

## 5.1 Eigenvalues and Eigenvectors

- Diagonalization
- Eigenvalues and Eigenvectors
- Characteristic Polynomial
- Properties

# Diagonalization

## Definition

A linear operator  $T$  on a finite-dimensional vector space  $V$  is diagonalizable if there is an ordered basis  $\beta$  for  $V$  such that  $[T]_{\beta}$  is a diagonal matrix. A square matrix  $A$  is diagonalizable if  $L_A$  is diagonalizable.

# Eigenvalues and Eigenvectors

## Definition

Let  $T$  be a linear operator on a vector space  $V$ . A nonzero vector  $v \in V$  is an eigenvector of  $T$  if there exists a scalar eigenvalue  $\lambda$  corresponding to the eigenvector  $v$  such that  $T(v) = \lambda v$ .

Let  $A \in M_{n \times n}(F)$ . A nonzero vector  $v \in F^n$  is an eigenvector of  $A$  if  $v$  is an eigenvector of  $L_A$ ; that is, if  $Av = \lambda v$  for some scalar eigenvalue  $\lambda$  of  $A$  corresponding to the eigenvector  $v$ .

# Eigenvalues and Eigenvectors: Example

## Example

Let  $A = \begin{bmatrix} 0 & -2 \\ -4 & 2 \end{bmatrix}$ ,  $\mathbf{u} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ , and  $\mathbf{v} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ . Examine the images of  $\mathbf{u}$  and  $\mathbf{v}$  under multiplication by  $A$ .

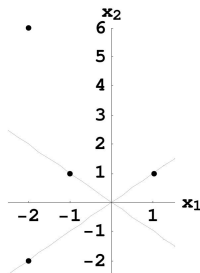
## Solution

$$\begin{aligned} A\mathbf{u} &= \begin{bmatrix} 0 & -2 \\ -4 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ -2 \end{bmatrix} = \\ &= -2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} = -2\mathbf{u} \end{aligned}$$

$\mathbf{u}$  is called an *eigenvector* of  $A$  since  $A\mathbf{u}$  is a multiple of  $\mathbf{u}$ .

$$A\mathbf{v} = \begin{bmatrix} 0 & -2 \\ -4 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ 6 \end{bmatrix} \neq \lambda\mathbf{v}$$

$\mathbf{v}$  is not an eigenvector of  $A$  since  $A\mathbf{v}$  is not a multiple of  $\mathbf{v}$ .



$$\begin{aligned} A\mathbf{u} &= -2\mathbf{u}, \text{ but} \\ A\mathbf{v} &\neq \lambda\mathbf{v} \end{aligned}$$

## Eigenvalues and Eigenvectors: Example

## Example

Show that 4 is an eigenvalue of  $A = \begin{bmatrix} 0 & -2 \\ -4 & 2 \end{bmatrix}$  and find the corresponding eigenvectors.

*Solution:* Scalar 4 is an eigenvalue of  $A$  if and only if  $A\mathbf{x} = 4\mathbf{x}$  has a nontrivial solution.

$$A\mathbf{x} - 4\mathbf{x} = \mathbf{0}$$

$$A\mathbf{x} - 4(\text{---})\mathbf{x} = \mathbf{0}$$

$$(A - 4I)\mathbf{x} = \mathbf{0}.$$

To solve  $(A - 4I)\mathbf{x} = \mathbf{0}$ , we need to find  $A - 4I$  first:

$$A - 4I = \begin{bmatrix} 0 & -2 \\ -4 & 2 \end{bmatrix} - \begin{bmatrix} 4 & 0 \\ 0 & 4 \end{bmatrix} = \begin{bmatrix} -4 & -2 \\ -4 & -2 \end{bmatrix}$$

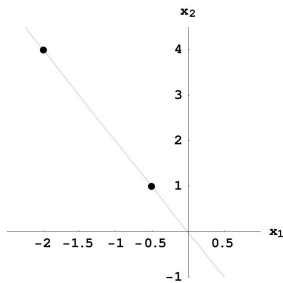
# Eigenvalues and Eigenvectors: Example

Now solve  $(A-4I)\mathbf{x} = \mathbf{0}$ :

$$\begin{bmatrix} -4 & -2 & 0 \\ -4 & -2 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & \frac{1}{2} & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow \mathbf{x} = \begin{bmatrix} -\frac{1}{2}x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix}.$$

Each vector of the form  $x_2 \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix}$  is an eigenvector corresponding to the eigenvalue  $\lambda = 4$ .



Eigenspace for  $\lambda = 4$

The set of all solutions to  $(A-\lambda I)\mathbf{x} = \mathbf{0}$  is called the **eigenspace** of  $A$  corresponding to  $\lambda$ .

# Diagonalization

## Theorem (5.1)

*A linear operator  $T$  on a finite-dimensional vector space  $V$  is diagonalizable if and only if there exists an ordered basis  $\beta$  for  $V$  consisting of eigenvectors of  $T$ . If  $T$  is diagonalizable,  $\beta = \{v_1, \dots, v_n\}$  is an ordered basis of eigenvectors of  $T$ , and  $D = [T]_\beta$ , then  $D$  is a diagonal matrix and  $D_{jj}$  is the eigenvalue corresponding to  $v_j$  for  $1 \leq j \leq n$ .*

# Diagonalization

To diagonalize a matrix or a linear operator is to find a basis of eigenvectors and the corresponding eigenvalues.



