

In simple terms, a matrix is a rectangular array of numbers, arranged in rows and columns. Each number in a matrix is called an element. For example, the matrix

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$$

has two rows and three columns. The element in the first row, second column is 2 (denoted as a_{01}), and the element in the second row, third column is 6 (denoted as a_{12}):

$$A = \begin{pmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \end{pmatrix}$$

We can use matrices to express linear systems of equations in condensed form. Consider the following example:

$$\begin{vmatrix} y_0 = a_{00}x_0 + a_{01}x_1 + a_{02}x_2 \\ y_1 = a_{21}x_0 + a_{22}x_1 + a_{23}x_2 \end{vmatrix}$$
 (1)

This exemplary system can be represented in matrix form as $y = A \cdot x$:

$$\begin{pmatrix} y_0 \\ y_1 \end{pmatrix} = \begin{pmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \end{pmatrix} \cdot \begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix} \tag{2}$$

In this notation, it holds that

- A ∈ R^{3×3} is the coefficient matrix, containing the coefficients a_{ij} of the variables
 x ∈ R³ is the column vector of variables (x₀, x₁, x₂)^T
 y ∈ R² is the column vector of constants (y₀, y₁)^T

Note: To refer to row i of A, we write $a_{i,:}$ or A(i,:) – and $a_{:,j}$ or A(:,j) for column j. The use of ":" reminds us of the slice operator that we would use in Python for arrays.

Besides collecting the numbers in matrices and vectors, the notation also implies a calculation scheme (dt.: Rechenschema): The multiplication of a matrix A with a column vector x is defined as:

- (Step 1): Row-wise multiplication of the elements of A with the elements of x
- (Step 2): Subsequent summation of the resulting products

We can express this also very compactly using the sum notation:

$$y_i = A(i,:) \cdot x = \sum_{j=0}^{N} a_{ij} \cdot x_j , \quad i \in \{0,1\}$$

If we apply this scheme, we indeed recover the original system of equations (1):

$$y_0 = A_{0,:} \cdot x = a_{00}x_0 + a_{01}x_1 + a_{02}x_2$$

$$y_1 = A_{1,:} \cdot x = a_{10}x_0 + a_{11}x_1 + a_{12}x_2$$

Notes:

- In the above formulas, we employ zero-based indices, meaning that we start counting the indices at zero: $x = (x_0, x_1, x_3)^T$. This choice corresponds to the convention used in Python and other programming languages, where vectors and matrices also use zero-based indices. Conversely, in mathematics and certain other coding languages such as R, it is common to use indices that start with the value 1: $x = (x_1, x_2, x_3)^T$. This is always a source of confusion! Here we use the computer scientist notation.
- With the operator T, we can transpose (flip) of a vector or matrix. For example:

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}^T = \begin{pmatrix} x_1 & x_2 & x_3 \end{pmatrix}$$
$$\begin{pmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \end{pmatrix}^T = \begin{pmatrix} a_{00} & a_{10} \\ a_{01} & a_{11} \\ a_{02} & a_{12} \end{pmatrix}$$

• We used matrix notation in equation (2) to multiply a matrix with a column vector. We can generalize this scheme to multiply two matrices A and B.

$$A \cdot B = \begin{pmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \end{pmatrix} \cdot \begin{pmatrix} b_{00} & b_{01} \\ b_{10} & b_{11} \\ b_{20} & b_{21} \end{pmatrix} = \begin{pmatrix} c_{00} & c_{01} \\ c_{10} & c_{11} \end{pmatrix} = C$$

$$C(i,j) = A(i,:) \cdot B(:,j)$$

As an example (see colored elements), by using the above scheme of multiplying and summing row A(0,:) with column B(:,1), we compute matrix element $C(0,1) = c_{01}$. Note that the matrices must have matching sizes for this to work!

There is a lot more to say about matrices. The mathematical discipline that studies the properties of matrices and related topics is called *linear algebra*.

Further reading

- A crash course on linear algebra for machine learners and Pythonists. Link
- Essential concepts of linear algebra, by Rob Taylor. Part 1, Part 2
- Videos on linear algebra by 3blue1brown, wonderfully produced. Link