

- **minimal solution** of $Ax = b$:
solution with the smallest norm: least energy, least power, etc.
- Theorem 6.13: $A \in M_{m \times n}(F)$; $b \in F^m$; and s is a minimal solution of $Ax = b$.

Then

- $s \in R(L_{A^*})$.
- s is the only solution in $R(L_{A^*})$.
That is, $AA^*u = b \Rightarrow s = A^*u$.
- s is unique.
- Compare $x = s + y$, $s \in R(L_{A^*})$, $y \in N(L_A)$ vs Theorem 3.9:
 $K = \{s\} + K_H$.

Normal and self-adjoint operator

- We now investigate diagonalizability of a linear operator on an inner product space.
- Lemma 6.14 V is an inner product space; $\dim(V) < \infty$; $T : V \rightarrow V$ is a linear operator. Then T has an eigenvector $\Rightarrow T^*$ has an eigenvector.
- Theorem 6.14 (Schur): $\dim(V) < \infty$; $T : V \rightarrow V$ is a linear operator; and $f_T(t)$ splits. Then \exists an orthonormal basis β such that $[T]_\beta$ is upper triangular.
 - Schur's theorem does not say that $[T]_\beta$ is invertible.
 - Neither does it say that $[T]_\beta$ is diagonalizable.
 - β in the theorem is not unique; nor $[T]_\beta$.

- T is diagonalizable $\Rightarrow [T]_{\beta}$ is diagonal for some β .

$$\Rightarrow [T^*]_{\beta} = [T]_{\beta}^* \text{ is diagonal.}$$

$$\Rightarrow [TT^*]_{\beta} = [T]_{\beta}[T]_{\beta}^* = [T]_{\beta}^*[T]_{\beta} = [T^*T]_{\beta} \text{ [diagonal]}$$

$$\Rightarrow TT^* = T^*T \text{ [rep is unique]}$$

Does this commutativity imply diagonalizability? NO

- **normal operator** T on an inner product space: $TT^* = T^*T$
- **normal matrix** A : $AA^* = A^*A$
- T is normal $\Leftrightarrow [T]_{\beta}$ is normal.

■ Theorem 6.15 : V is an inner product space; $T : V \rightarrow V$ is normal.

Then

$$T^*T = TT^*$$

1. $\forall x \in V, \|T(x)\| = \|T^*(x)\|.$

2. $\forall c \in F, T - cI$ is normal.

3. $T(x) = \lambda x \Leftrightarrow T^*(x) = \bar{\lambda}x$. That is, T and T^* have the same eigenvector x with the respective eigenvalues λ and $\bar{\lambda}$.

4. $T(x_1) = \lambda_1 x_1; T(x_2) = \lambda_2 x_2; \lambda_1 \neq \lambda_2 \Rightarrow \langle x_1, x_2 \rangle = 0.$

■ Theorem 6.16 : V is an inner product space over the "complex" field; $\dim(V) < \infty$; $T : V \rightarrow V$ is linear. Then T is normal if and only if there exists an orthonormal basis for V consisting of eigenvectors of T .

(diagonalizable)

■ [End of Review]

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- **self-adjoint (Hermitian) operator** T on an inner product space:
 $T = T^*$.
 - **self-adjoint (Hermitian) matrix** A : $A = A^*$
 - these definitions apply to both the real and complex field.
 - For **complex** matrices, self-adjoint means **conjugate symmetric**.
 - For **real** matrices, self-adjoint means **symmetric**.
 - For orthonormal β , $T = T^* \Leftrightarrow [T]_{\beta} = [T]_{\beta}^*$
 - self-adjoint \Rightarrow normal; normal $\not\Rightarrow$ self-adjoint

■ Lemma 6.17: V is an inner product space; $\dim(V) < \infty$; T *is self-adjoint*,

1. Every eigenvalue of T is real.
2. $f_T(t)$ splits over the real field.

proof: "1": Let $T(x) = \lambda x, x \neq 0$.

$$\Rightarrow \lambda \langle x, x \rangle = \langle T(x), x \rangle = \langle x, T^*(x) \rangle = \langle x, T(x) \rangle = \langle x, \lambda x \rangle = \bar{\lambda} \langle x, x \rangle$$

2 follows from 1.

■ Theorem 6.17: V is an inner product space over the "real" field; $\dim(V) < \infty$; $T : V \rightarrow V$ is linear. Then T is self-adjoint if and only if there exists an orthonormal basis β for V consisting of eigenvectors of T .

proof:

”only if”: Assume $T = T^*$.

$\Rightarrow f_T(t)$ splits (over the real field). [Lemma 6.17]

$\Rightarrow \exists \beta$, orthonormal, such that $[T]_\beta$ is upper triangular. [Schur’s Thm]

$\Rightarrow [T]_\beta^* = [T^*]_\beta = [T]_\beta$ [self-adjoint]

$\Rightarrow [T]_\beta$ is (real) diagonal.

