

(1)

7.5 SPECTRAL THM FOR NORMAL MATRICES

A is $n \times n$ real or complex matrix throughout.
Recall

SPECTRAL THM

A has basis of eectors of $\mathbb{C}^n \iff A = PDP^{-1}$ is diagonalizable

QW When does a diagonalizable matrix have an ONB of eectors?

SCAFFOLDED PROBLEM (A) Let $A = \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix}$

(1) Show $\lambda_1 = 1, \lambda_2 = 2$

(2) Show $N(A - \lambda_1 I) = \text{Span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}$

$$N(A - \lambda_2 I) = \text{Span} \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$$

(3) Explain why A ~~does not have~~ a basis $\{\vec{v}_1, \vec{v}_2\}$ of \mathbb{C}^2 so that

(a) \vec{v}_1, \vec{v}_2 are eectors of A

(b) $\vec{v}_1 \perp \vec{v}_2$

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SPECTRAL THM FOR NORMAL MATRICES

TF AE

① A has ONB of eigenvectors② $A = UDU^*$ where D is diagonal and U is unitary ($U^* = U^{-1}$)③ A is Normal, i.e. $AA^* = A^*A$ PF① \Leftrightarrow ②

Same as proof of original Spectral Thm

+ fact

 U unitary \Leftrightarrow Cols of U form ONB② \Rightarrow ③

$$AA^* = (UDU^*)(UDU^*)^*$$

$$= UDU^*UD^*U^*$$

$$= UDD^*U$$

$$= U D^* D U \quad \text{as Diagonal matrices commute}$$

$$= \dots = A^*A$$

③ \Rightarrow ②See ~~end of Lecture~~ test book.

(3)

LEMMA 1

Let A be normal and λ_1, λ_2 be distinct
eigenvalues of A

Then

$$N(A - \lambda_2 I) \perp N(A - \lambda_1 I)$$

PF

We will need the following

CLAIM [Meyer p 212 eqn (4.5.6)]

For any matrix B

$$N(B^* B) = N(B)$$

Suppose $A \vec{v}_1 = \lambda_1 \vec{v}_1$, $A \vec{v}_2 = \lambda_2 \vec{v}_2$ with $\lambda_2 \neq \lambda_1$

Then

$$N(A - \lambda_1 I) \stackrel{\text{claim}}{=} N[(A - \lambda_1 I)^* (A - \lambda_1 I)]$$

$$= N[(A - \lambda_1 I)(A - \lambda_1 I)^*]$$

as A and hence $A - \lambda_1 I$
is normal

$$\stackrel{\text{claim}}{=} N((A - \lambda_1 I)^*)$$

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So we know $(A - \lambda_1 I)^T \vec{v}_1 = \vec{0}$

So

$$\vec{v}_1^T (A - \lambda_1 I) = \vec{0}$$

$$\Rightarrow \vec{v}_1^T (A - \lambda_1 I) \vec{v}_2 = \vec{0}$$

$$\Rightarrow \vec{v}_1^T (\lambda_2 - \lambda_1) \vec{v}_2 = \vec{0} \quad \Leftrightarrow \quad A \vec{v}_2 = \lambda_2 \vec{v}_2$$

$$\Rightarrow \vec{v}_1^T \vec{v}_2 = 0 \quad \Leftrightarrow \quad \lambda_2 \neq \lambda_1$$

$$\Rightarrow \vec{v}_1 \perp \vec{v}_2 \quad \square$$

LEMMA 2

① If $A^* = A$ (A is real symmetric or Hermitian)
Then all eigenvalues of A are real

② If $A^T = -A$ and $A \in \mathbb{R}^{n \times n}$ Then all eigenvalues of A are pure imaginary.

