

NE 795 Advanced Topics in Nuclear Engineering Advanced Scientific Machine Learning

Fall 2023, 3 Credit Hours

1 Course Information

• Schedule

- *Time:* Tuesdays and Thursdays, 03:00 PM - 04:15 AM
- *Location:* Room 330, Dabney Hall
- *Course website:* **Moodle** will be used to post lecture notes, homework, computer projects and other materials.
<https://moodle-courses2324.wolfware.ncsu.edu/course/view.php?id=3901>
- *Lecture recordings:* This is the first time this course being offered. We will not record the lectures.

• Instructor

- *Instructor:* Dr. Xu Wu, Assistant Professor of Nuclear Engineering
- *Office:* Burlington Laboratory 2110
- *Office Hours:* Mondays, 1:00 - 3:00 PM, in-person, or Zoom (by appointment)
- *Phone:* 919-515-6570
- *Email:* xwu27@ncsu.edu
- *Website:* <https://www.ne.ncsu.edu/people/xwu27>

• Teaching Assistant

- *TA:* N/A
- *Email:* N/A
- *Office Hours:* N/A

2 Course Description

- **Scientific machine learning (SciML)** is a burgeoning discipline in Artificial Intelligence and Machine Learning (AI/ML) that blends scientific computing and ML. SciML is a core component of ML. It consists of computational technologies that can be trained with scientific data to augment or automate human skills.
- This course aims at augmenting the applications of AI/ML in scientific computing especially in the nuclear engineering area, and preparing the students for transformative solutions across various DOE missions.
- After this course, the students will learn advanced variations of deep neural networks and deep generative modeling techniques. The students will be able to implement advanced SciML algorithms in nuclear engineering scientific computing applications.

3 Student Learning Outcomes

- The students completing this course will be able to:
 - demonstrate understanding of information theory, variational inference and deep generative modeling;
 - evaluate the performance of advanced SciML models;
 - build ML models for advanced variations of neural networks and deep generative modeling;
 - use open source ML libraries such as scikit-learn, keras, TensorFlow and PyTorch.
 - solve research problems using the advanced SciML techniques learned in this class;
 - develop research skills that combine the complementary perspectives of computational science and computer science to craft a new generation of ML methods for complex applications across nuclear engineering.

4 Prerequisites

- Prior experience in Machine Learning is required, equivalent to the NE 795 - SciML class.
- **Python** is the recommended programming language.

5 Outline of Topics

1. Introduction and Short Review of ANNs (1 lecture)
2. Fundamentals of Information Theory (2 lectures)
3. Clustering (3 lectures)
4. Time Series Analysis and Forecasting (2 lectures)
5. Convolutional Neural Networks (ConvNets) (2 lectures)
6. Recurrent Neural Networks (RNNs) (2 lectures)
7. Variational Inference Theory (2 lectures)
8. Bayesian Neural Networks (BNNs) (1 lecture)
9. Generative Adversarial Networks (GANs) (2 lectures)
10. Variational Autoencoders (VAEs) (2 lectures)
11. Normalizing Flows (2 lectures)
12. Diffusion Models (2 lectures)

6 Course Schedule

- Detailed course schedule will be available in a separate document, including dates for each section/topic, homework/project assignment dates and deadlines, midterm exam date, etc.

7 Assignments and Grading Policy

(1) Computational projects (100%)

- The computer project will be evaluated by a **proposal**, a **progress report**, a **final report** and a **final presentation**. Each of the milestones will be submitted through Moodle. See Table 1 for the schedule.
- You may work by yourself or in teams (max. two students) for the project. There will be two options.
- **Option A (recommended)**: each group needs to propose a computer project based on the research of the group members. It is expected to be in the areas of SciML applications in prediction, uncertainty, sensitivity, calibration and validation, etc.
- **Option B**: if a team is unable to determine a proper computer project, the instructor will assign a published paper to the team. The team is expected to reproduce the results in the paper. If the codes are already available online from the original authors, the students will be asked to solve a different problem that is pertinent to Nuclear Engineering applications.

Table 1: Project schedules

Milestones	Weights of total	Deadlines
Proposal	10%	09/30/2023
Progress report	20%	10/29/2023
Final presentation	35%	11/xx/2023
Final report	35%	12/xx/2023

- The **proposal** should be 1-2 pages. It should contain the following information: (1) project title, (2) team member(s), (3) description of the problem to be solved with ML, (4) scope of the research to be performed and ML methodologies to be used, (5) anticipated results, and (6) reading list (important papers you will need to read).
- The **progress report** should include (1) a high quality introduction (including literature survey on similar work published in the literature), (2) what have you done so far, (3) what remains to be done and (4) a clear description of the division of work among teammates, if applicable.
- The **final report** should be 10 - 15 pages, including tables/figures/references/appendix. You should submit a PDF file electronically. The codes should be submitted separately in a Zip file. The report should use the format similar to a research paper. It should include (1) introduction, including the motivation, the issue to be addressed, the objective, the literature review and a quick summary of the problem/research/findings, (2) problem definition, (3) the methodologies, (4) ML implementation and major results, (5) discussions and conclusions, including comments on the results and open questions, and (6) major references.
- The **final presentation** will be in person or online via Zoom. Depending on how many teams we have, we will use the last two lectures' time for the presentation. Each project will have **15 minutes** to present including Q/A.