$\Rightarrow \beta$ consists of eigenvectors of T .

”if” : Assume an orthonormal basis β of eigenvectors of T .

$\Rightarrow [T]_\beta$ is (real) diagonal.

$\Rightarrow [T^*]_\beta = [T]_\beta^* = [T]_\beta$

$\Rightarrow T^* = T$ [rep is unique]

$$T^*T = TT^* = I, \Rightarrow T^{-1} = T^*$$

Unitary and orthogonal operator

$\mathbb{F} = \mathbb{C}$ ■ V is an inner product space; $\dim(V) < \infty$; $T : V \rightarrow V$ is linear.

■ **unitary operator:** $\forall x \in V(\mathbb{C}), \|T(x)\| = \|x\|$

■ **orthogonal operator:** $\forall x \in V(\mathbb{R}), \|T(x)\| = \|x\|$

■ Lemma 6.18: V is an inner product space; $\dim(V) < \infty$; $U : V \rightarrow V$ is self-adjoint. Then $\forall x \in V, \langle x, U(x) \rangle = 0 \Rightarrow U = T_0$. $\nexists U(x) \neq 0$
 $U(x) \perp x$

proof.

Assume U is self-adjoint and $\forall x \in V, \langle x, U(x) \rangle = 0$.

$\Rightarrow \exists \beta = \{v_1, \dots, v_n\}$ orthonormal and consisting of eigenvectors of U . [Thm 6.16, 6.17]

$\Rightarrow \bar{\lambda}_i \langle v_i, v_i \rangle = \langle v_i, \lambda_i v_i \rangle = \langle v_i, U(v_i) \rangle = 0, i = 1, \dots, n$

$\Rightarrow \lambda_i = 0, i = 1, \dots, n$

$\Rightarrow \forall x \in V,$

$$U(x) = U\left(\sum_{i=1}^n a_i v_i\right) = \sum_{i=1}^n a_i U(v_i) = \sum_{i=1}^n a_i \lambda_i v_i = 0$$

■ Theorem 6.18: V is an inner product space; $\dim(V) < \infty$; $T : V \rightarrow V$ is linear. Then these are all equivalent.

1. $TT^* = T^*T = I$ *unitary . orthogonal.*

2. $\forall x, y \in V, \langle T(x), T(y) \rangle = \langle x, y \rangle$

3. β is an orthonormal basis $\Rightarrow T(\beta)$ is an orthonormal basis.

4. There exists an orthonormal basis β such that $T(\beta)$ is an orthonormal basis.

5. $\forall x \in V, \|T(x)\| = \|x\|$

■ 5 means length-preserving; 2 means sort of angle-preserving.

■ rotation and reflection in \mathbb{R}^2 are orthogonal.

■ unitary or orthogonal \Rightarrow normal; not conversely.

proof: "1 \Rightarrow 2": Assume $TT^* = T^*T = I$

$$\Rightarrow \langle T(x), T(y) \rangle = \langle x, T^*T(y) \rangle = \langle x, y \rangle$$

"2 \Rightarrow 3" : Assume "2" and an orthonormal basis $\{v_1, \dots, v_n\}$.

$$\Rightarrow \langle T(v_i), T(v_j) \rangle = \langle v_i, v_j \rangle = \delta_{ij}$$

$\Rightarrow \{T(v_i), \dots, T(v_n)\}$ is an orthonormal basis.

"3 \Rightarrow 4" : Obvious.

"4 \Rightarrow 5" : Assume $\beta = \{v_1, \dots, v_n\}$ and $T(\beta)$ are orthonormal.

$$\Rightarrow \|T(x)\|^2 = \|T(\sum_{i=1}^n a_i v_i)\|^2 = \|\sum_{i=1}^n a_i T(v_i)\|^2$$

$$= \langle \sum_{i=1}^n a_i T(v_i), \sum_{j=1}^n a_j T(v_j) \rangle$$

$$= \sum_{i=1}^n \sum_{j=1}^n a_i \bar{a}_j \langle T(v_i), T(v_j) \rangle = \sum_{i=1}^n |a_i|^2 = \|x\|^2$$

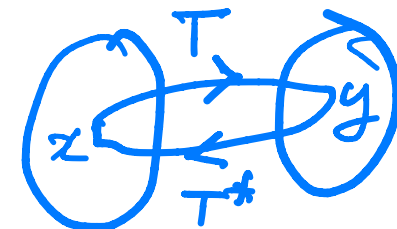
"5 \Rightarrow 1" : Assume $\forall x \in V, \|T(x)\| = \|x\|$.

