

Lecture Diagonalizability

- In the previous lecture, we saw that **not every** linear operator or matrix is diagonalizable.
- While we already have a **necessary and sufficient condition** for diagonalizability (**Theorem 5.1**), we haven't yet developed a **practical method** to test whether a given operator (or matrix) *can* be diagonalized—and if so, **how to do it**.

 This lecture fills that gap.

We'll develop a step-by-step method to:

-  Determine if a linear operator is diagonalizable.
-  Understand the role of algebraic vs geometric multiplicity.
-  Actually **construct a basis of eigenvectors** when diagonalizability holds.

By the end of this lecture, you will have a **clear test** (**Theorem 5.8**) and a good understanding of the **diagonalization process**.

Theorem 5.5.

Let T be a linear operator on a vector space, and let $\lambda_1, \lambda_2, \dots, \lambda_k$ be distinct eigenvalues of T . For each $i = 1, 2, \dots, k$, let S_i be a finite set of eigenvectors of T corresponding to λ_i . If each S_i ($i = 1, 2, \dots, k$), is linearly independent, then $S_1 \cup S_2 \cup \dots \cup S_k$ is linearly independent.

Proof: We prove this by induction.

- Base step. When $k=1$, it is obvious S_1 is linearly independent. by the assumption.
- Induction step. Assume the theorem is true for $k-1$ distinct eigenvalues, and $k \geq 2$

We will show the theorem is true for k .

For $1 \leq i \leq k$, consider $S_i = \{v_{i1}, v_{i2}, \dots, v_{in_i}\}$,

be linearly independent set of eigenvectors corresponding to λ_i . We want to show. $S = S_1 \cup S_2 \cup \dots \cup S_k$ is linearly independent.

Let a_{ij} be any scalar, where $1 \leq i \leq k$, $1 \leq j \leq n_i$.

s.t. $\sum_{i=1}^k \sum_{j=1}^{n_i} a_{ij} \vec{v}_{ij} = \vec{0}$ $\textcircled{*}$

As \vec{v}_{ij} is an eigenvector of T w.r.t. λ_i , $T\vec{v}_{ij} = \lambda_i \vec{v}_{ij}$.

or $(T - \lambda_i I_v) \vec{v}_{ij} = \vec{0}$

Apply $(T - \lambda_k I_v)$ on both sides of $\textcircled{*}$, we have.

$$\vec{0} = (T - \lambda_k I_v) \sum_{i=1}^k \sum_{j=1}^{n_i} a_{ij} \vec{v}_{ij} = \sum_{i=1}^k \sum_{j=1}^{n_i} a_{ij} (T - \lambda_k I_v) \vec{v}_{ij}$$

$$\begin{aligned}
 &= \sum_{i=1}^{k-1} \sum_{j=1}^{n_i} a_{ij} (T - \lambda_k I_v) \vec{v}_{ij} + \sum_{j=1}^{n_k} a_{kj} (T - \lambda_k I_v) \vec{v}_{kj} \xrightarrow{\text{0}} \\
 &= \sum_{i=1}^{k-1} \sum_{j=1}^{n_i} a_{ij} (T \vec{v}_{ij} - \lambda_k \vec{v}_{ij}) \\
 &= \sum_{i=1}^{k-1} \sum_{j=1}^{n_i} a_{ij} (\lambda_i - \lambda_k) \vec{v}_{ij} = \vec{0}
 \end{aligned}$$

As $S_1 \cup S_2 \cup \dots \cup S_{k-1}$ is linearly independent by the induction hypothesis, we must have.

$$a_{ij} (\lambda_i - \lambda_k) = 0, \text{ for } 1 \leq i \leq k-1, 1 \leq j \leq n_i$$

As $\lambda_1, \dots, \lambda_k$ are distinct, we know $\lambda_i - \lambda_k \neq 0$ for $1 \leq i \leq k-1$. Therefore $a_{ij} = 0$ for $1 \leq i \leq k-1$ and $1 \leq j \leq n_i$.

Then $\left(\sum_{i=1}^k \sum_{j=1}^{n_i} a_{ij} \vec{v}_{ij} = \vec{0} \right)$ reduces to

$$\sum_{j=1}^{n_k} a_{kj} \vec{v}_{kj} = \vec{0}.$$

Since S_k is also linearly independent $a_{kj} = 0$, $1 \leq j \leq n_k$.

Therefore $a_{ij} = 0$, where $1 \leq i \leq k$, $1 \leq j \leq n_i$.

So S is linearly independent. \square

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: Exercise.

