

Mini Hackathon Event

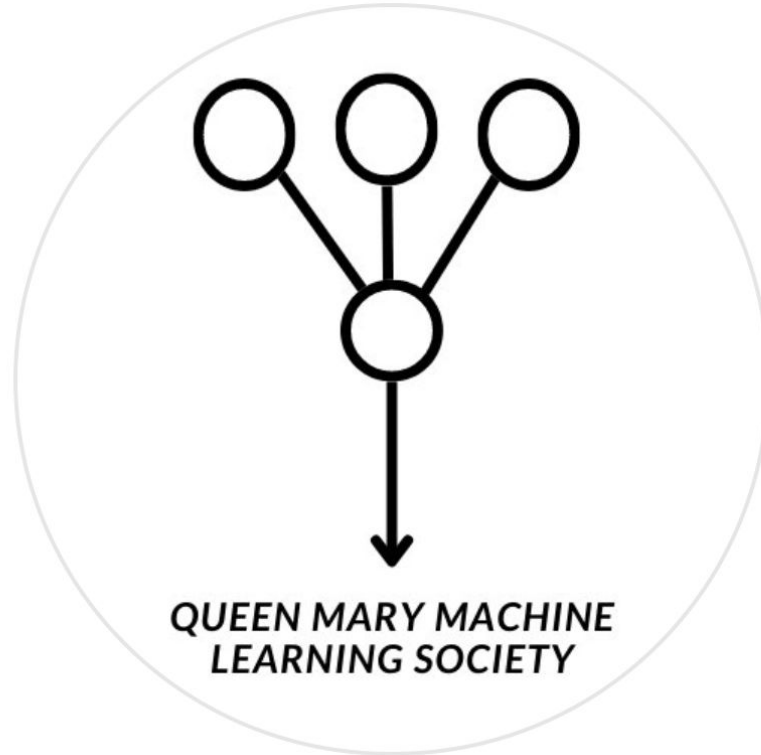
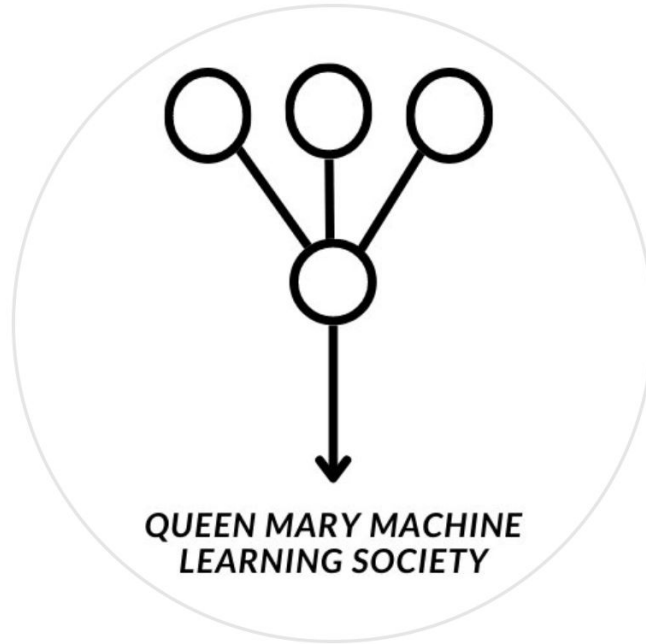


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The Machine Learning Roadmap



Let's talk about Chat-GPT...

Generative AI Can Harm Learning

Hamsa Bastani,^{1*} Osbert Bastani,^{2*} Alp Sungu,^{1*†}
Haosen Ge,³ Özge Kabakcı,⁴ Rei Mariman

¹Operations, Information and Decisions, University of Pennsylvania

²Computer and Information Science, University of Pennsylvania

³Wharton AI & Analytics, University of Pennsylvania

⁴Budapest British International School

*These authors (H.B., O.B., A.S.) contributed equally.

†To whom correspondence should be addressed; E-mail: alpsungu@wharton.upenn.edu.



