CS 4824/ECE 4424: Machine Learning

Website: https://people.cs.vt.edu/dbhattacharya/courses/cs4824/

Piazza: https://piazza.com/vt/spring2024/cs4824ece4424/home

Canvas: https://canvas.vt.edu/courses/185332 (CS 4824)

https://canvas.vt.edu/courses/185672 (ECE 4424)

Are you in the right place?

- This is CS 4824/ECE 4424 (ML): CRN 13430
 - Modality is "Face-to-Face Instruction"
- There is another section of the same course...
 - But with different CRN

Today

- What is Machine Learning, the discipline, about?
 - The *trifecta* of ML
- What is this class about?
 - What to expect?
 - Logistics

What are we here to discuss?

A wide range of topics Machine Learning covering their intuitions, mathematical foundations, analyses, and applications!

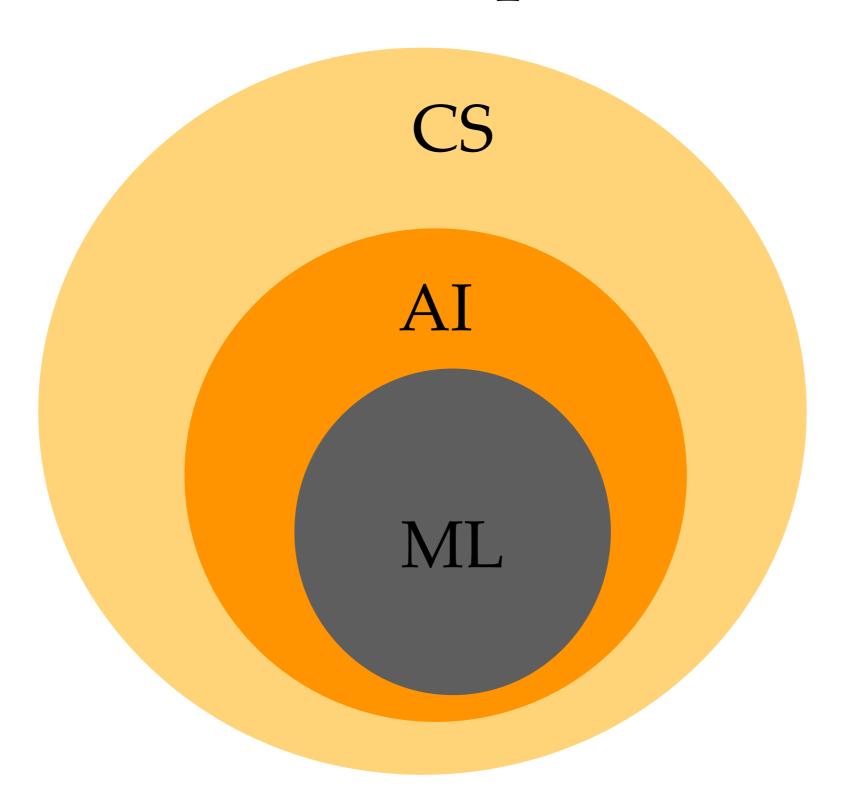
Demo time...

https://openai.com/blog/openai-codex/

Live demo:

https://youtu.be/SGUCcjHTmGY

Concepts



Intelligence

General Intelligence

 The ability to navigate in problem space.—Siddhartha Mukherjee, Columbia

Artificial Intelligence

 The science and engineering of making computers behave in ways that, until recently, we thought required human intelligence. —Andrew Moore, CMU

What is machine learning?

- Study of algorithms that
 - improve their performance (P)
 - at some task (T)
 - with experience (E)

Well-defined learning task: <P, T, E>

The ML niche is expanding...

- Machine learning is central to
 - Computer Vision
 - Natural Language Processing
 - Speech Recognition
 - Biology
 - Robotics
 - • •
- Machine learning has a conducive environment to grow in the 21st century
 - Accumulation of "Big Data"
 - Demand for Self-customization to User
 - Post Moore's Law Computing
 - 0

Spam classification

From: cheapsales@buystufffromme.com

To: ang@cs.stanford.edu

Subject: Buy now!

Deal of the week! Buy now!
Rolex w4tchs - \$100
Medlcine (any kind) - \$50
Also low cost M0rgages
available.

