

UNIVERSITÉ DE GENÈVE

ANALYSE ET TRAITEMENT DE L'INFORMATION
14X026

TP 1: Linear Algebra

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Exercise 1. Matrix

1. Find a quadratic polynomial, say $f(x) = ax^2 + bx + c$, such that

$$f(1) = 1, \quad f(2) = 9, \quad f(3) = 27$$

To find a , b and c we solve the following system of equations :

$$\begin{aligned} \begin{cases} a + b + c = 1 \\ 4a + 2b + c = 9 \\ 9a + 3b + c = 27 \end{cases} &\Leftrightarrow \begin{cases} a + b + c = 1 \\ 6b + 5c = -27 \\ 9a + 3b + c = 27 \end{cases} \Leftrightarrow \begin{cases} 3b + 4c = -9 \\ 6b + 5c = -27 \\ 9a + 3b + c = 27 \end{cases} \Leftrightarrow \begin{cases} c = 3 \\ 6b + 5c = -27 \\ 9a + 3b + c = 27 \end{cases} \Leftrightarrow \\ \begin{cases} c = 3 \\ b = -7 \\ 9a + 3b + c = 27 \end{cases} &\Leftrightarrow \begin{cases} c = 3 \\ b = -7 \\ a = 5 \end{cases} \end{aligned}$$

The quadratic function is $f(x) = 5x^2 - 7x + 3$

2. Let a , b be some fixed parameters. Solve the system of linear equations

$$\begin{cases} x + ay = 2 \\ bx + 2y = 3 \end{cases}$$

We solve the system of linear equations for x and y :

$$\begin{aligned} \begin{cases} x + ay = 2 \\ bx + 2y = 3 \end{cases} &\Leftrightarrow \begin{cases} bx + bay = 2b \\ bx + 2y = 3 \end{cases} \Leftrightarrow \begin{cases} (ba - 2)y = 2b - 3 \\ bx + 2y = 3 \end{cases} \Leftrightarrow \begin{cases} y = \frac{2b-3}{ba-2} \\ bx + 2\frac{2b-3}{ba-2} = 3 \end{cases} \Leftrightarrow \\ \begin{cases} y = \frac{2b-3}{ba-2} \\ x = \frac{3 - 2\frac{2b-3}{ba-2}}{b} \end{cases} & \end{aligned}$$

Exercise 2. The importance of the mathematical concept behind a code

1. Download, open and run several times the PYTHON file "TP1/some script.py".

Explain the function `def project_on_first(u, v)` in geometrical terms and then explain the results of this code with clear mathematical concepts (you have to speak about projections and scalar product). *It is very important to understand the mathematical concepts behind a program!!*

The function `project_on_first(u, v)` takes two column vectors \mathbf{u} and \mathbf{v} and it returns the projection of \mathbf{v} over \mathbf{u} . the projection \mathbf{v}' and \mathbf{u} are collinear ($\exists \alpha, \mathbf{v}' = \alpha \mathbf{u}$)

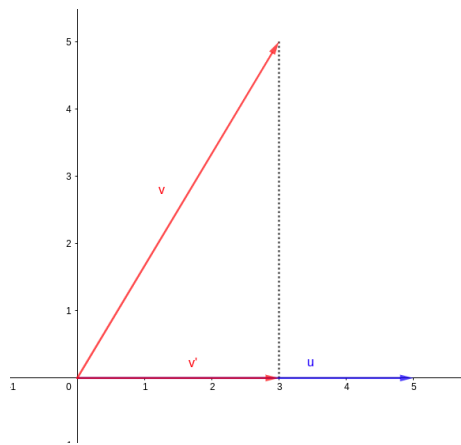


Figure 1: Projection of vector \mathbf{v} over \mathbf{u}

2. Let's consider $u = (u_i)_{1 \leq i \leq 5}$ and $v = (v_i)_{1 \leq i \leq 5}$, two vectors of length 5. How would you clearly write the following part of script using some mathematical notations.

```

1  # import numpy library as np in order to work on tensors
2  import numpy as np
3  # create 2 random vectors of length 5
4  u = np.random.randn(5)
5  v = np.random.randn(5)
6  # perform some computations
7  r = 0
8  for ui, vi in zip(u, v) :
9      r += ui * vi

```

Name the mathematics behind and rewrite the last 3 lines of code in one.

The last lines of the program perform a scalar product between the two vec-

tors u and v (taking two vectors and returns a scalar) $\mathbf{u} \cdot \mathbf{v} = \sum_i u_i v_i$ The `zip()` function takes two lists of same size and creates a list of tuples with the values of the two lists. The last line multiply the two values of each tuples and sum all these products (which corresponds indeed to the formula of the scalar product)
We can rewrite those three line as `r=np.dot(u,v)` (using the numpy library)

3. Complete the code to construct a vector v orthogonal to the vector u and of the same norm. Comment each line of your code.

The code can be found in the file `ex2.py`

Exercise 3. Computing Eigenvalues, Eigenvectors, and Determinants

- Find the determinant, eigenvectors, and eigenvalues of the matrix

$$\begin{bmatrix} 1 & 1 & 3 \\ 1 & 5 & 1 \\ 3 & 1 & 1 \end{bmatrix}$$

Compare your results with computer's one.

