



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  $\lambda$  of A is the set of all eigenvectors of A corresponding



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Let A be an  $n \times n$  square matrix. Then the equation  $\det (A - \lambda I_n) = 0$  is called the characteristic equation of the matrix A and the result of the determinant  $\det (A - \lambda I_n)$  is polynomial of the form  $P_A(\lambda) = \lambda^n + a_{n-1}\lambda^{n-1} + \ldots + a_1\lambda + a_0$  is called characteristic polynomial of A.

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Note:  $P_A(\lambda) = (-1)^n \lambda^n + (-1)^{n-1} Tr(A) \lambda^{n-1} + \ldots + \det(A)$ . For example, for a  $2 \times 2$  square matrix A

Mehmet E KÖRÖĞLÜ  $P_A(\lambda) = \lambda^2 - Tr(A)\lambda + \det(A)$ .

### Finding Eigenvalues and

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

#### Example

Let 
$$A = \begin{pmatrix} 3 & 6 & -8 \\ 0 & 0 & 6 \\ 0 & 0 & 2 \end{pmatrix}$$
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Solution (1) 
$$P_A(\lambda) = \det(A - \lambda I_n)$$

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

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$$\Rightarrow P_{A}(\lambda) = -\lambda (3 - \lambda) (2 - \lambda).$$

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

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

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

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

# Solution (3) For $\lambda_2 = 2$ .

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

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

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- 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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- 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

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

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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Note: The inverse or any power of a square matrix can be computed by using Cayley-Hamilton Theorem.

$$P_A(A) = A^n + a_{n-1}A^{n-1} + \dots + a_1A + a_0I_n = 0$$
  
 $\Rightarrow A^n = -(a_{n-1}A^{n-1} + \dots + a_1A + a_0I_n)$ 

#### Theorem (Cayley-Hamilton Theorem)

Every matrix A is a root of its characteristic polynomial, i.e.,  $P_A(A) = O_{n \times n}$ .

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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**

Let 
$$A = \begin{pmatrix} 1 & 4 \\ 2 & 3 \end{pmatrix}$$
. Find

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- 5.  $A^{-1}$  and  $A^{5}$  (by using Cayley-Hamilton Theorem).

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

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

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$$\Rightarrow P_A(\lambda) = (1 - \lambda)(3 - \lambda) - 8$$
$$= \lambda^2 - 4\lambda - 5.$$

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The eigenvalues are the roots of  $P_A(\lambda)$  such that

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 $\Rightarrow \lambda_1 = -1, \lambda_2 = 5.$ 

Solution (3) For 
$$\lambda_1 = -1$$
,

$$(A+I_2) = \begin{pmatrix} 2 & 4 \\ 2 & 4 \end{pmatrix} \sim \begin{pmatrix} 1 & 2 \\ 0 & 0 \end{pmatrix}$$

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

Solution (3) For 
$$\lambda_2 = 5$$
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$$(A-5I_2) = \begin{pmatrix} -4 & 4 \\ 2 & -2 \end{pmatrix} \sim \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix}$$

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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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?