

## Exercise 5: Time Series Analysis – Part V

- Fourier Transformation and filtering in the frequency domain -

Group:	Surname, Given Name:	Matriculation number:	Signature*:
<b>* With my signature I declare that I was involved in the elaboration of this exercise.</b>			
<b>Deadline: 28.01.2026</b>			

### Objective

This exercise deals with time series analysis in the frequency domain.

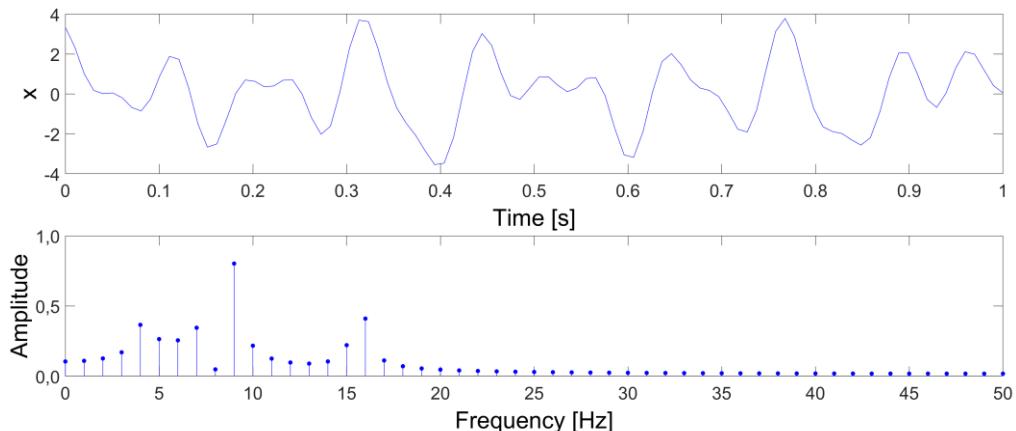


Figure 1: Time series (top) and related Fourier spectrum (bottom)

**in time domain the base function are the  
base vectors/base functions of “nomials”**

**more general the base functions is a  
function, defined in its different spaces/use  
cases**

**in fourier we can use the fourier basis, a set  
of harmonic basae functiobs with different  
coefficients**

### Task 1: Amplitude and phase spectra

- 1) Create a time series of  $n = 100$  observations using the following functional model and chose appropriate values for amplitude  $A$ , frequency  $f$  and phase shift  $\varphi$

$$y(t) = A \cdot \sin(2\pi ft + \varphi)$$

- 2) Calculate the Fast Fourier Transform of the time series using the implemented Matlab function `fft`
- 3) Plot the time series as well as the Fast Fourier Spectrum
- 4) Repeat steps 1) to 3) while changing the amplitude  $A$  and interpret the results
- 5) Repeat steps 1) to 3) while changing the phase shift  $\varphi$  and interpret the results

### Task 2: Fourier analysis

- Calculate the Fast Fourier Transform of the given time series
  - “*TimeSeries.txt*”
  - “*TimeSeries\_Offset.txt*”
  - “*TimeSeries\_Trend.txt*”
  - “*TimeSeries\_Trend\_Offset.txt*”
  - “*TimeSeries\_Data\_Gap.txt*”
- Plot each Fast Fourier spectrum
- Compare and interpret the results

### Task 3: Filtering

- 1) Use the file “*Exercise5-1.txt*” from Exercise 5 – Task 1
- 2) Implement a moving average filter in the frequency domain for different filter length  $\tau = [5 \ 15 \ 25]$
- 3) Compare the results with those from the filtering in the time domain in Exercise 5 – Task 1

#### Task 4: Frequency response of different periodic kernels

Consider a discrete, periodic signal with  $N = 127$  samples and the discrete index  $n = 0, \dots, N - 1$ . Furthermore, the signal is assumed to be periodic with the period  $N$ , with the centred frequency index  $k = -\frac{N-1}{2}, \dots, 0, \dots, \frac{N-1}{2}$ . For a periodic kernel  $h[n]$ , the frequency response (eigenvalues) is defined as:

$$\lambda_k = \text{FFT}\{h[n]\}_k$$

Construct the following kernels. In each case, build a length- $N$  kernel vector  $h$  by placing the nonzero coefficients centred around  $n=0$  (circularly), and setting all other entries to zero.

- Moving average,  $\tau = 11$

$$h_1[n] = \frac{1}{11} [1, 1, \dots, 1]$$

- Moving average,  $\tau = 21$

$$h_2[n] = \frac{1}{21} [1, 1, \dots, 1]$$

- Triangular kernel,  $\tau = 21$

Construct  $h_3[n]$  by convolving the length-11 moving average kernel  $h_1[n]$  with itself, and normalize such that  $\sum_n h_3[n] = 1$ .

- Random symmetric compact kernel,  $\tau = 21$

Create random coefficients on the support length 21, enforce symmetry  $h_4[-n] = h_4[n]$ , and normalize such that  $\sum_n h_4[n] = 1$ .

- 1) Compute the frequency response  $\lambda_k^{(i)}$  for each kernel  $h_i[n]$  and plot the magnitude spectrum  $|\lambda_k^{(i)}|$  versus  $k$ .
- 2) For each kernel, briefly discuss:
  - a. Is it mainly a low-pass filter?
  - b. How quickly does  $|\lambda_k^{(i)}|$  decay with increasing  $|k|$ .
  - c. Do you observe side lobes? What do they indicate?
  - d. Compare the rectangular and triangular kernels: which one has the smoother frequency response and why?

### Task 5: Ideal low pass filtering and the corresponding kernel

Consider a discrete, periodic signal with  $N = 127$  samples and the discrete index  $n = 0, \dots, N - 1$ . Furthermore, the signal is assumed to be periodic with the period  $N$ . A filter is defined in the frequency domain by the following frequency response:

$$\lambda_k = \begin{cases} 1, & |k| \leq k_0, \\ 0, & \text{otherwise,} \end{cases} \quad k = -\frac{N-1}{2}, \dots, 0, \dots, \frac{N-1}{2}$$

The parameter  $k_0$  is defined via the filter length

$$k_0 = \frac{\tau - 1}{2}, \quad \tau = 11.$$

Here,  $\lambda_k$  denotes the eigenvalues of the circular convolution operator associated with the filter.

- 1) Interpretation in the frequency domain
  - a. Identify the type of filter defined by  $\lambda_k$ .
  - b. Plot the frequency response  $\lambda_k$ .
  - c. Explain why this filter is referred to as an ideal low-pass filter.
- 2) Determination of the kernel
  - a. Determine the corresponding filter kernel  $h[n]$  by applying the inverse discrete Fourier transform to  $\lambda_k$  (e.g. using *ifft*). Ensure that the kernel is properly centred with respect to the discrete index  $n$ .
  - b. Plot the kernel  $h[n]$  over all  $N = 127$  samples.
  - c. Describe the characteristic shape of the kernel.
- 3) Analysis of the Kernel properties
  - a. Explain how the global, oscillatory shape of the kernel influences the filtered signal in terms of locality, ringing artifacts, and phase behaviour.