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

$$\Rightarrow \forall x \in V, \langle x, (I - T^*T)(x) \rangle = 0$$

$$\Rightarrow (I - T^*T) = T_0, [(I - T^*T) \text{ is self-adjoint; Lemma 6.18}]$$

$$\Rightarrow T^*T = I \Rightarrow TT^* = I \text{ [Thm 2.17c]}$$



- T is unitary or orthogonal; v is an eigenvector.

$$\Rightarrow \|v\| = \|T(v)\| = \|\lambda v\| = |\lambda| \|v\|$$

$$\Rightarrow |\lambda| = 1$$

$$\begin{cases} [T_\theta]_{\mathbb{R}^2} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \\ [T_\theta^*]_{\mathbb{R}^2} = \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} \\ = T_{-\theta} \end{cases}$$

- example: linear operators on \mathbb{R}^2

- T_θ : rotation by θ ; $T_\theta^* = T_{-\theta} \Rightarrow T_\theta T_\theta^* = T_\theta T_{-\theta} = I$

- T : reflection about a line or the origin; $T^* = T \rightarrow TT^* = T^*T = I$

- **unitary matrix**: $AA^* = A^*A = I \rightarrow A^{-1} = A^*$

- **orthogonal matrix** over the real field : $AA^t = A^tA = I$

- real unitary matrix = orthogonal matrix

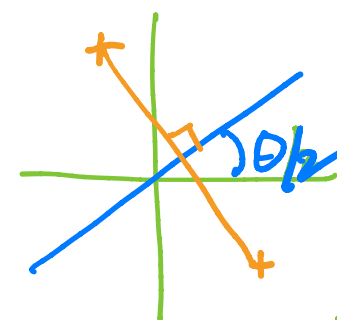
$$\hookrightarrow A^{-1} = A^t$$

- columns ~~form~~ ^{form} an orthonormal basis for F^n .

- rows form an orthonormal basis for F^n .

- T is unitary $\Leftrightarrow [T]_\beta$ is unitary for an orthonormal β .

- T is orthogonal $\Leftrightarrow [T]_\beta$ is orthogonal for an orthonormal β .



$$\begin{cases} [T]_{\mathbb{R}^2} = \begin{bmatrix} \cos\theta & \sin\theta \\ \sin\theta & -\cos\theta \end{bmatrix} \\ = [T]_{\mathbb{R}^2}^* \end{cases}$$

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- A matrix A is **unitarily equivalent** to B : \exists a unitary matrix Q such that $A = Q^* B Q$.
 - A real matrix A is ~~unitarily~~ orthogonally equivalent to real B : \exists an orthogonal matrix Q such that $A = Q^t B Q$.
 - A and B are unitarily equivalent \Rightarrow they are similar, but not conversely.
 - Theorem 6.19: A **complex** $n \times n$ matrix A is normal.
 $\Leftrightarrow A$ is unitarily equivalent to a diagonal matrix.
 - This is the matrix version of Theorem 6.16.

proof: " \Rightarrow ": Assume A is normal and γ is the std basis for F^n .

$\Rightarrow L_A$ is normal and $[L_A]_\gamma = A$.

$\Rightarrow \exists$ an orthonormal basis β of eigenvectors of L_A . [Thm 6.16]

$\Rightarrow [L_A]_\beta = D$, diagonal.

$$\Rightarrow A = [L_A]_\gamma = [I]_\beta^\gamma [L_A]_\beta [I]_\gamma^\beta = Q^{-1} D Q.$$

We show that $Q = [I]_\gamma^\beta$ is unitary if β and γ are orthonormal.

Let $\beta = (u_1, \dots, u_n)$ and $\gamma = (v_1, \dots, v_n)$, both orthonormal.

$$\Rightarrow v_j = \sum_{i=1}^n \langle v_j, u_i \rangle u_i \Rightarrow Q_{ij} = \langle v_j, u_i \rangle$$

$$\Rightarrow Q_{ij}^* = \overline{Q_{ji}} = \overline{\langle v_i, u_j \rangle} = \langle u_j, v_i \rangle = ([I]_\beta^\gamma)_{ij}$$

$$\Rightarrow Q Q^* = [I]_\gamma^\beta [I]_\beta^\gamma = I, \text{ and similarly, } Q^* Q = I.$$

” \Leftarrow ”: Assume Q , unitary, and D , diagonal, are such that $A = Q^* D Q$.

$$\Rightarrow A A^* = Q^* D Q (Q^* D Q)^* = Q^* D Q Q^* D^* Q$$

$$= Q^* D D^* Q = Q^* D^* D Q = Q^* D^* Q Q^* D Q = (Q^* D Q)^* Q^* D Q =$$

$$= A^* A \text{ [diagonals commute]}$$

⊛ Thm 6.17

T is self-adjoint

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- Theorem 6.20: A real $n \times n$ matrix A is symmetric.
 $\Leftrightarrow A$ is orthogonally equivalent to a real diagonal matrix.
 - This is the matrix version of Theorem 6.17.
 - Theorem 6.21 (Schur) : $A \in M_{n \times n}(F)$; $f_A(t)$ splits over F .
Then
 1. If $F = \mathbb{C}$, then A is unitarily equivalent to a complex upper triangular matrix.
 2. If $F = \mathbb{R}$, then A is orthogonally equivalent to a real upper triangular matrix.
 - This is the matrix version of Schur's Theorem 6.14.

