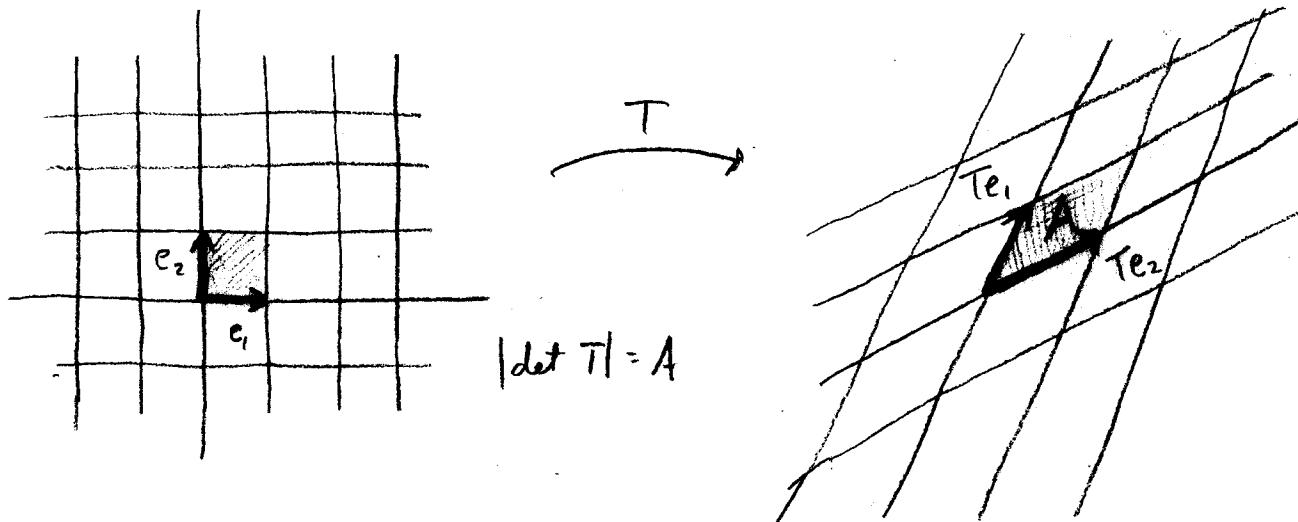


5. Determinant and trace.

The concept of the determinant of a linear map is simple - we will give an unofficial geometric definition as motivation, and then formalize it.

Def: (Unofficial). Let $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a linear map, and $[0,1]^n$ the unit n -cube. The determinant of T is the "signed volume" of $T([0,1]^n)$.



Our unofficial definition has some interesting properties:

- $\det T = 0$ iff T is not invertible
- $\det(TS) = (\det T)(\det S)$
- If T and S differ by swapping two columns, then $\det T = -\det S$.
- $\det T$ is "linear" in each column - i.e., if T is obtained from S by multiplying a column by c , then $\det T = c(\det S)$.

[2]

Permutations:

Def: Let $[n] := \{1, \dots, n\}$. A permutation is a bijection $\pi: [n] \rightarrow [n]$. The set of all permutations of n elements is denoted S_n , and is a group.

We can describe a permutation by a table, or more concisely, by cycle notation:

Example: $\pi: \{1, 2, 3, 4\} \rightarrow \{1, 2, 3, 4\}$.

i	1	2	3	4
$\pi(i)$	2	4	1	3

Table notation:

$$\pi = \begin{matrix} 1 & 2 & 3 & 4 \\ 2 & 4 & 1 & 3 \end{matrix}, \quad \pi^2 = \begin{matrix} 1 & 2 & 3 & 4 \\ 4 & 3 & 2 & 1 \end{matrix}, \quad \pi^3 = \begin{matrix} 1 & 2 & 3 & 4 \\ 3 & 1 & 4 & 2 \end{matrix} = \pi^{-1}, \quad \pi^4 = \begin{matrix} 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \end{matrix}$$

Cycle notation: $\pi = (1 \ 2 \ 4 \ 3)$ meaning $1 \xrightarrow{} 2 \xrightarrow{} 4 \xrightarrow{} 3$

$$\pi^2 = (1 \ 4)(2 \ 3), \quad \pi^3 = (1 \ 3 \ 4 \ 2), \quad \pi^4 = (1)(2)(3)(4).$$

By convention, we usually omit length-1 cycles, e.g., $(12)(3) = (12)$.

