

Engineering Tripos Part IIB, 4G3: Computational Neuroscience, 2025-26

Module Leader

[Prof Máté Lengyel](#) [1]

Lecturers

[Prof G Hennequin, Dr Y Ahmadian and Prof M Lengyel](#) [2]

Timing and Structure

Lent term. 16 lectures. Assessment: 100% coursework

Prerequisites

3G2 and 3G3 is useful but not essential

Aims

The aims of the course are to:

- develop an understanding of the fundamentals of reinforcement learning, and how they relate to neural and behavioural data on the ways in which the brain learns from rewards
- demonstrate the importance of internal models in neural computations, and provide examples for their behavioural and neural signatures
- introduce alternative ways of modelling single neurons, and the way these single neuron models can be integrated into models of neural networks.
- explain how the dynamical interactions between neurons give rise to emergent phenomena at the level of neural circuits
- describe models of plasticity and learning and how they apply to the basic paradigms of machine learning (supervised, unsupervised, reinforcement) as well as pattern formation in the nervous system
- demonstrate case studies of computational functions that neural networks can implement

Objectives

As specific objectives, by the end of the course students should be able to:

- understand how neurons, and networks of neurons can be modelled in a biomimetic way, and how a systematic simplification of these models can be used to gain deeper insight into them.
- develop an overview of how certain computational problems can be mapped onto neural architectures that solve them.
- recognise the essential role of learning is the organisation of biological nervous systems.
- appreciate the ways in which the nervous system is different from man-made intelligent systems, and their implications for engineering as well as neuroscience.

Content

The course covers basic topics in computational neuroscience, and demonstrates how mathematical analysis and ideas from dynamical systems, machine learning, optimal control, and probabilistic inference can be applied to gain

insight into the workings of biological nervous systems. The course also highlights a number of real-world computational problems that need to be tackled by any 'intelligent' system, as well as the solutions that biology offers to some of these problems.

Principles of Computational Neuroscience (9L, M Lengyel)

- introduction: the goals of computational neuroscience, levels of analysis, and module plan
- reinforcement learning: theoretical background and basic theorems, alternative algorithmic solutions and multiple learning & memory systems, model-based vs. model-free computations, the temporal difference learning theory of dopamine responses
- internal models: theoretical framework, internal models in perception, sensori-motor control, statistical learning, structure learning, neural correlates, neural representations of uncertainty, representational learning
- associative memory: the Hebbian paradigm, attractor neural networks, the Hopfield network, energy function, capacity, place cells, place cells, long-term plasticity, and navigation, place cell remapping

Network dynamics & Plasticity (4L, G Hennequin)

- linear and non-linear network dynamics
- spiking neural network dynamics
- excitatory-inhibitory balance
- chaotic dynamics
- network mechanisms of selective amplification
- orientation tuning in primary visual cortex

Plasticity & Biophysics (3L, Y Ahmadian)

- Hebbian plasticity
- spike timing-dependent plasticity
- learning receptive fields
- biophysical models of single neurons
- biophysical models of simple circuits

Further notes

See the [Moodle page for the course](#) [3] for more information (e.g. handouts, coursework assignments).

Examples papers

N/A

Coursework

Coursework	Format	Due date & marks
Coursework activity #1: network dynamics Most computations in the brain are implemented in networks of recurrently coupled neurons. In this coursework you will build simple neural network models and understand how they give rise to emergent dynamical and computational properties. <u>Learning objective:</u>	Individual report Anonymously marked	Posted week 6 Due week 6 [30/60]

Coursework	Format	Due date & marks
<ul style="list-style-type: none"> • implement simple neural networks and understand the effects of eigenvalues and eigenvectors on the resulting dynamics • implement balanced neural circuits and understand how asynchronous and irregular activity is generated 		
Coursework activity #2: synaptic plasticity and representational learning The brain constantly reconfigures itself via synaptic plasticity to develop useful representations of its inputs. In this coursework you will build and analyse simple models to understand some of the basic principles underlying this process <u>Learning objective:</u> <ul style="list-style-type: none"> • implement simple models of synaptic plasticity and analyse how they lead to pattern formation in feedforward and recurrent networks • implement a divisive normalisation model of visual cortical responses and analyse how it achieves efficient coding of natural image inputs and explains non-classical receptive field effects in the responses of simple cells in the primary visual cortex 	Individual Report Anonymously marked	Posted week Due two weeks [30/60]

See the [Moodle page for the course](#) [3] for more information (e.g. handouts, coursework assignments).

Booklists

Please refer to the Booklist for Part IIB Courses for references to this module, this can be found on the associated Moodle course.

Examination Guidelines

Please refer to [Form & conduct of the examinations](#) [4].

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Links

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[2] <mailto:gjuh2@cam.ac.uk>, ya311@cam.ac.uk, ml468@cam.ac.uk

[3] <http://tinyurl.com/cuedcompneuro>

[4] <https://teaching25-26.eng.cam.ac.uk/content/form-conduct-examinations>