



DSE

Data Science and Economics

## **Graph Theory, Discrete Mathematics and Optimization – *Module Graph Theory and Discrete Mathematics***

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Main Web page: [Ariel site](#)

**Exam:** 2 mid term (after I and II part + after III part) **OR**  
written exam with I+II+III part

## Topics:

- ❑ (some) Linear Algebra
- ❑ Introduction to Graph Theory
- ❑ Difference equations and dynamical systems

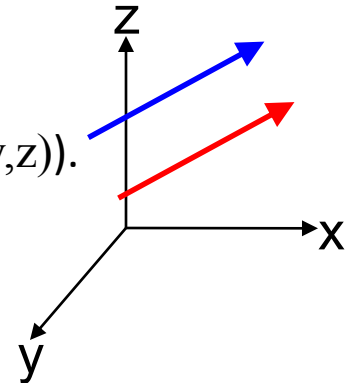
**Linear algebra** is an area of study in Mathematics that concerns itself primarily with the study of vector spaces and the linear transformations between them.

## What? And the data?

wait wait ... **Linear algebra is behind (almost) all the powerful machine learning algorithms and vectors/matrices/tensors are a basic data structure**

# Scalars, Vectors, and Tensors

- A Scalar
  - Has magnitude only (e.g.  $T$ =temperature)
  - Represented by a single number
- A Scalar Field
  - A scalar as function of position (e.g.  $T=T(x,y,z)$ )
  - Represented by a single number whose value varies in space.
- A Vector
  - Characterized by a magnitude and direction (e.g.  $v$ =velocity)
  - Represented by a set of numbers (e.g. in 3 dimensions 3 numbers)
  - Represented as an arrow with length and spatial orientation
  - Two vectors are said to be equal if they are Parallel (Pointed in same direction) and of equal length (magnitude).
- A Vector Field
  - A vector whose magnitude and direction vary in space (e.g.  $v=v(x,y,z)$ ).



Two Equal  
Vectors

# Scalars, Vectors, and Tensors

- A tensor (here we refer **only** to three-dimensional space)
  - Characterized by an *order*.
  - In general then:
    - Zeroth-order tensor is a scalar
    - First-order tensor is a vector
    - Second order tensor looks like a 3x3 matrix.
  - An  $n^{th}$  order tensor has  $3^n$  components
  - Usually, “tensor” refers to a second order tensor
    - Ordered set of nine numbers, each of which is associated with two directions
    - “Arrow-in-space” concept not helpful
    - Stress tensor a common example in fluid mechanics

## Abstract version

A **vector space** is a nonempty set  $V$  of objects, called *vectors*, on which are defined two operations, called *addition* and *multiplication by scalars* (real numbers), subject to the ten axioms (or rules) listed below.<sup>1</sup> The axioms must hold for all vectors  $\mathbf{u}$ ,  $\mathbf{v}$ , and  $\mathbf{w}$  in  $V$  and for all scalars  $c$  and  $d$ .

1. The sum of  $\mathbf{u}$  and  $\mathbf{v}$ , denoted by  $\mathbf{u} + \mathbf{v}$ , is in  $V$ .
2.  $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$ .
3.  $(\mathbf{u} + \mathbf{v}) + \mathbf{w} = \mathbf{u} + (\mathbf{v} + \mathbf{w})$ .
4. There is a **zero** vector  $\mathbf{0}$  in  $V$  such that  $\mathbf{u} + \mathbf{0} = \mathbf{u}$ .
5. For each  $\mathbf{u}$  in  $V$ , there is a vector  $-\mathbf{u}$  in  $V$  such that  $\mathbf{u} + (-\mathbf{u}) = \mathbf{0}$ .
6. The scalar multiple of  $\mathbf{u}$  by  $c$ , denoted by  $c\mathbf{u}$ , is in  $V$ .
7.  $c(\mathbf{u} + \mathbf{v}) = c\mathbf{u} + c\mathbf{v}$ .
8.  $(c + d)\mathbf{u} = c\mathbf{u} + d\mathbf{u}$ .
9.  $c(d\mathbf{u}) = (cd)\mathbf{u}$ .
10.  $1\mathbf{u} = \mathbf{u}$ .

**Example.** scalar: real numbers; vector: a force.

# For this course we will consider only space the Euclidean Vector Spaces

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**Note.** Symbol  $\mathbb{R}$  for the set of the real numbers.