Orthogonal projection and spectral theorem

- Let $V = W_1 \oplus W_2$ for subspaces W_1 and W_2 .
Then, $\forall x \in V, x = x_1 + x_2$ for some $x_1 \in W_1, x_2 \in W_2$.
- **projection** T on W_1 along W_2 : $T(x_1 + x_2) = x_1$
- $R(T) = W_1 = \{x \in V : T(x) = x\}$
- $N(T) = W_2 = \{x \in V : T(x) = 0\}$
- For a projection T on W_1 , we can choose various W_2 .
- If T is an "orthogonal" projection on W_1 , then W_2 is unique.
- T is a projection $\Leftrightarrow T = T^2$ [alt def]

proof: " \Rightarrow " : $\forall x \in V, T^2(x) = T(x_1) = T(x_1 + 0) = x_1$
 " \Leftarrow " : Assume that $T = T^2$.

First we show that $R(T) \cap N(T) = \{0\}$.

Assume $x \in R(T) \cap N(T)$.

$$x \in R(T) \Rightarrow \exists u \text{ such that } T(u) = x \Rightarrow T^2(u) = T(u) = x$$

$$x \in N(T) \Rightarrow T(x) = 0 \Rightarrow T^2(x) = T(x) = 0 \quad \xrightarrow{T^2(u) = T(x)}$$

$$\Rightarrow x = 0, (\because T^2(u) = T(x) = x = 0)$$

Now we show that $V = R(T) + N(T)$.

$$\forall x \in V, \text{ let } x_1 = T(x) \text{ and } x_2 = x - x_1. \Rightarrow x_1 \in R(T)$$

$$T(x_2) = T(x) - T(x_1) = T(x) - T^2(x) = 0 \Rightarrow x_2 \in N(T)$$

$$V = R(T) \oplus N(T)$$

■ **orthogonal projection T :**

$$R(T)^\perp = N(T) \text{ and } N(T)^\perp = R(T)$$

- If $\dim(V) < \infty$, $(R(T)^\perp = N(T) \Leftrightarrow N(T)^\perp = R(T))$.
- Given a subspace W , $T(y) = u$, [Thm 6.6]
where $y = u + z$, $u \in W$, $z \in W^\perp$,
defines an orthogonal projection on W .
- A truncated Fourier series, for $k < n$, $\underbrace{u}_{\text{blue circle}} = \sum_{i=1}^k \underbrace{\langle y, v_i \rangle}_{\text{blue underline}} \underbrace{v_i}_{\text{blue dot}}$ is an orthogonal projection on $\text{span}(\{v_1, \dots, v_k\})$.
- There is only one orthogonal projection on W .
- The orthogonal projection on W provides the best approximation.

- Theorem 6.24: V is an inner product space: $T : V \rightarrow V$ is linear. Then T is an orthogonal projection $\Leftrightarrow T^2 = T = T^*$. *orthogonal proj.*

Proof:

" \Rightarrow ": Assume T is an orthogonal projection.

$$\Rightarrow T^2 = T \text{ [projection]}$$

$$\Rightarrow V = R(T) \oplus N(T); R(T)^\perp = N(T)$$

$\Rightarrow \forall x, y \in V, x = x_1 + x_2$ and $y = y_1 + y_2$, for some $x_1, y_1 \in R(T)$ and $x_2, y_2 \in N(T)$.

$$\begin{aligned} \Rightarrow \langle x, T(y) \rangle &= \langle x_1 + x_2, y_1 \rangle = \langle x_1, y_1 \rangle + \langle x_2, y_1 \rangle = \langle x_1, y_1 \rangle = \\ &= \langle x_1, y_1 \rangle + \langle x_1, y_2 \rangle = \langle x_1, y_1 + y_2 \rangle = \langle T(x), y \rangle = \langle x, T^*(y) \rangle \end{aligned}$$

$\Rightarrow T = T^*$

" \Leftarrow ": Assume $T^2 = T = T^*$

$\Rightarrow T$ is a projection.

orthogonal. (2512, 2/2)
Show $R(T)^\perp = N(T), N(T)^\perp = R(T)$

Let $x \in R(T)$ and $y \in N(T)$. $\Rightarrow T(x) = x, T(y) = 0$.

