BEE 4850/5850 (Environmental Data Analysis and Simulation) Syllabus Spring 2024

Course Overview

This is a 3 credit course offered as an elective, which counts towards the Environmental Systems and Data Analytics focus area.

Instructor	TA	${f Meetings}$
• 🚨 Vivek Srikrishnan	• 🚨 TBD	• # MWF
• viveks@cornell.edu	• 🔀 TBD	• Q 1:25-2:15pm
• 🛱 318 Rilev-Robb	• 🛱 TBD	• 🏛 TBD

Course Description

Simulations from numerical and statistical models are a key tool used to quantify and represent our understanding of many different environmental systems, including but not limited to the climate, sea levels, air pollution, and the electric power system. Data analysis of environmental observations and model simulations is an integral part of developing and evaluating models used to 1) test our assumptions about environmental system dynamics; 2) develop new insights into environmental processes; and 3) project future environmental conditions and outcomes. understand system dynamics and project future conditions and outcomes. This course will provide an overview of the use of simulation model development and evaluation in environmental data analysis. The goal is to provide students with a framework and an initial toolkit of methods that they can use to formulate and update hypotheses about data and models. Students will actively analyze and use real data from a variety of environmental systems. In particular, over the course of the semester, we will:

- conduct exploratory analyses of environmental datasets;
- discuss best practices for and complexities of data visualization;
- calibrate statistical and process-based numerical models using environmental data;
- use simulations from calibrated models to identify key sources of uncertainty and model error;
- assess model fit and adequacy through predictive ability.

Learning Outcomes

After completing this class, students will be able to:

- create, interpret, and critique data visualizations;
- calibrate environmental models to observations, possibly including censored and missing data;
- simulate alternative datasets from models using statistical methods such as the bootstrap and Monte Carlo;
- assess model adequacy and performance using predictive simulations;
- apply and contextualize model selection criteria;
- evaluate evidence for and against hypotheses about environmental systems using model simulations;

emulate computationally-complex models with simpler representations.

Prerequisites & Preparation

The following courses/material would be ideal preparation:

- One course in programming (e.g. CS 1110, 1112 or ENGRD/CEE 3200)
- One course in probability or statistics (ENGRD 2700, CEE 3040, or equivalent)

In the absence of one or more these prerequisites, you can seek the permission of instructor.

What If My Programming or Stats Skills Are Rusty?

If your programming or statistics skills are a little rusty, don't worry! We will review concepts and build skills as needed.

Typical Topics

- Introduction to exploratory data analysis;
- Review of probability and statistics;
- Bayesian decision theory;
- Principles of data visualization;
- Model residuals and discrepancies;
- Censored, truncated, and missing data;
- Statistical methods for calibration;
- Predictive model assessment;
- Emulation with surrogate models

Course Meetings

This course meets MWF from 11:40-12:55 in 160 Riley-Robb Hall. In addition to the course meetings (a total of 42 lectures, 50 minutes each), the final project will be due during the university finals period. In addition to the work during the semester, students can expect to devote, on average, 4 hours of effort during the exam period.

Course Philosophy and Expectations

The goal of our course is to help you gain competancy and knowledge in the area of data analysis. This involves a dual responsibility on the part of the instructor and the student. As the instructor, my responsibility is to provide you with a structure and opportunity to learn. To this end, I will commit to:

- provide organized and focused lectures, in-class activities, and assignments;
- encourage students to regularly evaluate and provide feedback on the course;
- manage the classroom atmosphere to promote learning;
- schedule sufficient out-of-class contact opportunities, such as office hours;
- allow adequate time for assignment completion;
- make lecture materials, class policies, activities, and assignments accessible to students.

I encourage you to discuss any concerns with me during office hours or through a course communications channel! Please let me know if you do not feel that I am holding up my end of the bargain.

Students can optimize their performance in the course by:

- attending all lectures;
- doing any required preparatory work before class;
- actively participating in online and in-class discussions;
- beginning assignments and other work early;

• and attending office hours as needed.

Community

Diversity and Inclusion

Our goal in this class is to foster an inclusive learning environment and make everyone feel comfortable in the classroom, regardless of social identity, background, and specific learning needs. As engineers, our work touches on many critical aspects of society, and questions of inclusion and social justice cannot be separated from considerations of how data are generated, collected, and analyzed.