Vectors in the Euclidean  $n$ -space, usually denoted by  $\mathbb{R}^n$ , are the ordered  $n$ -tuples: sequence of  $n$  real numbers. The standard notation for a vector is a boldface lowercase letter and the  $n$ -tuple is represented as a column vector, as in

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

In general, scalar multiplication and addition in  $\mathbb{R}^n$  are, respectively, defined by

$$\alpha \mathbf{x} = \begin{pmatrix} \alpha x_1 \\ \alpha x_2 \\ \vdots \\ \alpha x_n \end{pmatrix} \quad \text{and} \quad \mathbf{x} + \mathbf{y} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \\ x_n + y_n \end{pmatrix}$$

for any  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$  and any scalar  $\alpha$ .

**Example** if  $\mathbf{x} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ ,  $\mathbf{y} = \begin{pmatrix} 3 \\ -4 \end{pmatrix}$ , then

$$-\mathbf{x} = \begin{pmatrix} -2 \\ -1 \end{pmatrix}, \quad 3\mathbf{x} = \begin{pmatrix} 6 \\ 3 \end{pmatrix}, \quad -2\mathbf{x} = \begin{pmatrix} -4 \\ -2 \end{pmatrix}$$

$$\mathbf{x} + \mathbf{y} = \begin{pmatrix} 5 \\ -3 \end{pmatrix} \quad \mathbf{x} - \mathbf{y} = \begin{pmatrix} -1 \\ 5 \end{pmatrix}$$

**Example** consider the following values for a Iris flower:

- sepal length
- sepal width
- petal length
- petal width

For the iris-setosa we have the vector:

$$\mathbf{is} = \begin{pmatrix} 5.1 \\ 3.5 \\ 1.4 \\ 0.2 \end{pmatrix}$$

The term *matrix* means simply a rectangular array of numbers. A matrix having  $m$  rows and  $n$  columns is said to be  $m \times n$ . A matrix is said to be *square* if it has the same number of rows and columns, that is, if  $m = n$ .

**Notation.** The name of a matrix is usually a capital letter, the elements (entries)  $a_{ij}$  have two indices:  $i$  row number,  $j$  column number,

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix},$$

We say that two matrices are **equal** if they have the same size (i.e., the same number of rows and the same number of columns) and if their corresponding entries are equal. If  $A$  and  $B$  are  $m \times n$  matrices, then the **sum**  $A+B$  is the  $m \times n$  matrix whose entries are the sum of the corresponding entries in  $A$  and  $B$ . The sum is defined only when  $A$  and  $B$  are the same size.

If  $r$  is a scalar and  $A$  is a matrix, then the **scalar multiple**  $rA$  is the matrix whose entries are  $r$  times the corresponding entries in  $A$ .



## Example

L. Eldén. Numerical linear algebra in data mining. *Acta Numerica*, 2006,

**Example 1.1** <sup>1</sup> *Term-document matrices* are used in *information retrieval*. Consider the following selection of four documents. Key words, which we call *terms*, are marked in boldface<sup>2</sup>.

Document 1: The **Google** **matrix**  $P$  is a model of the **Internet**.

Document 2:  $P_{ij}$  is nonzero if there is a **link** from **web** **page**  $j$  to  $i$ .

Document 3: The **Google** **matrix** is used to rank all **web** **pages**

Document 4: The **ranking** is done by solving a **matrix** **eigenvalue** problem. If we

Document 5: **England** dropped out of the top 10 in the **FIFA** **ranking**.

count the frequency of terms in each document we get the following result.

Term	Doc 1	Doc 2	Doc 3	Doc 4	Doc 5
eigenvalue	0	0	0	1	0
England	0	0	0	0	1
FIFA	0	0	0	0	1
Google	1	0	1	0	0
Internet	1	0	0	0	0
link	0	1	0	0	0
matrix	1	0	1	1	0
page	0	1	1	0	0
rank	0	0	1	1	1
web	0	1	1	0	0

Thus each each document is represented by a vector, or a point, in  $\mathbb{R}^8$ , and we can