(2) Grading

- The course will be graded on the letter grading scale listed in Table 2 and will count toward your GPA. More information can be found at <https://studentservices.ncsu.edu/your-grades/general-info/>.
- *Requirements for Credit-Only (S/U) Grading:* In order to receive a grade of S, students are required to take all exams and quizzes, complete all assignments, and earn a grade of C- or better. Conversion from letter grading to credit only (S/U) grading is subject to university deadlines. Refer to the Registration and Records calendar for deadlines related to grading. For more details refer to <http://policies.ncsu.edu/regulation/reg-02-20-15>.
- *Requirements for Auditors (AU):* Information about and requirements for auditing a course can be found at <http://policies.ncsu.edu/regulation/reg-02-20-04>.
- *Policies on Incomplete Grades:* If an extended deadline is not authorized by the instructor or department, an unfinished incomplete grade will automatically change to an F after either (a) the end of the next regular semester in which the student is enrolled (not including summer sessions), or (b) the end of 12 months if the student is not enrolled, whichever is shorter. Incompletes that change to F will count as an attempted course on transcripts. The burden of fulfilling an incomplete grade is the responsibility of the student. The university policy on incomplete grades is located at <http://policies.ncsu.edu/regulation/reg-02-50-3>.

Table 2: Letter grades.

Letter grades	Percentages	GPA's
A+	96-100%	4.333
A	92-96%	4.0
A-	89-92%	3.667
B+	86-89%	3.333
B	82-86%	3.0
B-	79-82%	2.667
C+	76-79%	2.333
C	72-76%	2.0
C-	69-72%	1.667
D+	65-69%	1.333
D	61-65%	1.0
D-	56-61%	0.667
F	0-56%	0

8 Recommended Texts

There is no required text but self-contained lecture notes will be posted. Some useful references are:

(1) For Statistics

- Robert, C., & Casella, G. (2013). Monte Carlo statistical methods. Springer Science & Business Media.

- Gelman, A., Carlin, J., Stern, H., and Rubin, D. (2014). Bayesian Data Analysis, 3rd Edition, Chapman & Hall.
<http://www.stat.columbia.edu/~gelman/book/>
- Sivia, D., & Skilling, J. (2006). Data analysis: a Bayesian tutorial. OUP Oxford.
<http://aprsa.villanova.edu/files/sivia.pdf>

(2) For Machine Learning and Deep Learning

- Goodfellow, I., Bengio, Y. and Courville, A. (2016). Deep Learning, MIT Press.
<https://www.deeplearningbook.org/>
- Hastie, T., Tibshirani, R., & Friedman, J. (2009). The elements of statistical learning: data mining, inference, and prediction. Springer Science & Business Media.
<https://web.stanford.edu/~hastie/ElemStatLearn/>
- Murphy, K. P. (2012). Machine learning: a probabilistic perspective. MIT press.
<https://www.cs.ubc.ca/~murphyk/MLbook/>
- Aston Zhang, Zack C. Lipton, Mu Li, Alex J. Smola (2020), Dive into Deep Learning - an interactive deep learning book with code, math, and discussions
<https://d21.ai/>
- Eli Stevens and Luca Antiga (2019), Deep Learning with PyTorch - Essential Excerpts, Manning Publications
<https://www.manning.com/books/deep-learning-with-pytorch#toc>

(3) For Python programming

- Shaw, Z. A. (2017). Learn python 3 the hard way: A very simple introduction to the terrifyingly beautiful world of computers and code. Addison-Wesley Professional.
- Scipy Lecture Notes - One document to learn numerics, science, and data with Python
<https://scipy-lectures.org/>
- McKinney, W. (2017). Python for data analysis: Data wrangling with Pandas, NumPy, and IPython. O'Reilly Media, Inc.
<https://www.oreilly.com/library/view/python-for-data/9781491957653/>
<https://github.com/wesm/pydata-book>

(4) For scientific computing

- Smith, R. C. (2013). Uncertainty quantification: theory, implementation, and applications (Vol. 12). SIAM.
https://rsmith.math.ncsu.edu/UQ_TIA/
- Saltelli, A., Ratto, M., Andres, T., Campolongo, F., Cariboni, J., Gatelli, D., ... & Tarantola, S. (2008). Global sensitivity analysis: the primer. John Wiley & Sons.

9 Others

(1) Late Assignments Policy

- Unless stated otherwise, assignments are due by the end of day on the designated due date. Assignments turned in within 24 hours of this time are considered late and will be assessed a 25% penalty. Assignments turned in after 24 hours will be marked and returned to the student, but no credit will be assigned.
- To allow for unforeseen circumstances, each student will be granted a **one-time exemption**. The student should contact the course instructor **at least three days** before the original deadline and explain the situation in order to get an extension. The assignment must be turned in by the end of the new designated due date.