Def: Let x_1, \dots, x_n be n variables. Their discriminant is defined as $P(x_1, \dots, x_n) = \prod_{i < j} (x_i - x_j)$.

For a permutation $\pi \in S_n$,

$$P(\pi(x_1, \dots, x_n)) = \prod_{i < j} (x_{\pi(i)} - x_{\pi(j)}) = \pm P(x_1, \dots, x_n).$$

Def: The signature $\text{sgn}(\pi)$ of a permutation $\pi \in S_n$ is defined as $P(\pi(x_1, \dots, x_n)) = \text{sgn}(\pi) P(x_1, \dots, x_n)$.

Clearly, $\text{sgn}(\pi) = \pm 1$.

Def: A transposition is a permutation $\tau \in S_n$ such that for some $j \neq k \in [n]$, $\tau(i) = i \quad i \neq j, k$
 $\tau(j) = k$
 $\tau(k) = j$.

Prop: (i) $\text{sgn}(\pi_1 \circ \pi_2) = \text{sgn}(\pi_1) \text{sgn}(\pi_2)$

(ii) $\text{sgn}(\tau) = -1$ for any transposition

(iii) Every permutation $\pi \in S_n$ can be written as a composition of transpositions: $\pi = \tau_k \circ \dots \circ \tau_1$.

(iv) This decomposition is not unique, but the parity of k , the number of transpositions is.

(v) If $\pi = \tau_k \circ \dots \circ \tau_1$, then $\text{sgn}(\pi) = (-1)^k$.

Proof: Exercise.

Multilinear forms

Def: A k -linear form is a function $f: X_1 \oplus \dots \oplus X_k \rightarrow K$ (we'll assume $X_1 = \dots = X_k = X$) that is linear in each coordinate, i.e., upon fixing any $k-1$ arguments, it remains linear in the remaining argument.

(4)

Example:

- (1) 1-linear forms are just functions
- (2) 2-linear forms are bilinear forms
- (3) A 3-linear form $f: X \oplus X \oplus X \rightarrow K$ has identities such as $f(a_1x_1 + a_2x_2, y, z) = a_1f(x_1, y, z) + a_2f(x_2, y, z)$, and similarly for $f(x, a_1y_1 + a_2y_2, z)$, $f(x, y, a_1z_1 + a_2z_2)$, etc.

Theorem 5.1: The set of k -linear forms is a vector space of dimension n^k .

Proof: (sketch). Verify that a basis consists of functions,

$$\{ f_{j_1 \dots j_k}(x_{i_1}, \dots, x_{i_k}) = \delta_{i_1 j_1} \dots \delta_{i_k j_k} : 1 \leq j_\ell \leq n \}.$$

For any permutation $\pi \in S_k$, write

$$(\pi f)(x_1, \dots, x_k) = f(x_{\pi(1)}, \dots, x_{\pi(k)}).$$

Note that for any k -linear form f , πf is also k -linear.

Def: A k -linear form is symmetric if $\pi f = f$, for all $\pi \in S_k$.

Example: (i) $f(x_1, x_2) = l_1(x_1)l_2(x_2) + l_1(x_2)l_2(x_1)$ for fixed $l_1, l_2 \in X'$.

$$(ii) f(x_1, \dots, x_k) = \sum_{\pi \in S_k} \pi f(x_1, \dots, x_k)$$

Def: A k -linear form is skew-symmetric if $\tau f = -f$ for every transposition $\tau \in S_k$.

Example: $f(x_1, x_2) = l_1(x_1)l_2(x_2) - l_1(x_2)l_2(x_1)$.

Def: A k -linear form is alternating if $f(x_1, \dots, x_k) = 0$ if $x_i = x_j$ for some $i \neq j$.

Prop: The set of alternating (resp. symmetric, or skew-symmetric) k -linear forms is a subspace.

Proof: Exercise.

Theorem 5.2: Every alternating multilinear form is skew-symmetric.

