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# Matrices characterize linear transformations

## Matrices as functions

In the context of matrix-vector multiplication, we can think of a matrix as a function between vectors spaces. That is, if we hold a matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  as fixed, this matrix maps vectors in  $\mathbb{R}^n$  to vectors in  $\mathbb{R}^m$ . Making this more explicit, we can define a function  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  as:

$$T(\mathbf{x}) := \mathbf{Ax}$$

where  $T$  uses the matrix  $\mathbf{A}$  to performing the mapping.

## Common matrix-defined functions

### The identity matrix defines the identity function

Recall an identity function  $f$  for a set  $S$  is the function  $f(x) := x$  for all  $x \in S$ . In the context of a function  $T$  over a vector space  $\mathbb{R}^n$ , the identity function  $T(\mathbf{x}) := \mathbf{x}$  for all  $\mathbf{x} \in \mathbb{R}^n$  is defined using the **identity matrix** for  $\mathbb{R}^n$ . The identity matrix for  $\mathbb{R}^n$ , denoted  $\mathbf{I}_n$  (or simply  $\mathbf{I}$  if the dimensionality is implied by the context), is a square matrix of all zeros except for ones along the diagonal (Definition 1). For example, the identity matrix for  $\mathbb{R}^3$  is defined as

$$\mathbf{I}_3 := \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

It can easily be shown that applying matrix-vector multiplication using an identity matrix  $\mathbf{I}_n$  with any vector  $\mathbf{x} \in \mathbb{R}^n$  will result in the same vector  $\mathbf{x}$ . Thus, a function  $T(\mathbf{x}) := \mathbf{Ix}$  is the identity function for  $\mathbb{R}^n$ .

**Definition 1** Each real-valued, Euclidean vector space  $\mathbb{R}^n$  is associated with an **identity matrix**, denoted  $\mathbf{I}_{n \times n}$  (or simply  $\mathbf{I}$  if the dimensionality is implied by the context), which is a square matrix consisting of zeros in the off-diagonal entries and ones along the diagonal.

### The zero matrix defines the zero function

Recall a zero-function  $f$  for a set  $S$  is the function  $f(x) := 0$  for all  $x \in S$ . In the context of a function  $T$  over a vector space  $\mathbb{R}^n$ , the zero function  $T(\mathbf{x}) := \mathbf{0}$  for all  $\mathbf{x} \in \mathbb{R}^n$  is

defined using the **zero matrix** for  $\mathbb{R}^n$ . The zero matrix for  $\mathbb{R}^n$ , denoted  $\mathbf{0}_n$  is a square matrix of all zeros (Definition 2). For example, the zero matrix for  $\mathbb{R}^3$  is defined as

$$\mathbf{0}_3 := \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

. It can easily be shown that applying matrix-vector multiplication as defined in Definition ?? using an identity matrix  $\mathbf{0}_n$  with any vector  $\mathbf{x} \in \mathbb{R}^n$  will result in the zero vector  $\mathbf{0}$ . Thus, a function  $T(\mathbf{x}) := \mathbf{0}_n \mathbf{x}$  is the zero function for  $\mathbb{R}^n$ .

**Definition 2** *Each real-valued, Euclidean vector space  $\mathbb{R}^n$  is associated with a **zero matrix**, denoted  $\mathbf{0}_{n \times n}$ , which is a square matrix consisting of all zeros.*

## A matrix characterizes a linear transformation between coordinate vector spaces

In fact, any function  $T$  in the form of  $T(\mathbf{x}) := \mathbf{A}\mathbf{x}$  is a linear function (Theorem 1). Thus, a matrix in  $\mathbb{R}^{m \times n}$  defines a linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ .

Moreover, given a linear transformation between coordinate vector spaces, the matrix that performs this linear transformation is unique. For a given linear transformation  $T$ , that matrix is called the **standard matrix** (Definition 3). Thus, each matrix uniquely defines a certain linear transformation between coordinate vector spaces and every linear transformation between coordinate vector spaces is defined by some unique matrix (Theorem 2).