EX $A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ has $\sigma(A) = \{\pm i\}$

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PF ① IF $A\vec{v} = \lambda \vec{v}$ WITH $\vec{v} \neq \vec{0}$ THEN

$$\lambda \|\vec{v}\|^2 = \lambda \langle \vec{v} | \vec{v} \rangle = \langle \vec{v} | \lambda \vec{v} \rangle = \langle \vec{v} | A\vec{v} \rangle \\ = \langle A^* \vec{v} | \vec{v} \rangle = \langle A\vec{v} | \vec{v} \rangle = \langle \lambda \vec{v} | \vec{v} \rangle = \bar{\lambda} \langle \vec{v} | \vec{v} \rangle = \bar{\lambda} \|\vec{v}\|^2$$

So $(\lambda - \bar{\lambda}) \|\vec{v}\|^2 = 0$

Since $\vec{v} \neq \vec{0}$, $\lambda = \bar{\lambda}$. So $\lambda \in \mathbb{R}$

② SIMILAR. YOU DO IT!

SPECTRAL THM FOR REAL SYMMETRIC MATRICES

Let $A \in \mathbb{R}^{n \times n}$, $A^T = A$

Then $A = PDP^T$ where D is diagonal and P is orthogonal.

SCAFFOLDED PROBLEM ⑤

Let $A \in \mathbb{R}^{3 \times 3}$, $A^T = A$ be

$$A = \begin{pmatrix} 7 & -1 & -2 \\ -1 & 7 & 2 \\ -2 & 2 & 10 \end{pmatrix}$$

THIS PROBLEM SHOWS
YOU HOW TO OBTAIN
S.T. FOR SYMMETRIC
MATRICES FROM S.T.
FOR NORMAL MATRICES

① SHOW $p(\lambda) = \det(A - \lambda I) = -(\lambda - 6)^2(\lambda - 12)$

② Use the S.T. for Normal Matrices to explain why $\dim N(A - 6I) = 2$ must hold.

③ Now calculate bases for $N(A - 6I)$, $N(A - 12I)$ and verify that $N(A - 6I) \perp N(A - 12I)$ holds.

④ For a general real symmetric matrix, A , explain why we can always obtain bases for each of the eigenspaces of A that consist of real (rather than complex) vectors. ⑥

⑤ Use fact that $\lambda = 6$ is real together with Gram-Schmidt to calculate an ONB for $N(A - 6I)$ consisting of real vectors.

⑥ Suppose λ is an eigenvalue of a real symmetric matrix, A , with $\text{Alg Mult } (\lambda) > 1$. Explain why the method in ⑤ can be used to obtain an ONB ~~for~~ of real vectors for $N(A - \lambda I)$. Why is $\dim N(A - \lambda I) = \text{Alg Mult } (\lambda)$?

⑦ For the 3×3 matrix, A , ~~in ⑤~~ above calculate an ONB for \mathbb{R}^3 consisting of eigenvectors of A (Hint: Use ③ and ⑥)

⑧ Finally show $A = PDP^T$ where $D = \text{Diag}(6, 6, 12)$

and P is orthogonal matrix $P = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{6}} \\ 0 & \frac{1}{\sqrt{3}} & -\frac{2}{\sqrt{6}} \end{pmatrix}$

⑨ Generalize these calculations to prove S.T. for Real Symmetric matrices.

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2ND DERIVATIVE TEST (CALCULUS 3)

Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, $z = f(\vec{x})$. Suppose \vec{x}_0 is a Critical point of f , i.e. $\nabla f(\vec{x}_0) = \vec{0}$.

Let $\Delta = \det(H)$ where $H = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix} \Big|_{\vec{x} = \vec{x}_0}$ HESSIAN

Then

| Δ | $f_{xx}(\vec{x}_0)$ | CLASSIFICATION |
|----------|---------------------|----------------|
| + | + | LOCAL MIN |
| + | - | LOCAL MAX |
| - | * | SADDLE POINT |

PF By Taylor's series

$$z = f(\vec{x}) \simeq f(\vec{x}_0) + \nabla f(\vec{x}_0) \cdot (\vec{x} - \vec{x}_0) + (\vec{x} - \vec{x}_0)^T H (\vec{x} - \vec{x}_0)$$

