

8. Spectral Theory

8.1 Matrix of Linear Transformation

Recall that a function between linear spaces is often referred to as a transformation. Let $T : X \rightarrow Y$ linear transformation where X and Y are finite dimensional linear spaces over a field F such that $\dim X = n$, $\dim Y = m$.

Let $s = \{x_1, x_2, \dots, x_n\}$ be an ordered basis for X so that each vector in X is expressible as linear combination of the elements of s , i.e. for every $x \in X$ has unique representation

$$x = \sum_{i=1}^n \alpha_i x_i, \quad \alpha_i \in F, \quad i = 1, 2, \dots, n$$

The vector $\alpha = (\alpha_1, \alpha_2, \dots, \alpha_n)$ is called the Coordinates Vector of x . Let $s' = \{y_1, y_2, \dots, y_m\}$ be an ordered basis for Y so that each vector in Y is expressible as linear combination of the elements of s' .

Let us choose nm scalars $a_{ij} \in F$ where $i = 1, 2, \dots, m$, $j = 1, 2, \dots, n$

Since $x_1 \in X \Rightarrow T(x_1) \in Y \Rightarrow T(x_1)$ can be expressible as linear combination of m vectors in s' .i.e.

$$T(x_1) = a_{11}y_1 + a_{21}y_2 + \dots + a_{m1}y_m = \sum_{i=1}^m a_{i1}y_i$$

Also

$$T(x_2) = a_{12}y_1 + a_{22}y_2 + \dots + a_{m2}y_m = \sum_{i=1}^m a_{i2}y_i$$

$$T(x_n) = a_{1n}y_1 + a_{2n}y_2 + \dots + a_{mn}y_m = \sum_{i=1}^m a_{in}y_i$$

We can write the above n equations in symbolic form as under

$$T(x_j) = \sum_{i=1}^m a_{ij}y_i, \quad j = 1, 2, \dots, n$$

The coefficient matrix in the above expression is

$$\begin{bmatrix} a_{11} & a_{21} & \dots & a_{m1} \\ a_{12} & a_{22} & \dots & a_{m2} \\ \vdots & \vdots & \vdots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{mn} \end{bmatrix}$$

Then the matrix of $T : X \rightarrow Y$ with respect to the given basis s and s' is the transpose of the above coefficient matrix which is obtained by changing the rows into columns and columns into rows of the coefficient matrix,

Matrix of T with respect to basis s and s' is

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} = [a_{ij}]$$

Above matrix is $m \times n$ matrix consisting of m rows and n columns. The above matrix is symbolically written as $[T : s, s']$ or simply $[T]$.

Remark

If $X = Y$, then $T : X \rightarrow X$ and $m = n$ so that the matrix of T with respect to basis s will be a $n \times n$ matrix and the rule for writing is same as expressed above.

Theorem(8.1.1)

Let $T : X \rightarrow Y$ linear transformation where X and Y are finite dimensional linear spaces over a field F such that $\dim X = n$, $\dim Y = m$. Let $s = \{x_1, x_2, \dots, x_n\}$ be an ordered basis for X , and let $s' = \{y_1, y_2, \dots, y_m\}$ be an ordered basis for Y . If $\{x\} = (\{x_1, x_2, \dots, x_n\})$ is the Coordinates Vector of $x \in X$ with respect to s and $r = (r_1, r_2, \dots, r_m)$ is the Coordinates Vector of $T(x) \in Y$ with respect to s' . Then the matrix A of T with respect to the given basis s and s' is satisfying $r = \{x\}A$

Proof :

Since $\{x\} = (\{x_1, x_2, \dots, x_n\})$ is the Coordinates Vector of $x \in X$ with respect to s

$$\Rightarrow x = \sum_{j=1}^n \{x_j\} x_j \Rightarrow T(x) = T\left(\sum_{j=1}^n \{x_j\} x_j\right) = \sum_{j=1}^n \{x_j\} T(x_j)$$

$$\text{Since } T(x_j) = \sum_{i=1}^m a_{ij} y_i, \quad j = 1, 2, \dots, n \Rightarrow T(x) = \sum_{j=1}^n \{x_j\} \left(\sum_{i=1}^m a_{ij} y_i\right) = \sum_{i=1}^m \left(\sum_{j=1}^n \{x_j\} a_{ij}\right) y_i$$

Since $r = (r_1, r_2, \dots, r_m)$ is the Coordinates Vector of $T(x) \in Y$ with respect to s'

$$\Rightarrow T(x) = \sum_{i=1}^m r_i y_i \Rightarrow \sum_{i=1}^m r_i y_i = \sum_{i=1}^m \left(\sum_{j=1}^n \{x_j\} a_{ij}\right) y_i$$

Since every $y \in Y$ has unique representation of linear combination of vectors in s' . Thus

$$y_i = \sum_{k=1}^m \{y_k\} a_{ki}$$

$$(r_1, r_2, \dots, r_m) = (\{x_1, x_2, \dots, x_n\}) \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \Rightarrow r = \{x\}A$$

Example(8.1.2)

(1) Let $X = \mathbb{R}^2, Y = \mathbb{R}^3$ and $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ defined by $T(x, y) = (x, x + y, 2x - y)$ for all $(x, y) \in \mathbb{R}^2$, then T is linear operator, the matrix A of T with respect to the given basis $s = \{(1,0), (0,1)\}$

and $s' = \{(1,0,0), (0,1,0), (0,0,1)\}$ is $A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & -1 \end{bmatrix}$.

