

Lecture 35

7.D Polar Decomposition

Last time: $T \in \mathcal{L}(V)$ is positive
if T is self-adjoint and $\langle Tu, u \rangle \geq 0$

Fact $T \in \mathcal{L}(V)$ positive \Leftrightarrow exists $R \in \mathcal{L}(V)$
such that $R^*R = T$

Cor For any $T \in \mathcal{L}(V)$, T^*T is a positive operator.

Proof Let $S = T^*T$, so S is positive by Fact \square

IB03/2LA3 P.O.V

Given any matrix A , (over \mathbb{R})

$A^T A$ is symmetric and $A^T A$ has nonnegative eigenvalues

Fact If $T \in \mathcal{L}(V)$ is positive, there exists a unique positive $R \in \mathcal{L}(V)$ such that $R^2 = T \leftarrow R$ is called the square root of T

Defⁿ If $T \in \mathcal{L}(V)$, let
 $\sqrt{T^*T}$ denotes the unique positive square
root of T^*T
↑ a positive op

Cor For any $T \in \mathcal{L}(V)$, $\sqrt{T^*T}$ exists

(Polar Decomposition)

For any $T \in \mathcal{L}(V)$, there exists an isometry $S \in \mathcal{L}(V)$
such that

$$T = S \sqrt{T^*T}$$

↑ ↑

positive operator

isometry

Analogy with \mathbb{C}

Very Roughly, \mathbb{C} and $\mathcal{L}(V)$ have similar properties

\mathbb{C}

complex number $z = a+bi$
 conjugate $\bar{z} = a-bi$
 $z = \bar{z} \iff z$ is real

$$|z| = \sqrt{a^2+b^2} = \sqrt{z \cdot \bar{z}}$$

$$|z|=1 \iff z \cdot \bar{z} = 1$$

$\mathcal{L}(V)$

operator T
 adjoint T^*
 self-adjoint $T^* = T$

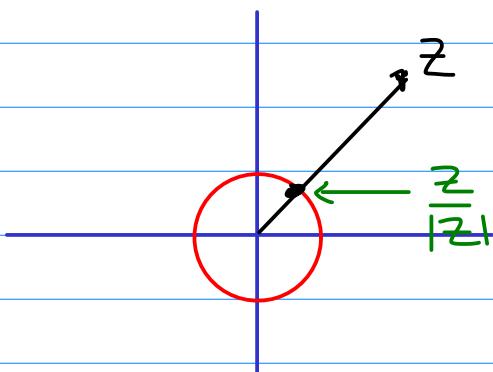
$$\text{isometry } T^* T = I$$

$$z = \left(\frac{z}{|z|} \right) |z| = \left(\frac{z}{|z|} \right) \sqrt{\bar{z} z} \quad \text{Polar Dec Thm}$$

\uparrow
 element on
 unit circle in
 complex plane

$T = S \sqrt{T^* T}$
 \uparrow
 rotation/
 scalar that
 pictures
 length

\uparrow
 Scaling
 like
 scaling
 (a positive
 operator)