organize them as a term-document matrix,

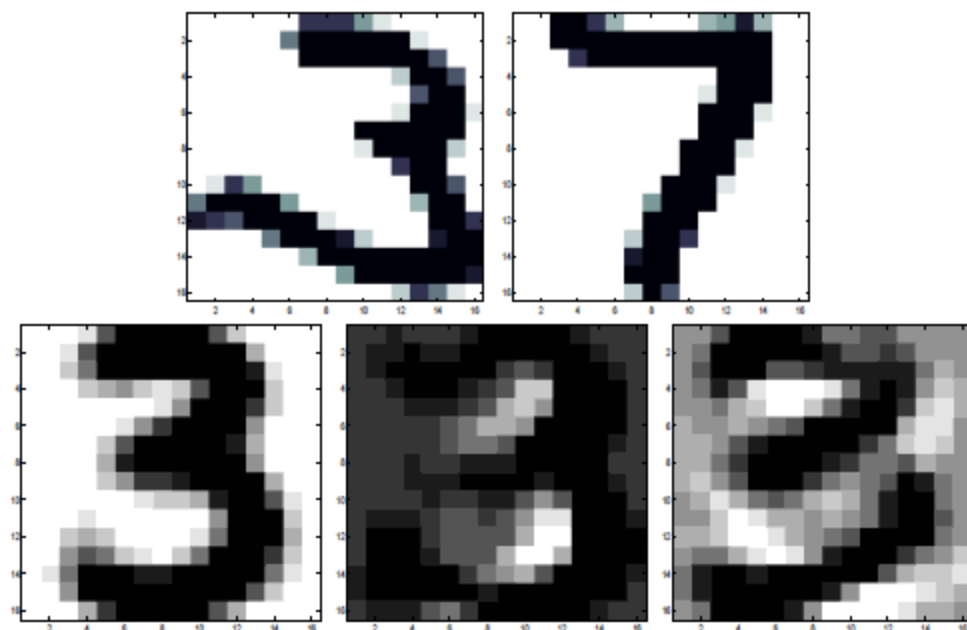
$$A = \begin{pmatrix} 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 & 0 \end{pmatrix}.$$

Now assume that we want to find all documents that are relevant to the query “ranking of web pages”. This is represented by *query* vector, constructed in an analogous way as the term-document matrix,

$$q = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 1 \\ 1 \end{pmatrix} \in \mathbb{R}^{10}.$$

Thus the query itself is considered as a document. The information retrieval task can now be formulated as a mathematical problem: *find the columns of  $A$  that are close to the vector  $q$* . To solve this problem we must use some distance measure in  $\mathbb{R}^{10}$ .

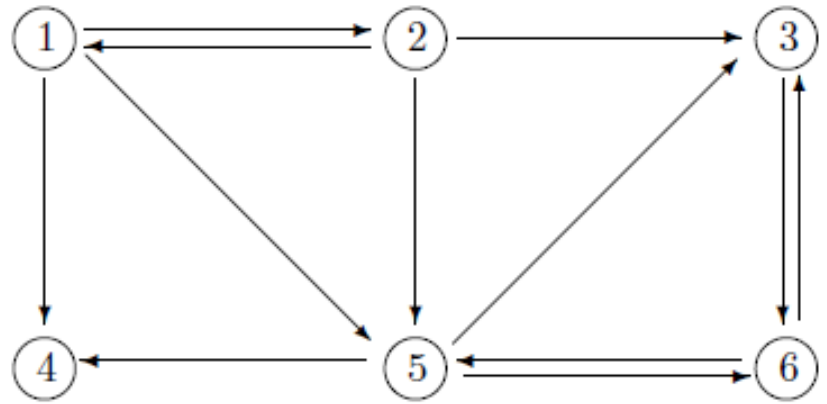
**Example 1.2** In handwritten digit recognition vectors are used to represent digits. The image of one digit is a  $16 \times 16$  matrix of numbers, representing grey scale. It can also be represented as a vector in  $\mathbb{R}^{256}$ , by stacking the columns of the matrix. A set of  $n$  digits (handwritten 3's, say) can then be represented by matrix  $A \in \mathbb{R}^{256 \times n}$ , and the columns of  $A$  span a subspace of  $\tilde{\mathbb{R}}^{256}$ .



*Handwritten digits from the US Postal Service data base*

**Example 1.3** The task of extracting information from all the web pages available on the Internet, is done by *search engines*. The core of the Google search engine is a matrix computation, probably the largest that is performed routinely. The matrix is constructed based on the link structure of the web,

The following small link graph illustrates a set of web pages with outlinks and inlinks.



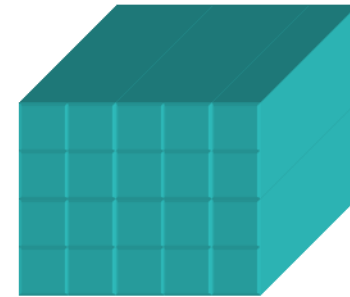
The corresponding matrix becomes

$$P = \begin{pmatrix} 0 & \frac{1}{3} & 0 & 0 & 0 & 0 \\ \frac{1}{3} & 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{3} & 0 & 0 & \frac{1}{3} & \frac{1}{2} \\ \frac{1}{3} & 0 & 0 & 0 & \frac{1}{3} & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & 0 & 0 & \frac{1}{2} \\ 0 & 0 & 1 & 0 & \frac{1}{3} & 0 \end{pmatrix}.$$

Tensor: “Generalization of an n-dimensional array”



Vector: order-1 tensor



Order-3 tensor

Matrix: order-2 tensor