Pandas



Foundational skills

$$\begin{bmatrix} 1 & x+1 & x^2+1 \\ 1 & y+1 & y^2+1 \\ 1 & z+1 & z^2+1 \end{bmatrix} \quad x = \sum_{i=1}^n x_i v_i = x_1 v_1 + x_2 v_2 + \dots + x_n v_n \quad v_k = y_k - \sum_{i=1}^{k-1} \frac{(v_i, y_k)}{(v_i, v_i)} v_i$$
$$\frac{p^T \nabla^2 F(x) p}{\|p\|^2} \quad F(x) = F(x^*) + \nabla F(x)^T |_{x=x^*} (x - x^*) + \frac{1}{2} (x - x^*)^T \nabla^2 F(x)^T |_{x=x^*} (x - x^*) + \dots$$
$$\nabla F(x) = \left[\frac{\partial}{\partial x_1} F(x) \quad \frac{\partial}{\partial x_2} F(x) \quad \dots \quad \frac{\partial}{\partial x_n} F(x) \right]^T$$
$$\begin{bmatrix} p_1^T \\ p_2^T \\ \vdots \\ p_Q^T \end{bmatrix}$$

LINEAR ALGEBRA

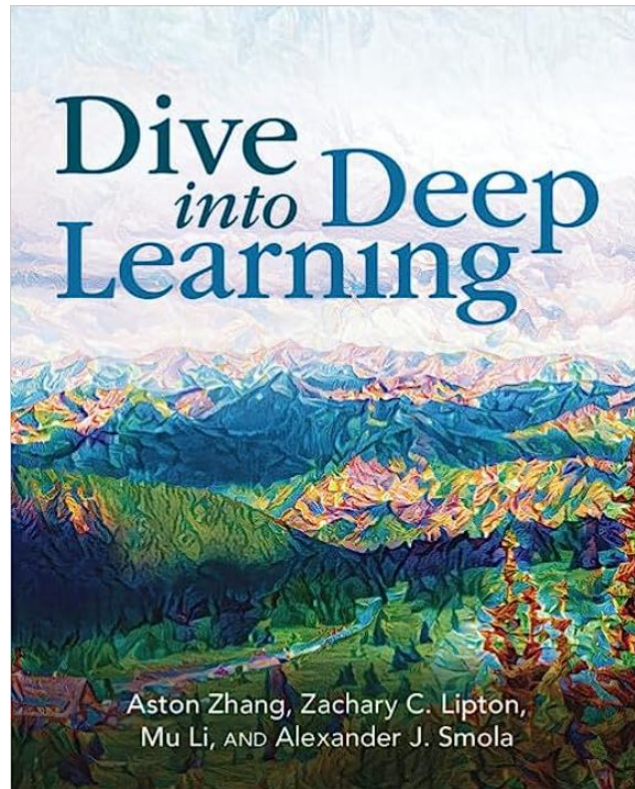
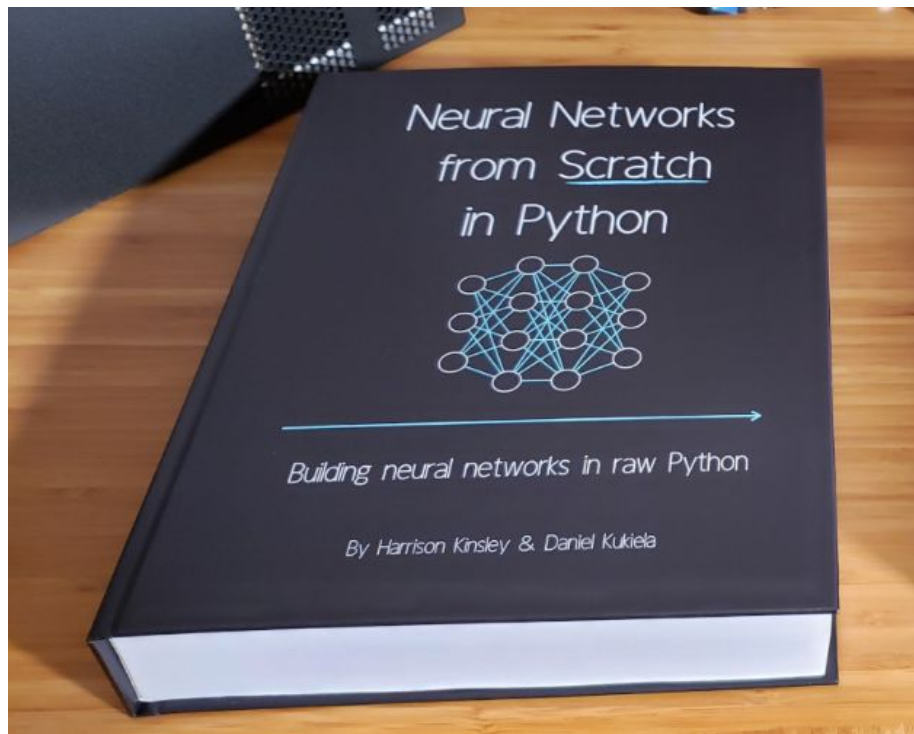
$$\begin{aligned} W^{new} &= (1 - \eta) W^{old} + \eta t_q p_q^T \\ W^{new} &= W^{old} + \eta (t_q - a_q) p_q^T \\ W^{new} &= W^{old} + \eta a_q p_q^T \end{aligned} \quad \begin{bmatrix} \frac{\partial}{\partial x_1} F(x) & \frac{\partial}{\partial x_2} F(x) & \dots & \frac{\partial}{\partial x_n} F(x) \\ \frac{\partial}{\partial x_1} F(x) & \frac{\partial}{\partial x_2} F(x) & \dots & \frac{\partial}{\partial x_n} F(x) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1} F(x) & \frac{\partial}{\partial x_2} F(x) & \dots & \frac{\partial}{\partial x_n} F(x) \end{bmatrix}$$