(2) Let $X = P_2(\mathbb{R}), Y = \mathbb{R}^2$ and $T : P_2(\mathbb{R}) \rightarrow \mathbb{R}^2$ defined by $T(a+bx+cx^2) = (2a, b-c)$ for all $a+bx+cx^2 \in P_2(\mathbb{R})$, then T is linear operator, the matrix A of T with respect to the given basis

$$s = \{5, 2x, x^2\} \text{ and } s' = \{(-1,0), (0,3)\} \text{ is } A = \begin{bmatrix} -10 & 0 \\ 0 & \frac{2}{3} \\ 0 & -\frac{1}{3} \end{bmatrix}.$$

Theorem (8.2.3)

If $s = \{x_1, x_2, \dots, x_n\}$ is an ordered basis for a finite dimensional linear space X over F . Then the function $T \rightarrow [T]$ which assigns to each operator T its matrix relative to s is an isomorphism of algebra $L(X)$ onto the total matrix algebra $M_n(F)$.

Proof :

Define a function $f : B(X) \rightarrow M_n(F)$ by $f(T) = [T]$ for all $T \in B(X)$

Let $T_1, T_2 \in B(X)$, then $f(T_1) = [T_1] = [a_{ij}]_{n \times n}$ and $f(T_2) = [T_2] = [b_{ij}]_{n \times n}$

$$T_1(x_j) = \sum_{i=1}^m a_{ij} y_i, \quad j=1,2,\dots,n, \quad T_2(x_j) = \sum_{i=1}^m b_{ij} y_i, \quad j=1,2,\dots,n$$

$$\text{Let } x = \sum_{j=1}^n \} x_j$$

To prove :

(1) f is one-one : Let $T_1, T_2 \in B(X)$ such that $f(T_1) = f(T_2)$

$$\Rightarrow [a_{ij}]_{n \times n} = [b_{ij}]_{n \times n} \Rightarrow \sum_{i=1}^n a_{ij} x_i = \sum_{i=1}^n b_{ij} x_i, \quad j=1,2,\dots$$

$$\Rightarrow T_1(x_j) = T_2(x_j), \quad j=1,2,\dots,n \Rightarrow \sum_{j=1}^n \} T_1(x_j) = \sum_{j=1}^n \} T_2(x_j) \Rightarrow T_1(\sum_{j=1}^n \} x_j) = T_2(\sum_{j=1}^n \} x_j)$$

$$\Rightarrow T_1(x) = T_2(x) \text{ for all } x \in X \Rightarrow T_1 = T_2. \text{ Hence } f \text{ is one-one.}$$

(2) f is onto : let $[c_{ij}]$ be any matrix in $M_n(F)$ the corresponding to this matrix there exists a

linear operator $T : X \rightarrow X$ such that $T(x_j) = \sum_{i=1}^n \} c_{ij} x_i, \quad j=1,2,\dots,n$

Above defines the operators T is extended by linearity to whole of X then the resulting operator has $[c_{ij}]$ as its matrix relative to s . Hence f is onto.

From (1),(2), we have f is bijective.

(3) f is preserves addition, i.e. $f(T_1+T_2) = f(T_1) + f(T_2)$

$$(T_1+T_2)(x_j) = T_1(x_j) + T_2(x_j) = \sum_{i=1}^n a_{ij}x_i + \sum_{i=1}^n b_{ij}x_i = \sum_{i=1}^n (a_{ij} + b_{ij})x_i$$

$$f(T_1+T_2) = [T_1+T_2] = [a_{ij} + b_{ij}]_{n \times n} = [a_{ij}]_{n \times n} + [b_{ij}]_{n \times n} = [T_1] + [T_2] = f(T_1) + f(T_2)$$

(4) f is preserves scalar multiplication , i.e. $f(\lambda T) = \lambda f(T)$

$$(\lambda T)(x_j) = \lambda T(x_j) = \lambda \sum_{i=1}^n a_{ij}x_i = \sum_{i=1}^n \lambda a_{ij}x_i = \sum_{i=1}^n (c_{ij})x_i$$

$$f(\lambda T) = [\lambda T] = [\lambda a_{ij}]_{n \times n} = \lambda [a_{ij}]_{n \times n} = \lambda [T] = \lambda f(T)$$

