

Review: Determinant

LA 2

24/3/2016

Def: $\det(A) = \sum_{\pi \in S_n} \text{sgn}(\pi) \cdot \prod_{i=1}^n a_{i\pi(i)}$

EROS!

THM 1: $\det(A) = \det(A^T)$

THM 2: \det is a linear function of each of its rows, columns

THM 3: swap of two lines changes the sign of the determinant.

THM 4: i) $A \in T^{n \times n}$ is invertible iff $\det(A) \neq 0$.

ii) $\det(A)$ can be computed by Gaussian elimination:

iii) For every $A, B \in T^{n \times n}$: $\det(B \cdot A) = \det(B) \cdot \det(A)$.

Theorem (Cramer's rule): Let $A \in T^{n \times n}$ be an invertible

matrix and $b \in T^n$. Then the unique solution x of

$Ax = b$ satisfies: $\forall i \in \{1, \dots, n\}$: $x_i = \frac{\det(A_{i \rightarrow b})}{\det(A)}$

where $A_{i \rightarrow b}$ is the matrix obtained from A by replacing its i -th column by b .

Proof: We know that the unique solution x

satisfies $x = A^{-1} \cdot b$.

For $i \in \{1, \dots, n\}$, let $I_i = I_{i \rightarrow A^{-1}b} = \begin{pmatrix} 1 & 2 & \dots & i & \dots & n \\ \vdots & \vdots & \vdots & \boxed{A_i^{-1}b} & \vdots & \vdots \\ 0 & \dots & \dots & \dots & \dots & 1 \end{pmatrix}$

i.e., I_i is obtained from I by replacing its i -th column by $A^{-1} \cdot b$.

- $A_{i \rightarrow b} = A \cdot I_i$
- $\det(I_i) = x_i$

(i -th column of $C \cdot D = C \cdot D_i$)

$\boxed{C \cdot D} = \boxed{C} \cdot \boxed{D}$

\Rightarrow By previous THM, part iii):

$\det(A_{i \rightarrow b}) = \det(A) \cdot \det(I_i) = \det(A) \cdot x_i$

i.e., $x_i = \frac{\det(A_{i \rightarrow b})}{\det(A)}$

\square

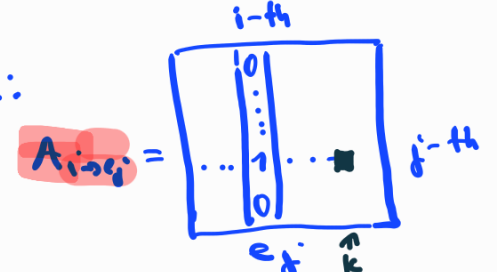
A FORMULA FOR AN INVERSE MATRIX

Given A , find B s.t. $A \cdot B = I$. Computation of j -th col. of A^{-1}

As $I = \begin{pmatrix} e_1 & \dots & e_n \\ \vdots & & \vdots \end{pmatrix}$, the solutions of $Ax = e_j$ for $j=1, \dots, n$ give the columns of A^{-1} .

For each e_j , by Cramer's rule:

$$x_i = \frac{\det(A_{i \rightarrow e_j})}{\det(A)}$$



$\det(A_{i \rightarrow e_j})$ does NOT depend on the (values of) row j of A :

Why: for each $\pi \in S_n$: if $\pi(j) \neq i$, then for some $l \neq j$, $\pi(l) = i \Rightarrow$ the entry on row l , column $\pi(l)$ is 0

i.e., only $\pi \in S_n$ s.t. $\pi(j) = i$ contribute to the det the minor of a_{ji}

\Rightarrow Let A^{ji} denote the $(n-1) \times (n-1)$ submatrix of A

obtained by the removal of row j and column i

Lemma: $\det(A_{i \rightarrow e_j}) = (-1)^{i+j} \det(A^{ji})$

Proof: Consider $(n-j)$ swaps of the row j of $A_{i \rightarrow e_j}$ with the row below it, and then $(n-i)$ swaps of the column i with the columns to the right of it. We obtain the matrix $B = \begin{pmatrix} A^{ji} & \begin{matrix} 0 \\ \vdots \\ 0 \end{matrix} \\ \dots & 1 \end{pmatrix}$.

