

Chapter 5. Eigenvalues & Eigenvectors.

$$Ax = \lambda x. \quad ; \quad \lambda: \text{eigenvalue.} \quad x: \text{eigenvector.}$$

\rightarrow row exchanges change the eigenvalue λ . \rightarrow we use determinants

$$\Rightarrow (A - \lambda I)x = 0.$$

• vector x : nullspace of $A - \lambda I$.

• λ : eigenvalue.

\Rightarrow for non-zero vector x , $A - \lambda I$ should be singular.

$$\det(A - \lambda I) = 0.$$

ex) $A = \begin{bmatrix} 4 & -5 \\ 2 & -3 \end{bmatrix} \quad A - \lambda I = \begin{bmatrix} 4 - \lambda & -5 \\ 2 & -3 - \lambda \end{bmatrix}$

$$\det(A - \lambda I) = (4 - \lambda)(-3 - \lambda) + 10 = \lambda^2 - \lambda - 2 = 0. \quad \lambda = 2 \text{ or } -1.$$

$$\left(\begin{array}{ll} \lambda_1 = -1 & (A - \lambda_1 I)x = \begin{bmatrix} 5 & -5 \\ 2 & -2 \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \end{bmatrix} & x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \end{array} \right.$$

$$\lambda_2 = 2 \quad (A - \lambda_2 I)x = \begin{bmatrix} 2 & -5 \\ 2 & -5 \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad x_2 = \begin{bmatrix} 5 \\ 2 \end{bmatrix}$$

1) Compute $\det(A - \lambda I)$ and find the roots of λ .

2) For each eigenvalue, solve the equation $(A - \lambda I)x = 0$.

Ex 3)

$$A = \begin{bmatrix} 1 & 4 & 5 \\ 0 & \frac{3}{4} & 6 \\ 0 & 0 & \frac{1}{2} \end{bmatrix}$$

$$\det(A - \lambda I) = \begin{vmatrix} 1-\lambda & 4 & 5 \\ 0 & \frac{3}{4}-\lambda & 6 \\ 0 & 0 & \frac{1}{2}-\lambda \end{vmatrix} = (1-\lambda)\left(\frac{3}{4}-\lambda\right)\left(\frac{1}{2}-\lambda\right)$$

• Eigenvalues are different from pivots in Gauss elimination.

$$\textcircled{a} \quad \prod \lambda_i = \prod \text{pivots} = \det A.$$

$$\textcircled{b} \quad \text{Trace of } A = \lambda_1 + \lambda_2 + \dots + \lambda_n = a_{11} + a_{22} + \dots + a_{nn}.$$

\Rightarrow assume $\det(A - \lambda I) = 0$ has n roots, $\lambda_1, \lambda_2, \dots, \lambda_n$, then.

$$\det(A - \lambda I) = (\lambda_1 - \lambda)(\lambda_2 - \lambda) \dots (\lambda_n - \lambda) = 0.$$

$$\hookrightarrow = (-\lambda)^n + (\lambda_1 + \lambda_2 + \dots + \lambda_n)(-\lambda)^{n-1} + \dots + \lambda_1 \lambda_2 \dots \lambda_n. \quad \textcircled{1}$$

$$\det(A - \lambda I) = \begin{vmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \dots & \dots & a_{nn} - \lambda \end{vmatrix} \rightarrow \lambda = 0 \rightarrow \det(A)$$

$\textcircled{1} \text{ if } \lambda = 0 \rightarrow \lambda_1 \lambda_2 \dots \lambda_n = \det A.$

\rightarrow to get the terms of $(-\lambda)^n$ and $(-\lambda)^{n-1}$, they are derived from

$(a_{11} - \lambda)C_{11}$ and consequently, $(-\lambda)^n$ and $(-\lambda)^{n-1}$ come from,

$$\begin{aligned} &\rightarrow (a_{11} - \lambda)(a_{22} - \lambda) \dots (a_{n-1, n-1} - \lambda)(a_{nn} - \lambda) \\ &= (-\lambda)^n + (a_{11} + a_{22} + \dots + a_{nn})(-\lambda)^{n-1} + \dots \end{aligned} \quad \textcircled{2}$$

by $\textcircled{1} = \textcircled{2}$ for $\forall (-\lambda)^1, (-\lambda)^2, \dots, (-\lambda)^n$,

$$\sum_{i=1}^n \lambda_i = \sum_{i=1}^n a_{ii}$$

HW 5.1 7, 10, 22, 25, 31

5.2 Diagonalization of a Matrix

⇒ The eigenvectors diagonalize a matrix.