In all communications and interactions with each other, members of this class community (students and instructors) are expected to be respectful and inclusive. In this spirit, we ask all participants to:

- share their experiences, values, and beliefs;
- be open to and respectful of the views of others; and
- value each other's opinions and communicate in a respectful manner.

Please let me know if you feel any aspect(s) of class could be made more inclusive. Please also share any preferred name(s) and/or your pronouns with me if you wish: I use he/him/his, and you can refer to me either as Vivek or Prof. Srikrishnan.

Please, Be Excellent To Teach Other

We all make mistakes in our communications with one another, both when speaking and listening. Be mindful of how spoken or written language might be misunderstood, and be aware that, for a variety of reasons, how others perceive your words and actions may not be exactly how you intended them. At the same time, it is also essential that we be respectful and interpret each other's comments and actions in good faith.

Student Accomodations

Let me know if you have any access barriers in this course, whether they relate to course materials, assignments, or communications. If any special accommodations would help you navigate any barriers and improve your chances of success, please exercise your right to those accommodations and reach out to me as early as possible with your Student Disability Services (SDS) accommodation letter. This will ensure that we have enough time to make appropriate arrangements.

! Important

If you need more immediate accommodations, but do not yet have a letter, please let me know and then follow up with SDS.

Course Communications

Most course communications will occur via Ed Discussion. Public Ed posts are generally preferred to private posts or emails, as other students can benefit from the discussions. If you would like to discuss something privately, please do reach out through email or a private Ed post (which will only be viewable by you and the course staff).

Announcements will be made on the course website and in Ed. Emergency announcements will also be made on Canvas.

l Ed Tips

- If you wait until the day an assignment is due (or even late the previous night) to ask a question on Ed, there is a strong chance that I will not see your post prior to the deadline.
- But if you see unanswered questions and you have some insight, please answer! This class will work best when we all work together as a community.

Mental Health Resources

We all have to take care of our mental health, just as we would our physical health. As a student, you may experience a range of issues which can negatively impact your mental health. Please do not ignore any of these stressors, or feel like you have to navigate these challenges alone! You are part of a community of students, faculty, and staff, who have a responsibility to look for one another's well-being. If you are struggling with managing your mental health, or if you believe a classmate may be struggling, please reach out to the course instructor, the TA, or, for broader support, please take advantage of Cornell's mental health resources.

Mental Health And This Class

I am not a trained counselor, but I am here to support you in whatever capacity we can. You should never feel that you need to push yourself past your limits to complete any assignment for this class or any other. If we need to make modifications to the course or assignment schedule, you can certainly reach out to me, and all relevant discussions will be kept strictly confidential.

Course Policies

Attendance

Attendance is not required, but in general, students who attend class regularly will do better and get more out of the class than students who do not. Your class participation grade will reflect both the quantity and quality of your participation, only some of which can occur asynchronously. I will put as many course materials, such as lecture notes and announcements, as possible online, but viewing materials online is not the same as active participation and engagement. Life happens, of course, and this may lead you to miss class. Let me know if you need any appropriate arrangements ahead of time.

What If I'm Sick?

Please stay home if you're feeling sick! This is beneficial for both for your own recovery and the health and safety of your classmates. We will also make any necessary arrangements for you to stay on top of the class material and if whatever is going on will negatively impact your grade, for example by causing you to be unable to submit an assignment on time.

Mask Policies

Masks are encouraged but not required in the classroom, per university policy. However, the University strongly encourages compliance with requests to mask from students, faculty, and staff who are concerned about the risk of infection. Please be respectful of these concerns and requests if you cannot wear a mask.

Masks Will Be Required In My Office

I will **require masks to be worn** in my office or during in-person office hours, as we are necessarily interacting in close quarters without great airflow.

Academic Integrity

Important

TL;DR: Don't cheat, copy, or plagiarize!