Let $z' = z - f(\vec{x}_0)$, $\vec{x}' = \vec{x} - \vec{x}_0$. Since $\nabla f(\vec{x}_0) = \vec{0}$

we just need to classify. $Q(\vec{x}) = \vec{x}'^T H \vec{x}'$

at $\vec{x} = \vec{0}$. Since H is real symmetric, $H = P^T D P$

where can choose $\det(P) = +1$. So P effects a rotation, $P = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$ for some θ

Since rotating coords does not change nature of CPT

we can look at $\tilde{Q}(\vec{y}) := Q(P^T \vec{y}) = Q(\vec{x}) = \vec{x}'^T P^T D P \vec{x}'$

$$= \vec{y}^T D \vec{y} = \lambda_1 y_1^2 + \lambda_2 y_2^2$$

$$\Delta = \det(H) = \det(D) = \lambda_1 \lambda_2$$

$$f_{xx} = (P^T D P)_{11} = \lambda_1 \cos^2 \theta + \lambda_2 \sin^2 \theta$$

| Sign (λ_1, λ_2) | Sign (λ_1) | CLASS |
|-------------------------------|--------------------|--------|
| + | + | MIN |
| + | - | MAX |
| - | * | SADDLE |

ISSUE Suppose choose $D = \text{Diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$ with $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$ and T repeated evolves.
Then T as many choices for P , i.e. P is NOT UNIQUE. (7)

For this reason it can be hard to define e^A .

BUT we have

THE SPECTRAL DECOMPOSITION OF A NORMAL MATRIX

Let A be $n \times n$ normal with $\sigma(A) = \{\lambda_1, \dots, \lambda_k\}$.

So $A = UDU^*$ where $U = [X_1 | X_2 | \dots | X_k]$

and X_j is $n \times a_j$ matrix whose cols are
ONB for $N(A - \lambda_j I)$.

$$\text{So } A = [X_1 | \dots | X_k] \begin{bmatrix} \lambda_1 I_{a_1} & & 0 \\ & \ddots & \\ 0 & & \lambda_k I_{a_k} \end{bmatrix} \begin{bmatrix} X_1^* \\ \vdots \\ X_k^* \end{bmatrix}$$

$$\boxed{A = \lambda_1 X_1 X_1^* + \dots + \lambda_k X_k X_k^*}$$

SPECIAL CASE If all eigenvalues have alg mult = 1
Then

$$A = \lambda_1 \vec{v}_1 \vec{v}_1^* + \dots + \lambda_k \vec{v}_k \vec{v}_k^*$$

$$(A \vec{v}_j = \lambda_j \vec{v}_j)$$

where $\{\vec{v}_1, \dots, \vec{v}_n\}$ is ONB for \mathbb{C}^n of vectors.

(7)

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$$A = \lambda_1 X_1 X_1^* + \dots + \lambda_k X_k X_k^*$$

SPECIAL CASE If all eigenvalues have alg mult = 1

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$$A = \lambda_1 \vec{v}_1 \vec{v}_1^* + \dots + \lambda_k \vec{v}_k \vec{v}_k^*$$

$$A \vec{v}_j = \lambda_j \vec{v}_j$$

where $\{\vec{v}_1, \dots, \vec{v}_n\}$ is ONB for \mathbb{C}^n of vectors.

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RECALL Let $\|\vec{v}\| = 1$. Then $P(\vec{w}) := \vec{v} \vec{v}^* \vec{w} = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\|^2} \vec{v}$
is a projection onto $\text{Span}(\vec{v})$.

Since $\text{Span}\{-\vec{v}\} = \text{Span}\{\vec{v}\}$ there are
2 possible choices for \vec{v} . However the L.T.
 P is the same for both.

GENERAL CASE : GEOMETRIC MEANING of $X_j X_j^*$

Since $A - \lambda_j I$ is NORMAL, The
Range-Nullspace Decomposition tells us

$$\mathbb{C}^n = N(A - \lambda_j I) \oplus R(A - \lambda_j I)$$

where $N(A - \lambda_j I)^\perp = R(A - \lambda_j I)$

Let G_j = Orthogonal Projector onto
 $N(A - \lambda_j I)$ along $R(A - \lambda_j I)$.

The L.T. G_j is well-defined indept of
choice of basis for these two
complementary spaces.

