

Determinants

$$A = a_{ij}$$

$$\det A = \sum_{i_1, \dots, i_n} \epsilon_{i_1, \dots, i_n} a_{1i_1} \dots a_{ni_n}$$

$\epsilon_{1,2,\dots,n} = 1$ and it is totally antisymmetric.

- $\det \lambda A = \lambda^n \det A$

- $\det A$ changes sign under the interchange of two rows

- determinant is linear in each row

$$\det \begin{pmatrix} \lambda a_{11} + \mu b_{11} & \dots & \lambda a_{1n} + \mu b_{1n} \\ a_{21} & & a_{2n} \\ \vdots & & \vdots \\ a_{n1} & & a_{nn} \end{pmatrix}$$

$$= \lambda \det \begin{pmatrix} a_{11} & \dots & a_{1n} \\ a_{21} & & a_{2n} \\ \vdots & & \vdots \\ a_{n1} & & a_{nn} \end{pmatrix} + \mu \det \begin{pmatrix} b_{11} & \dots & b_{1n} \\ a_{21} & \dots & a_{2n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix}$$

- the determinant is a skew symmetric n-linear form

$$\omega : V \times V \times \dots \times V \rightarrow \mathbb{F}$$

This space is denoted by $\Lambda^n(V^*)$

$$\omega(\lambda a + \mu b, c_2, \dots, c_n) = \lambda \omega(a, c_2, \dots, c_n) + \mu \omega(b, c_2, \dots, c_n)$$

$$\omega(\dots a \dots b \dots) = -\omega(\dots b \dots a \dots)$$

example $a_i = a_{ij} e_j$

$$\begin{aligned} \omega(a_1 \dots a_n) &= a_{1i_1} \dots a_{ni_n} \omega(e_{i_1}, \dots, e_{i_n}) \\ &= \underbrace{a_{1i_1} \dots a_{ni_n}}_{\det A} \sum_{i_1, \dots, i_n} \omega(e_{i_1}, \dots, e_{i_n}) \end{aligned}$$

definition of ω_A

$$\begin{aligned} \omega_A(x_1 \dots x_n) &= \omega(\underbrace{Ax_1 \dots Ax_n}_{A_{1k} x_k}) \\ &= \underbrace{\omega}_{\substack{\uparrow \\ \text{proportionality constant is } 1 \\ \text{(go to basis)}}} \det A \omega(x_1 \dots x_n) \end{aligned}$$

Can be used to prove properties of determinants.

$$\begin{aligned} \omega_{AB}(x_k) &= \omega(AB x_k) = \det AB \omega(x_k) \\ &= \det A \omega(B x_k) = \det A \det B \omega(x_k) \\ \implies \det AB &= \det A \det B \end{aligned}$$

We also have $\det A^T = \det A$

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$$\begin{aligned}\det A &= \sum_{\pi} \varepsilon_{\pi(1) \dots \pi(n)} A_{1 \pi(1)} \dots A_{n \pi(n)} \\ &= \sum_{\pi} \varepsilon_{\pi(1) \dots \pi(n)} A_{\pi^{-1}(1)} \dots A_{\pi^{-1}(n)}\end{aligned}$$

$$\varepsilon_{\pi(1) \dots \pi(n)} = \varepsilon_{\pi^{-1}(1) \dots \pi^{-1}(n)}$$

$$= \sum_{\pi} \varepsilon_{\pi^{-1}(1) \dots \pi^{-1}(n)} A_{\pi^{-1}(1)} \dots A_{\pi^{-1}(n)}$$

$$= \sum_{\pi^{-1}}$$

$$= \det A^T$$

Adjugate matrix

$$A = (a_{ij})$$

cofactor $C_{ij} = (-1)^{i+j} M_{ij}$
↑ determinant of matrix with row and column of a_{ij} deleted.

$$\sum_j a_{ij} C_{ij} = \delta_{ii} \det A$$

Laplace expansion of determinant
- follows from definition of determinant

Adjugate matrix $(\text{Adj } A)_{ij} = C_{ji}$

$$\text{then } A \text{ Adj } (A) = \det A$$

$$\Rightarrow A^{-1} = \frac{1}{\det A} \text{Adj } A$$

Cayley theorem

characteristic equation

$$\det(A - \lambda I) = 0$$

λ are the eigenvalues of A

$$\det(A - \lambda I) = (-1)^n (\lambda^n - \lambda^{n-1} \text{tr } A + \dots + (-1)^n \det A)$$

Cayley's theorem: A satisfies this equation

proof $\det(A - \lambda I) = (A - \lambda I) \text{Adj}(A - \lambda I)$
 $(-1)^n (\lambda^n + a_1 \lambda^{n-1} + \dots + a_n) = (A - \lambda I) (c_0 \lambda^{n-1} + \dots + c_{n-1})$

the rhs is zero for $\lambda = A$
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Derivative of a determinant

$$\det A = \sum_{i_1, \dots, i_n} \epsilon_{i_1, \dots, i_n} a_{1i_1} \dots a_{ni_n}$$

↑ ↑
functions of x

$$\frac{d}{dx} (\det A) = \sum_{i_1, \dots, i_n} \epsilon_{i_1, \dots, i_n} (a'_{1i_1} a_{2i_2} \dots a_{ni_n} + a_{1i_1} a'_{2i_2} \dots a_{ni_n} + a_{1i_1} a_{2i_2} \dots a'_{ni_n})$$

use Laplace expansion

$$= \sum_{i,j} a'_{ij} C_{ij} \leftarrow \text{cofactor}$$

$$= \sum_{i,j} a'_{ij} (A^{-1})_{ji} \det A$$

$$\Rightarrow \frac{1}{\det A} \frac{d}{dx} \det A = \text{Tr} \frac{dA}{dx} A^{-1}$$

$$\frac{d}{dx} \log \det A$$

$$\text{if } x = a_{ij} \text{ then } \frac{d}{da_{ij}} \log \det A = A^{-1}_{ji}$$

