# Chapter 5 Diagonalization

# 5.1 Eigenvalues and Eigenvectors

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

Remark. Want to determine if an linear operator T is diagonalizable and if so, ways to obtain the basis  $\beta = \{v_1, v_2, \dots, v_n\}$  for V such that  $[T]_{\beta}$  is a diagonal matrix. Note that if  $D = [T]_{\beta}$  is a diagonal matrix, i.e.  $D_{ij} = 0$  for  $i \neq j$ , then for each  $v_j \in \beta$ , we have

$$T(v_j) = \sum_{i=1}^{n} D_{ij}v_i = D_{jj}v_j = \lambda_j v_j$$

where  $\lambda_j = D_{jj}$ . Conversely, if  $\beta = \{v_1, v_2, \dots, v_n\}$  is an ordered basis for V such that  $T(v_j) = \lambda_j v_j$  for some scalars  $\lambda_1, \lambda_2, \dots, \lambda_n$ , then

$$[T]_{\beta} = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}$$

Definition. Eigenvalue and Eigenvector (characteristic/proper value or vector) Let T be a linear operator on a vector space V. A nonzero vector  $v \in V$  is called an eigenvector of T if there exists a scalar  $\lambda$  such that  $T(v) = \lambda v$ . The scalar  $\lambda$  is called the eigenvalue corresponding to the eigenvector v.

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

## Theorem. 5.1 Sufficient Condition for Diagonalizability

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 (i.e.  $v \in V$  is eigenvector if exists  $\lambda$  such that  $T(v) = \lambda v$ ). Furthermore, if T is diagonalizable,  $\beta = \{v_1, v_2, \cdots, v_n\}$  is an ordered basis of eigenvectors of T, and  $D = [T]_{\beta}$ , then D is diagonal matrix and  $D_{jj}$  is the eigenvalue corresponding to  $v_j$  for  $1 \leq j \leq n$ 

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

### Theorem. 5.2 Computing Eigenvalues

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

*Proof.* A scalar is an eigenvalue if and only if exists a nonzero vector  $v \in F^n$  such that  $Av = \lambda v$ , that is,  $(A - \lambda I_n)(v) = 0$ , which is true if and only if  $A - \lambda$  is not invertible (invertible and one-to-one, or  $N(A - \lambda) = \{0\}$  equivalent). This is equivalent to  $det(A - \lambda I_n) = 0$ 

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

- 1. The eigenvalues of a matrix are the zeros of its characteristic polynomial
- 2. To determine the eigenvalues of a matrix or linear operator, we normally compute its characteristic polynomial.

**Definition.** Characteristic Polynomial of a Linear Operator 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  $A = [T]_{\beta}$ . That is,

$$f(t) = det(A - tI_n) \qquad P_T(t) = det([T]_{\beta} - tI_n) = P_{[T]_{\beta}}(t)$$

We denote characteristic polynomial of an operator T by det(T-tI). Note the definition is independent of the choice of ordered basis  $\beta$ , the resulting characteristic polynomial is the same regardless the choice of basis.

*Proof.* Let  $\beta$  and  $\beta'$  be basis of V, let Q be change of basis matrix from  $\beta'$  to  $\beta$ , then we have  $[T]_{\beta'} = Q^{-1}[T]_{\beta}Q$ , so then characteristic polynomial of linear operator invariant of choice of basis

$$\det([T]_{\beta'} - tI_V) = \det(Q^{-1}([T]_{\beta} - tI_V)Q) = \det(Q^{-1})\det([T]_{\beta} - tI_V)\det(Q) = \det([T]_{\beta} - tI_V)$$

Theorem. 5.3 Properties of Characteristic Polynomial Let  $A \in M_{n \times n}(F)$ 

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

#### Theorem. 5.4 Computing Eigenvectors

Let T be a linear operator on a vector space T, 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  $v \in N(T - \lambda I)$ 