# Characteristic Polynomial

## Theorem (5.2)

*Let  $A \in M_{n \times n}(F)$ . Then a scalar  $\lambda$  is an eigenvalue of  $A$  if and only if  $\det(A - \lambda I_n) = 0$ .*

# Characteristic Polynomial

## Definition

Let  $A \in M_{n \times n}(F)$ . The polynomial  $f(t) = \det(A - tI_n)$  is called the characteristic polynomial of  $A$ .

# Characteristic Polynomial

## Definition

Let  $T$  be a linear operator on an  $n$ -dimensional vector space  $V$  with ordered basis  $\beta$ . We define the characteristic polynomial  $f(t)$  of  $T$  to be the characteristic polynomial of  $A = [T]_{\beta}$ :

$$f(t) = \det(A - tI_n).$$

# Properties

## Theorem (5.3)

Let  $A \in M_{n \times n}(F)$ .

- (a) *The characteristic polynomial of  $A$  is a polynomial of degree  $n$  with leading coefficient  $(-1)^n$ .*
- (b)  *$A$  has at most  $n$  distinct eigenvalues.*

# Properties

## Theorem (5.4)

*Let  $T$  be a linear operator on a vector space  $V$ , and let  $\lambda$  be an eigenvalue of  $T$ . A vector  $v \in V$  is an eigenvector of  $T$  corresponding to  $\lambda$  if and only if  $v \neq 0$  and  $v \in N(T - \lambda I)$ .*

## 5.2 Diagonalizability

- Diagonalizability
- Multiplicity
- Direct Sums

# Diagonalizability

## Theorem (5.5)

Let  $T$  be a linear operator on a vector space  $V$ , and let  $\lambda_1, \dots, \lambda_k$  be distinct eigenvalues of  $T$ . If  $v_1, \dots, v_k$  are the corresponding eigenvectors, then  $\{v_1, \dots, v_k\}$  is linearly independent.

## Corollary

Let  $T$  be a linear operator on an  $n$ -dimensional vector space  $V$ . If  $T$  has  $n$  distinct eigenvalues, then  $T$  is diagonalizable.

# Diagonalizability (cont.)

## Definition

A polynomial  $f(t)$  in  $P(F)$  splits over  $F$  if there are scalars  $c, a_1, \dots, a_n$  in  $F$  such that  $f(t) = c(t - a_1)(t - a_2) \cdots (t - a_n)$ .

## Theorem (5.6)

*The characteristic polynomial of any diagonalizable operator splits.*



# Multiplicity

## Definition

Let  $\lambda$  be an eigenvalue of a linear operator or matrix with characteristic polynomial  $f(t)$ . The (algebraic) multiplicity of  $\lambda$  is the largest positive integer  $k$  for which  $(t - \lambda)^k$  is a factor of  $f(t)$ .

## Definition

Let  $T$  be a linear operator on a vector space  $V$ , and let  $\lambda$  be an eigenvalue of  $T$ . Define  $E_\lambda = \{x \in V : T(x) = \lambda x\} = N(T - I_V)$ . The set  $E_\lambda$  is the eigenspace of  $T$  corresponding to the eigenvalue  $\lambda$ . The eigenspace of a square matrix  $A$  is the eigenspace of  $L_A$ .

# Multiplicity (cont.)

## Theorem (5.7)

*Let  $T$  be a linear operator on a finite-dimensional vector space  $V$ , and let  $\lambda$  be an eigenvalue of  $T$  having multiplicity  $m$ . Then  $1 \leq \dim(E_\lambda) \leq m$ .*

# Diagonalizability

## Lemma

Let  $T$  be a linear operator, and let  $\lambda_1, \dots, \lambda_k$  be distinct eigenvalues of  $T$ . For  $i = 1, \dots, k$ , let  $v_i \in E_{\lambda_i}$ . If

$$v_1 + v_2 + \dots + v_k = 0,$$

then  $v_i = 0$  for all  $i$ .

## Theorem (5.8)

*Let  $T$  be a linear operator on a vector space  $V$ , and let  $\lambda_1, \dots, \lambda_k$  be distinct eigenvalues of  $T$ . For  $i = 1, \dots, k$ , let  $S_i$  be a finite linearly independent subset of the eigenspace  $E_{\lambda_i}$ . Then  $S = S_1 \cup S_2 \cup \dots \cup S_k$  is a linearly independent subset of  $V$ .*

# Diagonalizability

## Theorem (5.9)

Let  $T$  be a linear operator on a finite-dimensional vector space  $V$  such that the characteristic polynomial of  $T$  splits. Let  $\lambda_1, \dots, \lambda_k$  be the distinct eigenvalues of  $T$ . Then

- (a)  $T$  is diagonalizable if and only if the multiplicity of  $\lambda_i$  is equal to  $\dim(E_{\lambda_i})$  for all  $i$ .
- (b) If  $T$  is diagonalizable and  $\beta_i$  is an ordered basis for  $E_{\lambda_i}$ , for each  $i$ , then  $\beta = \beta_1 \cup \beta_2 \cup \dots \cup \beta_k$  is an ordered basis for  $V$  consisting of eigenvectors of  $T$ .