Remark. The converse of the above corollary is not true. That is, even if T is diagonalizable, it does **not** necessarily have n distinct eigenvalues. For example, the identity operator is diagonalizable but has only one eigenvalue.

Example 1: Let

$$A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \in M_{2 \times 2}(\mathbb{R}).$$

The characteristic polynomial of A (thus L_A) is

$$\det(A - tI) = \det \begin{pmatrix} 1-t & 1 \\ 1 & 1-t \end{pmatrix} = t(t-2).$$

So the eigenvalues of A are 0 and 2. By the previous corollary, we know A (and hence L_A) is diagonalizable.

Definition.

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

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

Examples.

1. $t^2 - 1 = (t + 1)(t - 1)$ splits over \mathbb{R} .
2. $(t^2 + 1)(t - 2)$ does not split over \mathbb{R} since $t^2 + 1$ cannot be factored into a product of linear factors over \mathbb{R} .

However, $(t^2 + 1)(t - 2)$ does split over \mathbb{C} because it factors into the product $(t + i)(t - i)(t - 2)$.

Theorem 5.6.

The characteristic polynomial of any diagonalizable linear operator on a vector space V over a field F splits over F .

Proof: Let T be a diagonalizable linear

operator on V , where $\dim(V) = n$.

Let β be such a basis so

that.

$[T]_{\beta} = D$ is diagonal and

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

Then by the def of the char. poly. of T is

$$f(t) = \det(D - tI_n) = \begin{vmatrix} \lambda_1 - t & 0 & \cdots & 0 \\ 0 & \lambda_2 - t & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n - t \end{vmatrix}$$

would introduce a -1

$$= (\lambda_1 - t)(\lambda_2 - t) \cdots (\lambda_n - t)$$

$$= (-1)^n (t - \lambda_1)(t - \lambda_2) \cdots (t - \lambda_n)$$

Thus by def of $f(t)$ splits over F (previous page), $f(t)$ split over F .

Remark.

- The converse is false:
 - A characteristic polynomial may **split** (i.e., factor completely),
 - but T may still **not** be diagonalizable. (See **Exercise 2** below)

Definition: Determinant and Characteristic Polynomial of an Operator.

Let T be a linear operator on a finite-dimensional vector space V , and let β be any ordered basis for V .

We define:

- The **determinant** of T , denoted $\det(T)$, to be the determinant of the matrix $A = [T]_{\beta}$.
- The **characteristic polynomial** of T to be the characteristic polynomial of A . That is,

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

These definitions are well-defined because they do not depend on the choice of basis.

The following concept provides a criterion for determining when an operator with a split characteristic polynomial is diagonalizable.

Definition.

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

Remark

- If T is a diagonalizable linear operator on a finite-dimensional vector space V , then by **Theorem 5.1**:
 - The eigenvalues of T appear on the diagonal of $[T]_{\beta}$, each repeated according to its **algebraic multiplicity**.
- Therefore:
 - The number of **linearly independent eigenvectors** associated to an eigenvalue λ —also known as the **geometric multiplicity** of λ —plays a key role in determining whether T is diagonalizable.
- According to **Theorem 5.4**: $(\vec{v} \text{ is an eigenvector to } \lambda \Leftrightarrow (T - \lambda I)\vec{v} = \vec{0}) \Leftrightarrow \vec{v} \in \ker(T - \lambda I)$
 - Eigenvectors corresponding to an eigenvalue λ lie in $\ker(T - \lambda I_V)$.
 - This motivates the study of $\ker(T - \lambda I_V)$ in analyzing diagonalizability.

Want: $\dim(\ker(T - \lambda I_V)) = \text{multiplicity of } \lambda$ to ensure enough elements to form a basis.

Definition.

Let T be a linear operator on a vector space V , and let λ be an eigenvalue of T . The **eigenspace** of T corresponding to λ is defined as

$$E_{\lambda} = \{x \in V : T(x) = \lambda x\} = \ker(T - \lambda I_V).$$

Analogously, for a square matrix A , the eigenspace corresponding to the eigenvalue λ is defined to be the eigenspace of the associated linear operator L_A , that is,

$$E_{\lambda} = \ker(A - \lambda I).$$

So T is diagonal

The next theorem relates the **dimension of the eigenspace** E_λ (also called the **geometric multiplicity**) to the **algebraic multiplicity** of the eigenvalue λ .

Theorem 5.7.

Let T be a linear operator on a finite-dimensional vector space V , and let λ be an eigenvalue of T with algebraic multiplicity m . Then

$$1 \leq \dim(E_\lambda) \leq m.$$

Proof: See Page 264.

