

## 6. Exercise Sheet – Brain-Inspired Computing (WS 15/16)

Due date 23.11.16.

Name(s): \_\_\_\_\_ Group: \_\_\_\_\_ Points: \_\_\_\_/\_\_\_\_/\_\_\_\_/\_\_\_\_

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### 6.1 Poisson process with absolute refractoriness (40 Points)

Derive and plot the power spectrum for a Poisson process with absolute refractoriness and compare it to the one of the Poisson process. Which one would you choose for transmitting a slow (i.e., low-frequency) signal and why?

### 6.2 Inhibitory rebound with AdEx (30 Points)

Run a PyNN simulation (with NEST) to observe post-inhibitory rebound. The AdEx model is available as `EIF_cond_exp_isfa_ista`. First, think about which parameters influence the membrane in a way that is relevant for this phenomenon. Start from the default parameters and modify the relevant ones to get closer to the desired behaviour. Alternatively, you could also use a brute-force approach and sweep over an appropriate interval of parameter values to find the ones you need.

### 6.3 Power spectra (30 Points)

On the course website you find two example signal traces (named `signal_1.npy.gz` and `signal_2.npy.gz`). They were sampled with a  $dt$  of 1 ms and 0.1 ms respectively. After downloading, you can load them in the following way:

```
import gzip
import numpy as np

with gzip.GzipFile("signal_1.npy.gz") as f:
    signal_1 = np.load(f)

with gzip.GzipFile("signal_2.npy.gz") as f:
    signal_2 = np.load(f)
```

- Plot the first 100 ms of each signal.
- Plot the histogram over all values in the signal.
- Construct your own autocorrelation function (without using any convolution-related `numpy` functions).
- Plot the autocorrelation function of both signals.
- Based on your autocorrelation function, devise a function that computes the power spectra of a signal. (Hint: You may read up on and use both `numpy.fft.rfft` as well as `numpy.fft.fftfreq`.)
- Plot the power spectra of the two signals. Do both signals originate from the same source? Substantiate your answer.