$$\frac{d}{dx} \left[\frac{f(x)}{g(x)} \right] = \frac{g(x)f'(x) - f(x)g'(x)}{g(x)^2} \quad F = mg = ma = m \frac{d^2 h}{dt^2} \quad m \frac{d^2 x}{dt^2} = -kx$$
$$\frac{dA}{dt} = \frac{dB}{dt} = \frac{dC}{dt} = \frac{dD}{dt} = (A)AB - (C)CD \quad y = mx + b \quad \text{Geometric William Leibniz}$$
$$\frac{dA}{dx} = \frac{dB}{dy} = \frac{dy}{dx} \quad \text{Minin Gudana Agnesi} \quad f(x) = x^2 \quad \int \sin x dx = -\cos x + c$$
$$(\ln x)' = \frac{1}{x} \quad \int \frac{1}{x} dx = \ln|x| + c \quad \int_a^b f'(x) dx = f(b) - f(a) \quad m \frac{d^2 x}{dt^2} = -kx \quad \frac{df(x)}{dz}$$

Calculus



Start by coding from scratch



High Quality Machine Learning Resources

coursera

Stanford

 DeepLearning.AI

Machine Learning Specialization

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CS109: CS109: Probability for Computer Scientists:

<https://web.stanford.edu/class/archive/cs/cs109/cs109.1232/handouts/syllabus.html>

CS229: Machine Learning:

<https://cs229.stanford.edu/syllabus-spring2022.html>

CS230 Deep Learning:

<https://cs230.stanford.edu/> - <https://cs230.stanford.edu/files/>

CS224N: NLP with Deep Learning:

<https://www.youtube.com/playlist?list=PLoROMvodv4rMFqRtEuo6SGjY4XbRIVRd4>

CS231N: Deep Learning for Computer Vision:

<https://www.youtube.com/playlist?list=PL5-TkQAfAZFbzxjBHtzdVCWE0Zbhong7r>

CS224W: Machine Learning with Graphs:

<https://www.youtube.com/playlist?list=PLoROMvodv4rPLKxlpqhjhPgqQy7imNkDn>

CS330: Deep Multi-Task and Meta Learning:

<https://www.youtube.com/watch?v=bkVCAk9Nsss&list=PLoROMvodv4rNjRoawgt72BBNwL2V7doGI>

CS236: Generative Models:

https://www.youtube.com/watch?v=MJt_ahtO-to&list=PLoROMvodv4rPOWA-omMM6STXaWW4FvJT8

