

Exercise

Compute the following determinant by using elimination to reduce to upper (or lower) triangular form:

$$\begin{vmatrix} 2 & 1 & 4 \\ 1 & 3 & 2 \\ 1 & -7 & 1 \end{vmatrix}$$

One solution:

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Therefore from the last case, we have:

$$(-1) \left(\frac{-1}{5} \right) \det(A) = (1)(1)(-1)$$

so $\det(A) = -5$.

Today: Direct Computational Methods

Recall from before: *The determinant is unique!*. Therefore, so long as we have a computational method which satisfies the properties:

1. Results in 1 when applied to the identity matrix.
2. Is antisymmetric and multilinear.

Then it is a way to compute the determinant!

Combinatorial Method

A **permutation** on n elements is a one-to-one and onto map:

$$\sigma : \{1, 2, 3, \dots, n\} \rightarrow \{1, 2, 3, \dots, n\}$$

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It turns out that you can create a *signature* function:

$$\text{sgn} : S_n \rightarrow \{-1, 1\} \text{ by } \text{sgn}(\sigma) = \begin{cases} 1 & \sigma \text{ uses even flips} \\ -1 & \sigma \text{ uses odd flips} \end{cases}$$

Combinatorial Method:

Let $A = [a_{ij}]$ be an $n \times n$ matrix. Then

$$\det(A) = \sum_{\sigma \in S_n} \left(\operatorname{sgn}(\sigma) \prod_{i=1}^n a_{i\sigma(i)} \right)$$

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Do not worry so much about this definition, though if you are interested in (combinatorics, graph theory, decision theory, data mining, abstract/modern algebra) you might start thinking about how to use it and how to prove it is a determinant.

Cofactor Expansion Part 1: Minor Matrices

Let $A = [a_{ij}]$ be an $n \times n$ matrix.

Definition

The M_{ij} **minor** of A is an $(n-1) \times (n-1)$ matrix created by eliminating the i th row and j th column.

Example:

$$A = \begin{bmatrix} 3 & -1 & 2 \\ 4 & 5 & 6 \\ 7 & 1 & 2 \end{bmatrix}$$

Write out M_{11} and M_{23} .

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The Cofactor Expansion Method

Theorem

Let $A = [a_{ij}]$. Then for any fixed $1 \leq i \leq n$

$$|A| = \sum_{j=1}^n a_{ij}(-1)^{i+j} |M_{ij}|.$$

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- ▶ This definition is via *row expansion*. It turns out that you can create a similar definition for column expansions (just transpose the i and j).
- ▶ The computation does not depend on which row (or column) you choose!
- ▶ This definition is *recursive* in that we need to compute 'smaller' cofactor expansions ($|M_{ij}|$) to get 'larger' ones ($|A|$).

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Exercises

Compute

$$\begin{vmatrix} 3 & -1 & 2 \\ 4 & 5 & 6 \\ 7 & 1 & 2 \end{vmatrix}$$

using

- 1) cofactor expansion across the second row.
- 2) cofactor expansion across the third column.

Compute

$$\begin{vmatrix} 3 & -1 & 2 & 300 \\ 4 & 5 & 6 & -10 \\ 0 & 0 & 0 & 1 \\ 7 & 1 & 2 & \pi \end{vmatrix}$$

Hint: What row/column do you think is best to use for the expansion?