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Result: They are both 1. Note that the matrices are transposes.

Determinant Definition

Recall that the determinant is a function:

$$\det: \mathbb{R}^{n \times n} \to \mathbb{R}$$

so that

- 1. det(I) = 1.
- 2. det is antisymmetric on the rows of A.
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These properties can be applied to the elementary row operations:

Old Matrix	Operation	New Matrix	Det Prop
A	Switch two rows of A	В	B = - A
Α	Scale one row of A by k	В	B = k A
Α	Add/replace rows of A	В	B = A

Theorem

Let A be an $n \times n$ matrix then

$$\det(A) = \prod_{i=1}^n a_{ii}.$$

Theorem

Let A be an $n \times n$ matrix then

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- (1) tells us that if we row reduce A to a triangular form, we can easily compute the determinant.

One goal today is to strengthen (2.2) into an 'if and only if' statement.



Recall that each elementary row operation can be represented by left multiplication by some matrix E:

Operation Example E Determinant

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Operation Example
$$E$$
 Determinant Switch Rows $E = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ $|E| = -1$

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$$\begin{array}{lll} \text{Operation} & \text{Example E} & \text{Determinant} \\ \text{Switch Rows} & E = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} & |E| = -1 \\ \text{Scale row} & E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & k & 0 \\ 0 & 0 & 1 \end{bmatrix} & |E| = k \end{array}$$

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Lemma

Let A be a matrix and E be an elementary matrix. Then

$$det(EA) = det(E) det(A)$$
.



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Proof.

Suppose A is non-singular. Then there exist elementary row operation matrices E_i so that

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Therefore $det(A) \neq 0$.

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This has direct bearing on solving systems of equations Ax = b:

▶ If $|A| \neq 0$ then is Ax = b consistent? What types of solutions?

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 - Exercise: Prove this using $AA^{-1} = I$.

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Suppose there are two functions (\det_1) and (\det_2) which satisfy all the properties of the determinant. We need to show that $\det_1(A) = \det_2(A)$ for every matrix A.

Proof.

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- Case 2 A is non-singular. Then $A = E_1 \cdots E_n$ for some elementary matrices E_i . We have $\det_1(E_i) = \det_2(E_i)$. Therefore:

$$\begin{aligned} \det(A) &= & \det(E_1 \cdots E_n) \\ &= & \det(E_1) \cdots \det(E_n) \\ &= & \det_2(E_1) \cdots \det_2(E_n) \\ &= & \det_2(E_1 \cdots E_n) \\ &= & \det_2(A) \end{aligned}$$

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Now we know: if a computation satisfies all of the properties of the determinant, then it must be the determinant!

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Next: We will look at a computational interpretation for the determinant!