

# Computation of Some Determinants

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## Abstract

The purpose of this article is to highlight how the characterizing properties of determinants allow us to compute the determinants. Though the determinants are defined using normalization, mutli-linearity and skew-symmetry, students are not exposed to the utility of this definition in the computation of the determinants. Typically, students use the Laplace expansion when evaluating a determinant. We give a few examples to show how the abstract definition coupled/combined with Laplace expansion makes the computations easier. One should also note that Laplace expansion is computationally very inefficient, as it involves (asymptotically)  $n!$  operations whereas row operations involve  $n^3$  operations.

Let  $\mathbb{F}$  be a field (or a commutative ring with 1). We state the definition, the effect of elementary row/column operations on the determinant of a matrix,  $\det(A) = \det(A^T)$ , and the determinant of an upper/lower triangular matrix is the product of its diagonal entries.

**Example 1.** We compute the determinant of the following matrix.  $\begin{pmatrix} 2 & 4 & 5 & 0 \\ 1 & 0 & 1 & 0 \\ -2 & 3 & 1 & 2 \\ 6 & -5 & 4 & -3 \end{pmatrix}$ . By

carrying out the elementary row operations  $R_1 - 2R_2$ ,  $R_3 + 2R_2$ ,  $R_4 - 6R_2$ ,  $R_1 \leftrightarrow R_2$ ,  $R_2 - R_3$ ,  $R_3 - 3R_2$ ,  $R_3 + R_4$  and  $R_4 + 2R_3$ , we arrive at upper triangular matrix

$\begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & -2 \\ 0 & 0 & 1 & -5 \\ 0 & 0 & 0 & -23 \end{pmatrix}$  whose determinant is  $-23$ . Since we employed  $R_1 \leftrightarrow R_2$ , the determinant of the original matrix is 23.

**Example 2.** Consider the matrix  $A := \begin{pmatrix} 0 & 1 & 2 & 3 \\ 1 & 1 & 1 & 1 \\ -2 & -2 & 3 & 3 \\ 1 & -2 & -2 & -3 \end{pmatrix}$ . By carrying out the elementary

row operations  $R_1 \leftrightarrow R_2$ ,  $R_3 + 2R_1$ ,  $R_4 - R_1$ ,  $R_4 + 3R_2$ ,  $\frac{1}{5}R_3$ ,  $R_4 - 3R_3$ , we arrive the matrix

$B := \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 2 \end{pmatrix}$ . Hence we see that  $\det A = (-1) \times 5 \times \det B = -10$ .

**Example 3.** Consider the matrix  $A = \begin{vmatrix} 1 & 3 & 1 & 1 \\ 2 & 1 & 5 & 2 \\ 1 & -1 & 2 & 3 \\ 4 & 1 & -3 & 7 \end{vmatrix}$ . Carrying out the elementary row

operation  $R_2 - 2R_1$ ,  $R_3 - R_1$ , and  $R_4 - 4R_1$ , we arrive at the matrix  $B = \begin{pmatrix} 1 & 3 & 1 & 1 \\ 0 & -5 & 3 & 0 \\ 0 & -4 & 1 & 2 \\ 0 & -11 & -7 & 3 \end{pmatrix}$ .

Hence  $\det A = \det B = \det \begin{pmatrix} -5 & 3 & 0 \\ -4 & -1 & 2 \\ -11 & -7 & 3 \end{pmatrix}$ . Carrying out the elementary row operation

$2R_3 - 3R_2$ , we arrive at  $C = \begin{pmatrix} -5 & 3 & 0 \\ -4 & -1 & 2 \\ -10 & -17 & 0 \end{pmatrix}$ . Expanding by the 23-rd element, we obtain

$$\det C = (-1)^{2+3}[(-5) \times (-17) - (-10) \times 3] = -115.$$

**Example 4.** Consider  $D_n := \begin{vmatrix} 2 & 1 & 0 & \dots & 0 & 0 & 0 \\ 1 & 2 & 1 & \dots & 0 & 0 & 0 \\ \vdots & & \ddots & & \vdots & & \\ 0 & 0 & 0 & \dots & 1 & 2 & 1 \\ 0 & 0 & 0 & \dots & 0 & 1 & 2 \end{vmatrix}$ . We observe that  $D_1 = 2$ ,  $D_2 = \begin{vmatrix} 2 & 1 \\ 1 & 2 \end{vmatrix} =$

$3$  and  $D_3 = \begin{vmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{vmatrix} = 4$ . We claim that  $D_n = n + 1$ .