Diagonalization

A is linear map

$$Ax = \lambda x$$

↑
↑

 eigenvalue eigenvector

- Hermitian $A = A^+$

then all eigenvalues are real

$$\begin{aligned}
 \lambda (x, x) &= (x, \lambda x) = \langle x, Ax \rangle \\
 &= (A^+ x, x) \\
 &= (Ax, x) \\
 &= (\lambda x, x) \\
 &= \lambda^* (x, x)
 \end{aligned}$$

$$\Rightarrow \lambda^* = \lambda$$

- $\lambda_k \neq \lambda_e$ then $(x_k, x_e) = 0$
 $A = A^T$

$$\begin{aligned}
 (x_k, Ax_e) &= (Ax_k, x_e) \\
 \lambda_e (x_k, x_e) &= \lambda_k (x_k, x_e) \\
 \lambda_k \neq \lambda_e &\rightarrow (x_k, x_e) = 0
 \end{aligned}$$

- An operator is diagonalizable if \exists basis for which

$$A = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_n \end{pmatrix}$$

not all matrices are diagonalizable

IF the equation of lowest degree satisfied by A has repeated roots

A is not diagonalizable.

Reason: IF A is diagonalizable

$$A = \begin{pmatrix} \lambda_1 & & \\ & \lambda_1 & \\ & & \lambda_2 & \\ & & & \lambda_3 \end{pmatrix}$$

then minimal equation is $(A - \lambda_1)(A - \lambda_2)$

$$\times (A - \lambda_3) = 0$$

so all roots are different.

Jordan Canonical form

Any linear map may be brought into the Jordan canonical form with eigen values on the diagonal and 1 or 0 above the diagonal

Example ; $A = \begin{pmatrix} \lambda_1 & 1 & 0 & \dots & 0 \\ 0 & \lambda_1 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_2 \\ 0 & 0 & 0 & \dots & \lambda_2 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_2 \end{pmatrix}$

$$\det(A - \lambda) = (\lambda - \lambda_1)^3 (\lambda - \lambda_2)^3$$

$$\begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_1 & 1 \\ 0 & 0 & \lambda_1 \end{pmatrix} \begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_1 & 1 \\ 0 & 0 & \lambda_1 \end{pmatrix} = \begin{pmatrix} \lambda_1^2 & 2\lambda_1 & 1 \\ 0 & \lambda_1^2 & 2\lambda_1 \\ 0 & 0 & \lambda_1^2 \end{pmatrix}$$

$$\begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_1 & 1 \\ 0 & 0 & \lambda_1 \end{pmatrix}^3 = \begin{pmatrix} \lambda_1^3 & 3\lambda_1^2 & 3\lambda_1 \\ 0 & \lambda_1^3 & 3\lambda_1^2 \\ 0 & 0 & \lambda_1^3 \end{pmatrix}$$

$$\lambda^3 - 3\lambda^2\lambda_1 + 3\lambda\lambda_1^2 - \lambda_1^3$$

$$\left(\begin{array}{c} + \begin{pmatrix} -3\lambda_1^3 & -6\lambda_1^2 & -3\lambda_1 \\ 0 & -3\lambda_1^3 & -6\lambda_1^2 \\ 0 & 0 & -3\lambda_1^3 \end{pmatrix} \\ \parallel \\ \begin{pmatrix} 3\lambda_1^3 & 3\lambda_1^2 & 0 \\ 0 & 3\lambda_1^3 & 3\lambda_1^2 \\ 0 & 0 & 3\lambda_1^3 \end{pmatrix} \end{array} \right)$$

$$\begin{pmatrix} \lambda_1^3 & 3\lambda_1^2 & 3\lambda_1 \\ \lambda_1^3 & 3\lambda_1^2 & 3\lambda_1 \\ \lambda_1^3 & 3\lambda_1^2 & 3\lambda_1 \end{pmatrix}$$

$$\begin{pmatrix} \lambda_1 & 1 \\ 0 & \lambda_2 \end{pmatrix}^2 = \begin{pmatrix} \lambda_1^2 & 2\lambda_1\lambda_2 \\ 0 & \lambda_2^2 \end{pmatrix}$$

$$\lambda^2 - 2\lambda_1\lambda_2\lambda + \lambda_1\lambda_2^2$$

but for

$$\begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_1 & 0 \\ 0 & 0 & \lambda_2 \end{pmatrix}$$

$$\lambda^2 - 2\lambda_1^2\lambda + \lambda_1^2\lambda_2 = 0$$

⇒ minimal characteristic polynomial is of second order.

$$\begin{pmatrix} \lambda_1 & 1 \\ 0 & \lambda_2 \end{pmatrix}^2 = \begin{pmatrix} \lambda_1^2 & 2\lambda_1 \\ 0 & \lambda_2^2 \end{pmatrix}$$

$$\lambda^2 - 2\lambda_1\lambda + \lambda_2^2 = 0$$

but for

$$\begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_1 & 0 \\ 0 & 0 & \lambda_2 \end{pmatrix}$$

$$\lambda^2 - 2\lambda_1\lambda + \lambda_1 = 0$$

⇒ minimal characteristic polynomial is of second order.