(5) f is preserves multiplication, i.e. $f(T_1 T_2) = f(T_1) f(T_2)$

$$\begin{aligned} (T_1 T_2)(x_j) &= T_1(T_2(x_j)) = T_1\left(\sum_{k=1}^n b_{kj}x_k\right) = \sum_{k=1}^n b_{kj}T(x_k) = \sum_{k=1}^n b_{kj}\left(\sum_{i=1}^n a_{ik}x_i\right) = \sum_{i=1}^n \left(\sum_{k=1}^n a_{ik}b_{kj}\right)x_i \\ &= \sum_{i=1}^n \left(\sum_{k=1}^n a_{ik}b_{kj}\right)x_i + \sum_{i=1}^n b_{ij}x_i = \sum_{i=1}^n (c_{ij})x_i \end{aligned}$$

$$f(T_1 T_2) = [T_1 T_2] = [c_{ij}]_{n \times n} = \left[\sum_{k=1}^n a_{ik}b_{kj}\right] = [a_{ij}]_{n \times n} [b_{ij}]_{n \times n} = [T_1][T_2] = f(T_1)f(T_2)$$

Hence is $f : B(X) \rightarrow M_n(F)$ by $f(T) = [T]$ for all $T \in B(X)$ an isomorphism

Matrices of identity and zero operators

If T be a linear operators on linear space whose matrix relative to basis $S = \{x_1, x_2, \dots, x_n\}$ be

$$[a_{ij}] \text{ then is } T(x_j) = \sum_{i=1}^n a_{ij}y_i, \quad j = 1, 2, \dots, n$$

$$I(x_j) = x_j = 0x_1 + 0x_2 + \dots + 1 \cdot x_j + \dots + 0x_n \Rightarrow I(x_j) = \sum_{i=1}^n (u_{ij})x_i \text{ where } u_{ij} = \begin{cases} 0, & i \neq j \\ 1, & i = j \end{cases}$$

$\Rightarrow [I] = [u_{ij}]$, i.e. unit matrix.

$$\text{Again } 0(x_j) = 0 = 0x_1 + 0x_2 + \dots + 0x_n \Rightarrow 0(x_j) = \sum_{i=1}^n (0_{ij})x_i$$

$\Rightarrow [0] = [0_{ij}]$, i.e. null matrix.

Matrix of an inverse operator

Theorem(8.1.4)

Let T be linear operator on a linear space X whose matrix relative to basis $S = \{x_1, x_2, \dots, x_n\}$ is $[a_{ij}]$. Then T is non singular iff $[a_{ij}]$ is non singular and in this case $[a_{ij}]^{-1} = [T^{-1}]$

Proof :

Since T is non singular iff $T \circ T^{-1} = T^{-1} \circ T = I$

iff $[T][T^{-1}] = [T^{-1}][T] = [I]$

iff $[a_{ij}][T^{-1}] = [T^{-1}][a_{ij}] = u_{ij}$ unit matrix I_n

iff $[a_{ij}]$ is non singular and $[a_{ij}]^{-1} = [T^{-1}]$

8.2 Eigenvalues and Eigenvectors.

Definition(8.2.1)

Let X be a linear space over F , and let $T \in L(X)$.

(1) A scalar $\lambda \in F$ is called an eigenvalue of T , if there exists a non zero vector $x \in X$ such that $T(x) = \lambda x$.

(2) A non zero vector $x \in X$ is called an eigenvector of T , if there exists $\lambda \in F$ such that $T(x) = \lambda x$.

Form(1),(2), we say that x is an eigenvector of T associated with eigenvalue λ .

Eigenvalues are some times also called characteristic values, proper values, or spectral values. Similarly eigenvectors are called characteristic vectors, proper vectors, or spectral vectors.

The set of all eigenvalues of T is called the spectrum of T and we shall denote it by $\sigma(T)$.

Remark

If the linear space X has no non zero vectors at all, then T certainly has no eigenvectors. In this case the whole theory collapses into triviality. Therefore throughout the present lector we shall assume that $X \neq \{0\}$.

Examples(8.2.2)

(1) Let $X = \mathbb{R}^2$ and $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $T(x, y) = (-y, x)$ for all $(x, y) \in \mathbb{R}^2$, then T is linear operator has no eigenvalue.

(2) Let $X = \mathbb{R}^2$ and $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $T(x, y) = (x + 2y, 3x + 2y)$ for all $(x, y) \in \mathbb{R}^2$, then T is linear operator have eigenvalues $\lambda = -1, \lambda = 4$.

(3) Let $X = \ell^2$ and $T : \ell^2 \rightarrow \ell^2$ defined by $T(x_1, x_2, x_3, \dots) = (0, x_1, x_2, x_3, \dots)$ for all $(x_1, x_2, x_3, \dots) \in \ell^2$, then T is linear operator has no eigenvalue.