Expanding  $D_n$  by the first row, we get

$$D_n = 2D_{n-1} - \begin{vmatrix} 1 & 1 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & \dots & 0 & 0 & 0 \\ 0 & 1 & 2 & 1 & \dots & 0 & 0 & 0 \\ & & \vdots & \ddots & \vdots & & & \\ 0 & 0 & 0 & 0 & \dots & 1 & 2 & 1 \\ 0 & 0 & 0 & 0 & \dots & 0 & 1 & 2 \end{vmatrix}.$$

We expand the determinant on the RHS by the *first column* to get the recurrence relation  $D_n = 2D_{n-1} - D_{n-2}$ . Using induction, we show that  $D_n = n + 1$ .

**Example 5.** Evaluate  $D_n := \begin{vmatrix} 1 & 2 & 3 & \dots & 0 \\ -1 & 0 & 3 & \dots & n \\ -1 & -2 & 0 & \dots & n \\ \vdots & & & & \\ -1 & -2 & -3 & \dots & 0 \end{vmatrix}$ . Hint: Have a look at  $D_3$  and  $D_4$ , if

nothing strikes you.

**Example 6 (Vandermonde Determinant).** Let  $a_1, \dots, a_n$  be distinct nonzero elements of  $\mathbb{R}$ .

Consider the column vectors  $(a_1^k, a_2^k, \dots, a_n^k)^T$  for  $0 \leq k \leq n-1$ . Consider the determinant

$$D(a_1, \dots, a_n) := \begin{vmatrix} 1 & 1 & \dots & 1 \\ a_1 & a_2 & \dots & a_n \\ \vdots & \vdots & \ddots & \vdots \\ a_1^{n-1} & a_2^{n-1} & \dots & a_n^{n-1} \end{vmatrix}.$$

The determinant  $D$  is known as Vandermonde determinant and it is ubiquitous in diverse areas of mathematics. The value of the determinant is  $\prod_{j>i}(a_j - a_i)$ . You may start with  $n = 2$  and  $n = 3$  to get acquainted with this determinant. We show by induction that  $D(a_1, \dots, a_n) = \prod_{j>i}(a_j - a_i)$ .

We have  $D_2 = \begin{vmatrix} 1 & 1 \\ a & b \end{vmatrix} = b - a$ . Consider  $D_3 = \begin{vmatrix} 1 & 1 & 1 \\ a & b & c \\ a^2 & b^2 & c^2 \end{vmatrix}$ . Carry out the  $R_3 - aR_2$

to obtain  $\begin{vmatrix} 1 & 1 & 1 \\ a & b & c \\ 0 & (b-a)b & (c-a)c \end{vmatrix}$ . Now we do  $R_2 - aR_1$  to arrive at  $\begin{pmatrix} 1 & 1 & 1 \\ 0 & b-a & c-a \\ 0 & (b-a)b & (c-a)c \end{pmatrix}$ . We expand by the first column to get

$$D_3 = 1 \cdot \begin{vmatrix} b-a & c-a \\ (b-a)b & (c-a)c \end{vmatrix} = (b-a)(c-a) \begin{vmatrix} 1 & 1 \\ b & c \end{vmatrix} = (b-a)(c-a)(c-b).$$

Now to evaluate  $D_n$ , consider the following column operations in order:

$$R_n - a_1 R_{n-1} \rightarrow R_{n-1} - a_1 R_{n-2} \rightarrow \dots \rightarrow R_2 - a_1 R_1.$$

The resulting determinant is

$$\begin{aligned} D(a_1, a_2, \dots, a_n) &= \begin{vmatrix} 1 & 1 & 1 & \dots & 1 \\ 0 & (a_2 - a_1) & (a_3 - a_1) & \dots & (a_n - a_1) \\ 0 & (a_2 - a_1)a_2 & (a_3 - a_1)a_3 & \dots & (a_n - a_1)a_n \\ & & \vdots & & \\ 0 & (a_2 - a_1)a_2^{n-2} & (a_3 - a_1)a_3^{n-2} & \dots & (a_n - a_1)a_n^{n-2} \end{vmatrix} \\ &= (a_2 - a_1)(a_3 - a_1) \dots (a_n - a_1) \begin{pmatrix} 1 & 1 & \dots & 1 \\ a_2 & a_3 & \dots & a_n \\ & \vdots & & \\ a_2^{n-2} & a_3^{n-2} & \dots & a_n^{n-2} \end{pmatrix} \\ &= \prod_{j>1} (a_j - a_1) D(a_2, \dots, a_n) \\ &= \prod_{j>1} (a_j - a_1) \prod_{r>k \geq 2} (a_r - a_k) \\ &= \prod_{j>i} (a_j - a_i). \end{aligned}$$

We expanded by the first column and used the multi-linearity of the determinant in its column arguments and arrived at the second equality in the displayed array of equations.