Exercise 2: Let T be the linear operator on $P_2(\mathbb{R})$ defined by $T(f(x)) = f'(x)$.

In this exercise, you will determine whether T is diagonalizable.

(1) Write down the matrix representation of T with respect to the standard basis $\beta = \{1, x, x^2\}$ for $P_2(\mathbb{R})$.

$$T(1) = 1' = 0, \quad T(x) = x' = 1 + 0 \cdot x + 0 \cdot x^2, \quad T(x^2) = 0 \cdot 1 + 2 \cdot x + 0 \cdot x^2$$

Thus $[T]_\beta = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}$

(2) Compute the characteristic polynomial of T . Then identify the eigenvalues and their algebraic multiplicities.

$$\text{Let } \det([T]_\beta - \lambda I_3) = \begin{vmatrix} -\lambda & 1 & 0 \\ 0 & -\lambda & 2 \\ 0 & 0 & -\lambda \end{vmatrix} = -\lambda^3 = 0$$

$\Rightarrow \lambda = 0$ with algebraic multiplicity 3.

(3) Solve $T(f(x)) = f'(x) = 0$ to find the eigenspace $E_0 = \ker(T)$. What is the geometric multiplicity of $\lambda = 0$? Recall geometric multiplicity of an eigenvalue λ is $\dim(E_\lambda)$.

Then eigenvector to $\lambda=0$ satisfies

$$T(f(x)) = \lambda f(x) \Rightarrow f'(x) = 0, \text{ which are}$$

$$\overset{\parallel}{f'(x)} = \overset{\parallel}{0} \quad \text{precisely all constants.}$$

(4) Is T diagonalizable? Why?

No, Because there is

Thus $\{1\}$. so. $\dim(E_{\lambda}) = 1$.

no basis of $P_2(\mathbb{R})$ consisting of eigenvectors of T .

So T is not diagonalizable.

Example 3 Let T be the linear operator on \mathbb{R}^3 defined by

$$T \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} = \begin{pmatrix} 4a_1 + a_3 \\ 2a_1 + 3a_2 + 2a_3 \\ a_1 + 4a_3 \end{pmatrix}.$$

(1) Find the eigenvalues and corresponding eigenvectors of T .

(2) Decide whether T is diagonalizable.

(1) Let $\beta = \{e_1, e_2, e_3\}$ be the set of standard basis for \mathbb{R}^3 .

Then

$$T(e_1) = T \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \\ 1 \end{bmatrix}, \quad T(e_2) = T \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \\ 0 \end{bmatrix}, \quad T(e_3) = T \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \\ 2 \end{bmatrix}$$

Thus

$$A = [T]_{\beta} = \begin{bmatrix} 4 & 0 & 1 \\ 2 & 3 & 2 \\ 1 & 0 & 4 \end{bmatrix}$$

We compute.

$$\det(A - \lambda I_3) = \begin{vmatrix} 4-\lambda & 0 & 1 \\ 2 & 3-\lambda & 2 \\ 1 & 0 & 4-\lambda \end{vmatrix} = (4-\lambda) \begin{vmatrix} 3-\lambda & 2 \\ 0 & 4-\lambda \end{vmatrix}$$

$$+ \begin{vmatrix} 2 & 3-\lambda \\ 1 & 0 \end{vmatrix} = (4-\lambda)(3-\lambda) - (3-\lambda) = (3-\lambda)[(\lambda-4)^2 - 1]$$

$$= -(\lambda-3)(\lambda^2 - 8\lambda + 15) = -(\lambda-3)(\lambda-3)(\lambda-5) = 0$$

Thus $\lambda_1 = 5$, $\lambda_2 = 3$ with multiplicity 2.

- When $\lambda_1 = 5$, we solve.

$$(A - \lambda_1 I) \vec{v} = \vec{0} \Rightarrow \begin{bmatrix} -1 & 0 & 1 \\ 2 & -2 & 2 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\left[\begin{array}{ccc|c} -1 & 0 & 1 & 0 \\ 2 & -2 & 2 & 0 \\ 1 & 0 & -1 & 0 \end{array} \right] \sim \left[\begin{array}{ccc|c} 0 & 0 & 0 & 0 \\ 0 & -2 & 4 & 0 \\ 1 & 0 & -1 & 0 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 0 & -1 & 0 \\ 0 & 1 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$\Rightarrow \begin{cases} x_1 = x_3 \\ x_2 = 2x_3 \end{cases} \quad \text{Thus } \vec{v} = \begin{bmatrix} x_3 \\ 2x_3 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