(2) Course Attendance/Absence Policy

- Required; Active class participation is strongly encouraged.
- NC State attendance policies can be found at: REG 02.20.03 - Attendance Regulations - Policies, Regulations & Rules (<https://policies.ncsu.edu/regulation/reg-02-20-03-attendance-regulations/>). Please refer to the course's attendance, absence, and deadline policies for additional details.
- *Absences Policy*: Personal Problems: We understand that sometimes life makes it difficult to focus on schoolwork. If you are having a personal problem that affects your participation in this course, please talk to us to create a plan. Please do not wait until the end of the semester to share any challenges that have negatively impacted your engagement and academic performance. The sooner we connect, the more options we will have available to us to support your overall academic success. If you are not comfortable speaking with us directly, please utilize the other student resources provided below in order to understand how to best approach success in this course given your personal needs as soon as possible.

(3) Transportation

- This course will not require students to provide their own transportation. Non-scheduled class time for field trips or out-of-class activities is NOT required for this class.

(4) Safety & Risk Assumptions

- N/A

(5) Digital Course Components

- This on campus course will be captured and distributed via the Internet and/or electronic media as part of the Engineering Online (EOL) program for the distance students.
- These video recordings may contain an image of you entering the classroom, asking a question or being a part of the studio class.
- Please notify Dr. Linda Krute, Director of EOL, at ldkrute@ncsu.edu if you do NOT want your image to be included in the lecture presentation. If EOL does not hear from you after the first week of the class, we will assume that you are in agreement with this procedure.
- Students may be required to disclose personally identifiable information to other students in the course, via digital tools, such as email or web-postings, where relevant to the course. Examples include online discussions of class topics, and posting of student coursework. All students are expected to respect the privacy of each other by not sharing or using such information outside the course.

(6) Academic Integrity

- Students are required to comply with the university policy on academic integrity found in the Code of Student Conduct found at <https://policies.ncsu.edu/policy/pol-11-35-01/>
- Absolutely no collaboration is permitted during closed-book tests. All the tests are closed book unless otherwise specified.
- Collaboration on homework assignments is allowed, but the submitted work must be your own individual work. Homework assignments must not be treated as group assignments. Zero grade will be assigned for particular homework for the first offense. Second offense will be reported to the Office of Student Conduct.
- Violations of academic integrity will be handled in accordance with the Student Discipline Procedures (NCSU REG 11.35.02) at <https://policies.ncsu.edu/regulation/reg-11-35-02/>. 11.35.02)

(7) Additional NC State Rules and Regulations

- Students are responsible for reviewing the NC State University Policies, Rules, and Regulations (PRRs) which pertain to their course rights and responsibilities, including those referenced both below and above in this syllabus:
 - Equal Opportunity and Non-Discrimination Policy Statement <https://policies.ncsu.edu/policy/pol-04-25-05/> with additional references at <https://oied.ncsu.edu/divweb/policies/>.
 - Code of Student Conduct Policy <https://policies.ncsu.edu/policy/pol-11-35-01/>

(8) Use of Electronic Devices in Class

- Cell phones are to be turned OFF prior to entering the classroom/lab. No exceptions.
- Use of laptops/other electronic devices during class is permitted only for the purpose of following the posted lecture materials/taking electronic notes.

(9) Accommodations for Students with Disabilities

- Reasonable accommodations will be made for students with verifiable disabilities. In order to take advantage of available accommodations, students must register with Disability Services for Students at 1900 Student Health Center, Campus Box 7509, 515-7653.
- For more information on NC State's policy on working with students with disabilities, please see the "REG 02.20.01 - Academic Accommodations for Students with Disabilities" at <https://policies.ncsu.edu/regulation/reg-02-20-01/>.

(10) Non-Discrimination Policy

- NC State provides equal opportunity and affirmative action efforts, and prohibits all forms of unlawful discrimination, harassment, and retaliation ("Prohibited Conduct") that are based upon a person's race, color, religion, sex (including pregnancy), national origin, age (40 or older), disability, gender identity, genetic information, sexual orientation, or veteran status (individually and collectively, "Protected Status").

- Additional information as to each Protected Status is included in NCSU REG 04.25.02 (Discrimination, Harassment and Retaliation Complaint Procedure). NC State's policies and regulations covering discrimination, harassment, and retaliation may be accessed at <http://policies.ncsu.edu/policy/pol-04-25-05> or <https://oied.ncsu.edu/divweb/>.
- Any person who feels that he or she has been the subject of prohibited discrimination, harassment, or retaliation should contact the Office for Equal Opportunity (OEO) at 919-515-3148.

(11) Student Mental Health

- As a student you may experience a range of personal issues that can impede learning, such as strained relationships, increased anxiety, alcohol/drug concerns, feeling down, difficulty concentrating and/or lack of motivation. These mental health concerns or stressful events may lead to diminished academic performance and may impact your ability to participate in daily activities. It is very important that you have a support system and that you ask for help when you are struggling. The Counseling Center at NC State offers confidential mental health services for full time NC State students, including same-day emergency services. Please visit <https://counseling.dasa.ncsu.edu/> to get connected.