This class is designed to encourage collaboration, and students are encouraged to discuss their work with other students. However, I expect students to abide by the Cornell University Code of Academic Integrity in all aspects of this class. All work submitted must represent the students' own work and understanding, whether individually or as a group (depending on the particulars of the assignment). This includes analyses, code, software runs, and reports. Engineering as a profession relies upon the honesty and integrity of its practitioners (see e.g. the American Society for Civil Engineers' Code of Ethics).

External Resources

The collaborative environment in this class **should not be viewed as an invitation for plagiarism**. Plagiarism occurs when a writer intentionally misrepresents another's words or ideas (including code!) as their own without acknowledging the source. **All** external resources which are consulted while working on an assignment should be referenced, including other students and faculty with whom the assignment is discussed. You will never be penalized for consulting an external source for help and referencing it, but plagiarism will result in a zero for that assignment as well as the potential for your case to be passed on for additional disciplinary action.

AI/ML Resource Policy

As noted, all work submitted for a grade in this course must reflect your own understanding. The use and consulation of AI/ML tools, such as ChatGPT or similar, must be pre-approved and clearly referenced. If approved, you must:

- reference the URL of the service you are using, including the specific date you accessed it;
- provide the exact query or queries used to interact with the tool; and
- report the exact response received.

Failure to attain prior approval or fully reference the interaction, as described above, will be treated as plagiarism and referred to the University accordingly.

Late Work Policy

In general, late work will be subjected to a 10% penalty per day, which can accumulate to 100% of the total grade. However, sometimes things come up in life. Please reach out ahead of time if you have extenuating circumstances (including University-approved absences or illnesses) which would make it difficult for you to submit your work on time. Work which would be late for appropriate reasons will be given extensions and the late penalty will be waived.

Assessments

Technologies

We will use Canvas as a gradebook, and to distribute PDFs of readings (which also be made available through the website, via the Cornell library). Ed Discussion will be used for course communications. Assignments

will be submitted and graded in Gradescope.

Students can use any programming language they like to solve problems, though we will make notebooks and package environments available for Julia. If students use a language other than Julia, they may limited in the programming assistance they receive (though we're happy to try to help!).

Grading Scale

The following grading scale will be used to convert the numerical weighted average from the assessments (below) to letter grades

Grade	Range
A	93-100
A-	90 – 92
B+	87 - 89
В	83 - 86
В-	80 – 82
C+	77 - 79
\mathbf{C}	73 - 76
C-	70 - 72
D+	67 – 69
D	63 – 66
D-	60 – 62
F	< 59

Application Exercises: 10%

Most weeks, students will be given a set of exercises (typically involving analyzing a dataset, a model, or a figure) to complete. These will involve a small amount of programming, a minor calculation, and/or visual assessment of data or a figure. Application exercises will be provided the previous Monday and are intended to align with the content for the week, and solutions should be submitted by 9:00pm on the Friday at the end of the given week. These exercises will be given as quizzes on Gradescope and will be auto-graded. The lowest exercise score will be dropped automatically.

Readings: 10%

Readings will be assigned for discussion throughout the semester. Students in BEE 5850 will be expected to write one-page critiques of the papers; while students in BEE 4850 will be expected to contribute to the online discussions.

Paper Critique and Presentation: 20%

Students will select a peer-reviewed journal article related to an application of data analysis and will write a short discussion paper analyzing the hypotheses and statistical choices. Students enrolled in BEE 5850 will also write a referee report, as if they were a peer reviewer. Students will give an in-class presentation of their paper during one class period. Other students will be asked to submit evaluations of the presentations.

Homework Assignments: 30%

Approximately 6 homework assignments will be assigned throughout the semester (roughly one per course module). You will typically have 2 weeks to work on each assignment, though this depends on the module length. Students are encouraged to collaborate and learn from each other on homework assignments, but each student must submit their own solutions reflecting their understanding of the material. Consulting and

referencing external resources and your peers is encouraged (engineering is a collaborative discipline!), but plagiarism is a violation of academic integrity.