Proof: Pick $i \neq j$, and define $g(x_i, x_j) = f(x_1, \dots, x_k)$ (i.e., the other entries are fixed). Note that g is bilinear, alternating.

$$\begin{aligned} \text{Thus, } 0 &= g(x_i + x_j, x_i + x_j) = g(x_i, x_i) + g(x_j, x_j) + g(x_i, x_j) + g(x_j, x_i) \\ &= g(x_i, x_j) + g(x_j, x_i). \end{aligned}$$

$$\Rightarrow g(x_i, x_j) = -g(x_j, x_i)$$

$$\Rightarrow \tau f = -f \text{ for } \tau = (ij).$$

□

Remark: The converse "almost" holds.

Suppose $f = -f$. Then $(1+1)f = 0 \Rightarrow f = 0$ or $1+1=0$ (e.g., if $K = \mathbb{Z}_2$). Thus, the converse holds for $K \neq \mathbb{Z}_2$.

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Theorem 5.3: IF x_1, \dots, x_n are linearly dependent and f is an alternating k -linear form, then $f(x_1, \dots, x_n) = 0$.

Proof: IF $x_i = 0$, the result is trivial.

otherwise, write $x_h = \sum_{i=1}^{h-1} a_i x_i$ for some x_i , and replace x_h in $f(x_1, \dots, x_k)$ by this sum.

use linearity to break apart $f(x_1, \dots, x_n)$ into a sum, each one having some x_j in the h -position, $j \neq h$.

All of these terms are zero, since f is alternating. \square

The converse holds in one important special case: $k=n$.

Theorem 5.4: IF f is a non-zero alternating n -linear form, and x_1, \dots, x_n are linearly independent, then $f(x_1, \dots, x_n) \neq 0$.

Proof: By assumption, x_1, \dots, x_n is a basis for X .

Pick $y_1, \dots, y_n \in X$, and write $y_i = a_{i1}x_1 + \dots + a_{in}x_n$

$$\begin{aligned} \text{Now, } f(y_1, \dots, y_n) &= f\left(\sum a_{1j}x_1, \dots, \sum a_{nj}x_n\right) \\ &= \sum_i f(z_1, \dots, z_n) \quad z_{ij} \text{ distinct } x_i's. \\ &= \sum_{\text{lots of } \pi} f(x_{\pi(1)}, \dots, x_{\pi(n)}) \\ &= \sum \text{sgn}(\pi) f(x_1, \dots, x_n). \end{aligned}$$

IF $f(x_1, \dots, x_n) = 0 \Rightarrow f(y_1, \dots, y_n) = 0 \Rightarrow f = 0$ $\$$. \square

Remark: The proof of Theorem 5.4 yields an important corollary:

Theorem 5.5: Any two alternating n -linear forms are linearly dependent.

Proof: Let f, g be alternating n -linear forms, and let x_1, \dots, x_n be a basis of X .

Pick $y_1, \dots, y_n \in X$.

$$\text{As before, } f(y_1, \dots, y_n) = \sum_{\text{lots}} \text{sgn}(\pi) f(x_1, \dots, x_n) = a \in K.$$

$$g(y_1, \dots, y_n) = \sum_{\text{lots}} \text{sgn}(\pi) g(x_1, \dots, x_n) = b = \lambda a \in K.$$

$$\text{Thus, } g(x_1, \dots, x_n) = \lambda f(x_1, \dots, x_n).$$

□

* Thus, the space of alternating n -linear forms is at most, one-dimensional.

Theorem 5.6: The vector space of alternating n -linear forms is one-dimensional.

Proof: It suffices to show that there exists one such form. We'll use induction on $k \leq n$.

Base case ($k=1$): Any $\ell \neq 0$ in X' is a non-trivial 1-linear form.

Now, suppose F is a non-zero alternating k -linear form, for $k < n$. We will construct a non-zero alternating $(k+1)$ -linear form g .

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Pick y_1, \dots, y_k such that $f(y_1, \dots, y_k) \neq 0$, and find $y_{k+1} \in X$ not in the subspace $Y := \text{span}\{y_1, \dots, y_k\}$.

Pick $\ell \in Y^\perp$ such that $\ell(y_{k+1}) \neq 0$, and $\ell(y_1) = \dots = \ell(y_k) = 0$.

