

MATH 2040 Lecture 18 (10/11/2016)

Spectral Thm: T self-adjoint ($\mathbb{F} = \mathbb{R}$) or normal ($\mathbb{F} = \mathbb{C}$)
 $\Leftrightarrow \exists$ orthonormal eigenbasis

Question: How to understand it geometrically?

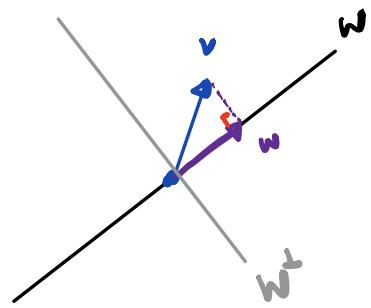
Recall: $W \subseteq V$ subspace

$$\Rightarrow \text{decompose } V = W \oplus W^\perp$$

$$v = w + w^\perp$$

Define $T = \text{proj}_W^\perp : V \rightarrow V$

$$v \mapsto w$$



Note: (1) T linear.

$$(2) T^2 = T$$

$$(3) R(T) = W, N(T) = W^\perp (= R(T)^\perp)$$

$$(4) T^* = T$$

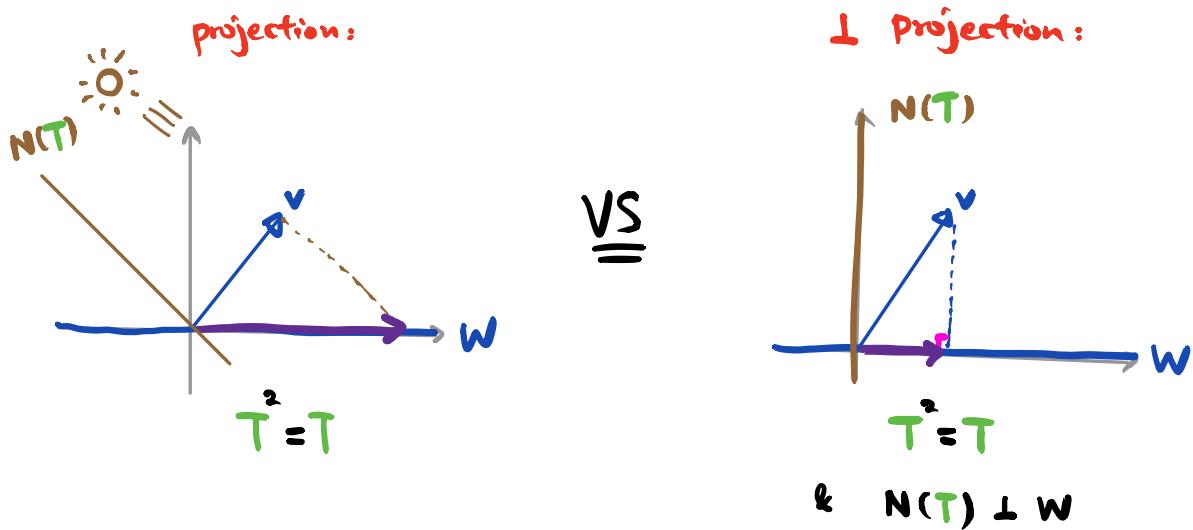
Reason: $V = W \oplus W^\perp$

On. basis $\beta = \gamma_1 \cup \gamma_2$

$$[T]_\beta = \begin{pmatrix} I & 0 \\ 0 & 0 \end{pmatrix} \quad \text{self-adjoint}$$

Defn: A linear map $T: V \rightarrow V$ is orthogonal projection

if ① $T^2 = T$ & ② $R(T)^\perp = N(T)$
 (projection)



Length-min. property: $\|Tv\| \leq \|v\|$ if T is \perp projection.

Prop: (1) T orthogonal projection

$$\Leftrightarrow (2) T^2 = T = T^*$$

$$\Leftrightarrow (3) \exists W \subset V \text{ subspace st. } T = \text{proj}_W^\perp.$$

Proof: (3) \Rightarrow (2) done!

$$(2) \Rightarrow (1) R(T)^\perp = N(T^*) \stackrel{\text{self adj.}}{=} N(T)$$

$$(1) \Rightarrow (3) \text{ Know: } T^2 = T \stackrel{(*)}{=} \text{ and } R(T)^\perp = N(T) \stackrel{(*)}{=}.$$

Define $W = R(T) \subseteq V$ subspace.

