

2008.01.14

①

Review of linear transformations in finite-dim'd inner product spaces

A good reference (among many) is

Linear Algebra by Friedberg, Insel, Spence
3rd ed. Prentice Hall 1997

The relevant chapter, Chapter 6, can be downloaded in pdf from the course website.

Defn Let V be a vector space over $\mathbb{C}$. An inner product in V is a function $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{C}$ with the following properties:

$$* \langle u+v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$$

$$* \langle cu, v \rangle = c \langle u, v \rangle$$

$$* \overline{\langle u, v \rangle} = \langle v, u \rangle$$

$$* \langle u, u \rangle > 0 \text{ if } u \neq 0$$

The following properties are implied by the above:

$$* \langle u, v+w \rangle = \langle u, v \rangle + \langle u, w \rangle$$

$$* \langle u, cv \rangle = \bar{c} \langle u, v \rangle$$

$$* \langle 0, u \rangle = 0$$

The map $V \rightarrow \mathbb{R} : v \mapsto \langle v, v \rangle^{1/2}$ is a norm on V .

[Recall the properties of a norm.]

2008.01.14

(2)

The adjoint T^* of a linear transformation T is defined through $\langle Tx, y \rangle = \langle x, T^*y \rangle \quad \forall x, y \in V$
 [Show that this is well-defined.]

Jordan canonical form (all transformations)

diagonalizable (min polyn. has simple roots)

orthogonally diagonalizable (T is normal: $TT^* - T^*T = 0$)

(T is self-adjoint $T^* = T$)

orthog. diag. with real e.vals

partial isometry

isometry ($|T(x)| = |x|$, or $TT^* = I$)

(T is positive $\langle T(x), x \rangle > 0 \quad \forall x \in V$)

orthog. diag. with real positive e.vals

unitary ($TT^* = T^*T = I$, or normal and $|\lambda| = 1 \quad \forall$ e.vals λ)

sq. rt. of Id. and orthog. diag. ($T^2 = I$)

identity ($T = I$)

Let V be a complex inner-product space.
 $\dim V = n < \infty$.

The Spectral Theorem for finite-dim'l inner-product spaces.

[see Friedberg, et al., for example]

A linear transformation $T: V \rightarrow V$ is normal if and only if it is orthogonally diagonalizable.

Let us briefly look at the main points in a proof.

First observe that, if $\mathcal{B} = \{e_i\}_{i=1}^n$ is a basis for V and $[T]_{\mathcal{B}}$ denotes the matrix for T with respect to $\mathcal{B}$, then if $\mathcal{B}$ is orthonormal, the adjoint T^* of T is represented by the conjugate of the transpose of $[T]_{\mathcal{B}}$. [The conj. transpose of a matrix A is denoted by A^*]:

$$[T^*]_{\mathcal{B}} = [T]_{\mathcal{B}}^*.$$

1. If T is orthogonally diagonalizable, there is an orthonormal basis $\mathcal{B}$ such that $A := [T]_{\mathcal{B}}$ is diagonal. Therefore $A^* = [T^*]_{\mathcal{B}}$ is diagonal and therefore commutes with A . Thus T commutes with T^* , so T is normal.
2. Let T be normal. First we use Thm. 6.14 in [FIS] to obtain an orthonormal basis for V such that $[T]_{\mathcal{B}} =: A$ is upper triangular. By computing successive diagonal elements of $AA^* - A^*A = 0$, one shows that the off-diagonal elements must be zero.

Our immediate objective is to present the spectral theorem for finite dimensional inner-product spaces in such a way that it can be easily generalized into the statement of the spectral theorem for (normal operators in) infinite-dimensional Hilbert spaces.

We begin with considering the meaning of the statement that T is orthogonally diagonalizable. This means that

- T has distinct eigenvalues $\lambda_1, \dots, \lambda_k$ (as does any operator!).
The spectrum of $T = \sigma(T) = \{\lambda_1, \dots, \lambda_k\}$ (closed)
The resolvent set of $T = \rho(T) = \mathbb{C} \setminus \sigma(T)$ (open)
- The corresponding eigenspaces $W_1, \dots, W_k$ are mutually orthogonal and

$$(*) \quad V = \bigoplus_{i=1}^k W_i \quad (\text{as a direct sum of inner prod. spaces})$$

Let P_i be the (orthogonal) projection onto W_i ($\text{Ran } P_i = W_i$)

The statement (*) is equivalent to the statement

$$\sum_{i=1}^k P_i = I,$$

Notice that $\text{Ran } P_j = W_j$ and $\text{Null } P_j = \bigoplus_{\substack{i=1, \\ i \neq j}}^k P_i =: V \ominus P_j$

$$\bullet \quad T = \sum_{i=1}^k \lambda_i P_i.$$

Note: The word "projection" P in an inner-product space refers to an orthogonal projection, that is $P^2 = P$ and $\text{Ran } P \perp \text{Null } P$.