We compute the determinant, the eigenvectors and the eigenvalues of the matrix

$$A = \begin{bmatrix} 1 & 1 & 3 \\ 1 & 5 & 1 \\ 3 & 1 & 1 \end{bmatrix}.$$

determinant : $\det(A) = 1 \times (5 \times 1 - 1 \times 1) + (-1) \times (1 \times 1 - 1 \times 3) + 3 \times (1 \times 1 - 5 \times 3) = 4 + 2 - 42 = -36$

$$\text{eigenvalues : } \det(A - \lambda I) = 0 \Leftrightarrow \det \begin{pmatrix} (1-\lambda) & 1 & 3 \\ 1 & (5-\lambda) & 1 \\ 3 & 1 & (1-\lambda) \end{pmatrix} = 0 = (1 -$$

$$\lambda) \times (4 - 6\lambda + \lambda^2) - 1 \times (2 + \lambda) + 3 \times (1 - 15 + 3\lambda) = -36 + 7\lambda^2 - \lambda^3 = -(\lambda + 2)(\lambda - 3)(\lambda - 6)$$

The eigenvalues are $\{-2, 3, 6\}$

eigenvectors : We solve $(A - \lambda I)\mathbf{v} = \mathbf{0}$ for each possible λ to find the eigenvectors and we find :

$\mathbf{v} = (-1, 0, 1)$ for $\lambda = -2$, $\mathbf{v} = (1, -1, 1)$ for $\lambda = 3$ and $\mathbf{v} = (1, 2, 1)$ for $\lambda = 6$

- The covariance matrix for n samples x_1, \dots, x_n , each represented by a $d \times 1$ column vector, is given by

$$C = \frac{1}{n-1} \sum_{i=1}^n (x_i - \mu)(x_i - \mu)^T,$$

where C is a $d \times d$ matrix and $\mu = \sum_{i=1}^n x_i / n$ is the *sample mean*. Prove that C is always positive semidefinite. (Note: A symmetric matrix C of size $d \times d$ is *positive semidefinite* if $v^T C v \geq 0$ for every $d \times 1$ vector v .)

- In this portion of the exercise, we will calculate the eigenvalues of the covariance matrices of six data sets listed as follows:

filename	n	d	description
tp1_artificialdata[1-3]	1024	100	Artificial data generated from various auto-regression (AR-1) models
tp1_artificialdata4	1024	100	Random Gaussian data
tp1_freyfaces	1965	560	Facial images of a man named Brendan
tp1_digit2	5958	784	Hand-written images of "2"

To access each data set, go to "TP1/data" and download `tp1_*`. Each file contains a $n \times d$ data matrix with rows representing n different samples in \mathbb{R}^d . For example, `tp1_artificialdata1` contains a data matrix of size 1024×100 (1024 samples, 100 features). Once each dataset has been imported into Python, complete the following tasks:

- Compute the covariance matrix of each dataset;
- Compute the eigenvalues for each covariance matrix;
- Compute the determinant of each covariance matrix;
- Compute the product of the eigenvalues for each covariance matrix;
- Display the eigenspectrum for each covariance matrix in a 2D plot, where the x -axis shows the rank of the eigenvalues, ranging from 1 (the largest eigenvalue) to 100 (the 100-th largest eigenvalue), and the y -axis shows the corresponding eigenvalue.

Describe the relationship between the product of the eigenvalues and the determinant of each covariance matrix. In addition, describe the observations you make regarding the plot of the spectrum of eigenvalues of each covariance matrix.

The code can be found in the file `ex3.py`

Exercise 4. Computing Projection Onto a Line

There are given a line $\alpha : 3x - 2y = -6$ and a point A with the coordinates $(5, 4)$.

1. Find a distance from the point A to the line α
2. Explain your code using a sketch and some mathematics (there should be a scalar product somewhere).

We want to find the distance between $A(5,4)$ and the line $\alpha : 3x - 2y = -6$. We start by rewriting α as the function $\alpha(x) = \frac{-6+2y}{3}$. Then we pick two values x_1 and x_2 and compute $\alpha(x_1 = y_1)$ and $\alpha(x_2 = y_2)$. We compute a vector \mathbf{u} collinear to α : $\mathbf{u} = \begin{bmatrix} x_2 - x_1 \\ y_2 - y_1 \end{bmatrix}$. Then we compute \mathbf{v} , a vector between a point of α and A : $\mathbf{v} = \begin{bmatrix} x_1 - A_x \\ y_1 - A_y \end{bmatrix}$. We compute two vectors, \mathbf{w}_1 the projection of \mathbf{v} over \mathbf{u} and $\mathbf{w}_2 = \mathbf{v} - \mathbf{w}_1$. Since \mathbf{w}_2 is collinear to \mathbf{u} , $\|\mathbf{w}_2\|$, corresponds to the distance between A and α .

The code can be found in the file `ex4.py`

Supplements

1. Define and explain the mean and the variance on some examples.
2. Define and explain what is a vector space, a projection and a scalar product.
3. Define and explain what is a vector space, a basis and a change of basis transformation matrix.
4. Present the eigen-value decomposition and the singular-value decomposition.
5. Be ready to answer questions from slide 37 *DS.02.highDimension*.