• for $n \times n$ matrix A , A has n linearly independent eigenvectors. (e_i)

$$S = \begin{bmatrix} e_1 & e_2 & \dots & e_n \end{bmatrix}$$

↳ eigenvector matrix,

$$Ae_i = \lambda_i e_i$$

$$S^{-1}AS = \Lambda = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix}$$

↳ eigenvalue matrix

$$AS = A \begin{bmatrix} e_1 & e_2 & \dots & e_n \end{bmatrix} = \begin{bmatrix} \lambda_1 e_1 & \lambda_2 e_2 & \dots & \lambda_n e_n \end{bmatrix}$$

$$= \begin{bmatrix} e_1 & e_2 & \dots & e_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix}$$

$$\begin{cases} S^{-1}AS = \Lambda \\ A = S\Lambda S^{-1} \end{cases}$$

S : invertible, n linearly independent eigenvectors

< Remark 1 >

If $\lambda_1, \lambda_2, \dots, \lambda_n$ are distinct, then n eigenvectors are independent.

pf \Rightarrow assume $x_1 = c_2 x_2 + \dots + c_n x_n \rightarrow$ (linearly dependent. x_1 independent $x_2 \sim x_n$)

$$Ax_1 = A(c_2 x_2 + \dots + c_n x_n)$$

$$\lambda_1 x_1 = c_2 \lambda_2 x_2 + \dots + c_n \lambda_n x_n$$

$$\lambda_1 (c_2 x_2 + \dots + c_n x_n) = c_2 \lambda_2 x_2 + \dots + c_n \lambda_n x_n$$

$$\Rightarrow c_2 (\lambda_1 - \lambda_2) x_2 + c_3 (\lambda_1 - \lambda_3) x_3 + \dots + c_n (\lambda_1 - \lambda_n) x_n = 0$$

$$\Rightarrow c_2 = \dots = c_n = 0$$

Remark 2.

$\Rightarrow S$ is not unique, An eigenvector can be multiplied by a constant,

Remark 3.

\Rightarrow The order of eigenvalues and eigenvectors in S and Λ are the same.

Remark 4.

\Rightarrow Not all matrices have n linearly independent eigenvectors, so not all matrices are diagonalizable.

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \rightarrow \lambda^2 = 0. \quad x = \begin{bmatrix} c \\ 0 \end{bmatrix}$$

\hookrightarrow double root.

\Rightarrow Diagonalization fails only if there are repeated eigenvalues.
 \hookrightarrow not always. (for $A = I$)

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- Invertibility of A depends on non-zero eigenvalues.

↳ if $\lambda = 0$ $\det A = 0$,

② Examples of Diagonalization

$$\text{ex 1) } A = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \rightarrow \det(A - \lambda I) = 0 \Rightarrow \lambda = 1 \text{ or } 0$$

$$S = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, \quad \Lambda = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \rightarrow S^{-1}AS = \Lambda,$$

ex 2) 90° rotation $K = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$, $\rightarrow \det(K - \lambda I) = 0$, $\lambda = i, -i$
($\lambda^2 + 1 = 0$)

$$S = \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix}, \quad \Lambda = \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix} \rightarrow S^{-1}KS = \Lambda \text{ in Complex domain}$$

⑨ Powers and Products : A^k and AB

\Rightarrow eigenvalues of $A^k \rightarrow \lambda_i^k$ $\Lambda^k = S^{-1} A^k S$
 eigenvectors of $A^k \rightarrow S_{ii}$

eg $A^2x = A \cdot Ax = A \cdot \lambda x = \lambda Ax = \lambda \cdot \lambda x = \lambda^2 x.$

$$A^k x = \lambda^k x \quad \rightarrow \text{eigenvectors are the same!}$$

$$\Lambda^k = S^{-1} A^k S.$$

• $A^T : Ax = \lambda x.$

$\Rightarrow A^T x = ? \quad A^T \lambda x \Leftrightarrow A^T x = \frac{1}{\lambda} x \rightarrow x^T$

The eigenvalue of A^T are $\frac{1}{\lambda_i}$.

optimal.

• usually AB (product of A, B) does not have same eigenvectors.

$\begin{cases} Ax_1 = \lambda_A x_1 & Bx_2 = \lambda_B x_2 \\ \Rightarrow AB = \lambda x ?? \end{cases}$

\Rightarrow iff $AB = BA$, \rightarrow diagonalization matrices share the same eigenvectors.

assume.

① $A = S \Lambda_1 S^{-1}$, and $B = S \Lambda_2 S^{-1}$

$\begin{cases} AB = S \Lambda_1 \Lambda_2 S^{-1} & , \quad BA = S \Lambda_2 \Lambda_1 S^{-1} \end{cases}$

since $\Lambda_1 \Lambda_2 = \Lambda_2 \Lambda_1 \Rightarrow$ then $AB = BA$

② assume $AB = BA$

$\rightarrow Ax = \lambda x.$

$\underline{B}Ax = B\lambda x = \lambda Bx. \quad \downarrow \quad AB = BA.$

$A\underline{B}x = \lambda Bx.$

$\rightarrow Bx$ is the eigenvector of A with λ

$\rightarrow x$ and Bx share λ , $\rightarrow Bx$ is multiple of x .

HW 5.2

15, 16, 22, 31