Span

From: Alfred Ng

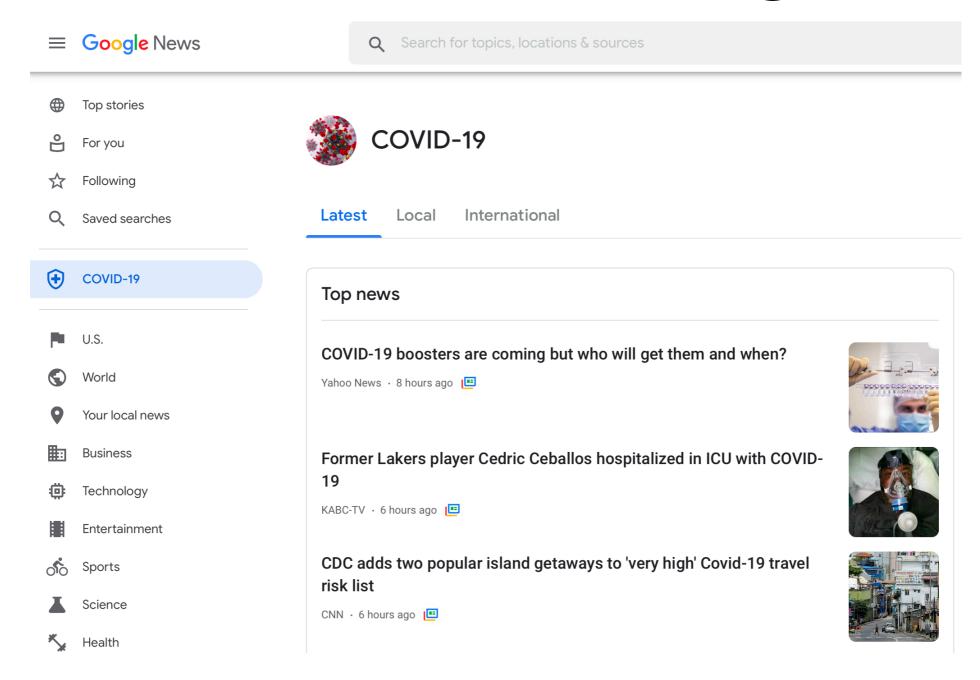
To: ang@cs.stanford.edu Subject: Christmas dates?

Hey Andrew,
Was talking to Mom about plans
for Xmas. When do you get off
work. Meet Dec 22?
Alf

Non-sbow

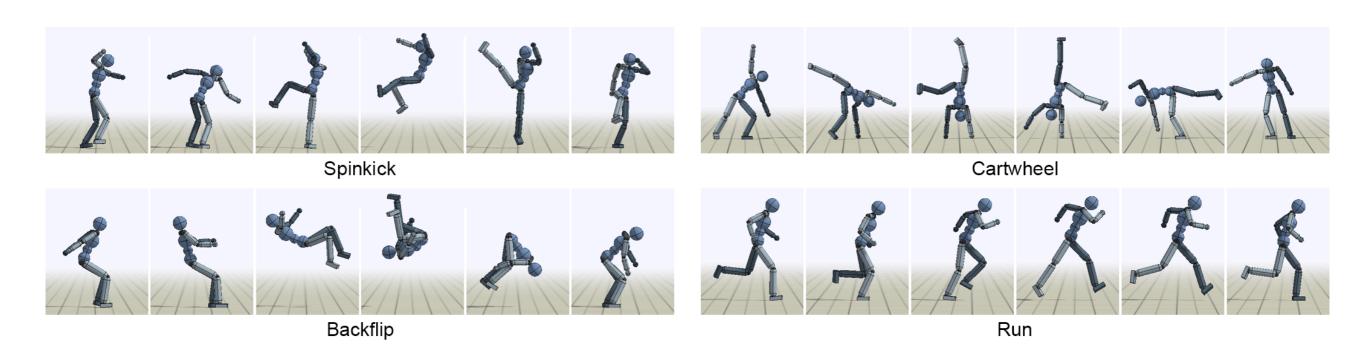
Supervised learning

News clustering



Unsupervised learning

Motion planning



Reinforcement learning

The trifecta of ML

Supervised learning

- Perform inductive inference and produce an inferred function to emulate mapping between input and output
- Supervisory signal in the form of example input-output pairs (a.k.a. labeled training data)

Unsupervised learning

- Group together data into categories having similar features without requiring pre-assigned category labels
- Self-discover any naturally occurring patterns in (unlabeled) data sets

Reinforcement learning

- Collect rewards for "good" actions while avoiding penalty for "bad" actions
- Take action to explore uncharted territory and exploit experience

What this class is about?

