**MA665: Introduction to Modeling and Data Analysis in Neuroscience (Fall 2018)**

**Instructor:** Mark Kramer (mak@math.bu.edu)

**Course Hours:** September 4, 2018 – October 18, 2018 [7 weeks]

Tuesday & Thursday, 12:30-1:45 PM, CAS 330

**Office Hours:** Math & Stats Room 224 by appointment

**Textbook:** None

**Course Website:**      https://github.com/Mark-Kramer/BU-MA665-MA666

**Prerequisites:** Graduate standing or consent of instructor.

This course is intended to introduce neuroscience graduate students to mathematical concepts in neuroscience. We will use experimental observations in neuroscience to motivate the study of mathematics, both to quantify the observed data and build computational models. The course will focus on six fundamental topics in mathematical neuroscience, with emphasis on quantifying neurophysiological time series and developing mathematical models of the activity observed. An important component of the course will include an introduction to scientific computing. Students completing this course will develop computational skills essential in interdisciplinary neuroscience research and in more advanced neuroscience courses offered at BU.

**Course goals**

To introduce mathematical concepts encountered in neuroscience research and more advanced neuroscience graduate courses. To teach basic programming skills. To think about neuroscience in quantitative ways.

**Course requirements**

The main requirement in this course is effort. I expect your full effort during our course meetings, and outside of the course, to meet the course objectives. As part of this course, you will work together in teams of 3-4. We’ll establish those teams during the first week of MA665. You will work with your team to complete a series of examinations in this course.

**Grades**

To earn an A in MA665, your team must pass oral examinations in six topics:

* Programming proficiency
* The integrate and fire neuron, and its extensions
* The Hodgkin-Huxley neuron
* The evoked response potential, and bootstrap resampling
* The power spectrum
* The coherence

You may choose to take one exam during any class. During an exam, I will ask your team a series of questions about your chosen topic. These questions will include: basic facts, conceptual questions, and practical programming implementation questions. Each question will be posed to one team member – the selected team member will answer the question, typically without assistance from other team members. Please do your best to answer each question. If your team successfully answers each question, then your team completes the exam. If not, I will recommend topics for additional study. You may repeat an exam as many times as you like during the course, but you may only take one exam per lab.

**Schedule**

Please begin with Week 1, and progress in the order of weeks. While you will complete this course at your own pace, I recommend you complete all 6 oral exams by the end of MA665. I recommend viewing online lectures and completing readings outside of lab. During lab, I recommend discussing the suggested assignments, asking questions, and completing the oral exams.

Completing the exams will provide you with the minimum requirements for an introduction to a subset of topics in computational neuroscience. If you complete all exams, then you may pursue a topic of your choice. This might include a survey of different topics in computational neuroscience, or a deep dive into a specific topic. For example, if you are new to Python, you might continue to build your knowledge in this area. Or, if you are interested in computational modeling, you might investigate other models or methods of modeling. Or, if you are interested in data analysis, you might explore traditional or novel methods, and their practical application. If you complete all six exams, let’s chat.

**Suggested Course Schedule**

Readings: These papers provide a general context and motivation for the application of mathematics and statistics to problems in neuroscience. Please read these papers at your leisure during the next 7 weeks.

* Gerstner, W., Sprekeler, H., & Deco, G. (2012). *Theory and simulation in neuroscience*. Science, 338(6103), 60–65.
* Marder, E. (2015). *Understanding brains: details, intuition, and big data*. PloS Biology, 13(5), e1002147
* Churchland, A. K., & Abbott, L. F. (2016). *Conceptual and technical advances define a key moment for theoretical neuroscience*. Nature Neuroscience, *19*(3), 348–349.

Advanced Reading: This paper provides a big-picture discussion of computational neuroscience and related topics. Important ideas to consider for your future career in neuroscience.

* Brain 2025: A Scientific Vision, 2014

Depressing reading:

* *Saving Science: Science isn’t self-correcting, it’s self-destructing*. Daniel Sarewitz, 2016.
* Button, K., et al (2013). *Power failure: why small sample size undermines the reliability of neuroscience.* Nature Reviews Neuroscience, 14(5), 365–376.

**Week 4**

Sept 25: Class discussion: The integrate and fire neuron, and its extensions

Sept 27: Computer challenges

Materials

Reading: (web) W. Gerstner, the integrate-and-fire model.