The Spectral Theorem (finite dim.), reformulated.

Let V be a finite-dim'l inner-product space and $T: V \rightarrow V$ a linear transformation. Then the following two statements are equivalent.

1. T is normal

2. There exists a finite set $\sigma(T) \subset \mathbb{C}$ and projections $\{P_\lambda: \lambda \in \sigma(T)\}$ such that

$$* P_\lambda P_\mu = \delta_{\lambda\mu} P_\lambda \quad [\text{Ran } P_\lambda \perp \text{Ran } P_\mu \text{ for } \lambda \neq \mu]$$

$$* \sum_{\lambda \in \sigma(T)} P_\lambda = I \quad [V = \underbrace{\bigoplus_{\lambda \in \sigma(T)} \text{Ran } P_\lambda}_{W_\lambda}]$$

$$* \sum_{\lambda \in \sigma(T)} \lambda P_\lambda = T \quad [“T \text{ is diagonal}”]$$

Notice that the statement $T = \sum_{\lambda \in \sigma(T)} \lambda P_\lambda$ is saying that

T is a “multiplication operator” in the sense that, with respect to the decomposition $V = \bigoplus_{\lambda \in \sigma(T)} W_\lambda$, T just

multiples the components by the eigenvalues — it does not couple the components :

$$T(v_{\lambda_1} + v_{\lambda_2} + \dots + v_{\lambda_k}) = \lambda_1 v_{\lambda_1} + \lambda_2 v_{\lambda_2} + \dots + \lambda_k v_{\lambda_k}$$

By choosing an orthonormal basis for each W_i and taking the union of all these bases, we obtain an orthonormal basis for V with respect to which T is diagonal, or a “multiplication operator”

By choosing an orthonormal basis for each W_i and taking the union of all these bases, we obtain an orthonormal basis $\mathcal{B}$ for V with respect to which T is diagonal, or a "multiplication operator", when elements of V are represented by n -tuples with respect to $\mathcal{B}$:

$$[T]_{\mathcal{B}} \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} \lambda_1 a_1 \\ \vdots \\ \lambda_n a_n \end{bmatrix}, \text{ or } [T]_{\mathcal{B}} = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix},$$

where the λ_i are not necessarily distinct.

Another way to say this is that the map

$$U: \mathbb{C}^n \rightarrow V \text{ such that } (0, 0, \dots, \overset{\substack{\uparrow \\ i\text{th place}}}{1}, \dots, 0) = e_i$$

is unitary and $U^{-1} T U$ is a diagonal matrix Λ :

The Spectral Theorem (fin. dim.), reformulated.

Let V be a fin. dim. inner-prod. space and $T: V \rightarrow V$ linear.

Then T is normal if and only if there exists a unitary operator $U: \mathbb{C}^n \rightarrow V$ and a diagonal matrix Λ such that

$$T = U \Lambda U^{-1}$$

Some observations

- $(T - \lambda)^* = T^* - \bar{\lambda}$
- $\sigma(T^*) = [\sigma(T)]^*$
- If T is normal, then $T - \lambda$ and $T^* - \bar{\lambda}$ have the same nullspaces (the eigenspaces of T and T^* coincide).
- If T is normal, then " $\sigma(T) \subset \mathbb{R}$ " is equivalent to $T = T^*$, that is, those normal operators that are self-adjoint are exactly those with real spectrum.

Note: All of these statements can be proven quite directly from the definitions of "adjoint" and "normal", without resorting to matrix representations. One of the key observations is that

Thm

If W_1 and W_2 are subspaces of V such that $W_1 = \{v \in V : T(v) \in W_2\}$ (W_1 is the preimage of W_2 under T),

then $W_2^\perp = \{v \in V : T^*(v) \in W_1^\perp\}$ ($W_2^\perp$ is the preimage of $W_1^\perp$

under T^*)
Furthermore, $T(W_1) = W_2$ and $T^*(W_2^\perp) = W_1^\perp$.

Actually, you only need the weaker statement

Cor

$$\text{Ran}(T^*) = [\text{Null}(T)]^\perp.$$

Since $T^{**} = T$, we have also $\text{Ran } T = [\text{Null}(T^*)]^\perp$.

This is a concise statement of the "Fredholm alternative" for operators in fin. dim. inner-prod. spaces.