Introduction to Machine Learning

Goal

 After finishing this course, you should be ready to get started on your first ML research project

Target audience

Senior undergrads, BS/MS students

What this class is NOT

NOT the target audience

- Grad students already working in ML area
- Folks looking to understand the most recent breakthroughs (e.g., AlphaGo, AlphaFold, etc.)

NOT the goal

Teaching a toolkit (e.g., TensorFlow/PyTorch)

Topics*

*Tentative and subject to change

- Basics of Statistical Learning
 - Loss functions, MLE, MAP, Bayesian estimation, bias-variance tradeoff, overfitting, regularization, cross-validation
- Supervised Learning
 - Decision Trees, Naïve Bayes, Logistic Regression, Linear Regression, Kernels and Kernel Regression, Support Vector Machines, Neural Networks
- Unsupervised Learning
 - EM, Clustering
- Graphical Models
 - Bayesian Networks, Hidden Markov Models
- Deep Learning
 - Convolutional Neural Networks, Recurrent Neural Networks, Attention and Transformer Networks, Autoencoders, Variational Autoencoders, Generative Adversarial Networks
- Advanced Topics
 - Generative AI, Diffusion Probabilistic Models

Textbook

- None required.
- Optional reference books (freely available online):
 - Machine Learning: a Probabilistic Perspective, Kevin Murphy, MIT Press,
 2012
 - Pattern Recognition and Machine Learning, Christopher Bishop, Springer,
 2006
 - The Elements of Statistical Learning, Trevor Hastie, Robert Tibshirani, and Jerome Friedman, Springer, 2009
 - Deep Learning, Ian Goodfellow, Yoshua Bengio, and Aaron Courville, MIT Press, 2016

Prerequisites

Probability and Statistics

- Distributions, densities, marginalization, moments, typical distributions.
- Calculus and Linear Algebra
- Matrix multiplication, eigenvalues, positive semi-definiteness, multivariate derivates.

Algorithms

Dynamic programming, basic data structures, complexity.

Programming

- This is a demanding class in terms of programming skills. HWs will involve a mix of Python and libraries. You are free to choose any programming language for the project.
- Ability to deal with abstract mathematical concepts.

Organization and Deliverables

Homeworks

- Hands-on implementation and analysis, covering various topics that complement and supplement the lecture topics.
- 4 individual HWs (15% \times 4 = 60%). Involve a mix of Python and libraries.
- 2 HWs before Spring break, 2 after Spring break.
- Start early, Start early

Course Project

- The course project is meant for students to (1) gain experience implementing machine learning models; and (2) try machine learning on problems that interest them.
- The project must be done individually in this semester (i.e., no double counting).
- Project Proposal (10%): Due on March 1 (i.e., before Spring break)
- Project Final Report (20%): Due on May 1 (i.e., on the last day of classes)
- Project midway progress check (not graded, but submission is mandatory): Due on April 24

Class participation and Pop Quiz

- Contribute to class discussions on Piazza
- Engage in class: ask questions, answer questions
- In-class pop quizes (10%) requiring your class presence and overall engagement in the classroom.

• Final letter grade

- Computed based on ceiling of the final percentage of points earned. The grade ranges are as follows:
 - A: 90%-100% B: 80%-89%. C: 70%-79%. D: 60%-69%. F: Below 60%

Late policy for deliverables

- Late homework policy is as follows:
 - Full credit when due.
 - Half credit next 48 hours.
 - Zero credit after that.

Avoid invoking penalties by starting early and seeking help. No penalties for medical reasons or emergencies.

 Late submissions are NOT allowed for the project proposal, midway or the final report.

Project

Goal

- Chance to try Machine Learning
- Encouraged to try out interesting applications of machine learning in various domains such as vision, NLP, speech, computational biology, etc.

Computing

Google colab: jupyter-notebook + free GPU instance

Policies

Collaboration Policy

- All assignments are individual assignments
- You may discuss the questions
- Each student writes their own answer/code for all assignments
- Project proposal/report will be plagiarism checked

Regarding Policy

Within 1 week of receiving the grades

Academic integrity

- Students enrolled in this course are responsible for abiding by the Honor Code
- · Zero-tolerance philosophy regarding plagiarism or other forms of cheating

Principles of Community

• The course will include in-class discussions, and we will adhere to Virginia Tech Principles of Community.