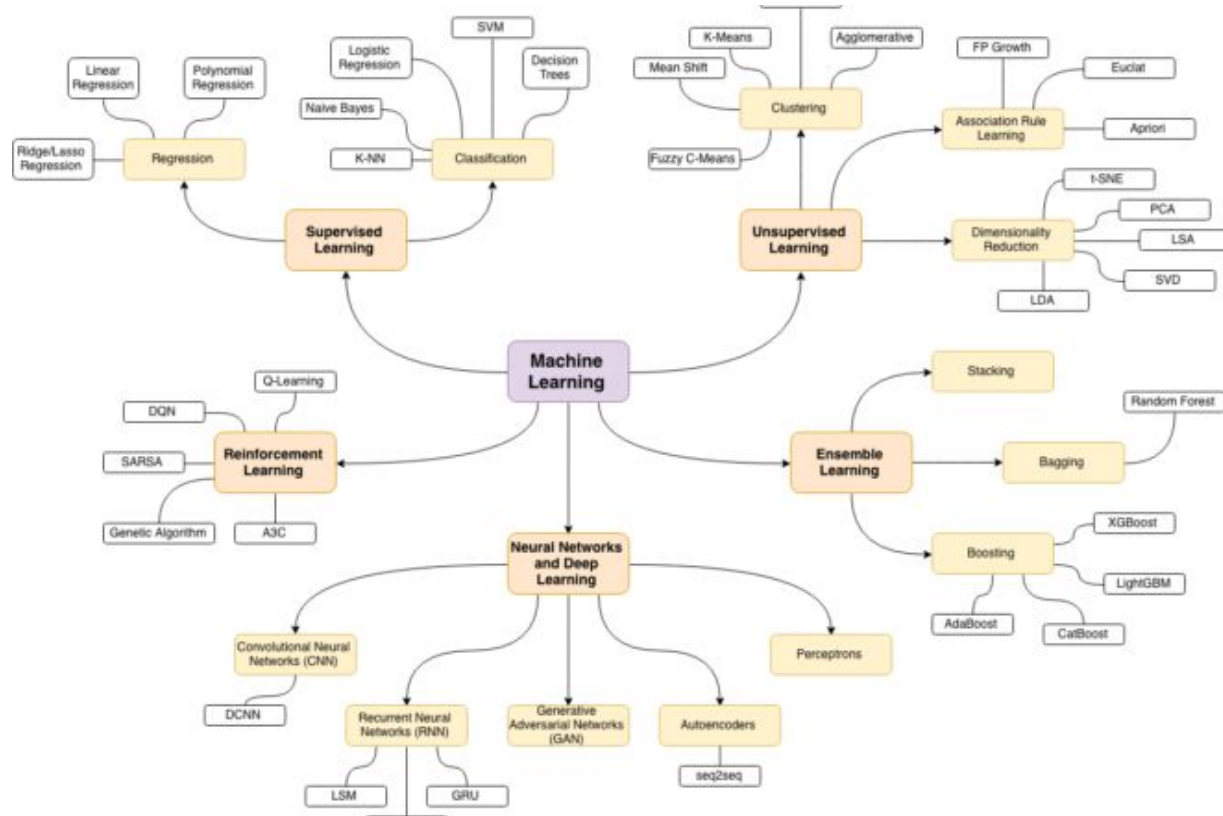
CS149: Parallel Computing:

<https://www.youtube.com/playlist?list=PLoROMvodv4rMp7MTFr4hQsDEcX7Bx6Odp>

CS231A: Computer Vision:

<https://www.youtube.com/watch?v=68wemjquj4o&list=PLoCMsyE1cudVnCcHk43vRy7PVTVWJ6WVR&index=18>

Try specializing in some of the following



Keep up to date



DeepSeek-R1: Incentivizing Reasoning Capability in LLMs via Reinforcement Learning

DeepSeek-AI

research@deepseek.com

Abstract

We introduce our first-generation reasoning models, L DeepSeek-R1-Zero, a model trained via large-scale reinforced fine-tuning (SFT) as a preliminary step, demonstrated Through RL, DeepSeek-R1-Zero naturally emerges with reasoning behaviors. However, it encounters challenges s