Check: $T = \text{proj}_W^\perp$.

$$V = R(T) \oplus R(T)^\perp \stackrel{(*)}{=}$$

$$= W \oplus N(T)$$

$$T(v = Tu + y) \Rightarrow Tv = T^2 u + 0$$

_____ .

T diagonalizable $\Leftrightarrow V = E_{\lambda_1} \oplus \cdots \oplus E_{\lambda_k}$

T diagonalizable
by an O.N.B.
 $\Leftrightarrow V = E_{\lambda_1} \oplus \cdots \oplus E_{\lambda_k}$
and $\perp \perp$

Spectral Decomposition Thm:

Assume $T : V \rightarrow V$ ^(C) normal / self-adjoint.

If T has distinct eigenvalues $\lambda_1, \dots, \lambda_k$, then

$$V = E_{\lambda_1} \oplus \cdots \oplus E_{\lambda_k} \quad \& \quad E_{\lambda_i} \perp E_{\lambda_j}$$

and

$$T = \lambda_1 T_1 + \cdots + \lambda_k T_k$$

$$T_i = \text{proj}_{E_{\lambda_i}}^{\perp}$$

matrix: β O.N.B.

$$[T]_{\beta} = \begin{pmatrix} & \lambda_1 & & \\ & & \ddots & \\ & & & \lambda_1 \\ \dots & & & \dots \\ & & & & \lambda_k & & \\ & & & & & \ddots & \\ & & & & & & \lambda_k \end{pmatrix}$$

$$V = E_{\lambda_1} \oplus \cdots \oplus E_{\lambda_k}$$

$$Tv = T v_1 + \cdots + T v_k = \lambda_1 v_1 + \cdots + \lambda_k v_k$$

$$T = \lambda_1 T_1 + \cdots + \lambda_k T_k$$

Applications:

Cor 1: $g(T) = g(\lambda_1)T_1 + \dots + g(\lambda_k)T_k$ Ex.

for any polynomial $g(x)$.

$$\text{Ex.: } T = \lambda_1 T_1 + \lambda_2 T_2 \Rightarrow T^2 + 2T = (\lambda_1^2 + 2\lambda_1)T_1 + (\lambda_2^2 + 2\lambda_2)T_2$$

$$[T]_P = \begin{pmatrix} \lambda_1 I & 0 \\ 0 & \lambda_2 I \end{pmatrix} \Rightarrow [T^k]_P = \begin{pmatrix} \lambda_1^k I & 0 \\ 0 & \lambda_2^k I \end{pmatrix}$$

Cor 2: $\mathbb{F} = \mathbb{C}$. T normal $\Leftrightarrow T^* = g(T)$ for some polynomial g .

Pf: (\Leftarrow) easy. $T^*T = TT^*$ ✓

(\Rightarrow) T normal

$$\Rightarrow T = \lambda_1 T_1 + \dots + \lambda_k T_k \quad (*)$$

$$\Rightarrow T^* = (\lambda_1 T_1 + \dots + \lambda_k T_k)^*$$

$$= \bar{\lambda}_1 T_1^* + \dots + \bar{\lambda}_k T_k^*$$

Q: \exists polynomial g

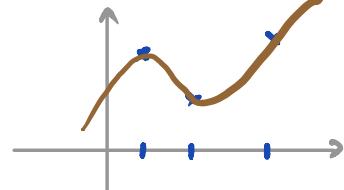
st. $g(\lambda_i) = \bar{\lambda}_i \forall i$.

A: Yes, if we allow large degree.

$$= \bar{\lambda}_1 T_1 + \dots + \bar{\lambda}_k T_k$$

$$\stackrel{?}{=} g(\lambda_1) T_1 + \dots + g(\lambda_k) T_k$$

$$= g(T)$$



"Lagrange Interpolation".

□