

Worksheet: spectral time series analysis

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The first five problems will examine oscillatory activity in local field potentials recorded from infralimbic cortex and from the basolateral amygdala of rats. These recordings are available in this [repository](#), and two publications related to this data are cited in this [README.md](#).

In the following exercises you will use a 45-minutes local field potential saved in the file `08102017/time_data_pre_45sec.mat` of the previous repository. To speed up its processing, I downsampled this file by a factor of ten. You can download the downsampled file from [here](#).

1. Plot data from one channel in this recording. You can adapt the Python script provided [here](#).
2. Estimate and plot the correlation of the data plotted above.
3. Estimate the spectral density using this data and periodogram method, as indicated in the section [Step 4: Power spectral density, or spectrum](#) of the lecture *The Power Spectrum (Part 1)*. You may want to complete the code provided [here](#).
4. The periodogram is a noisy estimator of the spectral density. Use the function `scipy.signal.welch` to re-estimate the previous power spectrum, and check if the Welch estimate is better than the periodogram one.
5. Both the periodogram and the Welch method assume that the time series is static. Use the spectrogram, as described in the section [Step 6: The spectrogram](#) of the lecture *The Power Spectrum (Part 1)*, to bypass this assumption and check if the spectral density changes with time.
6. (optional) We measure the LFP in human motor cortex with an Utah array. It is known that this LFP only has an oscillation at 11 Hz (i.e., $LFP(t) = \cos(\omega_0 t)$ with $\omega_0 = 2\pi f_0$ rad/sec, $f_0 = 11$ Hz). However, when we sample this LFP at a frequency of 10 Hz (i.e., $\omega_s = 2\pi f_s$ rad/sec, $f_s = 10$ Hz) we only observe an oscillation at 1 Hz (Figure [6](#)).
 - (a) explain the appearance of the 1 Hz oscillation using the sampling theorem.

Hints:

- the Fourier transform of a cosine is $\mathcal{FT}\{\cos(\omega_0 t)\} = \frac{1}{2}[\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$ and has the spectrum in the figure below.

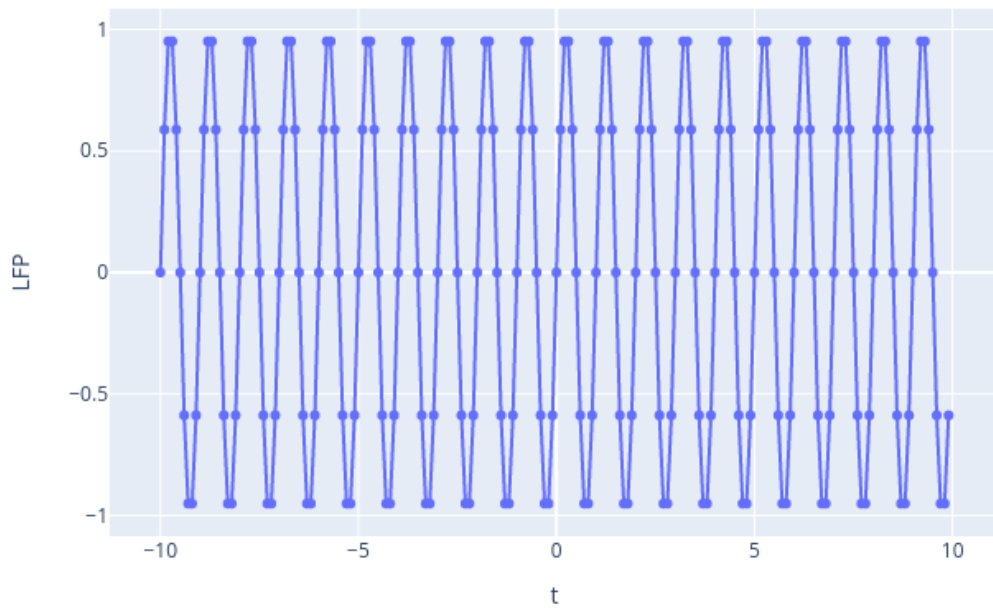
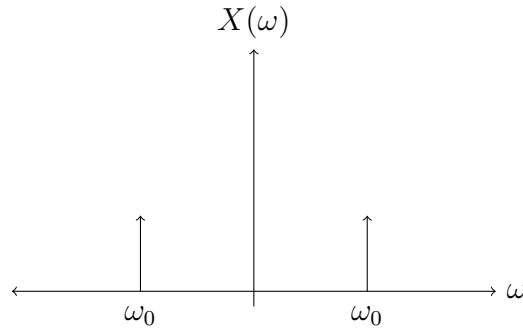


Figure 1: An LFP oscillating at 11 Hz (i.e., $LFP(t) = \cos(\omega_0 t)$ with $\omega_0 = 2\pi f_0$ rad/sec, $f_0 = 11$ Hz) when sampled at a frequency of 10 Hz (i.e., $\omega_s = 2\pi f_s$ rad/sec, $f_s = 10$ Hz) only displays an oscillation at 1 Hz. Use the sampling theorem to explain this observation. Code to generate this figure appears [here](#).



- replicate the above spectrum, as indicated by the sampling theorem, with replicas at multiples of the sampling frequency $\omega_s = 2\pi f_s$ ($f_s = \frac{1}{T_s} = 10 \text{ Hz}$).
- check if any of the above replicates adds signal at 1 Hz (i.e., $\omega = 2\pi f$ rad/sec, $f = 1 \text{ Hz}$).

Note: to avoid this type of problems where low-frequency oscillations appear due to frequencies in the signal above the Nyquist frequency (i.e., half of the sampling frequency), signals are low pass filtered with an analog filter at the Nyquist frequency before being sampled. This filter is called an **antialiasing filter**. Analog filters do not generate aliasing.

- use a sampling frequency above the Nyquist rate (i.e., $f_s > 2f_N$ where f_N is the largest frequency in the signal) and check that oscillations at 11 Hz appear in the sampled signal. You may want to use [this](#) code.
- build another example of an LFP having an oscillation at a high frequency that when sampled at a frequency below the Nyquist rate generates an oscillation at a lower frequency. You can use [this](#) code to verify that with your values of the LFP frequency and the sampling frequency an oscillation at a low frequency emerges.