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Attention Is All You Need

Ashish Vaswani*
Google Brain
avaswani@google.com

Noam Shazeer*
Google Brain
noam@google.com

Niki Parmar*
Google Research
nikip@google.com

Jakob Uszkoreit*
Google Research
usz@google.com

Llion Jones*
Google Research
llion@google.com

Aidan N. Gomez*[†]
University of Toronto
aidan@cs.toronto.edu

Łukasz Kaiser*
Google Brain
lukaszkaizer@google.com

Illia Polosukhin*[‡]
illia.polosukhin@gmail.com

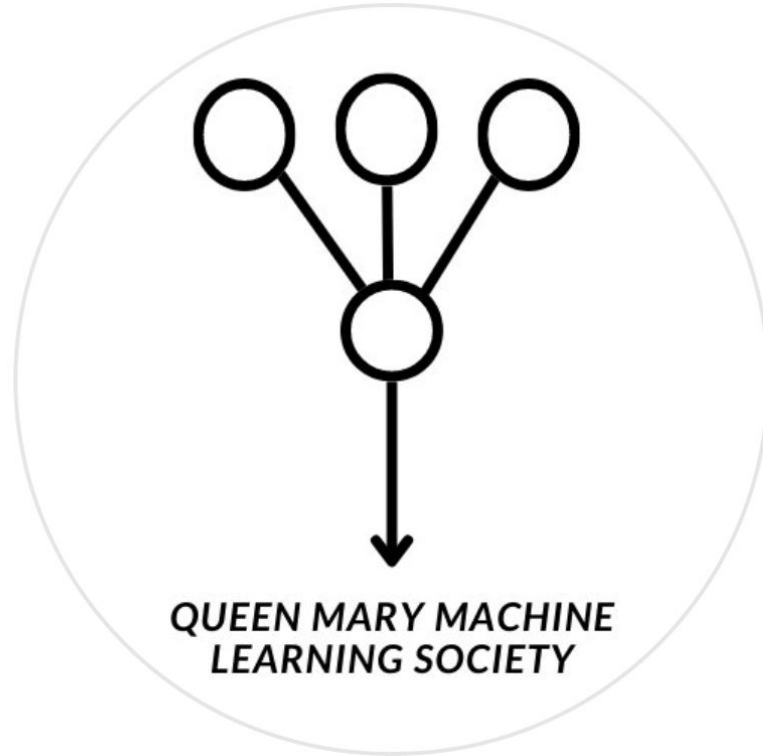
Generative Adversarial Nets

Ian J. Goodfellow, Jean Pouget-Abadie*, Mehdi Mirza, Bing Xu, David Warde-Farley, Sherjil Ozair,[‡] Aaron Courville, Yoshua Bengio[‡]
Département d'informatique et de recherche opérationnelle
Université de Montréal
Montréal, QC H3C 3J7

Abstract

For estimating generative models via an adversarial process, we have previously shown how to train a generative model G and a discriminative model D that estimates the probability of D making a mistake. This is a two-player game. In the space of arbitrary

Fake News Detection



Fake News Detection - Introduction

- What is fake news?
 - Factually false (GDP)
 - Misleading
- Why is it an important topic?
 - Anyone can publish information
 - Automated publication
- Importance of detection for society
 - Prevent spread of harmful misinformation
 - e.g., false health advice to sell useless drugs
 - Influence politics

kaggle

Fake News Detection - Detection Techniques

- NLP for text analysis
 - Extract language patterns, sentiment, or other linguistic features
- End-to-end classification models
- More advanced:
 - Network analysis of information dissemination

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Fake News Detection - Data Sources & Features

- Text content
 - Examples include
 - Exaggerated language
 - Emotional tone
- What is metadata?
 - Broad data
 - Source credibility
 - Publication timestamps



Fake News Detection - Zero-Shot Learning

- What Is Zero-Shot Learning?
 - Classify or understand new categories that were never seen during training
- How does it work?
 - Leverages generalizable representations
 - e.g., embeddings and language models to make predictions on unseen labels
- How to use it for fake news detection?