Theorem(8.2.3)

Let X be a linear space over F , and let $T \in L(X)$. If x is an eigenvector of T corresponding to the eigenvalue λ and r is any non zero scalar, then rx is also an eigenvector of T corresponding to the same eigenvalue λ .

Proof:

Since x is an eigenvector of T corresponding to the eigenvalue λ , then $x \neq 0$ and $T(x) = \lambda x$

Since $r \neq 0$ and $x \neq 0 \Rightarrow rx \neq 0 \Rightarrow T(rx) = rT(x) = r(\lambda x) = (r\lambda)x = (\lambda r)x = \lambda (rx)$

Therefore rx is an eigenvector of T corresponding to the eigenvalue λ .

Remark

Corresponding to an eigenvalue λ there may correspond more than one eigenvectors.

Theorem(8.2.4)

Let X be a linear space over F , and let $T \in L(X)$. If x is an eigenvector of T , then x cannot correspond to more than one eigenvalues of T .

Proof :

Let x be an eigenvector of T corresponding to two distinct eigenvalues λ_1 and λ_2 of T
 $T(x) = \lambda_1 x$ and also $T(x) = \lambda_2 x$. Therefore we have $\lambda_1 x = \lambda_2 x \Rightarrow (\lambda_1 - \lambda_2)x = 0$

Since $x \neq 0 \Rightarrow \lambda_1 - \lambda_2 = 0 \Rightarrow \lambda_1 = \lambda_2$

and r is any non zero scalar, then rx is also an eigenvector of T corresponding to the same eigenvalue λ .

Definition (8.2.5)

Let X be a linear space over F , $T \in L(X)$ and let λ be an eigenvalue of T . The set consisting of all eigenvectors of T which correspond to eigenvalue λ together with the vector 0 is called eigenspace of T corresponding to the eigenvalue λ and is denoted by M_λ .

(1) Since by definition an eigenvector is a non zero vector, therefore the set M_λ necessarily contains some non zero vectors.

(2) Since by definition of M_λ a non zero vector x is in M_λ iff $T(x) = \lambda x$. Also it is given that the vector 0 is in M_λ . the vector 0 definitely satisfies the equation $T(x) = \lambda x$. Therefore

$$M_\lambda = \{x \in X : T(x) = \lambda x\} = \{x \in X : (T - \lambda I)(x) = 0\}$$

Thus M_λ is null space(or kernel of) of linear operator $T - \lambda I$ on X .

Hence M_λ is a subspace of X .

(3) Let $x \in X$, since M_λ is a subspace of X and $\lambda \in F \Rightarrow \lambda x \in M_\lambda$

Since $x \in M_\lambda \Rightarrow T(x) = \lambda x \Rightarrow T(x) \in M_\lambda \Rightarrow M_\lambda$ is an invariant under T .

From(1),(2) and (3), we have M_λ is a non zero subspace of X invariant under T .

(4) If X is normed space, and $T \in B(X)$ then M_λ is closed subspace of X

M_λ is called eigenspace of T corresponding to the eigenvalue λ

Characteristic equation of operator

Theorem(8.2.6)

Let $s = \{x_1, x_2, \dots, x_n\}$ be an ordered basis for a finite dimensional linear space X over F , and let T be a linear operator on X whose matrix with respect to s be A and let $\lambda \in F$. Then λ is an eigenvalue of T iff $|A - \lambda I| = 0$

Proof :

Suppose that λ is an eigenvalue of T , then there exist non zero vector $x \in X$ such that $T(x) = \lambda x \Rightarrow T(x) = \lambda I(x) \Rightarrow T(x) - \lambda I(x) = 0 \Rightarrow (T - \lambda I)(x) = 0 \Rightarrow x \in \ker(T - \lambda I)$

Since $x \neq 0 \Rightarrow \ker(T - \lambda I) \neq \{0\} \Rightarrow T - \lambda I$ is not one-one $\Rightarrow T - \lambda I$ is not bijective.

i.e. $T - \lambda I$ is singular (not invertible)

Since A is the matrix of T with respect to the given basis s , then $A - \lambda I$ is the matrix of

$T - \lambda I$ with respect to the given basis s , $\Rightarrow A - \lambda I$ is singular (not invertible) $\Rightarrow |A - \lambda I| = 0$

Conversely : Suppose that $|A - \lambda I| = 0$

$\Rightarrow A - \lambda I$ is singular (not invertible) $\Rightarrow T - \lambda I$ is singular (not invertible)

$\Rightarrow \ker(T - \lambda I) \neq \{0\}$, there exists non zero $x \in X$ such that $(T - \lambda I)(x) = 0$

$\Rightarrow T(x) - \lambda I(x) = 0 \Rightarrow T(x) - \lambda x = 0 \Rightarrow T(x) = \lambda x \Rightarrow T(x) = \lambda x$

Remark

The equation $|A - \lambda I| = 0$ is called **characteristic equation** of T where A is the matrix of T with respect to s . Since $s = \{x_1, x_2, \dots, x_n\}$ and $A = [a_{ij}]_{n \times n}$, then $|A - \lambda I| = 0$ is an equation of n th degree in λ .

