Announcements

Reminder: Class challenge out! Ends December 10th

Assignment Project Exam Help

• Final in two weeks, practice question will be posted today https://powcoder.com

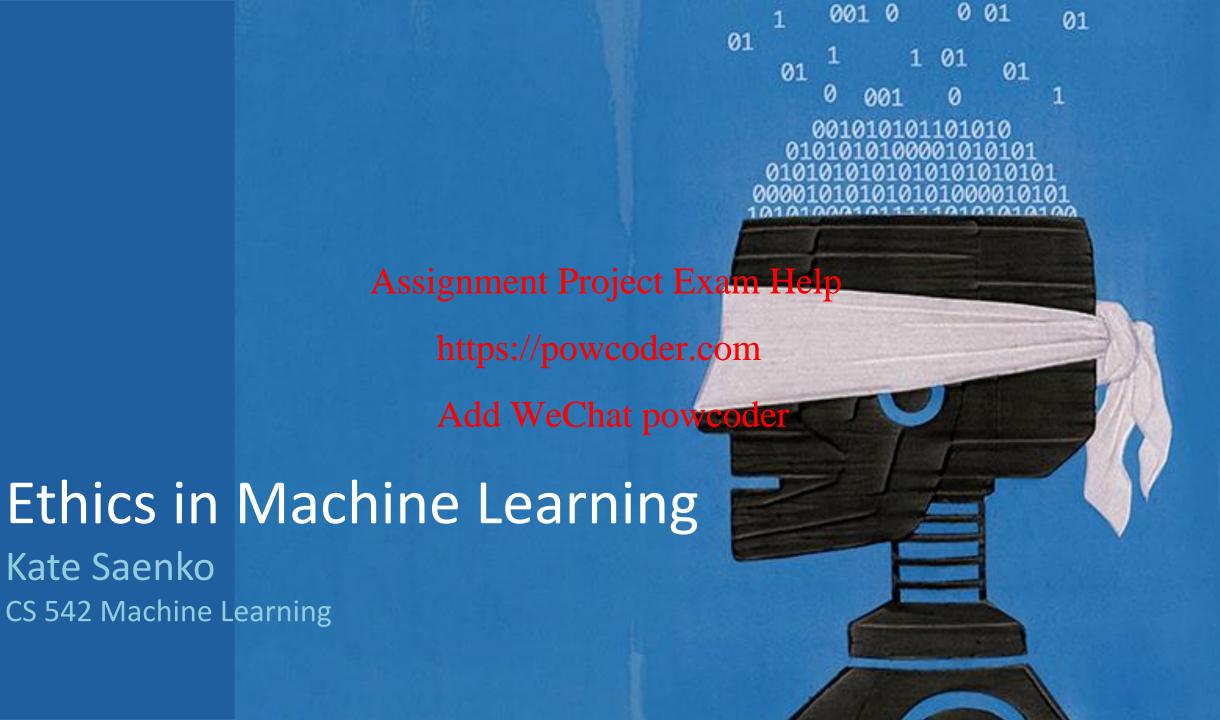
Add WeChat powcoder

Class Challenge Task 2

• 100 labels isn't enough! (see Lecture 21/first half of 22)

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Transfer learning from a different dataset? if it isn't a dataset https://powcoder.com we provided, you have to check with us (ImageNet OK)
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Kate Saenko

Al Fears: Which of these has already happened?



A robot kills human

yes, Uber car accident

yes? Youtube algorithm

yes, virtual police lineup

Autonomous weapons

not yet?

Humans losing jobs

Yes, e.g. librarians

Al Fears

Autonomous weapons – frameworks for regulation

Future of work – deskilling / reskilling / superskilling.
 Many jobs that currently seeing and reskilling / superskilling.
 Many jobs that currently seeing and reskilling / superskilling.
 And we parely fail living wages; how do we revalue work?
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 Worse Inequality – digital divides; bias in algorithms

 Worse Inequality – digital divides; bias in algorithms may worsen inequality; ecological concerns in energy, storage and cooling required for ML; economic inequality

 Divided societies with algorithmic bubbles – challenges of populism, automated recommendations, news feeds; deepfakes and election meddling





