

Practice exercises on the Fourier Transform

A. Periods and frequencies

1. Consider the following signals (assume t is in seconds):

- (a) A single sine wave:

$$f_1(t) = \sin(2\pi \cdot 3t).$$

- (b) The sum of two sine waves with different frequencies:

$$f_2(t) = \sin(2\pi \cdot 3t) + \frac{1}{2} \sin(2\pi \cdot 7t).$$

Questions:

- What frequencies (in Hz) are present in f_1 and f_2 ?
- What is the period of f_1 and f_2 ?

B. Discrete Fourier Analysis

1. Let

$$f(t) = \cos(2\pi t) + \frac{1}{2} \cos(4\pi t).$$

Questions:

- Which frequencies are present?
- Which Fourier coefficients are non-zero?
- Sketch the corresponding spectrum.

2. Plot the signal $f(t) = \sin(2\pi t)$ at $N = 8, 16, 32, \dots$ points over one period.

Questions:

- Compute the Discrete Fourier Transform numerically.
- How does the spectrum change as N increases?

C. Definitions

1. Fourier transforms (a decomposition of a discrete (f_n) or continuous ($f(t)$) signal into waves:

$$\hat{f}_k = \sum_{n=0}^{N-1} f_n e^{-2\pi i k n / N}, \quad \hat{f}(\omega) = \int_{-\infty}^{\infty} f(t) e^{-i\omega t} dt.$$

Inverse transforms:

$$f_n = \frac{1}{N} \sum_{k=0}^{N-1} \hat{f}_k e^{2\pi i k n / N}, \quad f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{f}(\omega) e^{i\omega t} d\omega.$$

Questions:

- What represents “frequency” in each case?
- What changes conceptually when going from the sum to the integral?

D. Simple Fourier Transforms

1. Compute the Fourier transform of $f(t) = \sin(\omega_0 t)$.

Questions:

- Where is the *support* of the spectrum? That is, of all the frequencies, which are the nonzero ones? What are their values?

2. Repeat for

$$f(t) = \sin(\omega_1 t) + \sin(\omega_2 t).$$

3. Consider the Gaussian

$$f(t) = e^{-t^2/2\sigma^2}.$$

Questions:

- Calculate the Fourier transform $\hat{f}(\omega)$ as a function of ω . What is the shape of its Fourier transform?
- Make sure you understand what the Fourier coefficients give you in terms of the
- How does it change as σ increases?

E. Spectral Density

1. In general, the Fourier coefficients are *complex numbers*. Given complex Fourier coefficients $\hat{f}(\omega)$ - essentially a function over the frequencies - we can have access to what a function looks like in frequency space (what you did before in the previous exercises, but in a systematic way instead of just counting the frequencies every time).

We are interested in seeing how much weight every frequency has in a given signal. As in general $\hat{f}(\omega)$ is a complex function (look at the definition of the transform), the quantity that shows the weight of frequency ω in $f(t)$ is not just the coefficient $f(\omega)$ but rather

$$|\hat{f}(\omega)|^2.$$

This is called the spectral energy. For a signal $f(t)$ it is literally: how much power (or weight) is in each frequency. How would it look for a pure sine wave?

2. Compare

$$f_1(t) = \cos(\omega t), \quad f_2(t) = \cos(\omega t + \phi).$$

Questions:

- Plot the two signals and compare their spectral densities.
- Compare their spectral densities $|\hat{f}_1(\omega)|$ and $|\hat{f}_2(\omega)|$. What information is lost?

E. Random signals

1. Try to construct some random signals from scratch. First, choose a power spectrum, for example

$$P(\omega) = |\omega|^{-2}.$$

2. For each *positive* frequency $\omega > 0$, sample a complex Fourier coefficient

$$\hat{f}(\omega) = a(\omega) + i b(\omega),$$

with

$$a(\omega) \sim \mathcal{N}\left(0, \frac{P(\omega)}{2}\right), \quad b(\omega) \sim \mathcal{N}\left(0, \frac{P(\omega)}{2}\right).$$

3. For each negative frequency, define the coefficient by

$$\hat{f}(-\omega) = \hat{f}^*(\omega).$$

4. For the zero-frequency mode, set $\hat{f}(0) = 0$.

5. After these steps, you have a full set of complex Fourier coefficients $\hat{f}(\omega)$. Use the Fourier expansion to reconstruct a signal $f(t)$ and plot it.

Questions:

1. Why was the relation $\hat{f}(-\omega) = \hat{f}^*(\omega)$ imposed?
2. Plot $|\hat{f}(\omega)|^2$. How is it related to $P(\omega) = |\omega|^{-2}$? How would you get from the former to the latter. The quantity $P(\omega)$ is called the *power spectrum*. This is an extremely important concept, so we should make sure that you are familiar with it.
3. What you have created here is a Gaussian random field (signal). What is the main difference with a non Gaussian one?

Now that you understand the concept of frequency and spectral density (or spectrum) of a given signal, you can go again to the beginning of the Jupyter Notebook and read a bit the 1-d examples.

Finally, try to see how this is generalised to images instead of 1-d signals. Instead of ω you have a vector $\mathbf{k} = (k_x, k_y)$ which is the frequency *for each dimension*. For images we have two dimensions, and so the “frequency” is a vector - we call it a *wavevector*.

Please try to convince yourself about what the relationship between *scales* and *frequency* or wavenumber is. Larger wavelength waves will capture the ‘larger’ features in an image, and the finer details will be captured by waves of high frequency. High frequencies \leftrightarrow small length scales, low frequencies \leftrightarrow high length scales, and vice versa.

I hope it’s more clear, and please do not hesitate to experiment on your own, and to send me any questions you might have :)