

MA-110 Linear Algebra and Differential Equations

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Summary: Eigenvalues and Characteristic Polynomial

Let A be $n \times n$.

- 1 The *characteristic polynomial* of A is $\det(A - \lambda I)$ (of degree n) and its roots are the *eigenvalues* of A .
- 2 For each eigenvalue λ , the associated *eigenspace* is $N(A - \lambda I)$. To find it, solve $(A - \lambda I)v = 0$. Any non-zero vector in $N(A - \lambda I)$ is an *eigenvector* associated to λ .
- 3 If A is a *diagonal matrix* with diagonal entries $\lambda_1, \dots, \lambda_n$, then its *eigenvalues* are $\lambda_1, \dots, \lambda_n$ with associated eigenvectors e_1, \dots, e_n respectively.
- 4 Write $\det(A - \lambda I) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda)$ and expand.

$$\begin{aligned}\text{Trace of } A &= a_{11} + \cdots + a_{nn} \quad (\text{sum of diagonal entries}) \\ &= \lambda_1 + \cdots + \lambda_n\end{aligned}$$

$$\det(A) = \lambda_1 \cdots \lambda_n$$

Thus: If $\lambda_1, \dots, \lambda_n$ are real numbers, then $\text{Tr}(A) = \text{sum of eigenvalues}$, and $\det(A) = \text{product of eigenvalues}$.

Similarity and Eigenvalues

Defn. The $n \times n$ matrices A and B are *similar*, if there exists an invertible matrix P such that $P^{-1}AP = B$.

Observe: If $B = P^{-1}AP$, then (i) $\det(A) = \det(B)$, and (ii) $B^n = P^{-1}A^nP$ for each n .

Theorem: If A and B are similar, then they have the same characteristic polynomial. In particular, they have the same eigenvalues, $\det(A) = \det(B)$ and $\text{Trace}(A) = \text{Trace}(B)$.

Proof. Given: $B = P^{-1}AP$. prove: $\det(A - \lambda I) = \det(B - \lambda I)$.

Note: It is enough to prove that $A - \lambda I$ and $B - \lambda I$ are similar! Indeed, $B - \lambda I = P^{-1}AP - \lambda P^{-1}P$ constant coefficient is the determinant of the matrix.
$$= P^{-1}(A - \lambda I)P.$$
 □

Write $\det(A - \lambda I) = (\lambda_1 - \lambda) \cdots (\lambda_n - \lambda)$. Compare constant coeff.: $\det(A) = \lambda_1 \cdots \lambda_n = \det(B)$; Compare coeff. of λ^{n-1} :

Sum of diagonal entries $= a_{11} + \cdots + a_{nn} = \text{Trace of } A = \lambda_1 + \cdots + \lambda_n = \text{Trace of } B$.

Ques: How are eigenvalues of A and B related?

Diagonalizability: Introduction

Note: Finding roots of characteristic polynomials (and hence eigenvalues) is difficult in general.

For $n \geq 5$, no formula exists for roots. (Abel, Galois)

For $n = 3, 4$, formulae for root exist, but not easy to use.

Defn. An $n \times n$ matrix A is diagonalizable if A is similar to a diagonal matrix Λ , i.e., there is an invertible matrix P and a diagonal matrix Λ such that $P^{-1}AP = \Lambda$.

Importance of Diagonalizability:

Let the $n \times n$ matrix A be diagonalizable, i.e., $P^{-1}AP = \Lambda$, where P is invertible and Λ is diagonal. If this happens,

- The eigenvalues of A are the diagonal entries of Λ ,
- $\det(A)$ is the product of the diagonal entries of Λ , and
- $\text{Trace}(A) = \text{sum of the diagonal entries of } \Lambda$.
- **Other Information:** e.g., what is $\text{Trace}(A^n)$?

Diagonalization: Example

Example: $A = \begin{pmatrix} 1 & 5 & 6 \\ 0 & 2 & -4 \\ 0 & 0 & 3 \end{pmatrix}$ is triangular.

$$\det(A - \lambda I) = (1 - \lambda)(2 - \lambda)(3 - \lambda).$$

Eigenvalues: $\lambda_1 = 1$, $\lambda_2 = 2$, $\lambda_3 = 3$.

Note: If A is triangular, its eigenvalues are on the diagonal

Eigenvectors: $v_1 = e_1$, $v_2 = \begin{pmatrix} 5 & 1 & 0 \end{pmatrix}^T$, $v_3 = \begin{pmatrix} -7 & -4 & 1 \end{pmatrix}^T$.

