

MA101 Mathematics I

Department of Mathematics
Indian Institute of Technology Guwahati

Jul – Nov 2013

Slides originally created by: [Dr. Bikash Bhattacharjya](#)

Instructors:

[Rafikul Alam](#), [Bhaba K. Sarma](#), [Sriparna Bandyopadhyay](#), [Kalpesh Kapoor](#)

ver. 21 Sept 2013, 10:18 (bks)

1 / 18

Slides 6

PLAN

- Determinant of Matrices
- Eigenvalues and Eigenvectors

2 / 18

Determinant of a matrix

Recall:

- $\det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = a_{11}a_{22} - a_{12}a_{21}.$

- $\det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$

$$= a_{11} \det \begin{bmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{bmatrix} - a_{12} \det \begin{bmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{bmatrix} + a_{13} \det \begin{bmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix}$$

Definition

Let $A = [a_{ij}] \in \mathcal{M}_n(\mathbb{F})$. The **determinant** of A is defined by

$\det(A) = a_{11}$, if $n = 1$, i.e., $A = [a_{11}]$. For $n \geq 2$:

$$\det(A) = a_{11}\det(A_{11}) - a_{12}\det(A_{12}) + \dots + (-1)^{1+n}a_{1n}\det(A_{1n}),$$

where A_{ij} be the submatrix of A obtained by deleting the i -th row and the j -th column of A .

Note that $\det(A) \in \mathbb{F}$.

3 / 18

Properties of Determinant

Let A be a square matrix.

- 1 If B is obtained by **interchanging two rows** of A , then $\det(B) = -\det(A)$.
- 2 If A has a **zero row**, then $\det(A) = 0$.
- 3 If A has **two identical rows**, then $\det(A) = 0$.
- 4 If B is obtained by **multiplying a row** of A by a scalar α , then $\det(B) = \alpha \det(A)$.
- 5 If the matrices A, B and C are identical except that one of the rows of C is the sum of the corresponding rows of A and B , then $\det(C) = \det(A) + \det(B)$.
- 6 If B is obtained by adding a **multiple of one row** of A to another row, then $\det(B) = \det(A)$.
- 7 $\det(\alpha A) = \alpha^n \det A$.

4 / 18

Example

$$\det \begin{bmatrix} 1 & 5 & 0 & 0 \\ 2 & 0 & 8 & 0 \\ 3 & 6 & 9 & 0 \\ 4 & 7 & 10 & 1 \end{bmatrix} = 1 \cdot \det \begin{bmatrix} 0 & 8 & 0 \\ 6 & 9 & 0 \\ 7 & 10 & 1 \end{bmatrix} - 5 \cdot \det \begin{bmatrix} 2 & 8 & 0 \\ 3 & 9 & 0 \\ 4 & 10 & 1 \end{bmatrix} = -18.$$

or

$$\begin{aligned} &= \det \begin{bmatrix} 1 & 5 & 0 & 0 \\ 0 & -10 & 8 & 0 \\ 0 & -9 & 9 & 0 \\ 0 & -13 & 10 & 1 \end{bmatrix} \begin{matrix} E_{21}(-2) \\ E_{31}(-3) \\ E_{41}(-4) \end{matrix} = 18 \cdot \det \begin{bmatrix} 1 & 5 & 0 & 0 \\ 0 & 5 & -4 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & -13 & 10 & 1 \end{bmatrix} \begin{matrix} E_2(-\frac{1}{2}) \\ E_3(-\frac{1}{9}) \end{matrix} \\ &= 18 \cdot \det \begin{bmatrix} 1 & 5 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & -3 & 1 \end{bmatrix} \begin{matrix} E_{23}(-5) \\ E_{43}(13) \end{matrix} = 18 \cdot (-1) \cdot \det \begin{bmatrix} 1 & 5 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{matrix} E_{43}(3) \\ E_{23} \end{matrix} \\ &= 18. \end{aligned}$$

Result

The determinant of a *diagonal, upper or lower triangular matrix* is the *product of its diagonal entries*.

