

# Brain Inspired Computing (SS 19): Exercise sheet 1

Hand in on 24.04.2019, 11:15

Name(s):

Group:

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Question:	1	2	3	4	Total
Points:	20	30	25	25	100
Score:					

## Exercise 1: Brain Power (20 points)

- (a) (5 points) Estimate the average energy consumption in the human brain using the approximate numbers given in the course
- per action potential and
  - per synaptic event
- (b) (5 points) For simulating 1 s in a network consisting of  $1.23 \times 10^9$  neurons, each with an average firing rate of 1 Hz and an average number of  $6 \times 10^3$  presynaptic partners, the K supercomputer needs 40 minutes ([http://www.riken.jp/en/pr/press/2013/20130802\\_1/](http://www.riken.jp/en/pr/press/2013/20130802_1/)). During this time, it consumes 12.6 MW of power. Calculate the energy consumption per spike and per synaptic event of this simulation.
- (c) (5 points) Scale up the simulation in b) to the size of the human brain (in terms of both, number of synapses and number of neurons). Assuming one could simply stack K computers to be able to perform this simulation. How much power would they require?
- (d) (5 points) Now assume that the energy/flop is constant and that you could magically speed up the K computer to real-time.
- How much power would it consume? Compare your result to the output of a typical nuclear power plant in Japan (1780 MW)<sup>1</sup>. How many reactors would you then need?

## Exercise 2: Stimulating currents (30 points)

Use the simplified equivalent circuit of a LIF neuron from the course to calculate the effect of various stimulus currents on the membrane potential of a cell. Sketch the results.

- (a) (10 points) A Dirac delta function:  $I(t) = I_0 \cdot \frac{cm}{g_l} \delta(t - t_0)$ .
- (b) (10 points) A step current:  $I(t) = I_0 \cdot \Theta(t - t_0)$ , where  $\Theta$  represents the Heaviside step function.

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<sup>1</sup>Sendai Nuclear Power Plant, c.f. [https://en.wikipedia.org/wiki/Nuclear\\_power\\_in\\_Japan#Nuclear\\_power\\_plants](https://en.wikipedia.org/wiki/Nuclear_power_in_Japan#Nuclear_power_plants)

- (c) (10 points) An exponential current:  $I(t) = I_0 \cdot \Theta(t - t_0) \cdot \frac{c_m}{\tau_s g_l} \cdot \exp(-\frac{t-t_0}{\tau_s})$ . Later in the lecture, we will discuss why this is a good approximation of the synaptic input current following a spike.

Hint: From your result from c) you should obtain the result from a) in the limit of very short synaptic time constants ( $\tau_s \rightarrow 0$ ).

### Exercise 3: Introduction to Python (25 points)

Get familiar with python for the course. There are several ways to run python code. You can directly install Python 3 on your computer (c.f. <https://wiki.python.org/moin/BeginnersGuide/Download>), or you can use a online source e.g. google colab notebooks (c.f. <https://colab.research.google.com>).

- (a) (10 points) If you have not yet worked with Python, reproduce the first 7 chapters of the Python tutorial <https://docs.python.org/3.7/tutorial/>. (You can omit sections 2.2, 4.7, 4.8, 5.8, 6.4 since they discuss more advanced topics) It is assumed that you successfully solved this task.
- (b) (10 points) Implement the Sieve of Eratosthenes [https://en.wikipedia.org/wiki/Sieve\\_of\\_Eratosthenes](https://en.wikipedia.org/wiki/Sieve_of_Eratosthenes), an ancient algorithm for finding prime numbers.
- (c) (5 points) Using your implementation, find the sum of all the primes below two million (<https://projecteuler.net/problem=10>).

### Exercise 4: Basic data analysis with Python (25 points)

Use Python to do the following:

- (a) (10 points) draw a sample set of 1000 samples from a Gaussian distribution with mean 1.0 and standard deviation 0.5.  
(Hint: Use the `numpy.random` package)
- (b) (5 points) check the mean and standard deviation of the set  
(Hint: Use `numpy.mean` and `numpy.std`)
- (c) (5 points) plot a histogram over the set  
(Hint: Use `matplotlib.pyplot.hist`)
- (d) (5 points) fit a Gaussian to the distribution of the set and compare its moments to those of the original distribution  
(Hint: Use `scipy.optimize.curve_fit`)