Define: $g(x_1, \dots, x_k, x_{k+1}) = -f(x_1, \dots, x_k)\ell(x_{k+1}) + \sum_{i=1}^k (i \leq k+1) f(x_1, \dots, x_k) \ell(x_{k+1})$

[For example, if $k=3$, then

$$\begin{aligned} g(x_1, x_2, x_3, x_4) &= f(x_4, x_2, x_3) \ell(x_1) + f(x_1, x_4, x_3) \ell(x_2) \\ &\quad + f(x_1, x_2, x_4) \ell(x_3) - f(x_1, x_2, x_3) \ell(x_4). \end{aligned}$$

It is elementary to prove that this is k -linear. We must show that it is non-zero and alternating.

Non-zero: $g(y_1, \dots, y_k, y_{k+1}) = -f(y_1, \dots, y_k)\ell(y_{k+1}) \neq 0$

Alternating: Consider vectors x_1, \dots, x_k, x_{k+1} such that $i < j$ but $x_i = x_j$. It suffices to prove $g(x_1, \dots, x_k, x_{k+1}) = 0$.

Note: x_i and x_j occur in all but two "terms" of g . All other (besides these two) terms vanish.

Case 1: ($j = k+1$):

$$\begin{aligned} g(x_1, \dots, x_{k+1}) &= (i \leq k+1) f(x_1, \dots, x_k) \ell(x_{k+1}) - f(x_1, \dots, x_k) \ell(x_{k+1}) \\ &= 0 \quad (\text{because } x_i = x_{k+1}). \end{aligned}$$

Case 2: ($j \leq k$):

$$\begin{aligned} g(x_1, \dots, x_{k+1}) &= [(i \leq j) f(x_1, \dots, x_k) - f(x_1, \dots, x_k)] \ell(x_{k+1}) \\ &= 0 \quad (\text{because } f \text{ is alternating.}) \end{aligned}$$

□

Let f be an alternating n -linear form on X , and $T: X \rightarrow X$ a linear map.

Note that $(\bar{T}f)(x_1, \dots, x_n) := f(Tx_1, \dots, Tx_n)$ is alternating and n -linear as well.

* Moreover, \bar{T} is a linear map on the (one-dimensional) space of all such forms. Thus, $\bar{T}: f \mapsto \lambda f$ for some $\lambda \in K$.

Def: This scalar is the determinant of T , denoted $\det T$.

Thus, the determinant satisfies the following:

Universal property: Given a linear map $T: X \rightarrow X$, there exists a unique scalar $\lambda \in K$ such that for every alternating n -linear form f ,

$$f(Tx_1, \dots, Tx_n) = \lambda f(x_1, \dots, x_n).$$

$$\begin{array}{ccc} V^n & \xrightarrow{T \oplus \dots \oplus T} & V^n \\ f \downarrow & & \downarrow f \\ K & \xrightarrow{\lambda} & K \end{array}$$

Remark: This definition is not only independent of matrices, but also independent of choice of basis!

Example: If $Tx = cx$, then

$$(\bar{T}f)(x_1, \dots, x_n) = f(cx_1, \dots, cx_n) = c^n f(x_1, \dots, x_n).$$

Thus, $\det T = c^n$. It follows that $\det 0 = 0$, $\det I = 1$.

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Theorem 5.7: For any two linear maps $A, B : X \rightarrow X$,

$$\det(AB) = (\det A)(\det B).$$

Proof: Write $C = AB$. For an alternating n -linear form f ,

$$\begin{aligned} (\bar{C}f)(x_1, \dots, x_n) &= f(ABx_1, \dots, ABx_n) \\ &= (\bar{A}f)(Bx_1, \dots, Bx_n) = (\bar{B}\bar{A}f)(x_1, \dots, x_n). \end{aligned}$$

Thus, $\bar{C} = \bar{B}\bar{A}$. We have $\bar{C}f = (\det C)f$

$$\begin{aligned} \text{and } \bar{C}f &= \bar{B}\bar{A}f = (\det \bar{B})\bar{A}f = (\det B)(\det A)f \\ \Rightarrow \det(AB) &= (\det A)(\det B). \end{aligned}$$

□

Corollary: $A : X \rightarrow X$ is invertible iff $\det A \neq 0$.

Proof: If A is invertible, then $AA^{-1} = I$.

$$1 = \det I = \det(AA^{-1}) = (\det A)(\det A^{-1}). \quad \checkmark$$

Conversely, if $\det A \neq 0$, and x_1, \dots, x_n is a basis of X , and f a non-zero alternating n -linear form on X , then

$$(\det A)f(x_1, \dots, x_n) \neq 0 \quad (\text{by Theorem 5.4})$$

$\Rightarrow \{Ax_1, \dots, Ax_n\}$ is linearly independent (by Theorem 5.3)

$\Rightarrow A$ is invertible.

