_k which satisfy conditions (i), (ii), and (iii) then T is diagonalizable c_1, \dots, c_k are distinct characteristic value of T and conditions (iv) and (v) are also satisfied.

Proof:- Suppose that T is diagonalizable, with distinct characteristic value c_1, \dots, c_k

let W_i be the space of characteristic vector associated with the characteristic value c_i

$V = W_1 \oplus \dots \oplus W_k$

let E_1, E_2, \dots, E_k be the Projections associated with this decomposition

Then (i), (ii), (iii), (iv) and (v) are satisfied

To verify (i) Proceed as follows

For each α in V , $\alpha = E_1 \alpha + E_2 \alpha + \dots + E_k \alpha$

$T\alpha = T E_1 \alpha + \dots + T E_k \alpha$

$$\therefore T\alpha = (c_1 E_1 + \dots + c_k E_k)\alpha$$

$$\therefore T = c_1 E_1 + \dots + c_k E_k$$

now suppose that we are given a linear operator T along with the distinct scalar c_i and non-zero operator E_i which satisfy

(i), (ii), and (iii)

$$\text{since } E_i E_j = 0 \text{ when } i \neq j$$

we multiply both side of $E = E_1 + \dots + E_k$ by E_i

$$\Rightarrow E_i = E_i E_1 + \dots + E_i E_i + \dots + E_i E_k$$

$$\Rightarrow E_i = E_i^2$$

Also multiplying $T = c_1 E_1 + \dots + c_k E_k$ by E_i

$$\Rightarrow T E_i = c_i E_i^2$$

since each E_i is a projection.

$$E_i^2 = E_i$$

$$T E_i = c_i E_i$$

$$T E_i - c_i E_i = 0$$

$$(T - c_i I) E_i = 0$$

which show that any vector in the range of E_i is in the null space of $(T - c_i I)$

Assume that $E_i \neq 0$

$$T - c_i I = 0$$

Hence c_i is the characteristic values of T

If c is any scalar then $T - cI = (c_1 - c)E_1 + \dots + (c_k - c)E_k$

$$(T - cI)\alpha = (c_1 - c)E_1\alpha + \dots + (c_k - c)E_k\alpha$$

If $(T - cI)\alpha = 0$, then $(c_i - c)E_i\alpha = 0$

If $\alpha \neq 0$, then $E_i\alpha \neq 0$ for some i

$$\therefore c_i - c = 0 \quad [\because E_i\alpha \neq 0]$$

Since T is diagonalisable

Every non zero vector in the range of E_i is a characteristic

value of T

And $I = E_1 + E_2 + \dots + E_k$ these characteristic vectors spans V .

The null space $(T - cI)$ is exactly the range of E_i

If $T\alpha = c\alpha$, then

$$\sum_{j=1}^k (c_j - c) E_j\alpha = 0$$

Hence $(c_j - c) E_j\alpha = 0$ for each j

and then $E_j\alpha = 0$ for $j \neq i$

Since $\alpha = E_1\alpha + \dots + E_k\alpha$ and $E_j\alpha = 0$ for $j \neq i$

$\Rightarrow \alpha = E_i\alpha$ which prove that α is in the

range of E_i .

Definition: let N be a linear operator on the vector space V . we

say that N is nilpotent if there is some positive integer r

$$\text{such that } N^r = 0$$

Cyclic Subspace and Annihilators

Definition:- If α is any vector in V , the T -cyclic subspace generated by α is the subspace $Z(\alpha; T)$ of all vectors of the form $g(T)\alpha$, g in $F[x]$. If $Z(\alpha; T) = V$, then α is called a cyclic vector for T .

Definition:- If α is any vector in V , the T -annihilator of α is the ideal $H(\alpha; T)$ in $F[x]$ consisting of all polynomials g over F such that $g(T)\alpha = 0$. The unique monic polynomial p_α which generates this ideal will also be called the T -annihilator of α .

Definition:- Let T be a linear operator on a vector space V and let W be a subspace of V . We say that W is T -admissible if

- (i) W is invariant under T
- (ii) if $f(T)\beta$ is in W , there exists a vector β in W such that $f(T)\beta = f(T)\beta$.