

## [MATH-06] The product of a matrix and a vector

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### The product of a matrix and a vector

The product of a matrix and a vector is defined in such a way that the product of two vectors is just a particular case, that in which the first factor is a 1-row matrix. If we write

$$\begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = x_1 y_1 + \cdots + x_n y_n,$$

the scalar product becomes

$$\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^\top \mathbf{y}.$$

Thus, the product of a row by a column is a number, that can be considered as a  $(1,1)$ -matrix. Next, we extend this definition to the product of an  $(n,m)$ -matrix  $\mathbf{A}$  and an  $m$ -vector  $\mathbf{x}$ . To do this, we take  $\mathbf{A}$  as a pack of row vectors, multiplying each row by  $\mathbf{x}$  and placing the numbers that result from these products as the coordinates of a vector, which is the product  $\mathbf{Ax}$ . In mathematical symbols,

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1m} \\ a_{21} & a_{22} & \cdots & a_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nm} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} = \begin{bmatrix} \sum_j a_{1j} x_j \\ \sum_j a_{2j} x_j \\ \vdots \\ \sum_j a_{nj} x_j \end{bmatrix}.$$

If  $\mathbf{a}_1, \dots, \mathbf{a}_n$  are vectors corresponding to the rows of  $\mathbf{A}$ , this can be written as

$$\begin{bmatrix} \mathbf{a}_1^\top \\ \mathbf{a}_2^\top \\ \vdots \\ \mathbf{a}_n^\top \end{bmatrix} \mathbf{x} = \begin{bmatrix} \mathbf{a}_1^\top \mathbf{x} \\ \mathbf{a}_2^\top \mathbf{x} \\ \vdots \\ \mathbf{a}_n^\top \mathbf{x} \end{bmatrix}.$$

### Matrices as linear operators

The definition of the product of a matrix and a vector allows an interesting interpretation of a matrix as a **linear operator**. Let us fix  $\mathbf{A}$  and define an operator (functions are called operators in this context)

$$\begin{aligned} T : E_m &\longrightarrow E_n \\ \mathbf{x} &\longmapsto \mathbf{Ax}. \end{aligned}$$

Such operator would be linear, i.e. it would transform a linear combination into a linear combination,

$$T(\alpha_1 \mathbf{u}_1 + \alpha_2 \mathbf{u}_2) = \alpha_1 T\mathbf{u}_1 + \alpha_2 T\mathbf{u}_2.$$

Conversely, any operator with this property can be associated to a matrix. It is easy to check that the columns of the matrix associated to a linear operator coincide with the images of the vectors of the canonical basis under the operator. A one-to-one linear operator  $T : E_n \rightarrow E_n$  is called a **linear transformation**.

**Example 1.** The matrix

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & -1 \\ 1 & 0 & 1 \\ 0 & 1 & -2 \end{bmatrix}$$

defines the operator  $T : E_3 \rightarrow E_4$  as

$$T \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 + x_2 \\ -x_3 \\ x_1 + x_3 \\ x_2 - 2x_3 \end{bmatrix}. \quad \square$$

### The null space

An interesting definition related to linear operators is the **null space** (also called kernel). The null space of  $\mathbf{A}$  (equivalently, of  $T$ ), that we denote by  $\mathcal{N}(\mathbf{A})$ , is the subspace of all vectors  $\mathbf{x}$  of  $E_m$  such that  $\mathbf{A}\mathbf{x} = \mathbf{0}$ . The vector  $\mathbf{0}$  always belongs to the null space. We may also say that  $\mathbf{x} \in \mathcal{N}(\mathbf{A})$  when  $\mathbf{x}$  is orthogonal to the rows of  $\mathbf{A}$ .

The subspace of all vectors  $\mathbf{A}\mathbf{x}$ , with  $\mathbf{x} \in E_n$ , is called the **range** of  $\mathbf{A}$ , denoted by  $\mathcal{R}(\mathbf{A})$ . Since the vectors of  $E_n$  are the linear combinations of the canonical basis, those of  $\mathcal{R}(\mathbf{A})$  are the linear combinations of the columns of  $\mathbf{A}$ . There are as many linearly independent column vectors as indicated by the rank, which thus coincides with the dimension of  $\mathcal{R}(\mathbf{A})$ . An interesting formula is (mind that, here,  $m$  is the number of columns)

$$\dim \mathcal{N}(\mathbf{A}) + \text{rank } \mathbf{A} = m.$$

In particular, when  $\mathbf{A}$  is a square matrix, the existence of vectors  $\mathbf{x} \neq \mathbf{0}$  such that  $\mathbf{A}\mathbf{x} = \mathbf{0}$  is equivalent to  $\det(\mathbf{A}) = 0$ .

**Example 2** The linear subspace  $S = \{\mathbf{x} \in E_3 : x_1 + x_2 = x_1 - x_3 = 0\}$  is the null space of the matrix

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & -1 \end{bmatrix}.$$

As the rank of this matrix is 2, the null space has dimension 1. This has been previously seen (lecture 3) by finding a basis of  $S$ .  $\square$

### Homework

**A.** Find the rank and a basis of the null space of

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \end{bmatrix}.$$