

Analytical Geometry and Linear Algebra. Lecture 1.

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Outline

- Part 1. About the course
- Part 2. Introduction. Vector spaces. Linear independence. Basis
- Part 3. Dot product

Main questions for today's lecture

- What is this course about?

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- How to use this course in your projects?

Main questions for today's lecture

What is this course about?

Topics of the course

- Vector spaces, matrices and transformations in 2D and 3D
- Lines and planes
- Conics or quadric curves
- Quadratic surfaces
- Polar and spherical coordinates

Goals of this course

What you will learn in this course?

- to use vectors and matrices to solve applied problems
- to change basis in a vector space
- to calculate determinants
- to recognise different transformations, such as rotation, reflection, shear, etc.
- to work with lines and planes in 2D and 3D
- to operate with quadric curves, such as ellipse, hyperbola and parabola
- many more + some examples in Python :)

Main questions for today's lecture

How to get a high grade in this course?

Grading in the course

- Labs 5%
- Test 1 15%
- Midterm 35%
- Test 2 15%
- Final Exam 30%

In total, 100 %

How to get the highest grade?

- Attend classes (either online or offline)
 - Labs
 - Tutorials
 - Lectures
- Solve assignments (also at home) on your own and in groups
- Read books (check the list in moodle)
- Come to office hours (either online or offline)

Repeat :)

What is the exact process you can follow?

● **Friday**

- attend lecture
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● **Tuesday - Thursday**

- apply your knowledge by some programming (yay!)
- do not forget about other courses

Team of the course and Materials

- Vladimir Ivanov (PhD), Principal Instructor, Lectures
- Mohammedreza Bahrami (PhD), Tutorials
- Anastasia Puzankova, Labs
- Oleg Bulichev, Labs
- Eugene Marchuk, Labs

Resources: Books, Assignments, Useful links, etc.

Please, check Moodle!

Main questions for today's lecture

Applications of Linear Algebra and Analytical Geometry

Applications of AGLA in Computer Science and Engineering

Areas:

- Computer Graphics and Computer Games
- Machine Learning, Data Analysis
- Natural Language Processing
- Robotics
- Computer Vision
- and many, many other areas...
- maybe, even in the backend...

Applications

Computer Graphics and Computer Games

- 2D/3D graphics
- Projective geometry, Homogeneous coordinates
- Collision detection in games. Calculation of trajectories

Machine Learning, Data Analysis

- Linear Regression
- Eigendecomposition
- Singular Value Decomposition
- Covariance matrix
- Linear Layers, Attention Mechanisms in Neural Networks

Agenda: Week 1

Vectors. Linear Independence

- Points and Vectors
- Vector Addition. Scalar Vector Multiplication
- Properties of Vector Arithmetic
- Vector spaces, Subspaces
- Span, Linear Independence
- Vector Bases and Vector Coordinates in Basis

Notation

- We denote points by capital italic letters, e.g., A, B, \dots, Q, \dots
- We denote numbers by Greek letters, e.g., $\alpha, \beta, \dots, \lambda, \theta, \dots$ and sometimes by Latin letters, $a, b, \dots, v, u, x, \dots$
- We denote vectors by **bold** letters, e.g., $\mathbf{a}, \mathbf{b}, \dots, \mathbf{v}, \mathbf{u}, \mathbf{x}, \dots$,
- and also we denote vectors by a letter with an arrow, e.g. $\vec{a}, \vec{b}, \vec{u}$
- and sometimes we denote vectors by end-points, e.g. $\overline{AB}, \overline{BC}, \overline{OA}$
- \mathbb{R} is the set of real numbers
- \mathbb{C} is the set of complex numbers

Introduction

Points and Vectors (informally). Direction

Vector. Geometrical point of view. Vectors as 'arrows' in plane or in 3D space

Let A and B be two points.

A directed line segment from A to B is denoted by: \overrightarrow{AB}

This directed line segment constitutes a vector.

Points and Vectors (**informally**). Direction

Vector. Geometrical point of view. Vectors as ‘arrows’ in plane or in 3D space

Let A and B be two points.

A directed line segment from A to B is denoted by: \overline{AB}

This directed line segment constitutes a vector.

Thus, each vector can be associated with a notion of *direction*. In this case, we can think of a vector as an “arrow” in space.

Points and Vectors (**informally**). Magnitude

Length (or Magnitude) of a Vector

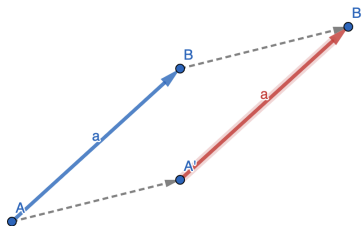
Also, often (**but not always!**) vector has a *length* (or a magnitude). The length of a vector is denoted by $\|\mathbf{v}\|$.