Accessibility

- If any student needs special accommodations because of any disabilities, please contact the instructor during the first week of classes.
- Such students are encouraged to work with The Office of Services for Students with Disabilities to help coordinate accessibility arrangements.

COVID-19 Policy

 Please follow the instructions posted at the University and public health guidelines for the latest COVID-19 Policy.

Course Team — TA



Weijie (Jack) Guan

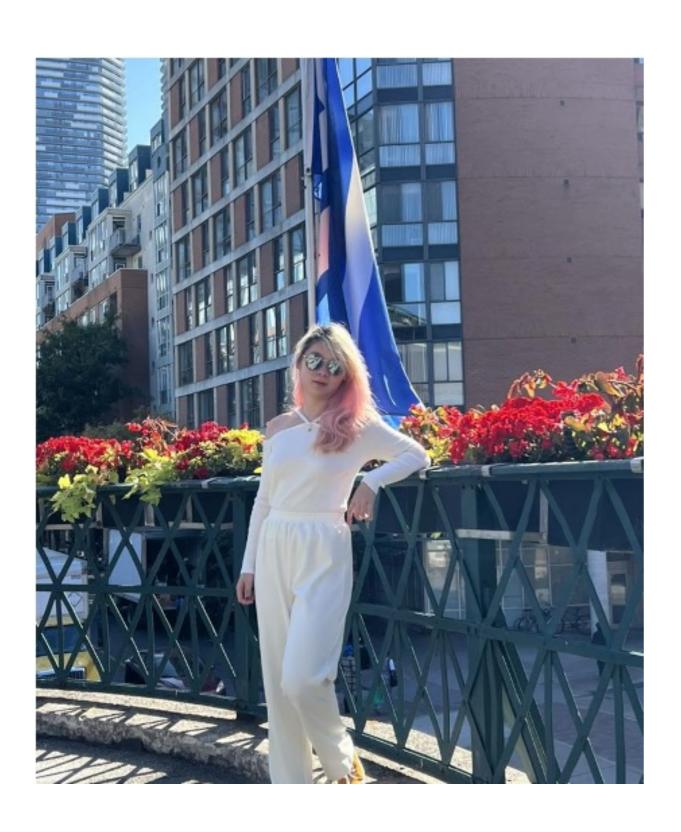
Ph.D. Student Computer Science Virginia Tech

Email: skjguan@vt.edu

Research interest lies in Graph Machine Learning, Trustworthy AI, General Out-of-distribution Detection

Office Hours: Tuesday and Thursday 1:00 pm - 2:00 pm **McBryde Hall 106** or via Zoom https://virginiatech.zoom.us/skype/9248463068

Course Team — TA



Jianan Nie

Ph.D. Student Computer Science Virginia Tech

Email: jianan@vt.edu

Research interest lies in AI for Science (Chemistry), which includes autonomous discovery, reaction prediction, molecular design, and drug discovery.

Office Hours: Tuesday and Thursday 2:00 pm - 3:00 pm at Gilbert Place 4112 or via Zoom https://virginiatech.zoom.us/j/2649120190

Course Team

Instructor

 Debswapna Bhattacharya (Office Hours: Monday and Wednesday 1:00 pm - 2:00 pm at Torgersen 3120B)

• TAs

- Weijie (Jack) Guan (Office Hours: Tuesday and Thursday 1:00 pm 2:00 pm McBryde Hall 106 or via Zoom https://virginiatech.zoom.us/skype/9248463068
- Jianan Nie (Office Hours: Tuesday and Thursday 2:00 pm 3:00 pm at Gilbert Place 4112 or via Zoom https://
 virginiatech.zoom.us/j/2649120190)

How to stay in touch?

• Primary means of communication —Piazza

- No direct email to instructor unless private information
- Instructor/TAs can provide answers to everyone on forum
- Extra credit for regularly (and correctly) answering questions on forum!
- No posting answers on forum. We will monitor.
- If you have a personal matter, create a private piazza post or send an email to the course staff.

Staff Mailing List

cs-4824-ece-4424-s24-staff-g@vt.edu

Todo: before next class

Go through the course webpage at: https://people.cs.vt.edu/dbhattacharya/courses/cs4824/

...and ask any questions in the next class.

Get into Piazza:

https://piazza.com/vt/spring2024/cs4824ece4424