# Proposition. Equivalent Eigenvector for Matrix and Linear Operators

Let  $T: V \to V$  be a linear operator and  $\beta$  be an ordered basis for V. Let  $A = [T]_{\beta}$  and note  $\phi_{\beta}(v) = [v]_{\beta}$ , the cooredinate vector of v relative to  $\beta$ . We could show that for all  $v \in V$  an eigenvector of T corresponding to an eigenvalue v if and only if v is an eigenvector of v corresponding to v. Now suppose v is an eigenvector of v corresponding to v, then

$$A\phi_{\beta}(v) = L_A\phi_{\beta}(v) = \phi_{\beta}T(v) = \phi_{\beta}(\lambda v) = \lambda\phi_{\beta}(v)$$

Note  $\phi_{\beta}(v) \neq 0$ , since  $\phi_{\beta}$  is an isomorphism, we have proved that  $\phi_{\beta}(v)$  is an eigenvector of A. Conversely, if  $\phi_{\beta}(v)$  is an eigenvector of A corresponding to  $\lambda$ . Equivalently, a

vector  $y \in F^n$  is an eigenvector of  $A = [T]_{\beta}$  corresponding to  $\lambda$  if and only if  $\phi_{\beta}^{-1}(y)$  is an eigenvector of T corresponding to  $\lambda$ . We have reduced the problem of finding the eigenvectors of a linear operator on a finite-dimensional vector space to the problem of finding the eigenvectors of a matrix.

Definition. Geometric Description of how a linear operator T acts on an eigenvector in the context of a vector space V over  $\mathbb{R}$ . Let v be eigenvector of T and  $\lambda$  be corresponding eigenvalue. Let  $W = span(\{v\})$ , the one-dimensional subspace of V spanned by v, a line passing through 0 and v. For any  $w \in W$ , w = cv for some  $c \in \mathbb{R}$ 

$$T(w) = T(cv) = cT(v) = c\lambda v = \lambda w$$

T acts on the vector in W by multiplying each such vector by  $\lambda$ 

# 5.2 Diagonalizability

#### Theorem. 5.5 Set of Eigenvectors is Linearly Independent

Let T be a linear operator on a vector space V, and let  $\lambda_1, \lambda_2, \dots, \lambda_k$  be **distinct** eigenvalues of T. If  $v_1, v_2, \dots, v_k$  are eigenvectors of T such that  $\lambda_i$  corresponding to  $v_i$  where  $1 \leq i \leq k$  (choose one eigenvector corresponding to each eigenvalue.), then  $\{v_1, v_2, \dots, v_k\}$  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.

*Proof.* Suppose T has n distinct eigenvalues  $\lambda_1, \dots, \lambda_n$ . For each i choose eigenvector  $v_i$  corresponding to  $\lambda_i$ . By previous theorem,  $\{v_1, \dots, v_n\}$  is linearly independent, and since dim(V) = n. the set is a basis for V. Thus, by theorem 5.1, T is diagonalizable  $\square$ 

1. Converse not true, if T is diagonalizable, then it need not have n distinct eigenvalues. For example,  $I_V$  is diagonalizable even though it has only 1 eigenvalue,  $\lambda = 1$ 

**Definition.** Splits Over A polynomial  $f(t) \in P(F)$  splits over F if there are scalars  $c, a_1, \dots, a_n$  (not necessarily distinct) in F such that

$$f(t) = c(t - a_1)(t - a_2) \cdots (t - a_n)$$

As an example  $t^2 = (t+1)(t-1)$  splits over  $\mathbb{R}$ , but  $(t^2+1)(t-2)$  does not split over  $\mathbb{R}$  but splits over  $\mathbb{C}$  since it factors into (t+i)(t-i)(t-2).

#### Theorem. 5.6 Diagonalizability implies f(t) Splits Completely

The characteristic polynomial of any diagonalizable linear operator splits. The converse is not true, i.e. that the characteristic polynomial of T may split but need not be diagonalizable.