Unit vector

A *unit vector*, \mathbf{u} is a vector with unit length (so $\|\mathbf{u}\|=1$). We can derive a unit vector as $\mathbf{u} = \mathbf{v}/\|\mathbf{v}\|$.

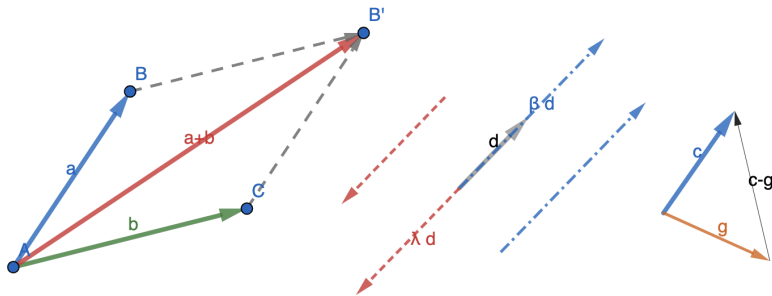
The length of a vector is closely related to the **dot product**, an operation which will be discussed in the next lecture. Therefore, $\mathbf{v}/\|\mathbf{v}\|$ is called a normalized vector.

If you move the line segment to another line segment with the same direction and length, they constitute **the same vector**.



Examples: Points and Vectors (informally)

Note that vector λd is either parallel ($\lambda > 0$) to or anti-parallel ($\lambda < 0$) to d .



$$\lambda, \beta \in \mathbb{R}$$

In this figure: $\lambda > 0$?

What if $\lambda = 0$?

Vector spaces

Vector space definition

Vector space

A *vector space* V over \mathbb{R} (or \mathbb{C}) is a collection of vectors, together with two operations:

- $\mathbf{a} + \mathbf{b}$, addition of two vectors and
- $\lambda \mathbf{a}$, multiplication by a scalar ($\lambda \in \mathbb{R}$)

A scalar is a number from \mathbb{R} or \mathbb{C} , respectively.

Addition and multiplication SHOULD satisfy following axioms

Vector addition axioms

Vector addition $\mathbf{a} + \mathbf{b}$ is defined $\forall \mathbf{a}, \mathbf{b} \in V$

Vector addition has to satisfy the following axioms:

- $\mathbf{a} + \mathbf{b} = \mathbf{b} + \mathbf{a}$ (commutativity)
- $(\mathbf{a} + \mathbf{b}) + \mathbf{c} = \mathbf{a} + (\mathbf{b} + \mathbf{c})$ (associativity)
- There is a vector $\mathbf{0}$ (zero vector) such that $\mathbf{a} + \mathbf{0} = \mathbf{a}$. (identity)
- For each vector \mathbf{a} , there exists a vector $(-\mathbf{a})$ such that $\mathbf{a} + (-\mathbf{a}) = \mathbf{0}$ (inverse)

Scalar multiplication axioms

$\lambda \mathbf{a}$ is defined $\forall \lambda \in \mathbb{R}, \forall \mathbf{a} \in V$

Scalar multiplication has to satisfy the following axioms:

- $\lambda(\mathbf{a} + \mathbf{b}) = \lambda \mathbf{a} + \lambda \mathbf{b}$.
- $(\lambda + \mu)\mathbf{a} = \lambda \mathbf{a} + \mu \mathbf{a}$.
- $\lambda(\mu \mathbf{a}) = (\lambda \mu)\mathbf{a}$.
- $1\mathbf{a} = \mathbf{a}$.

The scalar is called a *scalar*, because it **scales** a vector :)



Homework Assignment

Prove

The zero vector is unique.

Prove

The inverse vector $(-a)$ is unique for any vector a .

Vectors as lists of numbers

Column vectors. Examples

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$, $\begin{bmatrix} a \\ b \\ c \end{bmatrix}$ — we will use **this notation!** We represent vectors as **columns!**

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Row vectors. Examples

$[3 \ 4]$, $[3 \ 4 \ 5]$, $[x \ y \ z]$ Even though vectors can be represented as rows.

Vectors as lists of numbers

Column vectors. Examples

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$, $\begin{bmatrix} a \\ b \\ c \end{bmatrix}$ — we will use **this notation!** We represent vectors as **columns!**

Row vectors. Examples

$[3 \ 4]$, $[3 \ 4 \ 5]$, $[x \ y \ z]$ Even though vectors can be represented as rows.

$$[3 \ 4] \neq \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

Transposition

Transposition

$$\begin{bmatrix} 3 & 4 \end{bmatrix}^T = \begin{bmatrix} 3 \\ 4 \end{bmatrix} \quad (1)$$

$$\begin{bmatrix} 3 \\ 4 \end{bmatrix}^T = \begin{bmatrix} 3 & 4 \end{bmatrix} \quad (2)$$