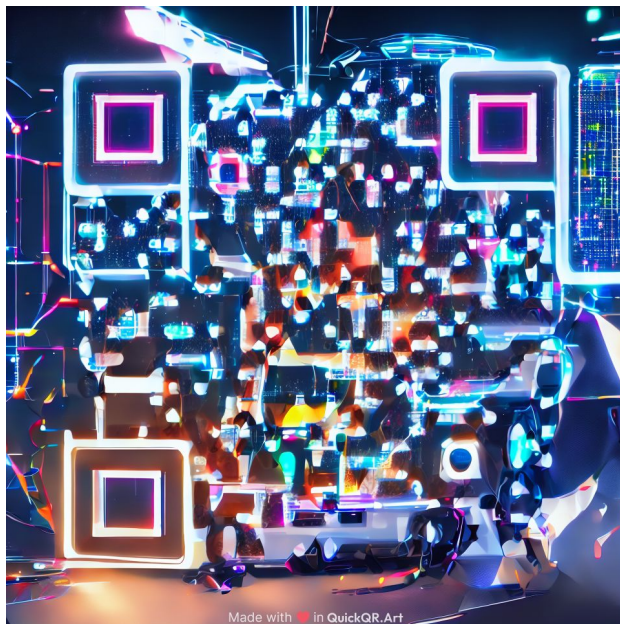
```
# Candidate labels to test (e.g., categories relevant to fake news detection)
candidate_labels = ["fake news", "reliable news", "misleading information"]

# Perform zero-shot classification
result = classifier(text, candidate_labels)
```



New Communications Platform

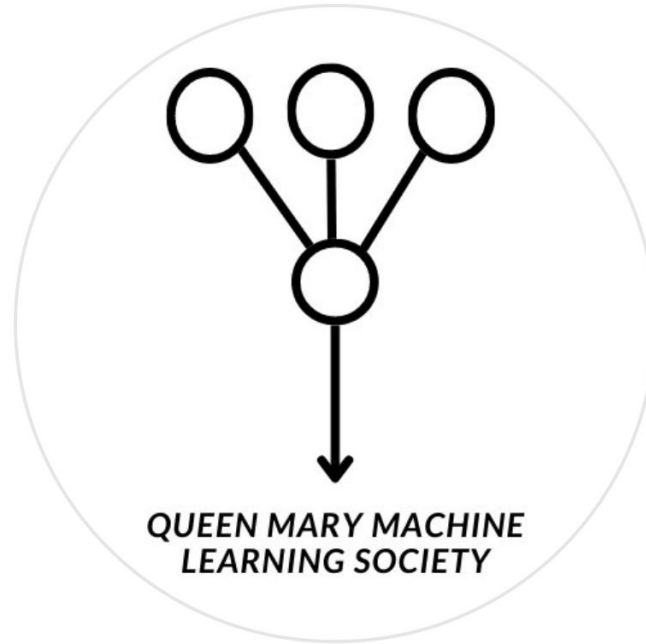
- Join our Discord:
 - Link: <https://discord.gg/xcfp4UHGWa>
 - QR-Code:



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Mini Hackathon Presentation

Fake news detection



Welcome to the Hackathon!

Manu's Machine Learning Lectures x Kaggle Teams



Showcase your skills! Compete in this exciting challenge to classify fake and real news using machine learning.



Your Goal: Build a model that can accurately distinguish between real and fake news articles using the provided dataset.



Prove your expertise and rank at the top!

Let's start by making teams of 2-4 people

Rules & Dataset Overview

Dataset:

- You will receive a dataset with **news articles** labeled as **Fake (1) or Real (0)**.
- Columns include **Title, Subject, Text, Date, and Target**.
- Use **only the first 75%** of the dataset for training. The final 25% is for testing only.

Rules:

1. **No external data** is allowed.
2. You must **only train on the first 75%** of the dataset.
3. **Submit predictions** for the test set.
4. Feature engineering is allowed.
5. Use any machine learning or deep learning approach.

Scoring System 🏆

You can earn up to **3 points** based on your performance:

🏆 **1 Point** - Highest overall accuracy (using any features you want).

📝 **1 Point** - Highest accuracy using **only the text** column.

📢 **1 Point** - Highest accuracy using **only the title** column.

♦ The leaderboard will be based on accuracy. Aim high! ♦

Best of Luck! 🚀

Thanks for Listening!