Theorem(8.2.7)

A non zero eigenvectors x_1, x_2, \dots, x_n corresponding to eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ of linear operator T on a linear space X over F are linearly independent.

Proof :

We shall prove linear independent by induction method

If $n = 1$

Let $r_1 x_1 = 0$, since $x_1 \neq 0 \Rightarrow r_1 = 0$. Thus the theorem is true for $n = 1$

Suppose the theorem is true for $n = m$. i.e. x_1, x_2, \dots, x_m are linearly independent

We shall prove that $x_1, x_2, \dots, x_m, x_{m+1}$ are linearly independent

Consider the relation $r_1 x_1 + r_2 x_2 + \dots + r_m x_m + r_{m+1} x_{m+1} = 0$ (1)

$\Rightarrow T(r_1 x_1 + r_2 x_2 + \dots + r_m x_m + r_{m+1} x_{m+1}) = T(0) = 0$

$\Rightarrow r_1 T(x_1) + r_2 T(x_2) + \dots + r_m T(x_m) + r_{m+1} T(x_{m+1}) = 0$

Since $T(x_i) = \lambda_i x_i$ for all $i = 1, 2, \dots, n \Rightarrow r_1 \lambda_1 x_1 + r_2 \lambda_2 x_2 + \dots + r_m \lambda_m x_m + r_{m+1} \lambda_{m+1} x_{m+1} = 0$ (2)

Multiplying (1) by λ_{m+1} and subtracting from (2) we get

$r_1 (\lambda_1 - \lambda_{m+1}) x_1 + r_2 (\lambda_2 - \lambda_{m+1}) x_2 + \dots + r_m (\lambda_m - \lambda_{m+1}) x_m = 0$

Since x_1, x_2, \dots, x_m are linearly independent and λ_i are all distinct and as such it follows from above that $r_1 = r_2 = \dots = r_m = 0$. Putting in (1) we get $r_{m+1} x_{m+1} = 0$

Since $x_{m+1} \neq 0 \Rightarrow r_{m+1} = 0$. Hence $x_1, x_2, \dots, x_m, x_{m+1}$ are linearly independent

Thus the theorem is true for all n .

Corollary(8.2.8)

If T is a linear operator on an n dimensional linear space X over F , then T can not have more than n distinct eigenvalues

Proof :

Suppose that T has more than n distinct eigenvalues, then these will form a linearly independent subset of X which will contain more than n vectors. But this is not possible as a n dimensional linear space can not have a linearly independent set containing more than n elements. Hence can not have more than n distinct eigenvalues.

Theorem(8.2.9)

If T be a self –adjoint operator on an n dimensional Hilbert space X over F , then the eigenvalues of T are real and the eigenvectors of T corresponding to distinct eigenvalues are orthogonal.

Proof :

(1) Let λ be the eigenvalue of T so that there exists non zero vector $x \in X$ such that $T(x) = \lambda x$
 Since T is self-adjoint , then $\langle T(x), x \rangle$ is real

$$\text{Now } \langle T(x), x \rangle = \langle \lambda x, x \rangle = \lambda \langle x, x \rangle = \lambda \|x\|^2 \Rightarrow \lambda = \frac{\langle T(x), x \rangle}{\|x\|^2}$$

Since $\|x\|^2 \geq 0$ and $\langle T(x), x \rangle$ is real it follows that λ is real.

(2) Let λ_1, λ_2 be two distinct eigenvalues of T and x_1, x_2 be the corresponding eigenvectors so that $T(x_1) = \lambda_1 x_1, T(x_2) = \lambda_2 x_2$ where λ_1, λ_2 are real . To prove that $x_1 \perp x_2$

Since T is self-adjoint $\Rightarrow T^* = T$, also since λ_1, λ_2 are real $\Rightarrow \overline{\lambda_1} = \lambda_1, \overline{\lambda_2} = \lambda_2$

$$\lambda_1 \langle x_1, x_2 \rangle = \langle \lambda_1 x_1, x_2 \rangle = \langle T(x_1), x_2 \rangle = \langle x_1, T^*(x_2) \rangle = \langle x_1, T(x_2) \rangle = \langle x_1, \lambda_2 x_2 \rangle = \overline{\lambda_2} \langle x_1, x_2 \rangle = \lambda_2 \langle x_1, x_2 \rangle$$

$$\Rightarrow \lambda_1 \langle x_1, x_2 \rangle = \lambda_2 \langle x_1, x_2 \rangle \Rightarrow (\lambda_1 - \lambda_2) \langle x_1, x_2 \rangle = 0$$

Since λ_1, λ_2 are distinct $\Rightarrow \lambda_1 - \lambda_2 \neq 0$, therefore $\langle x_1, x_2 \rangle = 0 \Rightarrow x_1 \perp x_2$, i.e. x_1, x_2 are orthogonal.