Link: <http://lcn.epfl.ch/~gerstner/SPNM/node26.html>

(pdf) Abbott, Brain Res Bull (1999) vol. 50 (5-6) pp. 303-4  
(pdf) Chapter 8, pages 267-269 @ E. Izhikevich, *Dynamical Systems in Neuroscience*, 2007.

(pdf) Chapter 1, pages 5-12 @ C. Koch, *Biophysics of computation*, 1998.

(pdf) Chapter 14, pages 330-341 @ C. Koch, *Biophysics of computation*, 1998.

Videos: (lec) M. Kramer, *Introduction to the integrate and fire model* (Neural Spike Train Analysis 4)  
 NOTE: Slides available as PDF on GitHub.

Link: <https://www.samsi.info/news-and-media/27-jul-drs-m-kramer-and-u-eden-samsi/>

(lec) W. Gerstner, *Passive membrane and Integrate-and-Fire model (a)*

Link: <http://klewel.com/conferences/epfl-neural-networks/index.php?talkID=1>

Code (pynb) [Integrate and Fire Model](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/beta%20versions/Integrate%20and%20Fire%20Model)

**Week 5**

Oct 2: Class discussion: The Hodgkin-Huxley neuron

Oct 4: Computer challenges

Materials

Reading: (pdf) Chapter 2, pages 25-42 @ E. Izhikevich, *Dynamical Systems in Neuroscience*, 2007.

(pdf) Chapter 6, pages 142-159 @ C. Koch, *Biophysics of computation*, 1998.

(pdf) Hodgkin-Huxley 1-page cheat sheet

(web) W. Gerstner, *Detailed Neuron Models* (sections 2.1 & 2.2)

Link: <http://lcn.epfl.ch/~gerstner/SPNM/node12.html>

(pdf) Advanced: Hodgkin and Huxley, J Physiol (Lond) (1952) vol. 117 (4) pp. 500-44.

Videos: (lec) M. Kramer, *Introduction to the Hodgkin-Huxley neuron* (**Neural Spike Train Analysis 5**)

NOTE: Slides available as PDF on GitHub.

Link: <https://www.samsi.info/news-and-media/27-jul-drs-m-kramer-and-u-eden-samsi/>

(lec) W. Gerstner, *Detailed Neuron Model (a)*

Link: <http://klewel.com/conferences/epfl-neural-networks/index.php?talkID=4>

(lec) W. Gerstner, *Detailed Neuron Model (b)*

Link: <http://klewel.com/conferences/epfl-neural-networks/index.php?talkID=5>

Code: (pynb) [Hodgkin Huxley Model](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/beta%20versions/Hodgkin%20Huxley%20Model)

**Week 6**

Oct 9: No class - Columbus Day, substitute Monday schedule

Oct 11: Computer challenges: The event-related potential

Materials

Readings: (pdf) Chapter 2 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

Tutorial: (pynb) [The Event-Related Potential](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/The%20Event-Related%20Potential)

**Week 7**

Oct 16: Class discussion: The power spectrum

Oct 18: Computer challenges

Materials

Readings: (pdf) Chapter 3 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

(pdf) Kramer, SFN Short Course Document.  
 (pdf) Chapter 10 @ M. X. Cohen, *Analyzing neural time series data*,2014.

(pdf) Chapter 11 @ M. X. Cohen, *Analyzing neural time series data*,2014.

(lib) **Advanced**: Chapter 4 @ Percival & Walden, *Spectral Analysis for Physical Applications.*

Tutorial: (pynb) [Analysis of Rhythmic Activity in the Scalp EEG](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/Analysis%20of%20Rhythmic%20Activity%20in%20the%20Scalp%20EEG)

**MA666: Advanced Modeling and Data Analysis in Neuroscience (Fall 2016)**

**Instructor:** Mark Kramer (mak@math.bu.edu)

**Course Hours:** October 23, 2018 – December 11, 2018 [7 weeks]

Tuesday & Thursday, 12:30-1:45 PM, CAS B25B

**Office Hours:** Math & Stats Room 224 by appointment

**Textbook:** None

**Course Website:**https://github.com/Mark-Kramer/BU-MA665-MA666

**Prerequisites:** Graduate standing or consent of instructor.

**Course goals**

The goal of this course is further study topics in computational neuroscience. The typical format of this course will be lecture on Tuesday, and computer lab on Thursday. We will continue to focus on three broad areas of computational neuroscience: (1) computer programming, (2) data analysis, (3) modeling. You are encouraged to continue working in teams.