□

Determinants and matrices

Let f be a non-zero alternating n -linear form.

Let $\{x_1, \dots, x_n\}$ be a basis of X , and, let $A = (a_{ij})$ be the matrix of a linear map $X \rightarrow X$ wrt this basis.

Recall that $Ax_j = \sum_{i=1}^n a_{ij} x_i$.

$$\begin{aligned} \text{Thus, } (\det A) f(x_1, \dots, x_n) &= f(Ax_1, \dots, Ax_n) \\ &= f\left(\sum_{i=1}^n a_{1i} x_i, \dots, \sum_{i=1}^n a_{ni} x_i\right) \end{aligned}$$

Now, expand this by multilinearity, to get a large sum of terms such as $f(z_1, \dots, z_n)$ where each z_i is a distinct x_i (if $z_i = z_j$ then $f(z_1, \dots, z_n) = 0$).

Moreover, every permutation $f(x_{\pi(1)}, \dots, x_{\pi(n)}) = \pi f(x_1, \dots, x_n)$ appears in this sum.

Remark: The coefficient of $\pi f(x_1, \dots, x_n)$ is $a_{\pi(1),1} \dots a_{\pi(n),n}$.

Since f is skew-symmetric, we have

$$\det A = \sum_{\pi \in S_n} (\text{sgn } \pi) a_{\pi(1),1} \dots a_{\pi(n),n}. \quad (*)$$

Also note that if $\pi, \sigma \in S_n$, then $a_{\pi(1),1} \dots a_{\pi(n),n}$ and $a_{\pi\sigma(1),\sigma(1)} \dots a_{\pi\sigma(n),\sigma(n)}$ differ in the order of the factors only.

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Let's apply this with $\sigma = \pi^{-1}$. (Note: $\text{sgn}(\pi) = \text{sgn}(\pi^{-1})$)

$$\det A = \sum_{\pi \in S_n} (\text{sgn } \pi) a_{\pi(1),1} \cdots a_{\pi(n),n} = \sum_{\pi \in S_n} (\text{sgn } \pi) a_{1,\pi(1)} \cdots a_{n,\pi(n)} = \det A^T.$$

* In summary, the determinant of A can be thought of as an n -linear function of its column vectors, or of its row vectors. In fact it is the unique "normalized" alternating n -linear form, in that $f(e_1, \dots, e_n) = 1$. ($\det I = 1$).

Lemma 5.8: Let $A = (c_1, \dots, c_n)$ be a matrix (c_i is a column vector) and let B be the matrix obtained from A by adding k times the i^{th} column of A to the j^{th} column, $i \neq j$. Then $\det A = \det B$.

Proof: Exercise.

Lemma 5.9: Let A be an $n \times n$ matrix whose first column is

$$e_1 : \quad A = \begin{pmatrix} 1 & * & * & * \\ 0 & \ddots & & \\ \vdots & & A_{11} & \\ 0 & & & \end{pmatrix}. \quad \text{Then } \det A = \det A_{11}$$

Proof: By Lemma 5.8, $\det A = \det \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & \ddots & & \\ \vdots & & A_{11} & \\ 0 & & & \end{pmatrix}$.

Define $f(A_{11}) = \det \begin{pmatrix} 1 & 0 \\ 0 & A_{11} \end{pmatrix}$, a function of an $(n-1) \times (n-1)$ matrix A_{11} .

Clearly, f is an alternating n -linear form with $f(I)=1$,
so it must be the determinant function. \square

Corollary 5.10: Let A be a matrix whose j^{th} column is e_j .

Then $\det A = (-1)^{i+j} \det A_{ij}$, where A_{ij} is the $(n-1) \times (n-1)$ matrix obtained by striking out the i^{th} row and j^{th} column of A (called the $(ij)^{\text{th}}$ minor of A).

Proof: Exercise.

Theorem 5.11: Let A be an $n \times n$ matrix, and $1 \leq j \leq n$.

Then $\det A = \sum_{i=1}^n (-1)^{i+j} a_{ij} (\det A_{ij})$. (The "Laplace expansion".)

