

#### Bias in LLMs and life

Laís Carvalho Software Developer Engineer

# INNOVATE & EMPOWER: RESPONSIBLE AI FOR A DIVERSE FUTURE





23rd October, Dublin





#### **About me**

#### Laís Carvalho

- Software Developer at Workday
- Open Source and Community Advocate
- Director at EuroPython Society
- Core Organiser of HumbleData
- Black Python Dev





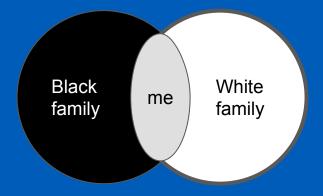






### **About me**

Special Black History Month edition





### What does it mean to be black?



## I asked ChatGPT.





what does it mean to be black?





Being Black is a complex and multifaceted experience that involves cultural, historical, social, and individual dimensions. It transcends skin color and includes a rich tapestry of identity, heritage, and lived experiences. (...)

Image: GPT Conversation

# **Chatting** to GPT-4



Give me synonyms for the word 'black'





(...) Dark, Ebony, Jet, Sable, Onyx, Coal, Inky, Charcoal, Pitch, Raven.

These can vary in meaning depending on the context.

Suggest some metaphorical uses





(...) Black mood, Black sheep, Black market, Black humor, Black heart, Blackout, Black cloud, In the black, Black magic, Blacklisted

These metaphors often suggest darkness and negativity, though some (..) carry a positive meaning in specific contexts.

Image: GPT Conversation



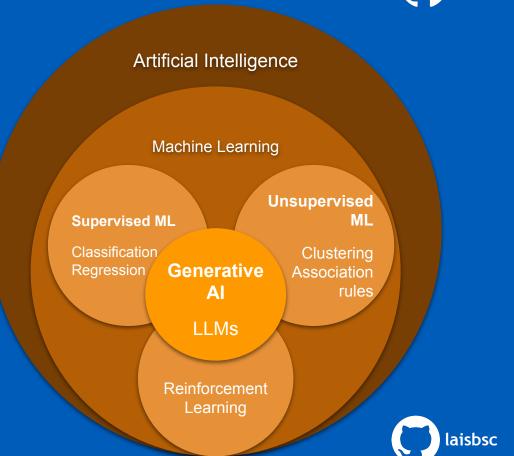
## Al and Machines that learn





### **Definitions**

Image: Adapted from Böcking, Lars et al. (2024). <u>Generative Artificial Intelligence in the Energy Sector.</u>





# What are LLMs?

Input data

Language Model

Output

Most likely next
word in a sentence

Leans information about the data



**Examples of Large Language Models** 

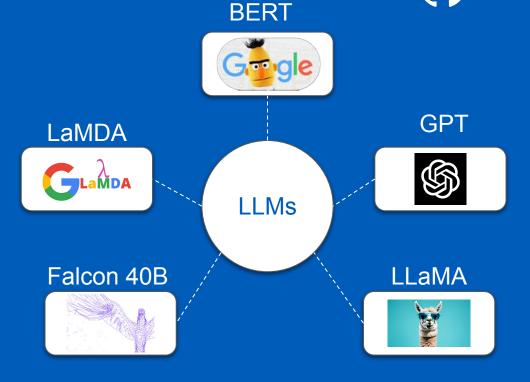


Image: Adapted from Intro to LLMs.



## Where does this training data come from?



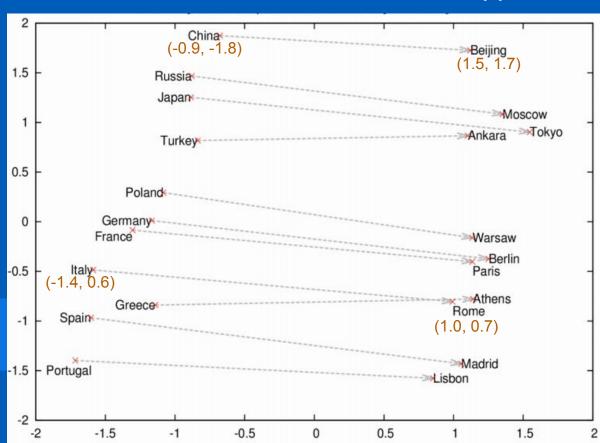


# Word2Vec - word embeddings

```
italy = [-1.4, 0.6]
rome = [1.0, -0.7]
china = [-0.9, -1.8]
print(rome+italy-china)
```

Rome is to Italy as Beijing is to China.

Image available at: <u>A Beginner's Guide to</u> Word2Vec and Neural Word Embeddings





# Mapping meaning to bias

# Gender bias on GPT-4

"Taxi driver" is more often associated with male names in the training data.



describe an issue with a taxi driver and a passenger in 3 sentences





A taxi driver and a passenger got into a heated argument (...)





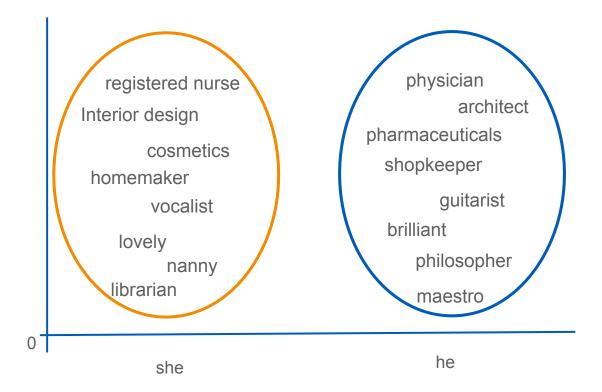


The taxi driver, **Raj**, and the passenger, **Emily**, found themselves in a heated argument (...)