Thus an eigenvector to $\lambda_1 = 5$ is a scalar of $\vec{v}_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$

So $E_1 = \text{span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \right\}$. $\dim E_1 = 1$

When $\lambda_2 = 3$, we solve

$$(A - \lambda_2 I) \vec{v} = \vec{0} \Rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 2 & 0 & 2 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x_1 = -x_3.$$

Thus $\vec{v} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -x_3 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ x_2 \\ 0 \end{bmatrix} + \begin{bmatrix} -x_3 \\ 0 \\ x_3 \end{bmatrix} = x_2 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$

So $\left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$ is a basis for $E_{\lambda_2} = E_3$.

and $\dim E_3 = 2$ which coincides with the algebraic multiplicity of $\lambda_2 = 3$.

By Thm 5.3. $\beta' = \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$ forms a linearly independent set with 3 vectors thus a basis for \mathbb{R}^3 consisting of eigenvectors of T.

So T is diagonalizable.

Exercise 2: $\dim(E_\lambda) = 1 <$ alg. multiplicity of λ . T not diagonalizable

Question:

What can you observe by comparing the geometric multiplicity $\dim(E_\lambda)$ and the algebraic multiplicity of each λ in **Exercise 2** and **Example 3**?

Example 3. $\dim(E_{\lambda_1}) = \text{alg. mul. of } \lambda_1$, and $\dim(E_{\lambda_2}) = \text{alg. mul. of } \lambda_2$,
 T is diagonalizable.

The guess from the observation is in general true. We have the following Theorem:

Theorem 5.8

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

(a) T is diagonalizable if and only if the algebraic multiplicity of each λ_i is equal to $\dim(E_{\lambda_i})$, where E_{λ_i} is the eigenspace corresponding to λ_i .

(b) If T is diagonalizable and β_i is an ordered basis of E_{λ_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 .

Proof. (a). $\forall i, i=1, \dots, k$, let m_i be the (algebraic) multiplicity of λ_i :
 $d_i = \dim(E_i)$, $n = \dim(V)$.

" \Rightarrow " If T is diagonalizable, β is a basis of eigenvectors of T . Let $\beta_i = \beta \cap E_{\lambda_i}$ and let $n_i = |\beta_i|$, i.e., the number of vectors in β_i .

Then

- $n_i \leq d_i = \dim(E_i)$, since β_i is linearly independent subset of β and E_i .
- $\dim(E_{\lambda_i}) = d_i \leq m_i$ by Thm 5.7.

Then

$$n = \sum_{i=1}^k n_i \leq \sum_{i=1}^k d_i \leq \sum_{i=1}^k m_i = n$$

(forces all the inequalities to be " $=$ ")

Since β has n vectors

Thus $\sum_{i=1}^k d_i = \sum_{i=1}^k m_i$

$$\Rightarrow \sum_{i=1}^k (m_i - d_i) = 0.$$

since the degree of the char. poly of T = sum of the (alg) multiplicity of eigenvalues.

As $(m_i - d_i) \geq 0 \forall i$, we know $m_i = d_i$.

Thus the algebraic multiplicity of λ_i (m_i) is the same as $\dim(E_{\lambda_i}) = d_i$.

" \Leftarrow " Assume $m_i = d_i, \forall i = 1, \dots, k$, we show T is diagonalizable.

$\forall i$, let β_i be an ordered basis for E_{λ_i} . and

$$\beta = \beta_1 \cup \beta_2 \cup \dots \cup \beta_k$$

By Thm 5.5, β is linearly independent.

Also as $d_i = m_i, \forall i$, β has.

$$\sum_{i=1}^k d_i = \sum_{i=1}^k m_i = n.$$

vectors.

Thus β is an ordered basis consisting of the eigenvectors so T is diagonalizable. This also proves (b)

Theorem 5.8 completes our study of the diagonalization problem. We summarize our results.

Test for Diagonalizability:

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:

1. The characteristic polynomial of T **splits** (i.e., it factors completely into linear terms over the field).
2. For each eigenvalue λ of T , the **algebraic multiplicity** of λ equals the **nullity** of $T - \lambda I$; that is,
$$\text{algebraic multiplicity of } \lambda = \text{geometric multiplicity of } \lambda = \text{nullity}(T - \lambda I) = n - \text{rank}(T - \lambda I).$$

These same conditions can be used to test whether a square matrix A is diagonalizable, since diagonalizability of A is equivalent to the diagonalizability of the associated linear operator L_A .