$$\Rightarrow \langle x, y \rangle = \langle T(x), y \rangle = \langle x, T^*(y) \rangle = \langle x, T(y) \rangle = \langle x, 0 \rangle = 0$$

$\Rightarrow x \in N(T)^\perp$ and $y \in R(T)^\perp$ (서로 수직)

$$\Rightarrow R(T) \subseteq N(T)^\perp \text{ and } N(T) \subseteq R(T)^\perp \quad (1)$$

Let $x \in N(T)^\perp$

$\Rightarrow x = x_1 + x_2, x_1 \in R(T), x_2 \in N(T)$ [projection]

$$\Rightarrow 0 = \langle x, x_2 \rangle = \langle x_1, x_2 \rangle + \langle x_2, x_2 \rangle = \|x_2\|^2 \quad [x_1 \in N(T)^\perp \quad (1)]$$

$\Rightarrow x_2 = 0 \Rightarrow x = x_1 \in R(T) \Rightarrow N(T)^\perp \subseteq R(T) \Rightarrow N(T)^\perp =$

$R(T)$ [(1), cf textbook]

Let $y \in R(T)^\perp$.

$\Rightarrow y = y_1 + y_2, y_1 \in R(T), y_2 \in N(T)$ [projection]

$$\Rightarrow 0 = \langle y, y_1 \rangle = \langle y_1, y_1 \rangle + \langle y_2, y_1 \rangle = \|y_1\|^2 \quad [y_2 \in R(T)^\perp \quad (1)]$$

$\Rightarrow y_1 = 0 \Rightarrow y = y_2 \in N(T) \Rightarrow R(T)^\perp \subseteq N(T)$

$\Rightarrow R(T)^\perp = N(T)$ [(1), cf textbook]

- **self-adjoint (Hermitian) operator** $T: T = T^*$.
- **self-adjoint (Hermitian) matrix** $A: A = A^*$
- A : Conjugate symmetric.
- **Theorem 6.17**: For finite V over the "real" field, T is self-adjoint **iff** \exists an orthonormal basis β for V consisting of eigenvectors of T .

proof:

"only if": Assume $T = T^*$. *(real eigen values)*

$\Rightarrow f_T(t)$ splits (over the real field). [Lemma 6.17]

$\Rightarrow \exists \beta$, orthonormal, s. t. $[T]_\beta$ is upper triangular. [Schur's Thm]

$\Rightarrow [T]_\beta^* = [T^*]_\beta = [T]_\beta$ [self-adjoint]

$\Rightarrow [T]_\beta$ is (real) diagonal.

$\Rightarrow \beta$ consists of eigenvectors of T .

$$T: \text{Normal} \quad , \quad TT^* = T^*T$$

Unitary and orthogonal operator $\Rightarrow TT^* = T^*T = I$

- V is an inner product space; $\dim(V) < \infty$; $T : V \rightarrow V$ is linear.
- **unitary operator:** $\forall x \in V(\mathbb{C}), \|T(x)\| = \|x\|$
- **orthogonal operator:** $\forall x \in V(\mathbb{R}), \|T(x)\| = \|x\|$
- Lemma 6.18: V is an inner product space; $\dim(V) < \infty$; $U : V \rightarrow V$ is self-adjoint. Then $\forall x \in V, \langle x, U(x) \rangle = 0 \Rightarrow U = T_0$.

proof.

Assume U is self-adjoint and $\forall x \in V, \langle x, U(x) \rangle = 0$.

$\Rightarrow \exists \beta = \{v_1, \dots, v_n\}$ orthonormal and consisting of eigenvectors of U . [Thm 6.16, 6.17]

$$\Rightarrow \bar{\lambda}_i \langle v_i, v_i \rangle = \langle v_i, \lambda_i v_i \rangle = \langle v_i, U(v_i) \rangle = 0, i = 1, \dots, n$$

$$\Rightarrow \lambda_i = 0, i = 1, \dots, n$$

$$\Rightarrow \forall x \in V,$$

$$U(x) = U\left(\sum_{i=1}^n a_i v_i\right) = \sum_{i=1}^n a_i U(v_i) = \sum_{i=1}^n a_i \lambda_i v_i = 0$$

■ Theorem 6.18: The followings are all equivalent.

