

# Signals and Systems

Lab experience 3

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Report deadline : 7<sup>th</sup> January 2025



**Politecnico  
di Torino**

# Exercise 1: DTF and FFT

- Load a music file of at least 20 seconds (you are free to choose the file) and determine the sampling frequency  $F_s$ 
  - To load the file you can use
    - The function `audioread` in Matlab
    - The function `scipy.io.wavfile.read()` in Python (or other equivalent functions depending on the file format)
  - Read carefully the manual to understand how these functions work
- Estimate the spectrum of energy over temporal windows, using two approaches:
  - Implementing yourself the DFT (Discrete Fourier Transform)
  - Using the built-in FFT (Fast Fourier Transform) function
- Compare the obtained results

# Exercise 1: DTF and FFT

- The music file should be divided into  $K$  temporal windows of length  $M$  seconds
  - Knowing the sampling frequency  $F_s$  you can determine how many samples  $N$  are contained in each window
  - Try different values for  $M$
- For each window, plot the resulting spectrum of energy
  - It is convenient to plot the spectrum in dB
  - Pay particular attention to the frequency axis that should be quoted in kHz
  - You may need to use the `fftshift()` function: read what it is and what it does

# Exercise 1: DTF and FFT

- Required outputs:
  - Plot of the spectra of energy for different windows and considering different  $M$  values
    - There is no need to put all the figures in the report, but only a subset of them which are more significant
  - Comparison of the computational time required by your implementation and the built-in FFT function
    - Use `tic()` and `toc()` functions

# Discrete Fourier Transform (DFT)

- Given a signal  $x(n)$  of  $N$  samples, the **discrete Fourier transform** is defined as

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi n \frac{k}{N}}, \quad \forall k = 0, 1, 2, \dots, N-1$$

- $X(k)$  can be interpreted as the DTFT  $X(e^{j2\pi f})$  evaluated at the  $N$  equi-spaced frequencies:

$$f_k = \frac{k}{N}, \quad \forall k = 0, 1, 2, \dots, N-1$$

$$X(k) = \text{DFT}[x(n)] = X(e^{j2\pi f_k})$$

- The **spectrum of energy** is defined as:  $S_x = |X(k)|^2$

# Exercise 2: DFT and circular convolution

- Consider a portion of the music file of exercise 1 (e.g. 10 seconds) and filter it using a given FIR filter
  - The impulse response  $h[n]$  of the filter is contained in the file «FIR\_impluse\_response.mat»
  - It is a raised-cosine with bandwidth equal to the 10% of the total bandwidth
- Perform the filtering in two ways:
  - In time domain using the convolution
  - Using the circular convolution, i.e. applying the DFT

# Exercise 2: DFT and circular convolution

- Required outputs:
  - Compare the filtered signals in time domain
  - Plot of the spectra of energy before and after the filtering for both implementations
- Suggestion: to play the music and listen to the impact of filtering, in Matlab you can use the function `soundsc`

# DFT and circular convolution

- Given a non-periodic signal  $x[n]$  and a filter describing an LTI system with impulse response  $h[n]$ , the filtered signal at the output of the system can be computed as:

$$y[n] = h[n] * x[n]$$

$$Y(k) = H(k)X(k)$$

- Where  $x[n]$  and  $h[n]$  have extensions  $[0, N_x - 1]$  and  $[0, N_h - 1]$ , respectively, with  $N_x \neq N_h$ , and  $x[n]$  has extension  $[0, N_y - 1]$  with  $N_y = N_x + N_h - 1$
- $X(k) = DFT\{x[n]\}$  and  $Y(k) = DFT\{y[n]\}$



# DFT and circular convolution

- To perform filtering in frequency domain using the DFT

$$Y(k) = H(k)X(k)$$

- $X(k)$  and  $H(k)$  should have the same size  $N$
- For this reason **zero-padding** can be performed, consisting in adding a sufficient number of zeros such that

$$N \geq N_x + N_h - 1$$

- If you use the FFT,  $N$  can be chosen equal to  $2^m$ , where  $m = \log_2(N_x + N_h - 1)$