Remark

If T is non negative or positive, then the eigenvalues of T are non negative or positive respectively.

Theorem(8.2.10)

If T be an unitary operator on an n dimensional Hilbert space X over F , then the eigenvalues of T are real unimodular and the corresponding distinct eigenvectors are orthogonal.

Proof :

(1) Let λ be a eigenvalue of T , so that there exists non zero $x \in X$ such that $T(x) = \lambda x$

Also since T is an unitary operator, then $\langle T(x), T(x) \rangle = \langle x, x \rangle$

$$\Rightarrow \langle \lambda x, \lambda x \rangle = \langle x, x \rangle \Rightarrow |\lambda|^2 \langle x, x \rangle = \langle x, x \rangle \Rightarrow |\lambda|^2 \|x\|^2 = \|x\|^2 \Rightarrow |\lambda|^2 = 1 \Rightarrow |\lambda| = 1$$

i.e. eigenvalues are unimodular.

(2) Let λ_1, λ_2 be two distinct eigenvalues of T and x_1, x_2 be the corresponding eigenvectors so that $T(x_1) = \lambda_1 x_1, T(x_2) = \lambda_2 x_2$, where λ_1, λ_2 are unimodular.

since T is an unitary operator, then $\langle T(x_1), T(x_2) \rangle = \langle x_1, x_2 \rangle$

$$\Rightarrow \langle \lambda_1 x_1, \lambda_2 x_2 \rangle = \langle x_1, x_2 \rangle \Rightarrow \lambda_1 \overline{\lambda_2} \langle x_1, x_2 \rangle = \langle x_1, x_2 \rangle. \text{ Since } |\lambda_2|^2 = 1 \Rightarrow \overline{\lambda_2} \lambda_2 = 1 \Rightarrow \overline{\lambda_2} = \frac{1}{\lambda_2}$$

$$\Rightarrow \lambda_1 \frac{1}{\lambda_2} \langle x_1, x_2 \rangle = \langle x_1, x_2 \rangle \Rightarrow \lambda_1 \langle x_1, x_2 \rangle = \lambda_2 \langle x_1, x_2 \rangle \Rightarrow (\lambda_1 - \lambda_2) \langle x_1, x_2 \rangle = 0$$

Since λ_1, λ_2 are distinct $\Rightarrow \lambda_1 - \lambda_2 \neq 0$, therefore $\langle x_1, x_2 \rangle = 0 \Rightarrow x_1 \perp x_2$.

Theorem(8.2.11)

Let T be a normed operator on a finite dimensional Hilbert space X over F

- (1) If $\lambda \in \sigma(T)$, then $T - \lambda I$ is normal
- (2) Every eigenvector of T is also a eigenvector for T^* .
- (3) the eigenspaces of T are pair wise orthogonal

Proof :

(1) Since T is normal $\Rightarrow T \circ T^* = T^* \circ T$

$$\begin{aligned} (T - \lambda I)^* &= T^* - \bar{\lambda} I^* = T^* - \bar{\lambda} I \Rightarrow (T - \lambda I)(T - \lambda I)^* = (T - \lambda I)(T^* - \bar{\lambda} I) = TT^* - \lambda T^* - \bar{\lambda} T + \lambda \bar{\lambda} \\ &\Rightarrow (T - \lambda I)^*(T - \lambda I) = (T^* - \bar{\lambda} I)(T - \lambda I) = T^*T - \lambda T^* - \bar{\lambda} T + \lambda \bar{\lambda} \Rightarrow TT^* - \lambda T^* - \bar{\lambda} T + \lambda \bar{\lambda} \\ &\Rightarrow (T - \lambda I)(T - \lambda I)^* = (T - \lambda I)^*(T - \lambda I) \Rightarrow T - \lambda I \text{ is normal} \end{aligned}$$

(2) Let x be an eigenvector of T corresponding to eigenvalue $\lambda \Rightarrow T(x) = \lambda x$

$$\|T(x)\|^2 = \langle T(x), T(x) \rangle = \langle \lambda x, \lambda x \rangle = \langle x, T^*(\lambda x) \rangle = \langle x, T(T^*(x)) \rangle = \langle T^*(x), T^*(x) \rangle = \|T^*(x)\|^2 \Rightarrow \|T(x)\| = \|T^*(x)\|$$

Since $T - \lambda I$ is normal, therefore $x \in X$, we have

$$\|(T - \lambda I)(x)\| = \|(T - \lambda I)^*(x)\| \Rightarrow \|T(x) - \lambda x\| = \|(T^* - \bar{\lambda} I)(x)\| = \|T^*(x) - \bar{\lambda} x\|$$