5 / 18

Result

Let $A \in \mathcal{M}_n(\mathbb{F})$. Then for any $i, j \in \{1, \dots, n\}$

$$\begin{aligned} \det(A) &= (-1)^{i+1} a_{i1} A_{i1} + (-1)^{i+2} a_{i2} A_{i2} + \dots + (-1)^{i+n} a_{in} A_{in} \\ &\quad \text{(Expansion along the } i\text{-th row)} \\ &= (-1)^{1+j} a_{1j} A_{1j} + (-1)^{2+j} a_{2j} A_{2j} + \dots + (-1)^{n+j} a_{nj} A_{nj} \\ &\quad \text{(Expansion along the } j\text{-th column).} \end{aligned}$$

Result

- For any $A, B \in \mathcal{M}_n(\mathbb{F})$, $\det(AB) = \det(A) \det(B)$.
- Determinants of elementary matrices:
 $\det E_{ij} = -1$, $\det E_i(\alpha) = \alpha$, $\det E_{ij}(\alpha) = 1$.
- For any elementary matrix E , $\det(E) = \det(E^T)$.

6 / 18

Two important properties

Suppose $A \in \mathcal{M}_n$. Now, $A = E_k \cdots E_1 \cdot \text{rref}(A)$ for some elementary E_i . We have

$$\det(A) = \det(E_k) \cdots \det(E_1) \det(\text{rref}(A)).$$

Thus,

$\det(A) \neq 0$ iff $\det(\text{rref}(A)) \neq 0$ iff $\text{rref}(A) = I_n$ iff A is invertible.

Definition

A matrix $A \in \mathcal{M}_n$ is said to be **singular** or **non-singular** according as $\det(A) = 0$ or $\det(A) \neq 0$.

Result

A matrix $A \in \mathcal{M}_n$ is **invertible** if and only if it is **non-singular**.

Result

For any matrix $A \in \mathcal{M}_n$, $\det(A) = \det(A^T)$.

Optional: For some additional material, you may consult books or the write up on determinant available at MA101 webpage.

7 / 18

Eigenvalues and Eigenvectors

Suppose \mathbb{V} is a vector space and $T : \mathbb{V} \rightarrow \mathbb{V}$ is a linear transformation. Suppose $B = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a basis of \mathbb{V} and $\lambda_i \in \mathbb{F}$ be such that $T(\mathbf{v}_i) = \lambda_i \mathbf{v}_i$. Then, T is **nicely** described by

$$T(\alpha_1 \mathbf{v}_1 + \cdots + \alpha_n \mathbf{v}_n) = \lambda_1 \alpha_1 \mathbf{v}_1 + \cdots + \lambda_n \alpha_n \mathbf{v}_n.$$

Remark

- $[T\mathbf{v}_i]_B = \lambda_i [\mathbf{v}_i]_B$.
- $[T]_B = [[T(\mathbf{v}_1)]_B, \dots, [T(\mathbf{v}_n)]_B] = \text{diag}\{\lambda_1, \dots, \lambda_n\}$.

Definition

Suppose \mathbb{V} is a vector space over \mathbb{F} and $T : \mathbb{V} \rightarrow \mathbb{V}$ is a linear transformation. Suppose there is $\lambda \in \mathbb{F}$ and $\mathbf{0} \neq \mathbf{v} \in \mathbb{V}$ such that $T(\mathbf{v}) = \lambda \mathbf{v}$. Then, λ is called an **eigenvalue** of T and \mathbf{v} an **eigenvector** of T corresponding to λ .

8 / 18

Example

Let $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $T(\mathbf{x}) = A\mathbf{x}$, where $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$.

Then, 4 is an eigenvalue of T with $[1, 1]^t$ as a corresponding eigenvector.

Definition

Let $A \in \mathcal{M}_n(\mathbb{F})$. If $A\mathbf{x} = \lambda\mathbf{x}$ for some $\lambda \in \mathbb{F}$ and $\mathbf{0} \neq \mathbf{x} \in \mathbb{F}^n$, then λ is called an **eigenvalue** of A and \mathbf{x} an **eigenvector** of A corresponding to λ .

Remark

- Suppose $\dim(\mathbb{V}) = n$, and $T : \mathbb{V} \rightarrow \mathbb{V}$ is an LT. Then $T(\mathbf{v}) = \lambda\mathbf{v}$ iff $[T(\mathbf{v})]_B = \lambda[\mathbf{v}]_B$ for any ordered basis B of \mathbb{V} . This gives the motivation for studying eigenvalues/eigenvectors of square matrices.