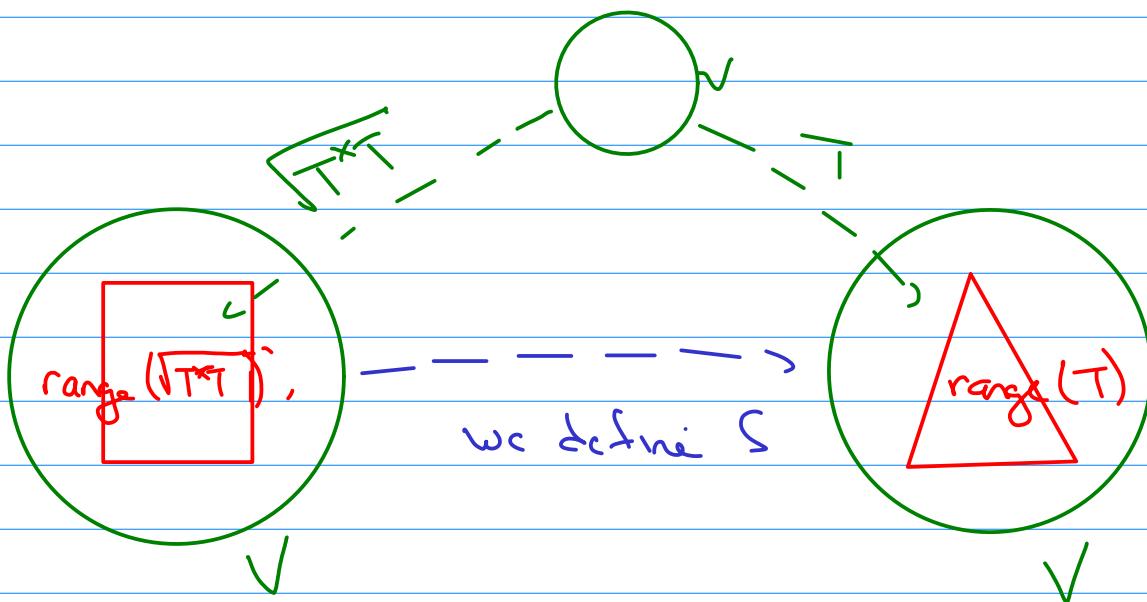


(Sketch of Proof)

Since $T, \sqrt{T^*T} \in \mathcal{L}(V)$

$$\text{range}(\sqrt{T^*T}) \subseteq V$$

$$\text{range}(T) \subseteq V$$



Define a map $S_1: \text{range}(\sqrt{T^*T}) \rightarrow \text{range}(T)$ by

$$(\sqrt{T^*T})v \mapsto Tv$$

(need to check this is well-defined)

Also define a map $S_2: (\text{range}(\sqrt{T^*T}))^\perp \rightarrow (\text{range}T)^\perp$
(Details stripped!)

Orthogonal complement

= all elements orthogonal to the given space

$$\underline{\text{Fact}} \quad V = (\text{range } \overline{T^*T})^\perp \oplus (\text{range } \overline{T^*T})$$

So $v \in V \Rightarrow v = u + w$ with $u \in \text{range } (\overline{T^*T})$
 and $w \in (\text{range } (\overline{T^*T}))^\perp$

We define $S: V \rightarrow V$
 $v = u + w \mapsto S_1u + S_2w$

$$\text{So } S(\overline{T^*T}v) = S_1(\overline{T^*T}v) + S_2 \cdot 0$$

\uparrow
 in range $(\overline{T^*T})$

$$= T_v$$

$$\Rightarrow S(\overline{T^*T}) = T$$

Many details to be checked:

- all maps are linear maps
- S is an isometry

Matrix Interpretation

In terms of matrices, the Polar Decomposition theorem tells us:

If A is an $n \times n$ matrix, can find an orthogonal matrix U and a positive semidefinite matrix R ($\leftarrow R$ has nonnegative eigenvalues)

such that $R^2 = A^T A$ and $A = UR$

matrix analog of isometry matrix analog of Sq-root

If A is an invertible matrix, "easy" to find this decomposition (over \mathbb{R})

① Compute $A^T A$ and find the diagonalization $A^T A = P D P^{-1}$

② Let $R = P \begin{bmatrix} \sqrt{d_1} & & \\ & \ddots & 0 \\ 0 & & \sqrt{d_n} \end{bmatrix} P^{-1}$ where $D = \begin{bmatrix} d_1 & & 0 \\ & \ddots & \\ 0 & & d_n \end{bmatrix}$

③ Since A is invertible, $U = AR^{-1}$

Ex $A = \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix}$ ← find the polar decomposition

① $A^T A = \begin{bmatrix} 2 & 2 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} = \begin{bmatrix} 8 & 2 \\ 2 & 5 \end{bmatrix}$

Diagonalize $A^T A = \begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 9 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix}^{-1}$
 $P \quad D \quad P^{-1}$

② Let $R = \begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 2.8 & 0.4 \\ 0.4 & 2.2 \end{bmatrix}$

③ $U = AR^{-1} = \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 2.8 & 0.4 \\ 0.4 & 2.2 \end{bmatrix}^{-1} = \begin{bmatrix} 0.8 & -0.6 \\ 0.6 & 0.8 \end{bmatrix}$

Thus $\begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} = \begin{bmatrix} 0.8 & -0.6 \\ 0.6 & 0.8 \end{bmatrix} \begin{bmatrix} 2.8 & 0.4 \\ 0.4 & 2.2 \end{bmatrix}$

Key Ideas: * Polar Decomposition