Since $T(x) = \lambda x \Rightarrow 0 = \|T^*(x) - \bar{\lambda} x\|$. Hence it follows that $T^*(x) = \bar{\lambda} x$, therefore x is eigenvector of T^* and corresponding eigenvalue is $\bar{\lambda}$

(3) Let x_i and x_j belong to M_i and M_j the eigenspaces of T and the corresponding eigenvalues be λ_i and λ_j respectively so that $T(x_i) = \lambda_i x_i$, $T(x_j) = \lambda_j x_j$ and $T^*(x_j) = \bar{\lambda}_j x_j$ as T is normal

$$\begin{aligned} \lambda_i \langle x_i, x_j \rangle &= \langle \lambda_i x_i, x_j \rangle = \langle T(x_i), x_j \rangle = \langle x_i, T^*(x_j) \rangle = \langle x_i, \bar{\lambda}_j x_j \rangle = \bar{\lambda}_j \langle x_i, x_j \rangle = \lambda_j \langle x_i, x_j \rangle \\ &\Rightarrow \lambda_i \langle x_i, x_j \rangle = \lambda_j \langle x_i, x_j \rangle \Rightarrow (\lambda_i - \lambda_j) \langle x_i, x_j \rangle = 0 \end{aligned}$$

Since λ_i, λ_j are distinct $\Rightarrow \lambda_i - \lambda_j \neq 0$, therefore $\langle x_i, x_j \rangle = 0 \Rightarrow x_i \perp x_j$, i.e. M_i and M_j are pair wise orthogonal.

Theorem(8.2.12)

If T be a normal operator on an n dimensional Hilbert space X over F , then each eigenspace reduces T .

Proof :

Let x_i belong to M_i the eigenspace of T and the corresponding eigenvalue be λ_i so that

$$T(x_i) = \lambda_i x_i$$

Since T is normal $\Rightarrow T^*(x) = \bar{\lambda} x$

Since M_i is a subspace $\Rightarrow \bar{\lambda}_i x \in M_i \Rightarrow T^*(x) \in M_i \Rightarrow M_i$ is invariant under T^* , but M_i is invariant under T . Hence M_i is reduces T .

8.3 Spectral Theorem for Normal Operators

Theorem(8.3.1) Spectral theorem for normal operators

Let T be an arbitrary linear operator on finite dimensional Hilbert space X , and $\lambda_1, \lambda_2, \dots, \lambda_n$ be eigenvalues of T with eigenspaces M_1, M_2, \dots, M_n . Further P_1, P_2, \dots, P_n are perpendicular Projections on the spaces M_1, M_2, \dots, M_n respectively. Then the spectral theorem states that the following statements are equivalent.

- (1) The subspaces M_1, M_2, \dots, M_n are pair wise orthogonal and span X
- (2) P_1, P_2, \dots, P_n are pair wise orthogonal and (i) $P_1 + P_2 + \dots + P_n = I$ (ii) $\lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n = T$
- (3) T is a normal operator.

Proof :

(1) \Rightarrow (2)

Since P_1, P_2, \dots, P_n are perpendicular Projections on the spaces M_1, M_2, \dots, M_n respectively. Also the subspaces M_1, M_2, \dots, M_n are pair wise orthogonal and span X , i.e. $X = M_1 \oplus M_2 \oplus \dots \oplus M_n$. Hence any $x \in X$ can be uniquely expressed as $x = x_1 + x_2 + \dots + x_n$, where $x_i \in M_i$

Since M_i, M_j are subspaces of X and P_i, P_j are perpendicular projections on M_i, M_j respectively then M_i, M_j are orthogonal iff $P_i \circ P_j = 0$ iff $P_j \circ P_i = 0$

Since M_i, M_j are orthogonal, then $P_i \circ P_j = 0, i \neq j$

P_i is projection on M_i and $x = x_1 + x_2 + \dots + x_n, x_i \in M_i$

Since $x_j \in M_j$ and $M_j \perp M_i \Rightarrow x_j \in M_i^\perp$. But M_i^\perp is null space of P_i and hence $P_i(x_j) = 0, i \neq j$

Thus $P_i(x_i) = x_i$ for all i and $P_i(x_j) = 0, i \neq j \Rightarrow P_i(x) = P_i(x_1 + x_2 + \dots + x_n) = P_i(x_i) = x_i$

Now $I(x) = x = x_1 + x_2 + \dots + x_n = P_1(x) + P_2(x) + \dots + P_n(x) = (P_1 + P_2 + \dots + P_n)(x)$

Since above is true for x It follows that $P_1 + P_2 + \dots + P_n = I$.