# Diagonalizability (cont.)

## Test for Diagonalization

Let  $T$  be a linear operator on an  $n$ -dimensional vector space  $V$ . Then  $T$  is diagonalizable if and only if both of the following conditions hold.

- The characteristic polynomial of  $T$  splits.
- The multiplicity of each eigenvalue  $\lambda$  equals  $n - \text{rank}(T - \lambda I)$ .

# Direct Sums

## Definition

The sum of the subspaces  $W_1, \dots, W_k$  of a vector space is the set

$$\sum_{i=1}^k W_i = \{v_1 + \dots + v_k : v_i \in W_i \text{ for } 1 \leq i \leq k\}.$$

## Definition

A vector space  $V$  is the direct sum of subspaces  $W_1, \dots, W_k$ , denoted  $V = W_1 \oplus \dots \oplus W_k$ , if

$$V = \sum_{i=1}^k W_i \text{ and } W_j \cap \sum_{i \neq j} W_i = \{0\} \text{ for each } j, 1 \leq j \leq k.$$

# Direct Sums (cont.)

## Theorem (5.10)

Let  $W_1, \dots, W_k$  be subspaces of finite-dimensional vector space  $V$ . The following are equivalent:

- (a)  $V = W_1 \oplus \dots \oplus W_k$ .
- (b)  $V = \sum_{i=1}^k W_i$  and for any  $v_1, \dots, v_k$  s.t.  $v_i \in W_i$  ( $1 \leq i \leq k$ ), if  $v_1 + \dots + v_k = 0$ , then  $v_i = 0$  for all  $i$ .
- (c) Each  $v \in V$  can be uniquely written as  $v = v_1 + \dots + v_k$ , where  $v_i \in W_i$ .
- (d) If  $\gamma_i$  is an ordered basis for  $W_i$  ( $1 \leq i \leq k$ ), then  $\gamma_1 \cup \dots \cup \gamma_k$  is an ordered basis for  $V$ .
- (e) For each  $i = 1, \dots, k$  there exists an ordered basis  $\gamma_i$  for  $W_i$  such that  $\gamma_1 \cup \dots \cup \gamma_k$  is an ordered basis for  $V$ .

# Direct Sums (cont.)

## Theorem (5.11)

*A linear operator  $T$  on finite-dimensional vector space  $V$  is diagonalizable if and only if  $V$  is the direct sum of the eigenspaces of  $T$ .*



## 5.3 Matrix Limits and Markov Chains

- Matrix Limits
- Existence of Limits

# Matrix Limits

## Definition

Let  $L, A_1, A_2, \dots$  be  $n \times p$  matrices with complex entries. The sequence  $A_1, A_2, \dots$  is said to *converge to the limit*  $L$  if  $\lim_{m \rightarrow \infty} (A_m)_{ij} = L_{ij}$  for all  $1 \leq i \leq n$  and  $1 \leq j \leq p$ . If  $L$  is the limit of the sequence, we write  $\lim_{m \rightarrow \infty} A_m = L$ .

## Theorem (5.12)

Let  $A_1, A_2, \dots$  be a sequence of  $n \times p$  matrices with complex entries that converges to  $L$ . Then for any  $P \in M_{r \times n}(\mathbb{C})$  and  $Q \in M_{p \times s}(\mathbb{C})$ ,

$$\lim_{m \rightarrow \infty} PA_m = PL \text{ and } \lim_{m \rightarrow \infty} A_m Q = LQ.$$

# Matrix Limits (cont.)

## Corollary

Let  $A \in M_{n \times n}(\mathbb{C})$  be such that  $\lim_{m \rightarrow \infty} A^m = L$ . Then for any invertible  $Q \in M_{n \times n}(\mathbb{C})$ ,

$$\lim_{m \rightarrow \infty} (QAQ^{-1})^m = QLQ^{-1}.$$

# Existence of Limits

Consider the set consisting of the complex number 1 and the interior of the unit disk:  $S = \{\lambda \in \mathbb{C} : |\lambda| < 1 \text{ or } \lambda = 1\}$ .

## Theorem (5.13)

*Let  $A$  be a square matrix with complex entries. Then  $\lim_{m \rightarrow \infty} A^m$  exists if and only if both of the following hold:*

- (a) Every eigenvalue of  $A$  is contained in  $S$ .*
- (b) If 1 is an eigenvalue of  $A$ , then the dimension of the eigenspace corresponding to 1 equals the multiplicity of 1 as an eigenvalue of  $A$ .*

# Existence of Limits (cont.)

## Theorem (5.14)

Let  $A \in M_{n \times n}(\mathbb{C})$ .  $\lim_{m \rightarrow \infty} A^m$  exists if

- (a) Every eigenvalue of  $A$  is contained in  $S$ .
- (b)  $A$  is diagonalizable.