## **Responsible Machine Learning**

#### Lecture 1: Intro

CS 4973-05

Fall 2023

Instructor: Avijit Ghosh ghosh.a@northeastern.edu Northeastern University, Boston, MA







evijit.io ∂`adept ID

## **Avijit Ghosh**

- Research Data Scientist at AdeptID
- Areas of study include:
  - Audits of real systems,
  - Measurement and evaluation of algorithmic fairness, and
  - Challenges of incorporating algorithmic fairness interventions in the real world.



jlgleason.github.io

## **Jeffrey Gleason**

- 3rd-year PhD Candidate working with Christo Wilson
- Areas of study include:
  - Algorithm auditing (controlled experiments and causal inference)
  - Platform power (e.g. Google + Amazon self-preferencing)

## Please Introduce Yourself

- What made you want to take this class?
- What problems are you excited to solve?



## **Class Schedule**

• Class: Ryder 277

Tuesdays 11:45am - 1:25pm

Thursdays 2:50pm - 4:30pm

• Office Hours:

Avijit: Thursdays 5pm-6pm Jeffrey: Wednesdays 4pm - 5pm



## **Class Format**

- **Reading assignments:** 30%
- **Coding assignments:** 15%
- **Quizzes:** 10%
- **Final project + Term paper:** 45%



#### **Reading Assignments**

I will assign portions of research papers to read to prepare for next day's lecture. A brief reading assignment quizlet will accompany the readings. You are required to upload your answers to the reading assignment questions **BEFORE** class starts.

Helpful guide: <u>How to read a research paper</u>



#### **Coding Assignments**

Less frequent than reading assignments, few and far between. Expected language: Python 3, preferably on Jupyter notebooks



#### Quizzes

Short, multiple choice quizzes, one for midterm and one for finals. Prep level needed low as long as you were paying attention in class.



#### Final Project + Term paper

The most important component of this class. You are expected to work on a final project in groups of two. This project involves three components:

- 1. Reading research papers/blogs on the topic you have chosen
- 2. Downloading relevant data and writing code to get results
- 3. A term paper describing what you did. The format is usually: Introduction > Related work > Methods > Results > Conclusion > Limitations and Future Work
  - Overleaf Latex Format

We will help you in every step of the way! While this is due at the end of the semester, it is good to start planning early and keep me updated on what you are doing.



# Al is powerful

## odo you guysever thinkabout BIAS

# [record scratch]

# What does Responsible ML mean?



# What does Responsible ML mean?

- Fairness/Bias
- Explainability
- Transparency
- Privacy
- Safety
- Regulations/Policy



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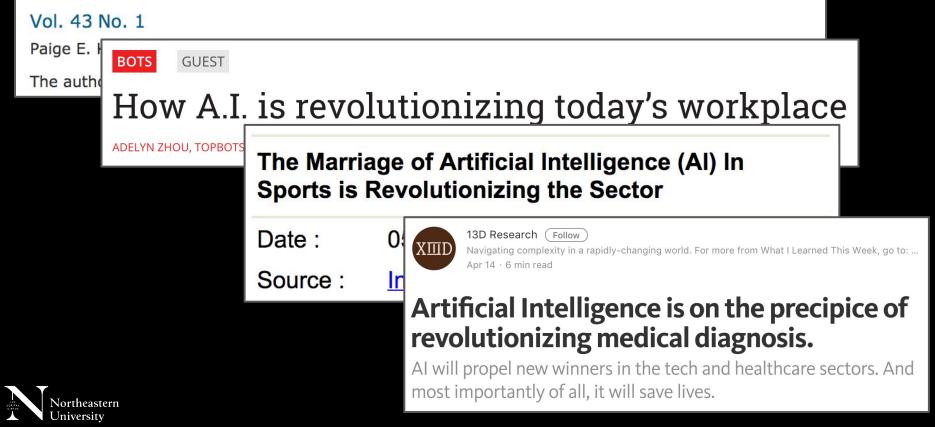




## Predictive systems are on the rise. Machine learning and AI technologies have generated enormous benefits for business.



#### How Artificial Intelligence Is Revolutionizing the Legal Practice



## **A Provocation:** Should we trust a machine to make the right decisions about a person's future?



### Skepticism is healthy.

Businesses have learned some painful lessons about the dangers of machines creating their own algorithms.



The Telegraph	HOME NEWS		
Technology		<b>US</b> politics world opinion sports soccer tech arts lifestyle fashion	
News   Reviews   Opinion   Internet security   Social me	dia   Apple   Google	home > tech	
↑ Technology		Google	
Microsoft deletes 'teen girl' AI after it became a Hitler- loving sex robot within 24		Google says sorry for racist auto-tag in	
		photo app	
hours			
		<ul> <li>Google Photos labelled a picture of two black people as 'gorillas'</li> <li>Google Maps and Flickr have also suffered from race-related problems</li> </ul>	
cm tech	BUSINESS CULT	LTURE GADGETS FUTURE STARTUPS	

#### Connect

Flickr's new auto-tags are racist and offensive



by David Goldman @DavidGoldmanCNN

## How do algorithms become biased?



# Let's play spot the bias...

if person.ethnicity == 'African-American':
 credit.deny()
else:

credit.grant()



# The bias is... Intentional. Obvious.

#### (and horrible!)

if person.ethnicity == 'African-American':
 credit.deny()
else:

credit.grant()



# Before machine learning, algorithms were written only by humans.

#### Only a malicious developer would write such a rule.

### But problems like these are easy to fix.



# Can you spot the bias here?

if person.zip\_code == 38131: credit.deny() else: credit.grant()



Seems harmless... until you learn that zip code 38131 is nearly 100% African-American

if person.zip\_code == 38131: credit.deny() else: credit.grant()