**Theorem 1** *The function  $T(\mathbf{x}) := \mathbf{A}\mathbf{x}$  is linear.*

**Proof:**

We show that for all  $\mathbf{u}, \mathbf{v}$  in the domain of  $T$  and for all scalars  $c$ , the following conditions hold:

- a.)  $\mathbf{A}(\mathbf{u} + \mathbf{v}) = \mathbf{A}\mathbf{u} + \mathbf{A}\mathbf{v}$
- b.)  $\mathbf{A}(c\mathbf{u}) = c(\mathbf{A}\mathbf{u})$

a.)

$$\begin{aligned}\mathbf{A}(\mathbf{u} + \mathbf{v}) &= \mathbf{a}_{*,1}(u_1 + v_1) + \mathbf{a}_{*,2}(u_2 + v_2) + \cdots + \mathbf{a}_{*,n}(u_n + v_n) \\ &= \mathbf{a}_{*,1}u_1 + \mathbf{a}_{*,1}v_1 + \mathbf{a}_{*,2}u_2 + \mathbf{a}_{*,2}v_2 + \cdots + \mathbf{a}_{*,n}u_n + \mathbf{a}_{*,n}v_n \\ &= (\mathbf{a}_{*,1}u_1 + \mathbf{a}_{*,2}u_2 + \cdots + \mathbf{a}_{*,n}u_n) + (\mathbf{a}_{*,1}v_1 + \mathbf{a}_{*,2}v_2 + \cdots + \mathbf{a}_{*,n}v_n) \\ &= \mathbf{A}\mathbf{u} + \mathbf{A}\mathbf{v}\end{aligned}$$

b.)

$$\begin{aligned}\mathbf{A}(c\mathbf{u}) &= \mathbf{a}_{*,1}(cu_1) + \mathbf{a}_{*,2}(cu_2) + \cdots + \mathbf{a}_{*,n}(cu_n) \\ &= c\mathbf{a}_{*,1}(u_1) + c\mathbf{a}_{*,2}(u_2) + \cdots + c\mathbf{a}_{*,n}(u_n) \\ &= c(\mathbf{a}_{*,1}u_1 + \mathbf{a}_{*,2}u_2 + \cdots + \mathbf{a}_{*,n}u_n) \\ &= c(\mathbf{A}\mathbf{u})\end{aligned}$$

□

**Definition 3** Given a linear transformation  $T$  defined as

$$T(\mathbf{x}) := \mathbf{A}\mathbf{x}$$

where  $\mathbf{A} \in \mathbb{R}^{m \times n}$  and  $\mathbf{x} \in \mathbb{R}^n$ , the matrix  $\mathbf{A}$  is called the **standard matrix** of  $T$ .

**Theorem 2** Given a linear transformation  $T$ , this transformation is characterized by a unique standard matrix  $\mathbf{A}$ . Furthermore,  $\mathbf{A}$  is defined as follows:

$$\mathbf{A} := \left[ T(\mathbf{i}_{*,1}), T(\mathbf{i}_{*,2}), \dots, T(\mathbf{i}_{*,n}) \right]$$

where  $\mathbf{i}_{*,i}$  is the  $i$ th column of the identity matrix  $\mathbf{I}_n$ .

**Proof:**

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$$\begin{aligned}
\mathbf{x} &= \mathbf{I}\mathbf{x} \\
&= \mathbf{i}_{*,1}x_1 + \mathbf{i}_{*,2}x_2 + \cdots + \mathbf{i}_{*,n}x_n \\
\Rightarrow T(\mathbf{x}) &= T(\mathbf{i}_{*,1}x_1 + \mathbf{i}_{*,2}x_2 + \cdots + \mathbf{i}_{*,n}x_n) \\
&= T(\mathbf{i}_{*,1}x_1) + T(\mathbf{i}_{*,2}x_2) + \cdots + T(\mathbf{i}_{*,n}x_n) && T \text{ is linear} \\
&= \left[ T(\mathbf{i}_{*,1}), T(\mathbf{i}_{*,2}), \dots, T(\mathbf{i}_{*,n}) \right] \mathbf{x}
\end{aligned}$$

This implies that the linear transformation  $T$  operates on a vector  $\mathbf{x}$  through the matrix

$$A := \left[ T(\mathbf{i}_{*,1}), T(\mathbf{i}_{*,2}), \dots, T(\mathbf{i}_{*,n}) \right]$$

That is, the standard matrix  $\mathbf{A}$  is formed by transforming each column of the identity matrix with  $T$  and using each resultant vector as a column of  $\mathbf{A}$ .

□