

(Axler §7B) last time:

- Thm if $T : V$ to V is normal, then:
- 1–2) $\ker(T^*) = \ker(T)$, $\text{im}(T^*) = \text{im}(T)$
 - 3) $T - \lambda$ is normal for all λ in F
 - 4) $\ker(T - \lambda) = \ker(T^* - \bar{\lambda})$

today, we work over $F = \mathbb{C}$

Spectral Thm

if V is finite-dim'l over \mathbb{C} and $T : V$ to V is normal
then T is diagonalizable

in fact:

V has a basis of orthonormal eigenvectors for T

Restatement in Matrices

let (e_1, \dots, e_n) be any orthonormal basis for V
A the matrix of T wrt $(e_i)_i$
let (u_1, \dots, u_n) be the basis of
orthonormal eigenvectors for T
 λ_i defined by $Tu_i = \lambda_i u_i$
P the $n \times n$ matrix defined by $Pe_i = u_i$
D the $n \times n$ diagonal matrix with diagonal
 $\lambda_1, \dots, \lambda_n$

[what's A in terms of P, D?] then $A = P^{-1}DP$

Note 1 we proved last time: if T is self-adjoint,
not just normal, then the λ_i 's are all real

Note 2 the cols of P expand the u_i 's into e_i 's
but the u_i 's are orthonormal, so

$$PP^* = I$$

that is: $Pu \cdot (Pv)^- = u^t PP^* v^- = u \cdot v^-$ for all u, v

Df recall:
an operator Q is orthogonal wrt \langle, \rangle
iff $\langle Qu, Qv \rangle = \langle u, v \rangle$ for all u, v

if $F = \mathbb{C}$, then we often say “unitary” rather than
“orthogonal”

we say a matrix P is unitary iff $Pu \cdot (Pv)^- = u \cdot v^-$
[which occurs] iff $PP^* = I$

Pf of Thm induct on $n := \dim V$
if $n = 0$, then done

suppose $n > 0$
then [recall:] T has an eigenvector v , say,
with eigenvalue λ

the line Cv is T -stable
let $W = (Cv)^\perp = \{w \text{ in } V \mid \langle w, v \rangle = 0\}$
recall that the Gram–Schmidt process shows

$$V = Cv + W \quad \text{and this sum is direct}$$

so it remains to show:

Claim W is T -stable

Claim Finishes Pf note $\dim W = n - 1$

by inductive hypothesis, W has a basis of orthonormal eigenvectors u_1, \dots, u_{n-1} all are orthogonal to v
now set $u_n = v/\|v\|$ \square

Pf of Claim pick w in W
want T_w in W : that is, $\langle T_w, v \rangle = 0$

know $\langle Tw, v \rangle = \langle w, T^*v \rangle$

but [recall!] $v \in \ker(T - \lambda) = \ker(T^* - \lambda^-)$

$$\begin{aligned} \text{now, } \langle w, T^*v \rangle &= \langle w, \lambda^-v \rangle \\ &= \lambda^- \langle w, v \rangle \\ &= 0 \end{aligned}$$

Rem claim + its proof generalize:

if $T : V \rightarrow V$ is normal and $U \subseteq V$ is T^* -stable
then U^\perp is T -stable

Applications

Cor if $TT^* = T^*T$ and all eigenvalues of T are real and nonnegative, then $T = S^*S$ for some $S : V$ to V

in particular, $S^* = S$

[because $(S^*S)^* = S^*S^{**} = S^*S$]

Pf pick a basis of orthonormal eigenvectors
the matrix of T in this basis is diagonal

call it D

let C be diagonal s.t. $C^2 = D$

let $S : V \rightarrow V$ be the op with matrix C in that basis

Cor TFAE for an $n \times n$ matrix A:

- 1) the pairing $\langle u, v \rangle := u^t A v^-$ is
an inner product
- 2) A is Hermitian and positive-definite
[pos-def: $v^t A v^- > 0$ for $v \neq \mathbf{0}$]
- 3) $A = B B^*$ for some invertible $B : V \rightarrow V$

Pf the direction 1) implies 2) implies 3)
are PS8, #8, part (1)

[use the previous corollary]

[why do we need B invertible?]

conversely, if $A = B^* B$, then:

A is Hermitian, via the argument earlier

$$[(B^* B)^* = B^* B^{**} = B^* B]$$

$$v^t A v^- = v^t B^* B v^- = (B^- v)^t (B^- v)^- > 0$$

since the skew-dot product is pos-def
so \langle, \rangle is pos-def and conj-symmetric

Df for general square B,
 $B^* B$ is called the Gram matrix
its eigenvals are all real and nonnegative
their sq roots are the singular vals of B

similar lingo for linear operators

[why useful? just as vector in an inner product
space get norms, so too do operators on it]

Df the L^2 operator norm of $S : V$ to V is

$$\|S\| = \max_{\{v \text{ s.t. } \|v\| = 1\}} \|Sv\|$$

i.e., the largest factor by which S rescales the norm of a vector

Cor $\|S\| = \max \{\text{singular values of } S\}$

Pf 1 pick a basis of orthonormal eigenvectors for S^*S

now, e.g., Lagrange multipliers show:

$\|Sv\|^2 = \langle Sv, Sv \rangle = \langle v, S^*Sv \rangle$ is maximized on $\{\|v\| = 1\}$ when v is an eigenvector for the largest eigenvalue of S^*S

Pf 2 in some orthonormal basis, the matrix of S^*S looks like $P^{-1}DP$ with P orthogonal and D diagonal [so, enough to show:]

$$\begin{aligned} & \max_{\{v \text{ s.t. } \|v\| = 1\}} \|P^{-1}DPv\| \\ &= \max_{\{v \text{ s.t. } \|v\| = 1\}} \|Dv\| \end{aligned}$$

indeed, the set $\{v \mid \|v\| = 1\}$ is stable under unitary ops like P , P^{-1}

more general result [has a “Min-Max” version]:

Thm (Max-Min) $\max_{\{\dim U = i\}} \min_{\{v \in U \mid \|v\| = 1\}} \|Sv\|$
= i th largest singular value of S

[since $\|v\| = 1$ iff $\|v^-\| = 1$ and $\dim U = \dim U^-$:]

Cor S and S^* have the same singular vals
i.e.,
 S^*S and SS^* have the same eigenvals