

# Eigenvalues and Eigenvectors

An **eigenvector** of an  $n \times n$  matrix  $A$  is a nonzero vector  $\mathbf{x}$  such that  $A\mathbf{x} = \lambda\mathbf{x}$  for some scalar  $\lambda$ . A scalar  $\lambda$  is called an **eigenvalue** of  $A$  if there is a nontrivial solution  $\mathbf{x}$  of  $A\mathbf{x} = \lambda\mathbf{x}$ ; such an  $\mathbf{x}$  is called an *eigenvector corresponding to  $\lambda$* .<sup>1</sup>

Hence for  $A = \begin{bmatrix} 2 & 1 \\ 0 & -1 \end{bmatrix}$ , the vector  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  is an eigenvector with eigenvalue 2 and  $\begin{bmatrix} -1 \\ 3 \end{bmatrix}$  is an eigenvector with eigenvalue -1.

**Question** Is  $\begin{bmatrix} 5 \\ 0 \end{bmatrix}$  also an eigenvector?

# How to find eigenvalue/vector?

$$Ax = \lambda x$$

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

$$\Rightarrow x \in \text{Null}(A - \lambda.I).$$

$$\text{Null}(A - \lambda.I) \neq \{0\} \Leftrightarrow \det(A - \lambda.I) = 0$$

Given a matrix  $A$ , the polynomial  $\det(A - \lambda.I) = 0$  is called characteristic polynomial of  $A$  (here  $\lambda$  is treated as a variable). Its roots are the eigenvalues of  $A$ .

## Theorem

*The eigenvalues of a triangular matrix are the entries on its main diagonal.*

# Determinant and eigenvalues

Recall for  $A$  an  $n \times n$  matrix, let  $U$  be any echelon form obtained from  $A$  by row replacements and row interchanges (without scaling), and let  $r$  be the number of such row interchanges. Then the determinant of  $A$ ,  $\det(A) = (-1)^r u_{11} u_{22} \cdots u_{nn}$  where  $u_{ii}$  are diagonal entries of  $U$ .

## Theorem (Invertible matrix theorem)

*Let  $A$  be an  $n \times n$  matrix. Then  $A$  is invertible (i.e.,  $\det(A) \neq 0$ ) iff  $0$  is not an eigenvalue of  $A$ .*

## Theorem

*If matrices  $A$  and  $B$  are similar ( $A = PBP^{-1}$ ), then they have the same characteristic polynomial and hence the same eigenvalues.*

## Theorem

*Let  $A$  be an  $n \times n$  matrix with eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$ . Then  $\det(A) = \prod_i \lambda_i$  and trace of  $(A) = \sum_i \lambda_i$*

# Diagonalization

## Definition

A square matrix  $A$  is said to be **diagonalizable** if  $A$  is similar to a diagonal matrix, i.e.,

$$A = PDP^{-1}$$

for some invertible matrix  $P$  and diagonal matrix  $D$ .

## Theorem

*An  $n \times n$  matrix  $A$  is diagonalizable ( $A = PDP^{-1}$ ) iff  $A$  has  $n$  linearly independent eigenvectors.*

In this case eigenvectors will form a basis of  $\mathbb{R}^n$ .

## Theorem

An  $n \times n$  matrix with  $n$  **distinct eigenvalues** is diagonalizable

**Proof:** Let  $v_1, \dots, v_n$  be eigenvectors corresponding to the  $n$  distinct eigenvalues of a matrix  $A$ . Then  $\{v_1, \dots, v_n\}$  is linearly independent set, hence basis of  $\mathbb{R}^n$ .

Let  $A$  be an  $n \times n$  matrix whose distinct eigenvalues are  $\lambda_1, \dots, \lambda_p$ .

- For  $1 \leq k \leq p$ , the dimension of the eigenspace for  $\lambda_k$  is less than or equal to the multiplicity of the eigenvalue  $\lambda_k$ .
- The matrix  $A$  is diagonalizable if and only if the sum of the dimensions of the eigenspaces equals  $n$ , and this happens if and only if (i) the characteristic polynomial factors completely into linear factors and (ii) the dimension of the eigenspace for each  $\lambda_k$  equals the multiplicity of  $\lambda_k$ .
- If  $A$  is diagonalizable and  $\mathcal{B}_k$  is a basis for the eigenspace corresponding to  $\lambda_k$  for each  $k$ , then the total collection of vectors in the sets  $\mathcal{B}_1, \dots, \mathcal{B}_p$  forms an eigenvector basis for  $\mathbb{R}^n$ .

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# Eigenvectors and Difference and Differential Equations

Consider a dynamical system  $x_{k+1} = Ax_k$ .

If  $x_0$  is an eigenvector with an eigenvalue  $\lambda$  then  $x_k = \lambda^k x_0$ .



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What if  $A$  is diagonalizable?

$$x_1' = a_{11}x_1 + \cdots + a_{1n}x_n$$

$$x_2' = a_{21}x_1 + \cdots + a_{2n}x_n$$

$$\vdots$$

$$x_n' = a_{n1}x_1 + \cdots + a_{nn}x_n$$