



Eigenvalues and Eigenvectors, Characteristic Polynomial, Diagonalization, Cayley-Hamilton Theorem

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Definition

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Note: Note that an eigenvector cannot be $\overrightarrow{\mathbf{0}}$, but an eigenvalue can be $0 \in \mathbb{R}$. If 0 is an eigenvalue of A, then there must be some nontrivial vector $\overrightarrow{\mathbf{x}}$ for which $A\overrightarrow{\mathbf{x}} = 0\overrightarrow{\mathbf{x}} = \overrightarrow{\mathbf{0}}$ which implies that A is not invertible.

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Note: The eigenspace of the $n \times n$ matrix A corresponding to the eigenvalue λ of A is the set of all eigenvectors of A corresponding



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Note: $P_A(\lambda) = \lambda^n + (-1)^{n-1} Tr(A) \lambda^{n-1} + \ldots + \det(A)$. For example, for a 2×2 square matrix A, $\frac{\text{Mehmet E. K\"{O}RO\'{G}bU}}{P_A(\lambda) = \lambda^2 - Tr(A) \lambda + \det(A)}$.

Finding Eigenvalues and

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Note 1: The eigenvalues of a triangular matrix are the entries on its main diagonal.

Note 2: If $P_A(\lambda)$ has multiple roots, then there exists multiple eigenvalues.

Example

Let
$$A = \begin{pmatrix} 3 & 6 & -8 \\ 0 & 0 & 6 \\ 0 & 0 & 2 \end{pmatrix}$$
. Find

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$$P_A(\lambda) = \det(\lambda I_3 - A)$$

$$\lambda I_3 - A = \lambda \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} - \begin{pmatrix} 3 & 6 & -8 \\ 0 & 0 & 6 \\ 0 & 0 & 2 \end{pmatrix}$$

Solution (1)
$$P_{A}(\lambda) = \det(\lambda I_{3} - A)$$

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$$= \begin{pmatrix} \lambda - 3 & -6 & 8 \\ 0 & \lambda & -6 \\ 0 & 0 & \lambda - 2 \end{pmatrix}$$

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$$= \begin{pmatrix} \lambda - 3 & -6 & 8 \\ 0 & \lambda & -6 \\ 0 & 0 & \lambda - 2 \end{pmatrix}$$
$$\Rightarrow P_{A}(\lambda) = \lambda (\lambda - 2) (\lambda - 3).$$

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Since, eigenvalues are the roots of $P_{A}(\lambda)$, we have

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 $\Rightarrow \lambda_1 = 0, \lambda_2 = 2, \lambda_3 = 3.$

Solution (3) For $\lambda_1 = 0$,

$$(0I_3 - A) = -A = \begin{pmatrix} -3 & -6 & 8 \\ 0 & 0 & -6 \\ 0 & 0 & -2 \end{pmatrix} \sim \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

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$$\Rightarrow \begin{pmatrix} x \\ y \\ z \end{pmatrix} = t \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}, t \in \mathbb{R} \Rightarrow \overrightarrow{\mathbf{v}}_1 = \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}.$$

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$$(2I_3 - A) = \begin{pmatrix} -1 & -6 & 8 \\ 0 & 2 & -6 \\ 0 & 0 & 0 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 10 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \end{pmatrix}$$

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Definition

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Note:
$$A = PDP^{-1} \Rightarrow A^k = PD^kP^{-1}$$

Theorem (The Diagonalization Theorem)

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- 1. An $n \times n$ matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.
- 2. If $\overrightarrow{\mathbf{v}}_1$, $\overrightarrow{\mathbf{v}}_2$, ..., $\overrightarrow{\mathbf{v}}_n$ are linearly independent eigenvectors of A and $\lambda_1, \lambda_2, \ldots, \lambda_n$ are their corresponding eigenvalues, then $A = PDP^{-1}$, where

$$P = \begin{pmatrix} \overrightarrow{\mathbf{v}}_1 & \dots & \overrightarrow{\mathbf{v}}_n \end{pmatrix} \text{ and } D = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{pmatrix}.$$

Example

We have found the eigenvalues and corresponding eigenvectors of

the matrix
$$A=\begin{pmatrix} 3 & 6 & -8 \\ 0 & 0 & 6 \\ 0 & 0 & 2 \end{pmatrix}$$
 as $\lambda_1=0, \lambda_2=2, \lambda_3=3$ and $\overrightarrow{\mathbf{v}}_1=\begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}, \overrightarrow{\mathbf{v}}_2=\begin{pmatrix} -10 \\ 3 \\ 1 \end{pmatrix}, \overrightarrow{\mathbf{v}}_3=\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$. Then

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- 1. show that $\overrightarrow{\mathbf{v}}_1$, $\overrightarrow{\mathbf{v}}_2$, $\overrightarrow{\mathbf{v}}_3$ are linear independent,
- 2. show that there exists matrices P and D such that $A = PDP^{-1}$,
- 3. and find A^{125} .

Solution

1.
$$P = \begin{pmatrix} \overrightarrow{\mathbf{V}}_1 & \overrightarrow{\mathbf{V}}_2 & \overrightarrow{\mathbf{V}}_3 \end{pmatrix} = \begin{pmatrix} -2 & -10 & 1 \\ 1 & 3 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$
 and since $\det(P) = 1 \neq 0$, $\overrightarrow{\mathbf{V}}_1$, $\overrightarrow{\mathbf{V}}_2$, $\overrightarrow{\mathbf{V}}_3$ are linearly independent.

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2. $P = \begin{pmatrix} -2 & -10 & 1 \\ 1 & 3 & 0 \\ 0 & 1 & 0 \end{pmatrix}$, $P^{-1} = \begin{pmatrix} 0 & 1 & -3 \\ 0 & 0 & 1 \\ 1 & 2 & 4 \end{pmatrix}$ and $D = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} \Rightarrow A = PDP^{-1}$.

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3.
$$A^{125} = PD^{125}P^{-1} = P \begin{pmatrix} 0 & 0 & 0 \\ 0 & 2^{125} & 0 \\ 0 & 0 & 3^{125} \end{pmatrix} P^{-1}$$
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Theorem (Cayley-Hamilton Theorem)

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If $a_0 \neq 0$, then

$$I_n = A \underbrace{\frac{-1}{a_0} \left(A^{n-1} + a_{n-1} A^{n-2} + \ldots + a_1 I_n \right)}_{A^{-1}}$$

Example

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- 4. matrices P and D such that $A = PDP^{-1}$, if any,
- 5. A^{-1} and A^{5} (by using Cayley-Hamilton Theorem).

Solution (1)
$$P_A(\lambda) = \det(\lambda I_n - A)$$

$$\lambda I_2 - A = \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} - \begin{pmatrix} 1 & 4 \\ 2 & 3 \end{pmatrix}$$

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$$\Rightarrow P_A(\lambda) = (\lambda - 1)(\lambda - 3) - 8$$
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 $\Rightarrow \lambda_1 = -1, \lambda_2 = 5.$

Solution (3) For
$$\lambda_1 = -1$$
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Solution (4)
$$P = \begin{pmatrix} -2 & 1 \\ 1 & 1 \end{pmatrix}, P^{-1} = \begin{pmatrix} -\frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{2}{3} \end{pmatrix} \text{ and } D = \begin{pmatrix} -1 & 0 \\ 0 & 5 \end{pmatrix} \Rightarrow A = PDP^{-1}.$$

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?