1. $TT^* = T^*T = I$
2. $\forall x, y \in V, \langle T(x), T(y) \rangle = \langle x, y \rangle$
3. \exists an orthonormal basis β s. t. $T(\beta)$ is an orthonormal basis.
4. $\forall x \in V, \|T(x)\| = \|x\|$

proof: "1": $TT^* = T^*T = I$

\Rightarrow "2": $\langle T(x), T(y) \rangle = \langle x, T^*T(y) \rangle = \langle x, y \rangle$

\Rightarrow "3" for $\beta = \{v_1, \dots, v_n\}$, $\langle T(v_i), T(v_j) \rangle = \langle v_i, v_j \rangle = \delta_{ij}$

$\Rightarrow \{T(v_i), \dots, T(v_n)\}$ is an orthonormal basis.

\Rightarrow "4": $\|T(x)\|^2 = \|T(\sum_{i=1}^n a_i v_i)\|^2 = \langle \sum_{i=1}^n a_i T(v_i), \sum_{j=1}^n a_j T(v_j) \rangle$

$= \sum_{i=1}^n \sum_{j=1}^n a_i \bar{a}_j \langle T(v_i), T(v_j) \rangle = \sum_{i=1}^n |a_i|^2 = \|x\|^2$

\Rightarrow "1": Assume $\forall x \in V, \|T(x)\| = \|x\|$.

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

$\Rightarrow T^*T = I \Rightarrow TT^* = I$ [Thm 2.17c]

- **unitary matrix:** $AA^* = A^*A = I$
 - **orthogonal matrix** over the real field : $AA^t = A^tA = I$
 - T is unitary $\Leftrightarrow [T]_\beta$ is unitary for an orthonormal β .
 - T is orthogonal $\Leftrightarrow [T]_\beta$ is orthogonal for an orthonormal β .
- A matrix A is **unitarily (or orthogonally) equivalent** to B : \exists a unitary matrix Q such that $A = Q^*BQ$ (or Q^tBQ).
- Theorem 6.19: A is a complex and normal matrix.
 $\Leftrightarrow A$ is unitarily equivalent to a diagonal matrix.

proof: " \Rightarrow ": L_A is normal, where $[L_A]_\gamma = A$.

$\Rightarrow \exists$ an orthonormal basis β of eigenvectors of L_A . [Thm 6.16]

$\Rightarrow [L_A]_\beta = D \Rightarrow A = [L_A]_\gamma = [I]_\beta^\gamma [L_A]_\beta [I]_\gamma^\beta = Q^{-1}DQ$

$\Rightarrow A = Q^*DQ$ since $Q = [I]_\gamma^\beta$ is unitary for orthonormal β and γ .

($\because Q_{ij} = \langle v_j, u_i \rangle, Q_{ij}^* = \overline{Q_{ji}} = \overline{\langle v_i, u_j \rangle} = \langle u_j, v_i \rangle = ([I]_\beta^\gamma)_{ij}$)

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- Theorem 6.20: A real $n \times n$ matrix A is symmetric(self-adjoint).
 $\Leftrightarrow A = Q^t D Q$, where D is a real diagonal matrix.
 - Theorem 6.21 (Schur) : $A \in M_{n \times n}(F)$; $f_A(t)$ splits over F . Then
 1. If $F = \mathbb{C}$, then $A = Q^* U Q$, where U is a complex upper triangular matrix.
 2. If $F = \mathbb{R}$, then $A = Q^t U Q$, where U is a real upper triangular matrix.

Orthogonal projection and spectral theorem

- **projection** T on W_1 along W_2 : $T(x_1 + x_2) = x_1$
 - $R(T) = W_1 = \{x \in V : T(x) = x\}$
 - $N(T) = W_2 = \{x \in V : T(x) = 0\}$
- $\Rightarrow V = R(T) \oplus N(T)$.

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- For a projection T on W_1 , we can choose various W_2 .
 - If T is an "orthogonal" projection on W_1 , then W_2 is unique.
 - T is a projection $\Leftrightarrow T = T^2$ [alt def]
 - **orthogonal projection T :**
 $R(T)^\perp = N(T)$ and $N(T)^\perp = R(T)$
 - If $\dim(V) < \infty$, $(R(T)^\perp = N(T) \Leftrightarrow N(T)^\perp = R(T))$.
 - Given a subspace W , $T(y) = u$, where $y = u + z$, $u \in W$, $z \in W^\perp$, is an orthogonal projection on W .
 - A truncated Fourier series, for $k < n$, $u = \sum_{i=1}^k \langle y, v_i \rangle v_i$ is an orthogonal projection on $\text{span}(\{v_1, \dots, v_k\})$.
 - The orthogonal projection on W provides the best approximation.
 - [End of Review]

- Theorem 6.24: V is an inner product space: $T : V \rightarrow V$ is linear. Then T is an orthogonal projection $\Leftrightarrow T^2 = T = T^*$.

projection orthogonal

Proof:

" \Rightarrow ": Assume T is an orthogonal projection.