**Course requirements**

The main requirement in this course is effort. I expect your full effort during our course meetings, and outside of the course, to meet the course objectives. As part of this course, you may work together in teams of 2-4. You will work with your team to complete assignments related to the topics in computational neuroscience

**Grades**

To earn an “A”, the general requirements are: attend lectures and labs, complete all assigned challenge problems; provide feedback on the GitHub repository.

**Schedule (tentative)**

**Week 8**

Oct 23: Class discussion: coherence

Oct 25: Dr. Eric Denovellis: Best programming practices.

Materials

Readings: (pdf) Chapter 5 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

(pdf) Kramer, SFN Short Course Document.

(pdf) Chapter 25 @ M. X. Cohen, *Analyzing neural time series data*,2014.

(pdf) Chapter 26 @ M. X. Cohen, *Analyzing neural time series data*,2014.

Tutorial: (pynb) [Analysis of Coupled Rhythms](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/Analysis%20of%20Coupled%20Rhythms)

**Week 9**

Oct 30: Computer challenges: coherence

Nov 1: Class discussion: Cross-frequency coupling

Materials

Readings: (pdf) Chapter 7 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

(pdf) Tort et al, J Neurophysiol, 2010.

(pdf) Hyafil et al, Trends Neurosci, 2015.

Tutorial: (pynb) [Cross-Frequency-Coupling](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/Cross-Frequency-Coupling)

**Week 10**

~~Nov 6: No class SFN~~

Nov 8: Class discussion: Cross-frequency coupling

Materials

Readings: (pdf) Chapter 7 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

(pdf) Tort et al, J Neurophysiol, 2010.

(pdf) Hyafil et al, Trends Neurosci, 2015.

Tutorial: (pynb) [Cross-Frequency-Coupling](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/Cross-Frequency-Coupling)

**Week 11**

Nov 13: Computer challenges: Cross-frequency coupling

Nov 15: Class discussion: Spike-field coherence

Materials

Readings: (pdf) Chapter 11 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

(pdf) [Advanced] Lepage et al, Neural Computation, 2011.

(pdf) Pesaran et al, Nat Neuro, 2018.

Tutorial: (pynb) [Spike-Field Coherence](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/Spike-Field%20Coherence)

**Week 12**

Nov 20: Computer challenges: Spike-field coherence

~~Nov 22: No class Thanksgiving~~

Materials

Readings: (pdf) Chapter 11 @ Kramer & Eden, *Case studies in neural data analysis*, 2016.

(pdf) [Advanced] Lepage et al, Neural Computation, 2011.

(pdf) Pesaran et al, Nat Neuro, 2018.

Tutorial: (pynb) [Spike-Field Coherence](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/Spike-Field%20Coherence)

**Week 13**

Nov 27: Class discussion: Neural networks and learning

Nov 29: Computer challenges: Neural networks and learning

Materials

Reading: (pdf) Lecture slides

(pdf) The Nature of Code, Chapter 10. Neural Networks

Videos: (mp4) [MIT 6.034 F10 Artificial Intelligence, Lecture 12A](https://archive.org/details/MIT6.034F10/MIT6_034F10_lec12A_300k.mp4)

(mp4) [MIT 6.034 F10 Artificial Intelligence, Lecture 12B](https://archive.org/details/MIT6.034F10/MIT6_034F10_lec12B_300k.mp4)

Tutorial: (pynb) [Training a Perceptron](https://github.com/Mark-Kramer/Case-Studies-Python/tree/master/beta%20versions/Training%20a%20Perceptron)

**Week 14 & 15**

Dec 4: Backpropagation

Dec 6 & 11: Filtering data & convolution  
OR Real-life models: gamma (ING, PING, sparse PING).

Materials

Reading: (pdf) Lecture slides

Videos: (mp4) [MIT 6.034 F10 Artificial Intelligence, Lecture 12](https://archive.org/details/MIT6.034F10/MIT6_034F10_lec12_300k.mp4)

Tutorial: (pynb) Backpropagation