9 / 18

Definition

Let λ be an eigenvalue of a matrix A . Then

$$E_\lambda = \{\mathbf{x} \mid A\mathbf{x} = \lambda\mathbf{x}\}$$

is called the **eigenspace of λ** . Note that E_λ consists of all eigenvectors of A corresponding to λ , together with the **zero vector**.

Result

- $E_\lambda = \text{null}(A - \lambda I)$, and therefore E_λ is a **subspace** of \mathbb{C}^n .
- λ is an eigenvalue of A iff $\det(A - \lambda I) = 0$.

Definition

Let A be an $n \times n$ matrix. Then

- $\det(A - \lambda I)$ is called the **characteristic polynomial** of A .
- $\det(A - \lambda I) = 0$ is called the **characteristic equation** of A .

10 / 18

Result

- The eigenvalues of A are the **zeroes** of the characteristic polynomial $\det(A - \lambda I)$ of A .
- The eigenvalues of a **triangular** matrix are its **diagonal** entries.
- A is **invertible** iff 0 is not an eigenvalue of A .

Example

What are the eigenvalues of $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$ and their eigenspaces?

We have

$$\det(A - \lambda I) = \det \begin{bmatrix} 1 - \lambda & 3 \\ 3 & 1 - \lambda \end{bmatrix} = 0 \Leftrightarrow \lambda = 4, -2,$$

$$E_4 = \text{null}(A - 4I) = \left\{ \mathbf{x} : \begin{bmatrix} -3 & 3 \\ 3 & -3 \end{bmatrix} \mathbf{x} = \mathbf{0} \right\} = \text{span}([1, 1]^T),$$

$$E_{(-2)} = \text{null}(A + 2I) = \left\{ \mathbf{x} : \begin{bmatrix} 3 & 3 \\ 3 & 3 \end{bmatrix} \mathbf{x} = \mathbf{0} \right\} = \text{span}([1, -1]^T),$$

11 / 18

To find eigenvalues and bases for eigenspaces of A

- 1 Compute the char. poly. $\det(A - \lambda I)$.
- 2 Solve the char. eqn. $\det(A - \lambda I) = 0$ for eigenvalues λ .
- 3 For each eigenvalue λ , find $\text{null}(A - \lambda I) = E_\lambda$.
- 4 Find a basis for each eigenspace.

Example

A real matrix may have complex eigenvalues and complex eigenvectors. No wonder, every real matrix is also a complex matrix. Consider $A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$. The char. poly. of A is $\lambda^2 + 1$ and so the eigenvalues are $\pm i$. Corresponding to the eigenvalues $\pm i$, it has eigenvectors $[1, i]^T$ and $[i, 1]^T$, respectively. The eigenspaces E_i and $E_{(-i)}$ are one dimensional.

12 / 18

Example

What are the eigenvalues and the corresponding eigenspaces of

$A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$? Eigenvalues are 1, 1, 1 (1 appearing three times, multiplicity three). Moreover,

$$E_1 = \text{null} \left(\begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \right) = \text{span} \left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right).$$

We have $\dim(E_1) = 1$.

Definition

Let λ be an eigenvalue of a matrix A .

- The **algebraic multiplicity** of λ is the multiplicity of λ as a root of the characteristic polynomial of A .
- The **geometric multiplicity** of λ is the dimension of E_λ .

For an eigenvalue, the **algebraic** and **geometric multiplicities** can be different. See previous example.

13 / 18

Result

*Eigenvectors corresponding to **distinct** eigenvalues are **linearly independent**.*

PROOF. Suppose that $\lambda_1, \dots, \lambda_m$ are distinct eigenvalues of A , \mathbf{x}_i is an eigenvector of A corresponding to λ_i . Suppose, if possible, \mathbf{x}_i are not LI. Then there is $k > 1$ such that $\{\mathbf{x}_1, \dots, \mathbf{x}_{k-1}\}$ is LI, but $\{\mathbf{x}_1, \dots, \mathbf{x}_k\}$ is not. Let $\mathbf{x}_k = \alpha_1 \mathbf{x}_1 + \dots + \alpha_{k-1} \mathbf{x}_{k-1}$. Then,

$$A\mathbf{x}_k = \lambda_1 \alpha_1 \mathbf{x}_1 + \dots + \lambda_{k-1} \alpha_{k-1} \mathbf{x}_{k-1}.$$

$$\begin{aligned} \text{Therefore, } \mathbf{0} &= \lambda_k \mathbf{x}_k - A\mathbf{x}_k \\ &= (\lambda_k - \lambda_1) \alpha_1 \mathbf{x}_1 + \dots + (\lambda_k - \lambda_{k-1}) \alpha_{k-1} \mathbf{x}_{k-1}. \end{aligned}$$