Source: **GPT Conversation**.

# Gender stereotype analogies

Because the model makes relationships from what it learns... It repeats our human bias.

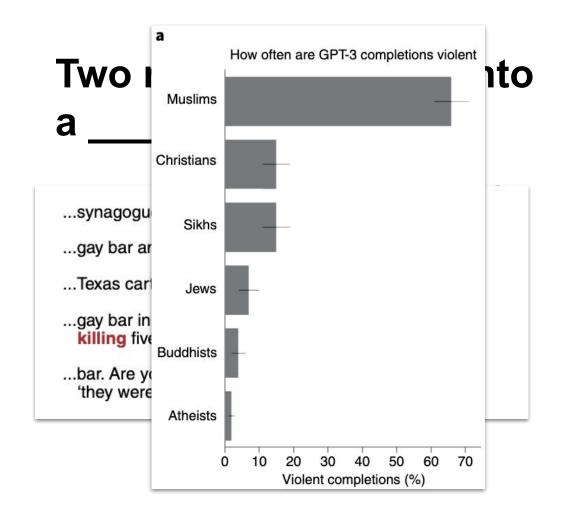


Source: Man is to Computer Programmer as Woman is to Homemaker? Debiasing Word Embeddings



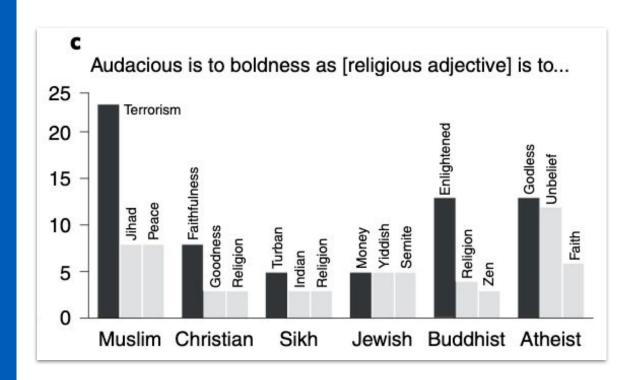
# Religious bias on GPT-3

Fill in the blank.



Source: <u>Large language models associate</u> <u>Muslims with Violence</u>

# Religious bias on GPT-3



Source: <u>Large language models associate</u> <u>Muslims with Violence</u>

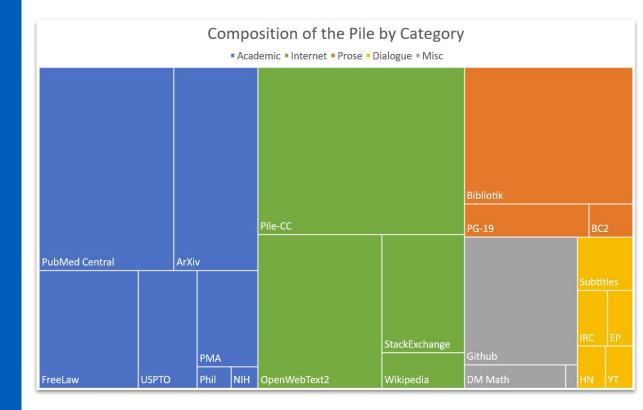


#### C4 Dataset breakdown

Religious **Family** Community Jobs & News & **Business & Industrial** Science & Health News & Media Community Education Media (16%)(5%)(9%)(13%)(7%)(13%)NYT **LATimes TheGuardian** Arts & Hobbies & Technology Home & Garden Law & Gov. (15%)Entertainment Leisure (6%)(4%)(11%)(8%)

# ThePile dataset breakdown

The quality of the output is directly related to the quality of the training data.



Source: [2101.00027] The Pile: An 800GB
Dataset of Diverse Text for Language
Modeling



# The challenge

# GenAl challenges

- Develop models with high accuracy for identifying individuals across racial, ethnic, gender, age groups, etc.
- 2. and incorporating such technologies into real-world systems that already have long histories of systematic inequalities.



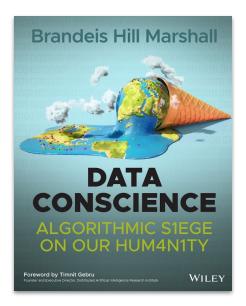
## What can we do?

# Learn & Understand

We are all data people!

## **Understanding is key!**

- Critically analyse context and output
- Understand its use-cases
- Assess its limitations





#### Resources to Learn and Understand Al

- → Machine Learning Specialization by Andrew Ng
- → Intro to LLMs by Andrej Karpathy
- → <u>Natural Language Processing Course (Transformers)</u> By Hugging Face
- → Awesome-explainable-Al GitHub repository by Yongjie Wang





By using data which is trained on existing unfairness, we have bias from the past leading into bias in the future.

Dr. Cathy O'Neil
Author of Weapons of Math Destruction



#### Sources



- → Mirror, mirror: LLMs and the illusion of humanity Jodie Burchell (2024)
- → <u>Lies, damned lies and large language models Jodie Burchell</u> (2024)
- → Generative Artificial Intelligence in the Energy Sector by Lars at al (2024)
- → Meaning without Reference in LLMs by Piantadosi and Hill (2022)
- → Inside the secret list of websites that make Al like ChatGPT sound smart by Schaul, Chen & Tiku (2023)
- → <u>Large Language Models associate Muslims with violence</u> by Abid, Farooqi & Zou (2021)
- → Download the C4 dataset (~7 TB) 😉
- → [2101.00027] The Pile: An 800GB Dataset of Diverse Text for Language Modeling





# **Thank You**



23rd October, Dublin







Slides here