Each of the swaps changes the sign of the det (THM 3).

$$\Rightarrow \det(A_{i \rightarrow e_j}) = (-1)^{n-i+j} \det(B) = (-1)^{i+j} \det(B)$$

$$\begin{aligned} \det(B) &= \sum_{\pi \in S_n} \text{sgn}(\pi) \prod_{k=1}^n B_{k, \pi(k)} = \sum_{\substack{\pi \in S_n \\ \pi(n) = n}} \text{sgn}(\pi) \prod_{k=1}^{n-1} B_{k, \pi(k)} \\ &= \sum_{\pi \in S_{n-1}} \text{sgn}(\pi) \prod_{k=1}^{n-1} B_{k, \pi(k)} = \det(A^{ji}) \end{aligned}$$

Theorem: For an invertible $A \in T^{n \times n}$

$$(A^{-1})_{ij} = \frac{(-1)^{i+j} \det(A^{ji})}{\det(A)}$$

Theorem (Laplace Expansion): For every $i \in \{1, \dots, n\}$:

$$\det(A) = \sum_{j=1}^n (-1)^{i+j} \cdot A_{ji} \cdot \det(A^{ji}).$$

Expansion by column i .

Proof: The i -th column of A equals $\sum_{j=1}^n A_{ji} \cdot e_j$.

By linearity of determinant, we get

$$\det(A) \stackrel{\downarrow}{=} \sum_{j=1}^n A_{ji} \cdot \det(A_{i \rightarrow e_j}) \stackrel{\uparrow}{=} \sum_{j=1}^n (-1)^{i+j} a_{ji} \det(A^{ji}) \quad \square$$

Lemma \triangle

Def: The adjugate of $A \in T^{n \times n}$, $n \geq 2$, is

the matrix $(\text{adj}(A))_{ij} = (-1)^{i+j} \cdot \det(A^{ji})$ $\begin{matrix} i=1, \dots, n \\ j=1, \dots, n \end{matrix}$.

Note: For an invertible $A \in T^{n \times n}$. $A^{-1} = \frac{\text{adj}(A)}{\det(A)}$

EIGENVALUES AND EIGENVECTORS

• Consider linear transformation $f: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ given

by $f(x) = A \cdot x$ for $A = \begin{pmatrix} 1 & 2 \\ 3 & 2 \end{pmatrix} = \underset{\substack{\uparrow \\ \text{matrix}}} [f]_K$

• Note: A is also the matrix of f w.r.t. canonical basis K .

• Consider vectors $b_1 = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$, $b_2 = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$, basis $B = (b_1, b_2)$

and the matrix of f w.r.t. the basis B .

By definition:

$$[f]_B = \left([f(b_1)]_B, [f(b_2)]_B \right) =$$

$$= \left(\begin{bmatrix} -1 \\ 1 \end{bmatrix}_B, \begin{bmatrix} 8 \\ 12 \end{bmatrix}_B \right) = \begin{pmatrix} -1 & 0 \\ 0 & 4 \end{pmatrix} = A'$$

simple
" a nice matrix - diagonal

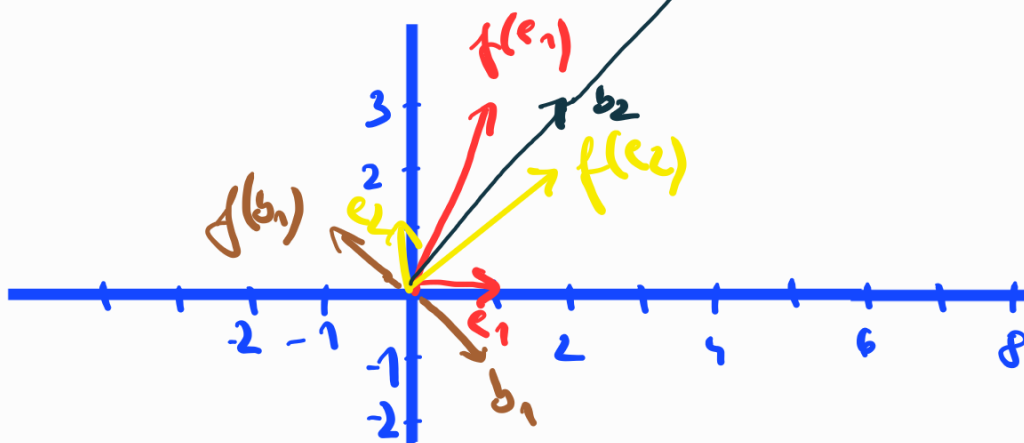
←
conjunctions

Observe:

$$\left. \begin{aligned} f(b_1) &= -b_1 \\ f(b_2) &= 4 \cdot b_2 \end{aligned} \right\}$$

i.e., b_1, b_2 is mapped onto its multiple

$$f(b_2) = \begin{pmatrix} 8 \\ 12 \end{pmatrix}$$



Note: $\det(A) = \det(A') = -4$

What's our goal? For a given lin. trans. $f: V \rightarrow V$
(or for a square matrix A)

- to find a basis B s.t. ${}_B[f]_B$ is simple (\sim diagonal)
- to find directions unaffected by f

Recall: the transition matrix from $B' = (b'_1, \dots, b'_n)$ to B

$$\text{is } [id]_{B'}^B = \begin{pmatrix} [b'_1]_B & \dots & [b'_n]_B \\ \vdots & & \vdots \end{pmatrix}.$$

It holds: ${}_B[id]_{B'} = [id]_{B'}^{-1}$, ${}_B[f]_B = [id]_{B'}^{-1} [f]_{B'} [id]_B$

In our case:

$$\begin{pmatrix} -1 & 0 \\ 0 & 4 \end{pmatrix} = \underbrace{\frac{1}{5} \begin{pmatrix} 3 & -2 \\ 1 & 1 \end{pmatrix}}_{\uparrow \det T = 5} \underbrace{\begin{pmatrix} 1 & 2 \\ 3 & 2 \end{pmatrix}}_A \underbrace{\begin{pmatrix} 1 & 2 \\ -1 & 3 \end{pmatrix}}_T$$

Def: Square matrices A and A' are similar if there exists an invertible R s.t. $A' = R \cdot A \cdot R^{-1}$


Def: A matrix A is diagonalizable if it is similar to a diagonal matrix.

E.g.: A and A' are similar, A is diagonalizable

Def: If $f: V \rightarrow V$ is a lin. trans. (V is VS over T), then $\lambda \in T$ is an eigenvalue of f if there exists non-zero vector $v \in V$ s.t. $f(v) = \lambda \cdot v$.

An eigenvector corresponding to λ is every $v \in V, v \neq 0$, s.t. $f(v) = \lambda \cdot v$.

Def: If $A \in T^{n \times n}$, then $\lambda \in T$ is an eigenvalue of A if there exists a non-zero vector v s.t. $Av = \lambda v$.
 An eigenvector corresponding to λ is every $v \in V, v \neq 0$, s.t. $Av = \lambda v$.

 For an eigenvalue λ of LT f (or of a matrix A), the set $U = \{u : f(u) = \lambda \cdot u\}$ is a vector subspace.
 set of all eigenvectors corresponding to λ and 0 .


Proof: We have to check that U is closed under $+$, \cdot .
 For any $u, v \in U, t \in T$, it holds:

$$f(t \cdot u) = t \cdot f(u) = t \cdot \lambda \cdot u = \lambda \cdot (t \cdot u)$$

$$f(u+v) = f(u) + f(v) = \lambda \cdot u + \lambda \cdot v = \lambda(u+v) \quad \bullet$$

Theorem (Characterization of eigenvalues & determinants)
 For $A \in T^{n \times n}$, λ is an eigenvalue of A $\Leftrightarrow \det(A - \lambda I) = 0$.

Proof: λ is eigenvalue $\Leftrightarrow \exists v \neq 0$ s.t. $Av = \lambda v = \lambda I v$
 $\Leftrightarrow \exists v \neq 0 (A - \lambda I)v = 0 \Leftrightarrow A - \lambda I$ is singular
 $\Leftrightarrow \det(A - \lambda I) = 0 \quad \square$

Def: Characteristic polynomial of the matrix $A \in T^{n \times n}$ is the polynomial $p_A(t) = \det(A - t \cdot I)$ where $I = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix}$.
 Roots of $p_A(t)$ are the eigenvalues of A .