(6)

Observations, continued

Let T be normal and $\sigma(T)$ and P_λ for $\lambda \in \sigma(T)$ be defined as in the spectral theorem. Observe that

$$\bullet \|v\|^2 = \sum_{\lambda \in \sigma(T)} \|P_\lambda v\|^2$$

$$\bullet \|Tv\|^2 = \sum_{\lambda \in \sigma(T)} |\lambda|^2 \|P_\lambda v\|^2 \leq \left(\max_{\lambda \in \sigma(T)} |\lambda|^2 \right) \sum_{\lambda \in \sigma(T)} \|P_\lambda v\|^2 \\ = \left(\max_{\lambda \in \sigma(T)} |\lambda|^2 \right) \|v\|^2$$

with equality if and only if $v \in W_{\lambda^*}$ ($\lambda^* = \text{e.val of max modulus}$)

$$\bullet \Rightarrow \|T\| = \sup_{\|v\|=1} \|T(v)\| = \max_{\lambda \in \sigma(T)} |\lambda|.$$

Polynomial functions of T (normal)

Let $\mathcal{P}$ denote the space of polynomials over $\mathbb{C}$ in one indeterminate: $\mathcal{P} = \mathbb{C}[x]$

$$T = \sum_{\lambda \in \sigma(T)} \lambda P_\lambda \Rightarrow T^n = \sum_{\lambda \in \sigma(T)} \lambda^n P_\lambda$$

$$\Rightarrow \text{for } p \in \mathcal{P},$$

$$p(T) = \sum_{\lambda \in \sigma(T)} p(\lambda) P_\lambda$$

Consequences:

- $p(T)$ is normal with adjoint $\bar{p}(T^*)$, where $\bar{p}$ is obtained from p by conjugating the coefficients.

$$\bullet \sigma(p(T)) = p[\sigma(T)] := \{p(\lambda) : \lambda \in \sigma(T)\}.$$

This is the "spectral mapping theorem".

(**)

$$\bullet \|p(T)\| = \max_{\mu \in \sigma(p(T))} |\mu| = \max_{\lambda \in \sigma(T)} |p(\lambda)| = \|p\|_{L^\infty(\sigma(T))}$$

Continuous functions of a normal operator $T: V \rightarrow V$, $\dim V < \infty$

This presentation will seem too technical for the simple case of finite dimension, but remember that it is for the purpose of setting the framework for the infinite-dim. case.

$$C(\sigma(T)) = \overline{P}(\sigma(T)) = \mathbb{C}^{\sigma(T)} = \text{set of all complex-valued functions on } \sigma(T) \text{ with the "uniform" norm } \|f\| = \max_{\lambda \in \sigma(T)} |f(\lambda)|$$

(continuous fns) (all fns)

Three coincide because $\sigma(T)$ is finite.

We now define a map $\varphi: C(\sigma(T)) \rightarrow \mathcal{L}(V)$ that is an isometric homomorphism

$\mathcal{L}(V)$ is the space of linear transformations from V to V with the operator norm.

- Given the spectral theorem, we can do this directly by

$$\varphi: f \mapsto \sum_{\lambda \in \sigma(T)} f(\lambda) P_\lambda$$

- A more abstract framework arises from this:

$$\Phi = \Phi_k \oplus \text{Null}(\Phi)$$

Φ is defined by restricting a polynomial to $\sigma(T)$:

$$[\Phi(p)](\lambda) = p(\lambda) \text{ for } \lambda \in \sigma(T)$$

Φ is defined by applying a polynomial to T :

$$\Phi(p) = p(T).$$

$\mathcal{L}(V) \leftarrow \varphi \leftarrow C(\sigma(T))$

$\|S\| = \max_{\|v\|=1} |S(v)|$

$\|f\| = \max_{\lambda \in \sigma(T)} |f(\lambda)|$

By (**), we see that $\|\Phi(p)\| = \|\Psi(p)\|$, so $\text{Null } \Phi = \text{Null } \Psi$, and since Ψ is surjective, the homomorphism φ exists:

$$\varphi(f) = \Phi(p), \text{ where } p \text{ is chosen in the preimage } \Phi^{-1}[f]$$

We denote $\varphi(f)$ by $f(T)$.

You may show that φ is a $*$ -homomorphism of algebras. This means that it preserves all of the algebraic structure and, in addition, takes the conjugates to adjoints:

$$\varphi: C(\sigma(\tau)) \longrightarrow \mathcal{L}(V)$$

$$\varphi(af + bg) = a\varphi(f) + b\varphi(g)$$

$$\varphi(fg) = \varphi(f)\varphi(g)$$

$$\varphi(1) = I$$

$$\varphi(\bar{f}) = (\varphi(f))^*$$