

# Real matrices and their properties

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"Introduction to linear algebra", 1/6

08/20

# Rotations 1/3

Problem: what's the formula of the rotation of angle  $t$ ?

## Rotations 2/3

The points in the plane  $\mathbb{R}^2$  can be represented as vectors  $\begin{pmatrix} x \\ y \end{pmatrix}$ . The  $2 \times 2$  matrices “act” on such vectors, as follows:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax + by \\ cx + dy \end{pmatrix}$$

Many simple transformations (symmetries, projections..) can be written in this form. What about the rotation of angle  $t$ ?

## Rotations 3/3

A quick picture shows that we must have:

$$\begin{pmatrix} * & * \\ * & * \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}$$

Also, by paying attention to positives and negatives:

$$\begin{pmatrix} * & * \\ * & * \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -\sin t \\ \cos t \end{pmatrix}$$

Thus, the matrix of our rotation can only be:

$$R_t = \begin{pmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{pmatrix}$$

By "linear algebra", this is the correct answer.

## Linear maps 1/4

Theorem. The maps  $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  which are linear, in the sense that they map lines through 0 to lines through 0, are:

$$f \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax + by \\ cx + dy \end{pmatrix}$$

Remark. If we make the multiplication convention

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax + by \\ cx + dy \end{pmatrix}$$

the theorem says  $f(v) = Av$ , with  $A$  being a  $2 \times 2$  matrix.

## Linear maps 2/4

Examples. The identity and null maps are given by:

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix} \quad , \quad \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

The projections on the horizontal and vertical axes are given by:

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \end{pmatrix} \quad , \quad \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ y \end{pmatrix}$$

The symmetry with respect to the  $x = y$  diagonal is given by:

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ x \end{pmatrix}$$

We have as well the rotation of angle  $t \in \mathbb{R}$ , studied before.

## Linear maps 3/4

Theorem. The maps  $f : \mathbb{R}^N \rightarrow \mathbb{R}^N$  which are linear, in the sense that they map lines through 0 to lines through 0, are:

$$f \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} a_{11}x_1 + \dots + a_{1N}x_N \\ \vdots \\ a_{N1}x_1 + \dots + a_{NN}x_N \end{pmatrix}$$

Remark. With the matrix multiplication convention

$$\begin{pmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{N1} & \dots & a_{NN} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} a_{11}x_1 + \dots + a_{1N}x_N \\ \vdots \\ a_{N1}x_1 + \dots + a_{NN}x_N \end{pmatrix}$$

the theorem says  $f(v) = Av$ , with  $A$  being a  $N \times N$  matrix.

## Linear maps 4/4

Example. Consider the all-1 matrix. This acts as follows:

$$\begin{pmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} x_1 + \dots + x_N \\ \vdots \\ x_1 + \dots + x_N \end{pmatrix}$$

But this formula can be written as follows:

$$\frac{1}{N} \begin{pmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = \frac{x_1 + \dots + x_N}{N} \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

And this latter map is the projection on the all-1 vector.



## Theory 1/4

Definition. We can multiply  $M \times N$  matrices with  $N \times K$  matrices,

$$\begin{pmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{M1} & \dots & a_{MN} \end{pmatrix} \begin{pmatrix} b_{11} & \dots & b_{1K} \\ \vdots & & \vdots \\ b_{N1} & \dots & b_{NK} \end{pmatrix}$$

the product being a  $M \times K$  matrix, given by the formula

$$\begin{pmatrix} a_{11}b_{11} + \dots + a_{1N}b_{N1} & \dots & a_{11}b_{1K} + \dots + a_{1N}b_{NK} \\ \vdots & & \vdots \\ \vdots & & \vdots \\ a_{M1}b_{11} + \dots + a_{MN}b_{N1} & \dots & a_{M1}b_{1K} + \dots + a_{MN}b_{NK} \end{pmatrix}$$

obtained via the rule “multiply rows by columns”.

## Theory 2/4

Better definition. The matrix multiplication is given by

$$(AB)_{ij} = \sum_k A_{ik} B_{kj}$$

with  $A_{ij}$  being the entry on the  $i$ -th row and  $j$ -th column.

Theorem. The linear maps  $f : \mathbb{R}^N \rightarrow \mathbb{R}^M$  are those of the form

$$f_A(v) = Av$$

with  $A$  being a  $N \times M$  matrix.

Remark. Size check  $(N \times 1) = (N \times M)(M \times 1)$ , ok.

## Theory 3/4

Theorem. With the above convention  $f_A(v) = Av$ , we have

$$f_A f_B = f_{AB}$$

"the product of matrices corresponds to the composition of maps".

Theorem. A linear map  $f : \mathbb{R}^N \rightarrow \mathbb{R}^N$  is invertible when the matrix  $A \in M_N(\mathbb{R})$  which produces it is invertible, and we have:

$$(f_A)^{-1} = f_{A^{-1}}$$

## Theory 4/4

Theorem. The inverses of the  $2 \times 2$  matrices are given by:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Proof. When  $ad = bc$  the columns are proportional, so the matrix cannot be invertible. When  $ad - bc \neq 0$ , let us solve:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} * & * \\ * & * \end{pmatrix}$$

We must solve the following equations:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} * & * \\ * & * \end{pmatrix} = \begin{pmatrix} ad - bc & 0 \\ 0 & ad - bc \end{pmatrix}$$

But this leads to the formula in the statement.

## Eigenvectors 1/4

Definition. Let  $A \in M_N(\mathbb{R})$  be a square matrix, and assume that  $A$  multiplies by  $\lambda \in \mathbb{R}$  in the direction of a vector  $v \in \mathbb{R}^N$ :

$$Av = \lambda v$$

In this case, we say that:

- (1)  $v \in \mathbb{R}^N$  is an eigenvector of  $A$ .
- (2)  $\lambda \in \mathbb{R}$  is the corresponding eigenvalue.

## Eigenvectors 2/4

Examples. The identity has all vectors as eigenvectors, with  $\lambda = 1$ :

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix}$$

The same goes for the null matrix, with  $\lambda = 0$  this time:

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

For the projection on the horizontal axis,  $P \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ 0 \end{pmatrix}$ , we have:

$$Pv = \lambda v \iff v = \begin{pmatrix} 0 \\ y \end{pmatrix}, \lambda = 0 \quad \text{or} \quad v = \begin{pmatrix} x \\ 0 \end{pmatrix}, \lambda = 1$$

A similar result holds for the projection on the vertical axis.

## Eigenvectors 3/4

More examples. For the symmetry  $S \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ x \end{pmatrix}$ , we have:

$$Sv = \lambda v \iff v = \begin{pmatrix} x \\ x \end{pmatrix}, \lambda = 1 \quad \text{or} \quad v = \begin{pmatrix} x \\ -x \end{pmatrix}, \lambda = -1$$

For the transformation  $T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix}$  we have:

$$Tv = \lambda v \iff v = \begin{pmatrix} x \\ 0 \end{pmatrix}, \lambda = 0$$

For the rotation of angle  $t \neq 0$ , we must have  $v = 0, \lambda = 0$ .

## Eigenvectors 4/4

Definition. We say that a matrix  $A \in M_N(\mathbb{R})$  is diagonalizable if it has  $N$  eigenvectors  $v_1, \dots, v_N$  which form a basis of  $\mathbb{R}^N$ .

Remark. When  $A$  is diagonalizable, in that basis we can write:

$$A = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_N \end{pmatrix}$$

This means that we have  $A = PDP^{-1}$ , with  $D$  diagonal.

Problems. Which matrices are diagonalizable? And, how to diagonalize them?