$\Rightarrow T^2 = T$ [projection]

$\Rightarrow V = R(T) \oplus N(T); R(T)^\perp = N(T)$

$\Rightarrow \forall x, y \in V, x = x_1 + x_2$ and $y = y_1 + y_2$, for some $x_1, y_1 \in R(T)$ and $x_2, y_2 \in N(T)$.

$\Rightarrow \langle x, T(y) \rangle = \langle x_1 + x_2, y_1 \rangle = \langle x_1, y_1 \rangle + \langle x_2, y_1 \rangle = \langle x_1, y_1 \rangle = \langle x_1, y_1 \rangle + \langle x_1, y_2 \rangle = \langle x_1, y_1 + y_2 \rangle = \langle T(x), y \rangle = \langle x, T^*(y) \rangle$

$\Rightarrow T = T^*$

" \Leftarrow ": Assume $T^2 = T = T^*$

$\Rightarrow T$ is a projection.

\Rightarrow orthogonal $\frac{1}{2} \frac{1}{2} \frac{2}{2} \frac{1}{2}$
 $R(T)^\perp = N(T)$

Let $x \in R(T)$ and $y \in N(T)$. $\Rightarrow T(x) = x, T(y) = 0$.

$$\Rightarrow \langle x, y \rangle = \langle T(x), y \rangle = \langle x, T^*(y) \rangle = \langle x, T(y) \rangle = \langle x, 0 \rangle = 0$$

$$\Rightarrow x \in N(T)^\perp \text{ and } y \in R(T)^\perp$$

$$\Rightarrow R(T) \subseteq N(T)^\perp \text{ and } N(T) \subseteq R(T)^\perp \quad (1)$$

Let $x \in N(T)^\perp$

$$\Rightarrow x = x_1 + x_2, x_1 \in R(T), x_2 \in N(T) \text{ [projection]}$$

$$\Rightarrow 0 = \langle x, x_2 \rangle = \langle x_1, x_2 \rangle + \langle x_2, x_2 \rangle = \|x_2\|^2, [\because x_1 \in N(T)^\perp \quad (1)]$$

$$\Rightarrow x_2 = 0 \Rightarrow x = x_1 \in R(T) \Rightarrow N(T)^\perp \subseteq R(T) \Rightarrow N(T)^\perp = R(T) \quad [(1)]$$

Let $y \in R(T)^\perp$.

$$\Rightarrow y = y_1 + y_2, y_1 \in R(T), y_2 \in N(T) \text{ [projection]}$$

$$\Rightarrow 0 = \langle y, y_1 \rangle = \langle y_1, y_1 \rangle + \langle y_2, y_1 \rangle = \|y_1\|^2, [\because y_2 \in R(T)^\perp \quad (1)]$$

$$\Rightarrow y_1 = 0 \Rightarrow y = y_2 \in N(T) \Rightarrow R(T)^\perp \subseteq N(T)$$

$$\Rightarrow R(T)^\perp = N(T) \quad [(1)]$$

■ **Theorem 6.25 (The spectral theorem):** V is an inner product space over F ; $\dim(V) < \infty$; $T : V \rightarrow V$ is a linear operator with

distinct eigenvalues: spectrum	$\lambda_1 \cdots \lambda_k$
corresponding eigenspaces	$W_1 \cdots W_k$
orthogonal projection on W_i	$T_1 \cdots T_k$

and T is normal if $F = \mathbb{C}$ and self-adjoint if $F = \mathbb{R}$. Then the following statements are true.

1. $V = W_1 \oplus \cdots \oplus W_k.$

2. $W_i^\perp = \bigoplus_{j=1, j \neq i}^k W_j.$

3. $T_i T_j = T_0, i \neq j.$

4. $T_1 + \cdots + T_k = I.$

5. $T = \lambda_1 T_1 + \cdots + \lambda_k T_k$: **spectral decomposition**

$[T]_{\mathcal{B}} = \begin{bmatrix} \lambda_1 & & & & \\ & \lambda_1 & & & \\ & & \lambda_2 & & \\ & & & \ddots & \\ & & & & 0 \\ & & & & & \ddots \\ & & & & & & \lambda_k \\ & & & & & & & \lambda_k \end{bmatrix}$

proof of 5: $x \in V$

$$\Rightarrow x = x_1 + \cdots + x_k, x_i \in W_i \text{ [1]}$$

$$\begin{aligned} T(x) &= T(x_1) + \cdots + T(x_k) \\ &= \lambda_1 x_1 + \cdots + \lambda_k x_k \\ &= \lambda_1 T_1(x) + \cdots + \lambda_k T_k(x) \\ &= (\lambda_1 T_1 + \cdots + \lambda_k T_k)(x) \end{aligned}$$