## The bias is... Maybe Intentional. Less obvious.

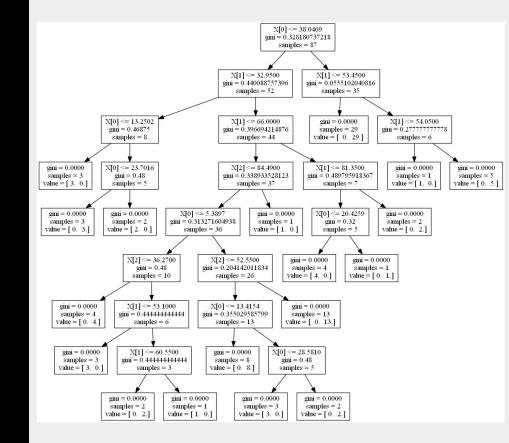


# A statistical model might suggest such a rule — less obvious, but equally bad.

## A principled developer or statistician can still catch these if they are vigilant. This is still relatively easy to fix.



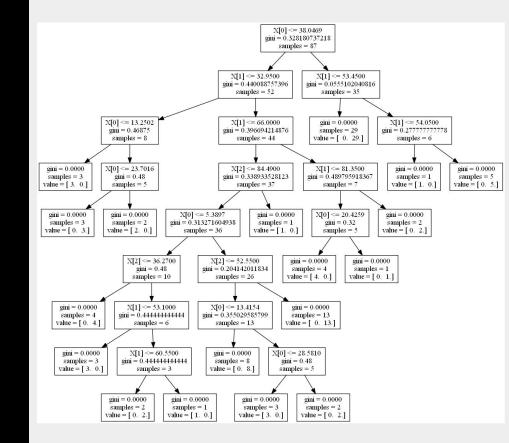
# What about this one?





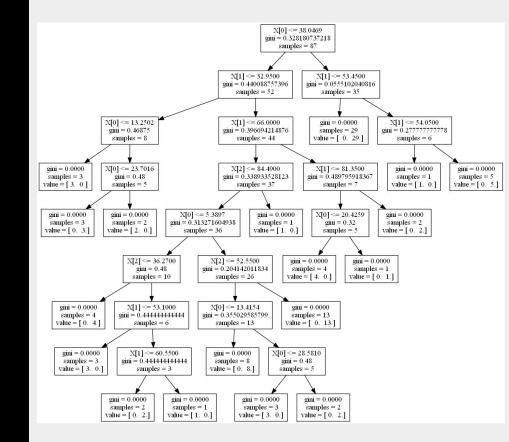
Credits: Chris Walker - Data Science @ Illuminate Education

## This is a small section of a decision tree built with machine learning





## The bias is... Unintentional. And far less obvious.



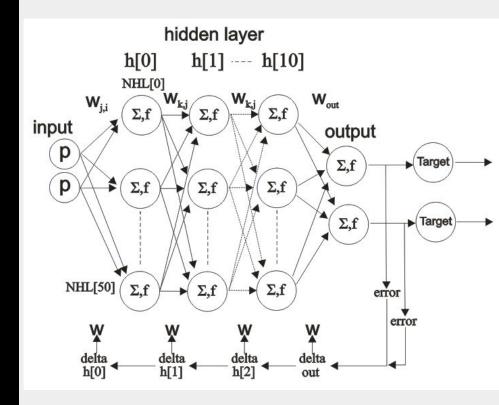


Now our algorithm is made of hundreds of decision points. Each one might be biased.

## Combinations of decision points might be biased too. That's millions of possibilities! Not so easy to fix.

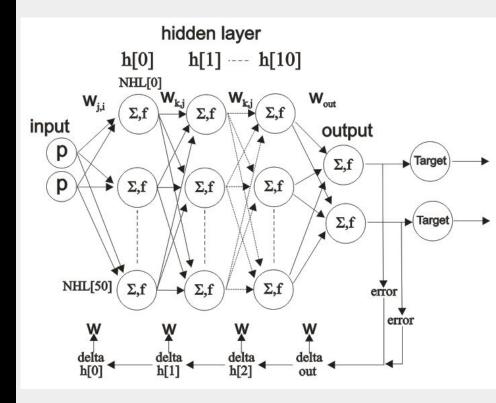


# Okay, last one...



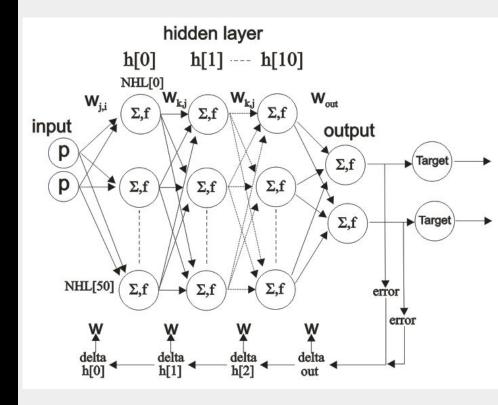


# **Welcome to Artificial** Intelligence





# The bias is... Unintentional. Totally obscured.





**Another Provocation:** If the builder of a system can't spot the bias, what hope do we have of correcting it?



Yet Another **Provocation:** How accurately do you think vendors report their system's capabilities?



### **Principles Worth Defending**

#### Fairness

How can a computer judge a person on something they haven't even done yet?

#### Transparency

How exactly was the decision made? What data were used?

#### Remediation

To whom does a person complain when things go sideways?

How can we make it right?



## **Thank You!**

#### **Readings for Next Class:**

 <u>Machine learning: Trends,</u> <u>perspectives, and prospects</u>
 - M. I. Jordan and T. M. Mitchell