Now $x_i \in M_i$ the eigenspace of T corresponding to eigenvalue $\lambda_i \Rightarrow T(x_i) = \lambda_i x_i$

$T(x) = T(x_1 + x_2 + \dots + x_n) = T(x_1) + T(x_2) + \dots + T(x_n) = \lambda_1 x_1 + \lambda_2 x_2 + \dots + \lambda_n x_n$

$T(x) = \lambda_1 P_1(x) + \lambda_2 P_2(x) + \dots + \lambda_n P_n(x) = (\lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n)(x)$

Since above holds for all x we have $T = \lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n$

(2) \Rightarrow (3)

Since $T = \lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n \Rightarrow T^* = (\lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n)^* = (\lambda_1 P_1)^* + (\lambda_2 P_2)^* + \dots + (\lambda_n P_n)^*$

Since $(\lambda_i P_i)^* = \overline{\lambda_i} P_i^*$ and $P_i^* = P_i = P_i^2 \Rightarrow T^* = \overline{\lambda_1} P_1^* + \overline{\lambda_2} P_2^* + \dots + \overline{\lambda_n} P_n^* = \overline{\lambda_1} P_1 + \overline{\lambda_2} P_2 + \dots + \overline{\lambda_n} P_n$

$T \circ T^* = (\lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n)(\overline{\lambda_1} P_1^* + \overline{\lambda_2} P_2^* + \dots + \overline{\lambda_n} P_n^*) = \sum_{i=1}^n \lambda_i \overline{\lambda_i} P_i^2 + \sum_{i \neq j} \lambda_i \overline{\lambda_j} P_i P_j, i \neq j$

$\Rightarrow T \circ T^* = \sum_{i=1}^n |\lambda_i|^2 P_i + 0 = \sum_{i=1}^n |\lambda_i|^2 P_i$ in a similar manner we can show that $T^* \circ T = \sum_{i=1}^n |\lambda_i|^2 P_i$

Hence $T \circ T^* = T^* \circ T$ and as such T is normal operator .

(3) \Rightarrow (1)

Theorem(5.3.2) Uniqueness of spectral resolution of a normal operator

The spectral resolution of a normal operator on a finite dimensional Hilbert space X is unique.

Another form. Let T be a normal operator on finite dimensional Hilbert space X . If $\sum_{i=1}^n \lambda_i P_i$ is the spectral form of T , then λ_i are all the distinct eigenvalues of T . If more $1 \leq k \leq n$ then there exists polynomials P_k with complex coefficients such that $P_k(\lambda_i) = 0$ whenever $i \neq k$ and $P_k(\lambda_k) = 1$. For all such polynomials $P_k(T) = Q_k$, i.e. each Q_k is a polynomial in T .

Proof :

Theorem(5.3.3)

Let T be a normal operator on a finite dimensional Hilbert space X , then X has an orthonormal basis s consisting of eigenvectors of T . Consequently the matrix of T relative to s is a diagonal matrix.

5.4 Spectral theorem for Self adjoint Operators

Theorem(5.4.1)

Let T be a self-adjoint operator on finite dimensional Hilbert space X , then there exists n real numbers $\lambda_1, \lambda_2, \dots, \lambda_n$ and perpendicular projections P_1, P_2, \dots, P_n , (where $n > 0$, and $n \leq$ the dimension of X) such that

- (1) $\lambda_i, i = 1, 2, \dots, n$ are pair wise distinct
- (2) P_1, P_2, \dots, P_n are pair wise orthogonal and different from zero.
- (3) $P_1 + P_2 + \dots + P_n = I$
- (4) $\lambda_1 P_1 + \lambda_2 P_2 + \dots + \lambda_n P_n = T$

Proof :

Theorem(5.4.2)

Let T be a self-adjoint operator on finite dimensional Hilbert space X . If $\sum_{i=1}^n \lambda_i P_i$ is the spectral form of T , then λ_i are all the distinct eigenvalues of T . If moreover $1 \leq k \leq n$ then there exists polynomials P_k with real coefficients such that $P_k(\lambda_i) = 0$ whenever $i \neq k$ and $P_k(\lambda_k) = 1$. For all such polynomials $P_k(T) = Q_k$, i.e. each Q_k is a polynomial in T .

Theorem(5.4.3)

Let T be a self-adjoint operator on finite dimensional Hilbert space X such that $T = \sum_{i=1}^n \lambda_i P_i$.

If S is any linear transformation on X , then S commutes with T iff S commutes with each P_i for $i = 1, 2, \dots, n$

Proof :

Suppose that S commutes with each P_i , i.e. $S \circ P_i = P_i \circ S$ for all i . To prove that : $S \circ T = T \circ S$

Since $T = \sum_{i=1}^n P_i$

$$\Rightarrow S \circ T = S \circ \left(\sum_{i=1}^n P_i \right) = \sum_{i=1}^n (S \circ P_i) = \sum_{i=1}^n (P_i \circ S) = \sum_{i=1}^n (P_i \circ S) = \left(\sum_{i=1}^n P_i \right) \circ S = T \circ S$$

Conversely : Suppose that S commutes with T , i.e. $S \circ T = T \circ S$. To prove that : $S \circ P_i = P_i \circ S$ for all i

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