⑨

By 5.9, 5.11 we get

$$\begin{aligned} E_j &= [X_j | 0] [X_j | Y_j]^{-1} \\ &= [X_j | 0] \begin{bmatrix} X_j^* \\ Y_j^* \end{bmatrix} \\ &= X_j X_j^* \end{aligned}$$

← DWRS for N, R

SPECTRAL DECOMPN THM

Let A be Normal, E_j as above. Then

① $A = \lambda_1 E_1 + \dots + \lambda_k E_k$ SPECTRAL DECOMPOSITION OF A

② $I = E_1 + \dots + E_k$

③ $E_i E_j = 0$ if $i \neq j$

NOTE A generalization of this Thm holds for
Any $n \times n$ MATRIX !!

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PF

① See above

$$\textcircled{2} \quad I = UV^* = [X_1 \dots X_k] \begin{bmatrix} X_1^* \\ \vdots \\ X_k^* \end{bmatrix} = X_1 X_1^* + \dots + X_k X_k^* = E_1 + \dots + E_k$$

$$\textcircled{3} \quad E_1 E_2 = X_1 X_1^* X_2 X_2^* = 0$$

||
0

since by Lemma 1 if $A \vec{v}_j = \lambda_j \vec{v}_j \quad j=1,2$

Then $\vec{v}_1 \perp \vec{v}_2$ So $\vec{v}_1^* \vec{v}_2 = 0$.

SCAFFOLDED PROBLEM (C)

① Compute the Spectral Decomposition for the matrix $A = UDU^*$ where

$$U = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad \text{and} \quad D = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

↑
SLIGHTLY DIFFERENT
FROM S.P. (B)

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② You should find that for $\lambda_1 = 2$

we have $E_1 = X_1 X_1^*$ where X_1 is 3×1

and for $\lambda_2 = 3$, $E_2 = X_2 X_2^*$ where X_2 is 3×2

Calculate the matrices E_1 and E_2

③ Check that $E_1 + E_2 = I$
 $E_1 E_2 = 0$

and $A = 2E_1 + 3E_2$.

MAGIC FORMULA FOR E_j

CASE $k=3$

Suppose $\sigma(A) = \{\lambda_1, \lambda_2, \lambda_3\}$

So $A = \lambda_1 E_1 + \lambda_2 E_2 + \lambda_3 E_3$ ①

$I = E_1 + E_2 + E_3$ ②

$E_i E_j = 0$ $i \neq j$, $E_j^2 = E_j$ ③

AS
A PROJECTOR

Let solve for E_j !

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By ① and ③

$$A - \lambda_1 I = (\lambda_2 - \lambda_1) \underline{e}_2 + (\lambda_3 - \lambda_1) \underline{e}_3$$

$$A - \lambda_2 I = (\lambda_1 - \lambda_2) \underline{e}_1 + (\lambda_3 - \lambda_2) \underline{e}_3$$

So

$$(A - \lambda_1 I)(A - \lambda_2 I)$$

$$= [(\lambda_2 - \lambda_1) \underline{e}_2 + (\lambda_3 - \lambda_1) \underline{e}_3] [(\lambda_1 - \lambda_2) \underline{e}_1 + (\lambda_3 - \lambda_2) \underline{e}_3]$$

$$= (\lambda_3 - \lambda_1)(\lambda_3 - \lambda_2) \underline{e}_3 \quad \text{by } ③$$

SCALAR!

So

$$\underline{e}_3 = \frac{(A - \lambda_1 I)(A - \lambda_2 I)}{(\lambda_3 - \lambda_1)(\lambda_3 - \lambda_2)}$$

SCAFFOLDED PROBLEM ④

- ① What are $\underline{e}_1, \underline{e}_2$ when $k=2$?
- ② Use ① to rederive formula in SP ②.
- ③ Show in general

$$\underline{e}_i = \frac{\prod_{\substack{j=1 \\ j \neq i}}^k (A - \lambda_j I)}{\prod_{\substack{j=1 \\ j \neq i}}^k (\lambda_i - \lambda_j)}$$