*Proof.* Let T be linear operator on n-dimensional vector space V, and let  $\beta$  be an ordered basis for V such that  $[T]_{\beta} = D$  is a diagonal matrix. Suppose

$$D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}$$

and let f(t) be characteristic polynomial of T, then

$$f(t) = det(D - tI) = \begin{pmatrix} \lambda_1 - t & 0 & \cdots & 0 \\ 0 & \lambda_2 - t & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & \lambda_n - t \end{pmatrix} = (-1)^n (t - \lambda_1)(t - \lambda_2) \cdots (t - \lambda_n)$$

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

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

### Theorem. 5.7 Dimension of Eigenspace is Bounded by Multiplicity

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  $1 \leq \dim(E_{\lambda}) \leq m$ 

**Lemma.** Let T be a linear operator, and let  $\lambda_1, \lambda_2, \dots, \lambda_k$  be distinct eigenvalues of T. For each  $i = 1, 2, \dots, k$ , let  $v_i \in E_{\lambda_i}$ , the eigenspace corresponding to  $\lambda_i$ . If

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

then  $v_i = 0$  for all i.

#### Theorem. 5.8 Union of l.i. Subsets of Eigenspaces are l.i.

Let T be a linear operator on a vector space V, and elt  $\lambda_1, \lambda_2, \dots, \lambda_k$  be distinct eigenvalues of T. for each  $i = 1, 2, \dots, k$ , let  $S_i$  be a finite linearly independent subset of eigenspace  $E_{\lambda_i}$ . Then  $S = S_1 \cup S_2 \cup \dots \cup S_k$  is a linearly independent subset of V.

# Theorem. 5.9 Construct Bases of Eigenvectors in Eigenspace to Form a Basis for the Entire Space

Let T be a linear operator on a finite-dimensional vector space V such that the characteristic polynomial of T splits. Let  $\lambda_1, \lambda_2, \dots, \lambda_k$  be distinct eigenvalues of T. Then

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

**Definition.** Test for Diagonalization Let T be a linear operator on n-dimensional vector space V. Then T is diagonalizable if and only if both of conditions hold

- 1. characteristic polynomial of T splits
- 2. For each eigenvalue  $\lambda$  of T, the multiplicity of  $\lambda$  equals  $n rank(T \lambda I) = dim(E_{\lambda})$

Same condition can be used to test a square matrix A is diagonalizable because diagonalizability of A is equivalent to diagonalizability of  $L_A$ . To test T for diagonalizability, usually pick a basis  $\alpha$  and let  $B = [T]_{\alpha}$ . If characteristic polynomial of B splits, then use condition 2 to check if the multiplicity of each repeated eigenvalue of B equals  $n - rank(B - \lambda I)$  (dont need to check for eigenvalues with multiplicity 1 by theorem 5.7). If so, then B, and hence T, is diagonalizable. If T is diagonalizable, we can find a basis  $\beta$  for V consisting of eigenvectors of T by taking the union of basis for each eigenspace of B. Furthermore, if A is  $n \times n$  diagonalizable matrix, we can find an invertible  $n \times n$  matrix Q, and a diagonal matrix  $n \times n$  matrix D such that  $Q^{-1}AQ = D$  with Q having its columns the vectors in the basis of eigenvectors of A, and D having its jth column entry the eigenvalue of A corresponding to jth column of Q.

