

# EE P 567 A Wi 24: Machine Learning For Cybersecurity

## EEP 567 2024: Machine Learning for Cybersecurity

Class will meet in person every Wednesday from 6pm to 9pm, with online meeting available.

Classroom: ECE 269

Zoom link for instruction: <https://washington.zoom.us/j/92244122483>

Join the course Slack for updates and Q&As: <https://shorturl.at/cdfor>

Instructor: Prof. Radha Poovendran ([rp3@uw.edu](mailto:rp3@uw.edu))

Instructor Office Hours: By appointment and via Zoom

TAs: Qifan Lu (Lead TA; [lqf96@uw.edu](mailto:lqf96@uw.edu)), Fengqing Jiang ([fqjiang@uw.edu](mailto:fqjiang@uw.edu)),  
Yuchen Wu ([yuchenw@uw.edu](mailto:yuchenw@uw.edu))

TA Office Hours (Tentative): 6pm to 8pm every Tuesday and Thursday, and 2pm to 4pm every Sunday

Zoom link for TA Office Hours: (TBA)

## Course Introduction

This course will study the use of machine learning for cybersecurity applications. Many security applications have large amounts of data related to the system and adversarial actions. Our ability to identify the type of machine learning algorithms useful for specific security applications can help us improve the defense against attacks and anticipate the potential attack variants that may arise. Even when one does not know the type of attack, if the machine learning algorithms can identify any anomaly, then the next level of security checks can be performed by other means or experts.

But nothing comes in life for free, and this holds for machine learning in the context of cyber, too! Another point to remember is, "Beware of what you add to your tool bag! You may have an adversary manipulating your machine learning itself." This leads to adversarial machine learning, where the machine learning could be tricked into failing the detection!

Attacks on Google Video, Toxic Comments, and Google Vision are fun examples of simple modifications that make the machine learning algorithms fail!

With the advent of transformers and LLMs, new opportunities are available for defenders and attackers. We will look into some of the materials along the course.

We will start with setups where machine learning will be helpful in cybersecurity. As indicated in the course title, it will be a hands-on course on applying machine learning to cybersecurity applications. Machine learning algorithms will be introduced as needed.

## Lectures

1. Introduction to Machine Learning (ML) for CyberSecurity
2. Email Spam Detection using Supervised Learning (**Scheduled Release of Homework #1**)
3. Knocking down CAPTCHAs
4. Transformers in CyberDefense (**Scheduled Release of Homework #2**)
5. Efficient Network Anomaly Detection Using k-means
6. Use of NLP for Instruction Set Architecture Identification from Binaries (**Scheduled Release of Homework #3**)
7. Malicious Event Detection with Decision Tree
8. Financial Fraud and How Deep Learning Can Mitigate It
9. Adversarial Machine Learning
10. Student Presentations of Course Projects

## Textbook

***Hands-On Machine Learning for Cybersecurity by Soma Halder and Sinan Ozdemir; Packt Publisher.*** You can [read it online](#) for free with your UW NetID. Alternatively, you can obtain physical copies from the UW library or the University bookstore.

## References

1. ***AI for CyberSecurity by Alessandro Parisi; Packt Publishers***
2. ***Machine Learning for Cybersecurity by Chiheb Chebbi; Packt Publishers***
3. ***Machine Learning for Penetration Testing by Emmanuel Tsukerman; Packt Publishers***
4. ***Adversarial Machine Learning by Vorobeychik and Kantarcioglu; Morgan and Claypool Publishers***

## Labs

Throughout the course, we will have eight labs where you will learn and practice various machine-learning algorithms and use them to solve cybersecurity problems. Below is a tentative schedule for all lab sessions:

1. Python basics review and introduction of typical data analysis libraries
2. Machine learning pipeline for cybersecurity problems
  - Case study: spam email detection
3. A small step into deep learning and convolutional neural network (CNN)

- Case study: breaking Captchas with neural network
- 4. Dimensionality reduction and data visualization
  - Case study: network anomaly detection and visualization
  - Dataset: KDD Cup 1999 dataset (We will reuse these in lab 6)
- 5. Autoencoder and clustering algorithm
- 6. Data oversampling and decision tree algorithm
  - Case study: detecting and categorizing network attacks
  - Dataset: Kaggle credit card fraud detection dataset (We will reuse these in lab 8)
- 7. Ensemble learning

## Homeworks / Project

We plan to have three homework throughout the whole course. Throughout the quarter, you will also get to form groups with others to work on a course project **we may suggest or your team could suggest**. See some preliminary [Course Project](#) pages for ideas, requirements, and details. **We have no homework in the last two weeks so you can focus on the project.**

## Grading

- Homework: 45% (15% each)
- Project: 55%
  - Proposal: 5%
  - Checkpoint report: 10%
  - Final presentation: 15%
  - Final report: 25%

## Course Announcements and Discussions

- Announcements
  - We will post course announcements and updates to the **class mailing list** ([multi\\_eep567a\\_wi24@uw.edu](mailto:multi_eep567a_wi24@uw.edu)) and occasionally **Canvas**. These announcements will also be forwarded to the **Slack workspace** used by this course (see below).
- Discussions – Slack
  - We will use Slack workspace "EEP567-24" for course discussions. Feel free to use this Slack to post your questions on the course material, lectures, homework, and projects, or answer questions from others if you can help. Instructors and TAs will also answer these questions promptly.
  - **Join the course Slack:** <https://shorturl.at/cdfor>

## Course Policy

- Please complete the homework by yourself and do not copy code from others or the internet without understanding what it is doing. Suppose your homework is identical

to others or any sample code snippets online; in that case, you will receive zero scores, and we must report it to the College of Engineering.

- You are encouraged to discuss lab and homework content with your classmates offline and on the discussion board. However, please **limit the scope of your question** and do not discuss about the answers directly. Specifically, **do not copy-paste any homework answers** into the discussion board.
- You are allowed to use AI assistants (such as ChatGPT or Google Bard) for help. In this case, you should **acknowledge your use of AI assistant**. Again, you should **not copy-paste from the response of the AI assistant**.
- You should submit homework and project materials online by the posted due date. Throughout the quarter, **we will provide you with six late-day credits for the three homework assignments**, which you can use to submit one or more homework without a penalty. However, if you have used all six days, each additional late day will result in a 20% penalty in the corresponding homework.
- You must submit project reports and presentations strictly on time. **Any overdue project materials will receive a zero score**. The project does not have any credit days.
- Washington state law requires that UW develop a policy for the accommodation of student absences or significant hardship due to reasons of faith or conscience or for organized religious activities. The University of Washington policy, including information about requesting accommodation, is available at Faculty Syllabus Guidelines and Resources. Accommodations must be requested within the first two weeks of this course using the Religious Accommodations Request form available at <https://registrar.washington.edu/students/religious-accommodations-request/>.