# **Laboratory Test**

#### **DEDP 2020-2021**

### Information

- The test will last for 1 hour
- You will upload the Matlab files on Moodle (or send by email)
- General Matlab stuff you need to know is listed in the Syllabus section
- Template subjects (i.e. exercises extracted from the labs) are in **Template Subjects** section
- The test will be roughly based on these templates, with modifications

## **Syllabus**

Things to know in Matlab:

- Load and save a \*.mat file
- Generate random numbers with Gaussian and uniform distributions
  - single numbers, vectors or matrices
  - with various distribution parameters (mu, sigma, a, b)
- Generate a random vector with 0's and 1's in variable proportions (or some value A instead of 1)
- Generate a sin or cos signal of a certain length, with a specified amplitude, frequency and initial phase
- Generate a vector with N values equally spaced between a start and a stop value (e.g. linspace())
- Compute mean and variance of vectors or columns

- Operate with columns (or rows) of a matrix:
  - extract one or more columns
  - arithmetic operations: add columns, divide element by element, etc
  - same with rows instead of columns
- Count how many values of 1 are in a vector (or maybe how many values equal to A)
- Count pair of values in two vectors (e.g. when there is a 0 in a vector and 1 in another vector, like for false alarms, misses etc)
- Compute Euclidean distance between vectors
- Sort a vector and find the original positions of the smallest k values
- Find the maximum value in a vector and its position
- Find the minimum value in a vector and its position
- Plot a vector
- Plot a vector as a function of another vector
- Create a histogram plot
- Create and use a Matlab function
- Display a message with fprintf()
- Create and use simple cell arrays
- General instructions: if, for, etc

### **Template Subjects**

#### Lab 1

- 1. Create a Matlab function myPDF() that estimates the probability density function from a vector of data.
- the function requires three arguments and returns one value: p = myPDF(v,x,epsilon)
- v is a vector, and x, epsilon are scalar numbers
- the function computes how many elements from  $\mathbf{v}$  are in the interval  $[x \epsilon, x + \epsilon]$ , divided to the total number of elements of  $\mathbf{v}$ , and also divided to 2 times epsilon
- 2. Plot the probability density function estimated from a vector of data

- generate a vector v with 100000 values from the normal distribution  $\mathcal{N}(2,2)$  and plot the values
- generate a vector n of 50 values uniformly spread between -5 to 15
- apply myPDF() on v to estimate the probability density at every value from n (use epsilon = 0.1)
- plot the results of the function against the values of n

#### Lab 2

- 1. Load the file ElectionsData.mat. It contains election data for the local elections in the city of Iasi held on 27.09.2020 (data taken from https://prezenta.roaep.ro). The file contains two variables:
- names: a cell array with the names of the voting centers
- values: a matrix with the voting numbers for each center

The structure of the values matrix is as follows:

- first column: total number of registered voters on permanent lists
- second column: total number of registered voters on complementary lists
- third column: number of votes from permanent lists
- fourth column: number of votes from complementary lists
- fifth column: number of votes from supplementary lists
- sixth column: number of votes with mobile urns
- a. Compute the **turnout** for every voting center, defined as: total number of votes / total number of registered voters on all lists.
- b. Plot the turnout vector
- c. Compute the mean and the variance for the turnout across the city of Iasi.

#### Lab 3

- 1. Simulate threshold-based detection with a single sample, as follows:
  - Generate a vector of 100000 values 0 or A, with equal probability (consider A=5)
  - Add over it a random noise with normal distribution  $\mathcal{N}(0, \sigma^2 = 1)$
  - Pick a value of T = A/2 and compare each element with T to decide which sample is logical 0 or logical 1 (A)
  - Compare the decision result with the true original vector, and count how many correct detections and how may false alarms have been
  - Estimate the four probabilities by dividing the above numbers to the size of the vector

**Variant:** Same exercise, but written as a function which accepts T as input and return the four values as outputs. Running the function for 100 values of T uniformly spaced between 0 and A, and plotting the resulting vector **phit** against **pfa**.

#### Lab 4

- 1. Implement a function [class] = myKNN(signal, k, trainset) for performing k-NN classification of a signal:
  - the function takes as input an unclassified signal signal, the parameter value k, and the training set matrix trainset
  - the function computes the Euclidean distance between signal and each vector from the training set
  - the output class is defined by the majority of the k nearest neighbours of the signal
- 2. Call the function myKNN for the first signal from the testing set and determine its class. Use different values for k: k = 1, then k = 5, then k = 15.

Note: the training set matrix can be loaded from the file ECG\_train.mat, and the test set from ECG\_test.mat

#### Lab 6

- 1. Generate a 500-samples long sinusoidal signal  $s_{\Theta} = A * \sin(2\pi f n)$  with frequency f = 0.02, and add over it normal noise with distribution  $\mathcal{N}(0, \sigma^2 = 0.5)$ . Name the resulting vector  $\mathbf{r}$ . Plot the  $\mathbf{r}$  vector.
- 2. Estimate the frequency  $\hat{f}$  of the signal via Maximum Likelihood estimation from the **r** vector:
  - Generate 1000 candidate frequencies  $f_k$  equally spaced from 0 to 0.5
  - $\bullet$  Compute the Euclidean distance between r and the sine signal with each candidate frequency
  - Maximum Likelihood: choose  $\hat{f}_{ML}$  as the candidate frequency which minimizes the Euclidean distance
  - Display the estimate value  $\hat{f}_M L$
  - Plot a sinusoidal with the estimated frequency  $\hat{f}_{ML}$ , and the original vector  $\mathbf{r}$ , on the same figure

**Variant**: estimate amplitude A instead of frequency f, in the same way