(**How?**) Further, $\{v_1, v_2, v_3\}$ is a basis of \mathbb{R}^3 . Hence

$P = \begin{pmatrix} v_1 & v_2 & v_3 \end{pmatrix}$ is invertible, and

$AP = \begin{pmatrix} Av_1 & Av_2 & Av_3 \end{pmatrix} = \begin{pmatrix} v_1 & 2v_2 & 3v_3 \end{pmatrix} = P\Lambda$, where

$\Lambda = \begin{pmatrix} 1 & & \\ & 2 & \\ & & 3 \end{pmatrix}$. Thus $P^{-1}AP = \Lambda$, i.e., A is diagonalizable.

Example: If $\mathcal{B} = \{v_1, v_2, v_3\}$, and $T(v) = Av$, then

$[T]_{\mathcal{B}}^{\mathcal{B}} = ____$.

Eigenvalue Decomposition (EVD)

Question: What is the advantage of a basis of \mathbb{R}^n consisting of eigenvectors?

Let A be an $n \times n$ matrix with n eigenvectors v_1, \dots, v_n , associated to eigenvalues $\lambda_1, \dots, \lambda_n$. If $\mathcal{B} = \{v_1, \dots, v_n\}$ is a basis of \mathbb{R}^n , then the matrix $P = (v_1 \ \cdots \ v_n)$ is invertible.

Moreover, $AP = A(v_1 \ \cdots \ v_n) = (Av_1 \ \cdots \ Av_n)$

$$= (\lambda_1 v_1 \ \cdots \ \lambda_n v_n) = P\Lambda, \text{ where } \Lambda = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}.$$

Therefore $P^{-1}AP = \Lambda$, i.e., A is similar to a diagonal matrix.

Thus: Eigenvectors diagonalize a matrix

Eigenvalue Decomposition (EVD): Let A be diagonalizable. With notation as above, we have $A = P\Lambda P^{-1}$.

This is called as the **eigenvalue decomposition (EVD)** of A .

Diagonalizability and Eigenvectors

Theorem A is diagonalizable $\Leftrightarrow A$ has n linearly independent eigenvectors. In particular, \mathbb{R}^n has a basis consisting of eigenvectors of A .

Proof. (\Leftarrow): Done! To prove (\Rightarrow), assume $P = (v_1 \ \cdots \ v_n)$ is an invertible matrix such that $P^{-1}AP = \Lambda = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$.

Then $AP = P\Lambda$, i.e. $(Av_1 \ \dots \ Av_n) = (\lambda_1 v_1 \ \dots \ \lambda_n v_n)$. Therefore v_1, \dots, v_n are eigenvectors of A . They are linearly independent since P is invertible. □

Question: Is every matrix is diagonalizable? **A:** No.

Examples: $Q = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$ no eigenvalues (over \mathbb{R})!

$P = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ not enough eigenvectors!

When is A Diagonalizable?

Ques: When does A have n linearly independent eigenvectors?

- If v_1, \dots, v_r are eigenvectors of A associated to distinct eigenvalues $\lambda_1, \dots, \lambda_r$, then v_1, \dots, v_r are linearly independent.

Proof. Suppose v_1, \dots, v_r are linearly dependent. Choose a linear relation involving minimum number of v_i 's, say

$$(1) \quad a_1 v_1 + \dots + a_t v_t = 0. \quad (1 < t \leq r, t \text{ is minimal, } a_i \neq 0)$$

Apply A to get
$$a_1 \lambda_1 v_1 + \dots + a_t \lambda_t v_t = 0 \quad (2)$$

$$\lambda_1(1) - (2) \text{ gives } a_2(\lambda_1 - \lambda_2)v_2 + \dots + a_t(\lambda_1 - \lambda_t)v_t = 0,$$

which contradicts the minimality of t . □

- If A has n distinct eigenvalues, then A is diagonalizable.

Proof. If v_1, \dots, v_n are eigenvectors associated to distinct eigenvalues $\lambda_1, \dots, \lambda_n$, then $\{v_1, \dots, v_n\}$ is linearly independent. Then $P = (v_1 \ \dots \ v_n)$ is invertible, and $P^{-1}AP = \Lambda$ as seen earlier. Hence A is diagonalizable. □

Reading Slide - Eigenvalues of AB and $A+B$

- If λ is an eigenvalue of A , μ is an eigenvalue of B , is $\lambda\mu$ an eigenvalue of AB ?

False Proof. $ABx = A(\mu x) = \mu(Ax) = \lambda\mu x.$

This is false since A and B may not have same eigenvector x .

- **Example:** $A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$, $B = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$, $AB = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}.$

The eigenvalues of A and B are 0,0 and that of AB are 1,0.

- Eigenvalues of $A+B$ are NOT $\lambda + \mu$.

In above example, $A+B = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ has eigenvalues 1, -1.

- If A and B have same eigenvectors associated to λ and μ , then $\lambda\mu$ and $\lambda + \mu$ are eigenvalues of AB and $A+B$ respectively.

Question: When do A and B have the same eigenvectors?