# Application: Closed Formula for Exponential of Diagonalizable Matrix Application: System of Differential Equations

**Definition.** Matrix Exponential For  $A \in M_{n \times n}(C)$ , define  $e^A = \lim_{m \to \infty} B_m$ , where

$$B_m = I + A + \frac{A^2}{2!} + \dots + \frac{A^m}{m!}$$

So  $e^A$  is the sum of infinite series

$$I + A + \frac{A^2}{2!} + \cdots$$

Note

$$e = \sum_{n=0}^{\infty} \frac{1}{n!} \qquad e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

# 5.4 Invariant Subspaces and the Cayley-Hamilton Theorem

**Definition.** T-invariant Subspace Let T be a linear operator on a vector space V. A subspace W of V is called a T-invariant subspace of V if  $T(W) \subseteq W$ , that is, if  $T(v) \in W$  for all  $v \in W$ 

**Definition.** T-cyclic Subspace of V Generated by x Let T be a linear operator on a vector space V, and let x be a nonzero vector in V. The subspace

$$W = span(\{x, T(x), T^2(x), \dots\})$$

is called the T-cyclic subspace of V generated by x, denoted by  $\langle v \rangle_T$ .

- 1. W is a T invariant subspace
- 2. W is the smallest subspace of V containing x; any T-invariant subspace of V containing x must also contain W

Proof on how  $T_W$  is a linear operator on W

**Theorem. 5.21** Characteristic Polynomial of  $T_W$  Divides That of T Let T be a linear operator on a finite-dimensional vector space V, and let W be a T-invariant subspace of V. Let  $T_W$  be restriction of T to W. Then the characteristic polynomial of  $T_W$  divides the chracteristic polynomial of T.

- 5.4:12 Proof Supplementary to theorem 5.21
- 4.3:12 Proof Supplementary to theorem 5.21

Theorem. 5.22 Basis and Characteristic Polynomial For T-Invariant Subspace are Readily Computable Let T be a linear operator on a finite-dimensional vector space V, and let W denote the T-cyclic subspace of V generated by a nonzero vector  $v \in V$ . Let k = dim(W). Then

- 1.  $\{v, T(v), T^2(v), \dots, T^{k-1}(v)\}\$ is a basis for W
- 2. If  $a_0v + a_1T(v) + \cdots + a_{k-1}T^{k-1}(v) = 0$ , then the characteristic polynomial of  $T_W$  is

$$f(t) = (-1)^k (a_0 + a_1 t + \dots + a_{k-1} t^{k-1} + t^k)$$

Idea is we can easily compute the characteristic polynomial by computing  $T^k(v)$  and express it as a linear combination of the basis  $\{v, T(v), \cdots, T^{k-1}(v)\}$ , and use the 2nd claim of the theorem. Of course, we can derive chracteristic polynomial by computing determinants. By theorem 5.21, we can use characteristic polynomial of  $T_W$  to gain information about the characteristic polynomial of T itself

5.4:12 for theorem 5.22, proves the second claim by induction

**Definition.** Linear Operator over a Polynomial Let  $f(x) = a_0 + a_1x + \cdots + a_nx^n$  be a polynomial with coefficients from a field F. If T is a linear operator on a vector space V over F, or similarly for  $A \in M_{n \times n}(F)$ , we define

$$f(T) = a_0 I + a_1 T + \dots + a_n T^n$$
  $f(A) = a_0 I + a_1 A + \dots + a_n A^n$ 

**Theorem. 5.23** Cayley-Hamilton Theorem Let T be a linear operator on a finite-dimensional vector space V, and let f(t) be the characteristic polynomial of T. Then  $f(T) = T_0$ , the zero transformation. That is T satisfies its characteristic equation.

Proof. Prove f(T)(v) = 0 for all  $v \in W$ . Idea is to consider a T-invariant subspace W for v chosen. Write down kth element as a linear combination of preivous basis vectors and compute its characteristic polynomial using method in theorem 5.22 and notice  $P_{T_W}(T) = T_0$ , i.e.  $P_{T_W}(T)(v) = 0$  for all  $v \in W$ . Also  $P_{T_W}$  divides  $P_T$ , from here results follows  $\square$ 

Corollary. Cayley-Hamilton Theorem for Matrices Let A be  $n \times n$  matrix, and let f(t) be the characteristic polynomial of A. Then f(A) = 0, the  $n \times n$  zero matrix.