This operation transforms a row vector a to column vector and back

For any vector

$$(\mathbf{v}^T)^T = \mathbf{v}$$

Examples

Example

Vector space V consisting of all functions $f(x)$ that are continuous on \mathbb{R}

$$V = \{f(x), \text{ such that } f(x) \text{ is continuous on } \mathbb{R}\}$$

Linear combination and linear independence

Linear combination

Vector $\mathbf{w} \in V$ is a linear combination of vectors $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$ with coefficients $c_k \in \mathbb{R}; (k = 1..n)$ such that

$$\mathbf{w} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n = \sum_{k=1}^n c_k \mathbf{v}_k$$

Span

Span

Let $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\} \subset V$.

$$\text{span}(S) \equiv \left\{ \mathbf{w} \in V : \mathbf{w} = \sum_{k=1}^n c_k \mathbf{v}_k, \quad \forall c_k \in \mathbb{R} \right\}$$

Basically, $W = \text{span}(S)$ is the set of all (possible) linear combinations of the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$.

Note that W is a subspace of V .

Subspace

Definition

W is a subspace of V if

- a) $W \subset V$ (subset)
- b) $\mathbf{u}, \mathbf{v} \in W \Rightarrow \mathbf{u} + \mathbf{v} \in W$ (closure under addition)
- c) $\mathbf{u} \in W, \lambda \in \mathbb{R} \Rightarrow \lambda \mathbf{u} \in W$ (closure under scalar multiplication)

Examples

Linear independence in \mathbb{R}^2 and in \mathbb{R}^3

Linearly independent vectors in \mathbb{R}^2

Two vectors \mathbf{a} and \mathbf{b} are *linearly independent*

if for $\alpha_1, \alpha_2 \in \mathbb{R}$, $\alpha_1 \mathbf{a} + \alpha_2 \mathbf{b} = \mathbf{0}$ if and only if $\alpha_1 = \alpha_2 = 0$.

Linear independence in \mathbb{R}^2 and in \mathbb{R}^3

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Linearly independent vectors in \mathbb{R}^3

Vectors \mathbf{a} , \mathbf{b} and \mathbf{c} are *linearly independent*

if for $\alpha_1, \alpha_2, \alpha_3 \in \mathbb{R}$, $\alpha_1 \mathbf{a} + \alpha_2 \mathbf{b} + \alpha_3 \mathbf{c} = \mathbf{0}$ if and only if $\alpha_1 = \alpha_2 = \alpha_3 = 0$.

Linear independence in \mathbb{R}^2 and in \mathbb{R}^3

Linearly independent vectors in \mathbb{R}^2

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Linearly independent vectors in \mathbb{R}^3

Vectors \mathbf{a} , \mathbf{b} and \mathbf{c} are *linearly independent*

if for $\alpha_1, \alpha_2, \alpha_3 \in \mathbb{R}$, $\alpha_1 \mathbf{a} + \alpha_2 \mathbf{b} + \alpha_3 \mathbf{c} = \mathbf{0}$ if and only if $\alpha_1 = \alpha_2 = \alpha_3 = 0$.

Try to give a definition for Linearly independent vectors in \mathbb{R}^n

Basis of a vector space

Basis

A **set** of vectors is a *basis* of a vector space if it spans a vector space and this set is **linearly independent**.

Basis in \mathbb{R}^2 and \mathbb{R}^3 Basis in \mathbb{R}^2

A **set** of vectors is a *basis* of \mathbb{R}^2 if it spans \mathbb{R}^2 and this set is **linearly independent**.

Standard basis in \mathbb{R}^2

$\{\hat{\mathbf{i}}, \hat{\mathbf{j}}\} = \{(1, 0), (0, 1)\}$ is a basis of \mathbb{R}^2 . They are the standard basis in \mathbb{R}^2 .

Standard basis in \mathbb{R}^3

$\{\hat{\mathbf{i}}, \hat{\mathbf{j}}, \hat{\mathbf{k}}\} = \{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$ is a basis of \mathbb{R}^3 . They are the standard (canonical) basis in \mathbb{R}^3 .

Examples

Representation of a Vector in Vector Space

Theorem

Let V be a vector space over \mathbb{R}^n and let $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ be a basis. Then each vector \mathbf{u} can be identified with its coordinates $\{u_1, \dots, u_n\}$ in the basis.

$$\mathbf{u} = \sum_{k=1}^n u_k \mathbf{e}_k$$

$$\mathbf{u} = \begin{bmatrix} u_1 \\ u_2 \\ \dots \\ u_n \end{bmatrix}$$

Homework Assignment

Let P_3 , be a set of all polynomials of degree 3 or less.

Show that P_3 is a vector space over \mathbb{R} .

Hint: check axioms of vector space.

What could be a basis of P_3 ?

Give examples of two bases in P_3 .

Express the polynomial $x^3 - 2x^2 + 3$ in the basis.

End of Lecture 1.

Useful links

- <https://www.geogebra.org>
- https://youtu.be/fNk_zzaMoSs
- <http://immersivemath.com/ila>
- <http://brilliant.com>