**Ex. 7.** If  $\alpha_1, \dots, \alpha_n$  are distinct complex numbers then for any complex numbers  $b_1, \dots, b_n$  there exists a polynomial  $P$  of degree at most  $n - 1$  uniquely determined by the conditions  $P(\alpha_j) = b_j$ ,  $1 \leq j \leq n$ . *Hint:* If  $P(X) := c_0 + c_1X + \dots + c_{n-1}X^{n-1}$  is the required polynomial, then  $(c_0, c_1, \dots, c_{n-1})^T$  is the unique solution of a linear system  $D(\alpha_1, \dots, \alpha_n)(x_1, \dots, x_n)^T = (b_1, \dots, b_n)^T$ . Here  $D(\alpha_1, \dots, \alpha_n)$  denotes the vandermonde matrix (and not the determinant).

**Example 8.** We show that

$$\begin{vmatrix} 1 + a_1 & a_2 & \dots & a_n \\ a_1 & 1 + a_2 & \dots & a_n \\ \vdots & \vdots & \ddots & \vdots \\ a_1 & a_2 & \dots & 1 + a_n \end{vmatrix} = 1 + a_1 + a_2 + \dots + a_n.$$

The given determinant is  $\det(e_1 + a_1v, e_2 + a_2v, \dots, e_n + a_nv)$ , where  $v = e_1 + \dots + e_n$ . Let us look at  $n = 3$ .

$$\begin{aligned} \det(e_1 + a_1v, e_2 + a_2v, e_3 + a_3v) &= \det(e_1, e_2 + a_2v, e_3 + a_3v) + \det(a_1v, e_2 + a_2v, e_3 + a_3v) \\ &= [\det(e_1, e_2, e_3) + \det(e_1, a_2v, e_3) + \det(e_1, e_2, a_3v)] \\ &\quad + \det(a_1v, e_2, e_3) \\ &= (1 + a_2 + a_3) + a_1 = 1 + a_1 + a_2 + a_3. \end{aligned}$$

Now proceed by induction.

**Example 9.** The matrix of this example is from differential geometry and the computation of its determinant is required while computing the volume element of a (hyper) surface given as a graph. The matrix is

$$\begin{pmatrix} 1 + x_1^2 & x_1x_2 & \dots & x_1x_n \\ x_2x_1 & 1 + x_2^2 & \dots & x_2x_n \\ \vdots & \vdots & \ddots & \vdots \\ x_nx_1 & x_nx_2 & \dots & 1 + x_n^2 \end{pmatrix}.$$

The trick here is, as in the last case, to realize the  $i$ th column vector  $C_i$ , which is the vector  $e_i + x_iv$ , where

$$v = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \sum_i x_i e_i \in \mathbb{R}^n.$$

Again, it may be worthwhile to start looking at the cases  $n = 2, 3$  and then complete the solution by induction.

**Example 10.** Let  $A \in M(n, \mathbb{F})$ . Consider the associated linear map  $T_A: \mathbb{F}^n \rightarrow \mathbb{F}^n$  given by  $T_Ax = Ax$  where  $x \in \mathbb{F}^n$  is considered as a column vector. Let  $\{v_1, \dots, v_n\}$  be an eigenbasis of  $\mathbb{F}^n$ . Assume that  $T_Av_j = Av_j = \lambda_jv_j$ . Then  $\det A = \lambda_1 \cdots \lambda_n$ .

Note that by the definition of the determinant,  $\det A$  is the unique element of  $\mathbb{F}$  such that for any  $u_1, \dots, u_n \in \mathbb{F}^n$ , we have

$$\det(Au_1, \dots, Au_n) = \det(A) \det(u_1, \dots, u_n).$$

Apply this with  $u_j = v_j$  to get the result.