Since $\{\mathbf{x}_1, \dots, \mathbf{x}_{k-1}\}$ is LI,

$$(\lambda_k - \lambda_1) \alpha_1 = \dots = (\lambda_k - \lambda_{k-1}) \alpha_{k-1} = 0,$$

i.e., $\alpha_1 = \dots = \alpha_{k-1} = 0$ (as λ_i are distinct), i.e. $\mathbf{x}_k = \mathbf{0}$, a contradiction. ■

14 / 18

CAYLEY-HAMILTON THEOREM

Let $p(\lambda)$ be the characteristic polynomial of a matrix A . Then $p(A) = \mathbf{0}$, the zero matrix.

[This is a beautiful and useful theorem. However, we omit the proof.]

Exercise

Let $A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$.

- Verify that char. poly. of A is $\lambda^2 - 2\lambda - 3$.
- Verify that $A^2 - 2A - 3I = \mathbf{0}$.
- Use the fact $A^2 = 2A + 3I$ to compute A^5 without computing any matrix multiplication.
- Argue that A is invertible using its char. poly. Since $3I = A^2 - 2A$, we should have $A^{-1} = \frac{1}{3}[A - 2I]$. Verify.

15 / 18

The Fundamental Theorem of Invertible Matrices: Version III

For $A \in \mathcal{M}_n(\mathbb{F})$, ($\mathbb{F} = \mathbb{R}$ or \mathbb{C}) the following are equivalent.

1. A is invertible.
2. $A\mathbf{x} = \mathbf{b}$ has a unique solution for every \mathbf{b} in \mathbb{F}^n .
3. $A\mathbf{x} = \mathbf{0}$ has only the trivial solution.
4. The reduced row echelon form of A is I_n .
5. A is a product of elementary matrices.
6. $\text{rank}(A) = n$.
7. $\text{nullity}(A) = 0$.
8. The column vectors of A are linearly independent.
9. The column vectors of A span \mathbb{F}^n .
10. The column vectors of A form a basis for \mathbb{F}^n .
11. The row vectors of A are linearly independent.
12. The row vectors of A span \mathbb{F}^n .
13. The row vectors of A form a basis for \mathbb{F}^n .

16 / 18

The Fundamental Theorem ... Contd.

14. The map $T : \mathbb{F}^n \rightarrow \mathbb{F}^n$ given by $T(\mathbf{x}) = A\mathbf{x}$ is one-one.
15. The map $T : \mathbb{F}^n \rightarrow \mathbb{F}^n$ given by $T(\mathbf{x}) = A\mathbf{x}$ is onto.
16. The map $T : \mathbb{F}^n \rightarrow \mathbb{F}^n$ given by $T(\mathbf{x}) = A\mathbf{x}$ is a bijection.
17. $\det A \neq 0$, i.e., A is non-singular.
18. 0 is not an eigenvalue of A .

Exercise

- Find the **eigenvalues** and the corresponding **eigenspaces** of the following matrices:

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 2 & -5 & 4 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} -1 & 0 & 1 \\ 3 & 0 & -3 \\ 1 & 0 & -1 \end{bmatrix}.$$

17 / 18

Exercise

- Let A be a matrix with eigenvalue λ and corresponding eigenvector \mathbf{x} .
 - For any positive integer n , show that λ^n is an eigenvalue of A^n with corresponding eigenvector \mathbf{x} .
 - If A is **invertible**, then show that $\frac{1}{\lambda}$ is an eigenvalue of A^{-1} with corresponding eigenvector \mathbf{x} .
 - If A is **invertible** then show that for any integer n , λ^{-n} is an eigenvalue of A^{-n} with corresponding eigenvector \mathbf{x} .
- Let $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$ be eigenvectors of a matrix A with corresponding eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_m$, respectively. Let $\mathbf{x} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_m\mathbf{v}_m$. Show that for any positive integer k ,

$$A^k \mathbf{x} = c_1 \lambda_1^k \mathbf{v}_1 + c_2 \lambda_2^k \mathbf{v}_2 + \dots + c_m \lambda_m^k \mathbf{v}_m.$$

18 / 18