Proof: To simplify notation, take $j=1$.

Write $a_1 = a_{11}e_1 + \dots + a_{1n}e_n$, where $A = (a_1, \dots, a_n)$.

$$\begin{aligned} \text{By multilinearity, } \det A &= f(a_1, \dots, a_n) \\ &= f(a_{11}e_1 + \dots + a_{1n}e_n, a_2, \dots, a_n) \\ &= a_{11}f(e_1, a_2, \dots, a_n) + \dots + a_{1n}f(e_n, a_2, \dots, a_n) \end{aligned}$$

Now apply Corollary 5.10. \square

Application: Systems of equations.

Consider a system $Ax = u$, A is invertible and $x = \sum_{j=1}^n x_j e_j$,
 $A = (a_1, \dots, a_n) = (Ae_1, \dots, Ae_n)$.

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Thus, our system can be written as $\sum_{j=1}^n x_j a_j = u$.

Let $A_k = (a_1, \dots, a_{k-1}, u, a_{k+1}, \dots, a_n)$ for each k .

$$= (a_1, \dots, a_{k-1}, \sum_{j=1}^n x_j a_j, a_{k+1}, \dots, a_n).$$

$$\begin{aligned} \text{By multilinearity, } \det A_k &= \sum_{j=1}^n x_j \det(a_1, \dots, a_{k-1}, a_j, a_{k+1}, \dots, a_n) \\ &= x_k \det(a_1, \dots, a_{k-1}, a_k, a_{k+1}, \dots, a_n) \\ &= x_k \det A. \end{aligned}$$

$$\text{Since } A \text{ is invertible, } \det A \neq 0 \Rightarrow x_k = \frac{\det A_k}{\det A}.$$

Now apply the Laplace Expansion (Thm 5.11) to A_k :

$$\det A_k = \sum_{i=1}^n (-1)^{i+k} \det A_{ik} u_i$$

$$\Rightarrow x_k = \sum_{i=1}^n (-1)^{i+k} \frac{\det A_{ik}}{\det A} u_i. \text{ This is "Cramer's rule".}$$

Theorem 5.12: If A is invertible, then its inverse A^{-1} has the form $(A^{-1})_{ki} = (-1)^{i+k} \frac{\det A_{ik}}{\det A}$.

Proof: Consider a system $Ax = u \Rightarrow x = A^{-1}u$. By Cramer's rule:

$$\sum_{i=1}^n (-1)^{i+k} \frac{\det A_{ik}}{\det A} u_i = x_k = (A^{-1}u)_k = \sum_{i=1}^n (A^{-1})_{ki} u_i.$$

□

Note: This formula isn't "practical" for computing inverses.

Def: The trace of an $n \times n$ matrix is $\text{tr } A = \sum_{i=1}^n a_{ii}$.

Theorem 5.13:

$$(a) \text{ Trace is linear: } \text{tr}(kA) = k(\text{tr } A)$$

$$\text{tr}(A+B) = \text{tr } A + \text{tr } B.$$

$$(b) \text{ Trace is "commutative": } \text{tr}(AB) = \text{tr}(BA)$$

Proof: (a) is obvious.

$$\text{For (b): } (AB)_{ii} = \sum_{k=1}^n a_{ik} b_{ki} \quad \text{and} \quad (BA)_{ii} = \sum_{k=1}^n b_{ik} a_{ki}$$

$$\text{Thus, } \text{tr}(AB) = \sum_{i,k} a_{ik} b_{ki} = \sum_{i,k} b_{ik} a_{ki} = \text{tr}(BA). \quad \square$$

Theorem 5.14: Similar matrices have the same determinant and trace.

Proof: Suppose that $A = SBS^{-1}$.

$$\begin{aligned} \det A &= \det(SBS^{-1}) = (\det S)(\det B)(\det S^{-1}) = (\det B)(\det S)(\det S^{-1}) \\ &= (\det B) \det(SS^{-1}) = \det B. \quad \checkmark \end{aligned}$$

$$\text{tr } A = \text{tr}(SBS^{-1}) = \text{tr}(S^{-1}S B) = \text{tr } B. \quad \checkmark$$

Since similar matrices represent the same linear map but with a different choice of basis, it is well-founded to speak of the determinant and trace as functions of linear maps, not just matrices.