Some notes on assignment and grading logistics:

- Homework assignments will be distributed using GitHub Classroom. Students should make sure they
 update their GitHub repositories as they work on the assignments; this helps with answering questions
 and gives you a backstop in case something goes wrong and you can't submit your assignment on time.
- Homeworks are due by 9:00pm Eastern Time on the designed due date. Your assignment notebook (which include your writeup and codes) should be submitted to Gradescope as a PDF with the answers to each question tagged (a failure to do this will result in deductions).
- Rubrics will be provided for the homeworks as part of the assignments.
- Students in 5850 will be asked to complete additional homework problems which go more deeply into the underlying concepts or apply more advanced techniques.
- Your lowest homework grade will be dropped. We can discuss arrangements if multiple assignments will be missed for university-approved reasons, preferably ahead of time.
- Regrade requests for specific problems must be made within a week of the grading of that assignment. However, note that regrades can cut both ways: the TA can take away points as well!

Term Project: 30%

Throughout the semester, students will apply the concepts and methods from class to a data set of their choosing. If a student does not have a data set in mind, we will find one which aligns with their interests.

The term project can be completed individually or in groups of 2. Students will provide updates throughout the semester corresponding to the various tasks discussed in each module and will submit a final report at the end of the semester. The deliverables are:

- A proposal describing the research question and hypotheses, the data set, and the numerical or statistical models the student would like to use to test the hypotheses;
- A simulation study applying simulation methods to the models;
- A calibration exercises applying calibration and uncertainty quantification methods from class;
- Assessments of the model performance based on the data;
- A final presentation and report.

Tentative Schedule

#	Date	Day	Topic
			Introduction to Exploratory Data Analysis
1	01-22	Mon	Class overview
2	01-24	Wed	Intro to Julia and GitHub
3	01-26	Fri	Lab 1: Julia Basics
4	01-29	Mon	Exploratory data analysis and data-generating processes
5	01-31	Wed	Simulation-based workflows for data analysis
			Probability and Statistics Background and Review
6	02-02	Fri	Uncertainty and probability basics
7	02 - 05	Mon	Probability distributions
8	02 - 07	Wed	Probabilistic models
9	02-09	Fri	Lab 2: Probability Models in Julia
10	02 - 12	Mon	Bayes' Theorem
11	02 - 14	Wed	Bayesian decision analysis
Data Visualization and Graphical Checks			
12	02 - 16	Fri	Psychology of memory and vision
13	02 - 19	Mon	Information channels and Gestalt principles
14	02-21	Wed	Class discussion of figures

#	Date	Day	Topic	
15	02-23	Fri	Lab 3: Plotting in Julia	
	02-26	Mon	February Break	
16	02-28	Wed	Approaches to visualizing data and uncertainty	
17	03-01	Fri	Graphical checks for model diagnostics	
			Model Calibration	
18	03-04	Mon	Understanding model residuals	
19	03-06	Wed	Model-data discrepancy	
20	03-08	Fri	Missing and censored data	
21	03-11	Mon	Lab 4: Missing Data	
22	03-13	Wed	Modeling complex residual structures	
$Simulation \ Methods$				
23	03-15	Fri	Introduction to Monte Carlo	
24	03-18	Mon	Monte Carlo error analysis	
25	03-20	Wed	Monte Carlo simulation: Examples	
26	03-22	Fri	Sampling distributions and the bootstrap	
27	03-25	Mon	Lab 5: Monte Carlo	
28	03-27	Wed	The bootstrap: Examples	
29	03-29	Fri	Overview of Bayesian computation	
	04-01	Mon	Spring Break	
	04-03	Wed	Spring Break	
	04-05	Fri	Spring Break	
30	04-08	Mon	Markov chains	
31	04-10	Wed	Markov chain Monte Carlo	
32	04-12	Fri	Markov chain Monte Carlo and convergence	
33	04 - 15	Mon	Lab 6: Markov chain Monte Carlo	
34	04 - 17	Wed	Markov chain Monte Carlo: Examples	
Model Selection				
35	04-19	Fri	Model assessment via predictive accuracy	
36	04-22	Mon	Cross-validation	
37	04-24	Wed	Lab 7: Model Assessment	
38	04-26	Fri	Predictive information criteria	
39	04-29	Mon	Model assessment and selection: Examples	
			Emulation and Surrogate Modeling	
40	05-01	Wed	Tradeoffs between complexity and ensemble size	
41	05-03	Fri	Emulation	
42	05-06	Mon	Emulation, cont'd	