■ example: consider $T = L_A$,

$$\text{where } A = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 5 \end{pmatrix}.$$

$$\lambda_1 = 2, W_1 = \{(a_1, 0, 0, 0)^t : a_1 \in \mathbb{R}\}, T_1 = L_{A_1},$$

$$\lambda_2 = 3, W_2 = \{(0, a_2, a_3, 0)^t : a_2, a_3 \in \mathbb{R}\}, T_2 = L_{A_2},$$

$$\lambda_3 = 5, W_3 = \{(0, 0, 0, a_4)^t : a_4 \in \mathbb{R}\}, T_3 = L_{A_3},$$

$$\text{where } A_1 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}, A_2 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}, \text{ and } A_3 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

$$T = \lambda_1 T_1 + \lambda_2 T_2 + \lambda_3 T_3 : L_A = 2L_{A_1} + 3L_{A_2} + 5L_{A_3}$$

- example: consider $T = L_A$, where $A = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$. $(\lambda_1 I - A)v = 0$
 $\lambda_1 = 3, W_1 = \{(a_1, a_2)^t \in \mathbb{R}^2 : a_1 = a_2\}, T_1 = L_{A_1}, \begin{bmatrix} 2 & -2 \\ -2 & 2 \end{bmatrix} v = 0$
 $\lambda_2 = -1, W_2 = \{(a_1, a_2)^t \in \mathbb{R}^2 : a_1 + a_2 = 0\}, T_2 = L_{A_2}, v_1 = v_2$
 where $A_1 = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$ and $A_2 = \frac{1}{2} \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}$. \vdots
 $T = \lambda_1 T_1 + \lambda_2 T_2 : L_A = 3L_{A_1} - L_{A_2}$

- **Corollary 6.25.2:** V is an inner product space over \mathbb{C} ; $\dim(V) < \infty$; $T : V \rightarrow V$ is unitary $\Leftrightarrow T$ is normal and $|\lambda| = 1$ for every eigenvalue λ of T .

Proof.

” \Rightarrow ” If T is unitary, then T is normal and every eigenvalue of T has absolute value 1 ($\because \|T(x)\| = \|x\|$).

” \Leftarrow ” Let $T = \lambda_1 T_1 + \lambda_2 T_2 + \cdots + \lambda_k T_k$ be the spectral decomposition of T . If $|\lambda| = 1$ for every eigenvalue λ of T , then by 3. of the spectral theorem,

$$\begin{aligned} TT^* &= (\lambda_1 T_1 + \lambda_2 T_2 + \cdots + \lambda_k T_k)(\bar{\lambda}_1 T_1^* + \bar{\lambda}_2 T_2^* + \cdots + \bar{\lambda}_k T_k^*) \\ &= |\lambda_1|^2 T_1 T_1^* + |\lambda_2|^2 T_2 T_2^* + \cdots + |\lambda_k|^2 T_k T_k^* \\ &= (T_1 + T_2 + \cdots + T_k)(T_1 + T_2 + \cdots + T_k)^* \\ &= I \end{aligned}$$

Hence T is unitary.

■ **Corollary 6.25.3:** V is an inner product space over \mathbb{C} ; $\dim(V) < \infty$; $T : V \rightarrow V$ is normal. Then T is self-adjoint \Leftrightarrow every eigenvalue of T is real.

Proof.

" \Rightarrow " $T^* = T \Rightarrow \bar{\lambda}_1 T_1 + \bar{\lambda}_2 T_2 + \cdots + \bar{\lambda}_k T_k = \lambda_1 T_1 + \lambda_2 T_2 + \cdots + \lambda_k T_k \Rightarrow \lambda_i$ is real.

" \Leftarrow " Let $T = \lambda_1 T_1 + \lambda_2 T_2 + \cdots + \lambda_k T_k$ be the spectral decomposition of T . Suppose that every eigenvalue of T is real. Then $T_i^*(v_i) = \lambda_i v_i = T_i(v_i)$, so $T^* = \lambda_1 T_1^* + \lambda_2 T_2^* + \cdots + \lambda_k T_k^* = \lambda_1 T_1 + \lambda_2 T_2 + \cdots + \lambda_k T_k = T$.

- Reflection is an example that is both self adjoint and unitary.

T^*T
 $=TT^*$

$T^* = T$
 $\lambda_i \text{ real}$

$T = \begin{bmatrix} \cos 2\theta & \sin 2\theta \\ \sin 2\theta & -\cos 2\theta \end{bmatrix}$

normal

self-adjoint (λ_i real) unitary ($|\lambda_i| = 1$)

$T^* = T$

$T^*T = I = TT^*$

$\begin{bmatrix} \cos 2\theta & \sin 2\theta \\ \sin 2\theta & -\cos 2\theta \end{bmatrix}^*$

$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$

$T^*T = TT^* = I$

$\|Tx\| = \|x\|$
 $(|\lambda_i| = 1)$