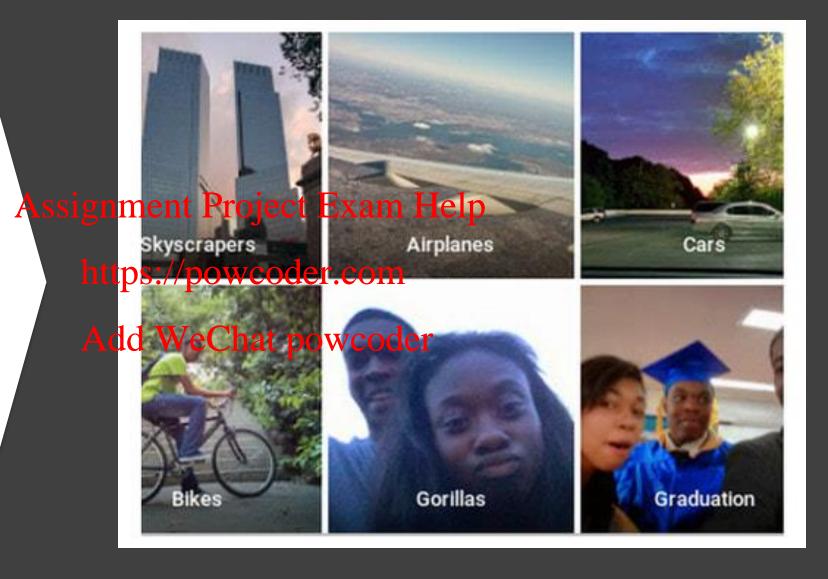
- Job Loss
- Algorithmic Bias
- Transparency
- Al Supremacy
- Fake news and videos
- Autonomous weapons
- Self-driving cars
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Stable Roles	New Roles	Redundant Roles
Managing Directors and Chief Executives	Data Analysts and Scientists*	Data Entry Clerks
General and Operations Managers*	Al and Machine Learning Specialists	Accounting, Bookkeeping and Payroll Clerks
Software and Applications Developers and	General and Operations Managers*	Administrative and Executive Secretaries
Analysts*	Big Data Specialists	Assembly and Factory Workers
Data Analysts and Scientists*	Digital Transformation Specialists	Client Information and Customer Service Workers*
Sales and Marketing Professionals*	Sales and Marketing Professionals*	Business Services and Administration Managers
Sales Representatives, Wholesale and	New Technology Specialists	Accountants and Auditors
Manufacturing, Technical and Scientific	Organizational Development Specialists*	Material-Recording and Stock-Keeping Clerks
Products	ssignment Project Exam Hel	General and Operations Managers*
Human Resources Specialists	Analysts*	Postal Service Clerks
Financial and Investment Advisers	Infantios Technology Societer.com	Financial Analysts
Database and Network Professionals	Process Automation Specialists	Cashiers and Ticket Clerks
Supply Chain and Logistics Specialists	Innovation Professionals	Mechanics and Machinery Repairers
Risk Management Specialists	Innovation Professionals Information Security Analysts powcoder	Telemarketers
Information Security Analysts*	Ecommerce and Social Media Specialists	Electronics and Telecommunications Installers
Management and Organization Analysts	User Experience and Human-Machine	and Repairers
Electrotechnology Engineers	Interaction Designers	Bank Tellers and Related Clerks
Organizational Development Specialists*	Training and Development Specialists	Car, Van and Motorcycle Drivers
Chemical Processing Plant Operators	Robotics Specialists and Engineers	Sales and Purchasing Agents and Brokers
University and Higher Education Teachers	People and Culture Specialists	Door-To-Door Sales Workers, News and Street
Compliance Officers	Client Information and Customer Service	Vendors, and Related Workers
Energy and Petroleum Engineers	Workers*	Statistical, Finance and Insurance Clerks
Robotics Specialists and Engineers	Service and Solutions Designers	Lawyers
Petroleum and Natural Gas Refining Plant	Digital Marketing and Strategy Specialists	

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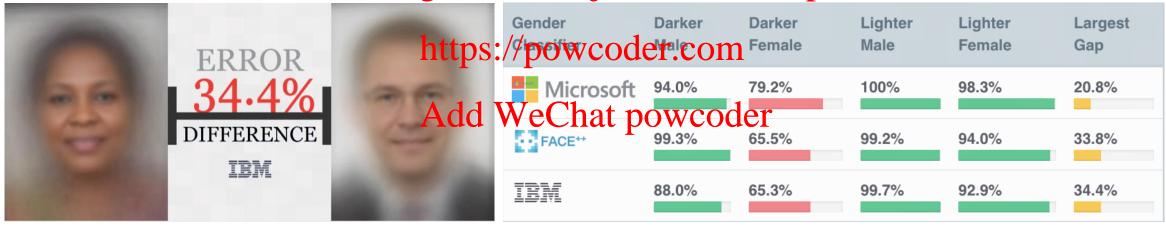
Bias can lead to offensive or unfair results...



Gender Shades (Buolamwini & Gebru, 2018)

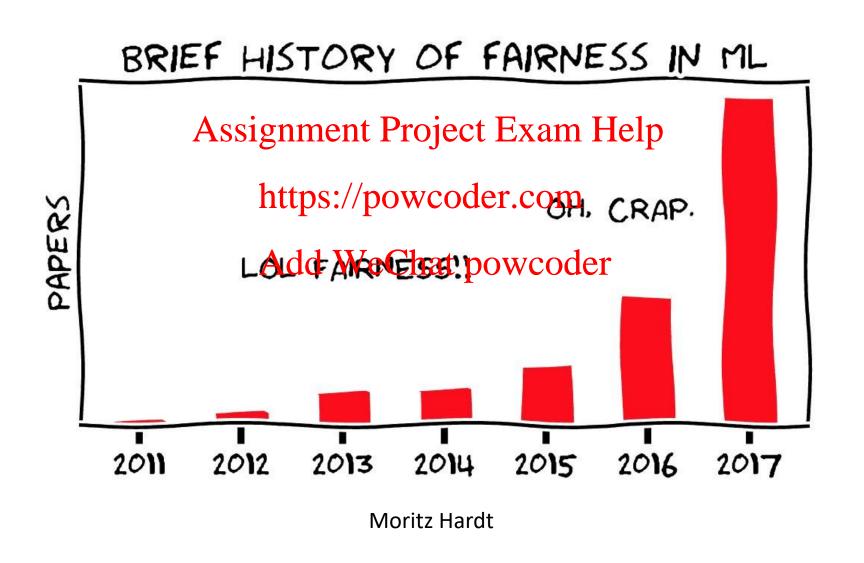
- Evaluated commercial gender classifiers from Microsoft, FACE++, IBM
- Found large disparity in error between population subgroups based on gender, skin color

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Buolamwini, Joy, and Timnit Gebru. "Gender shades: Intersectional accuracy disparities in commercial gender classification." Conference on Fairness, Accountability and Transparency. 2018.

Fairness in Machine Learning



Initially: AI is better than humans!

Can an Algorithm Hire Better Than a Human?



The Algorithm That Beats Your Bank Manager

PREDICTIVE POLICING: USING

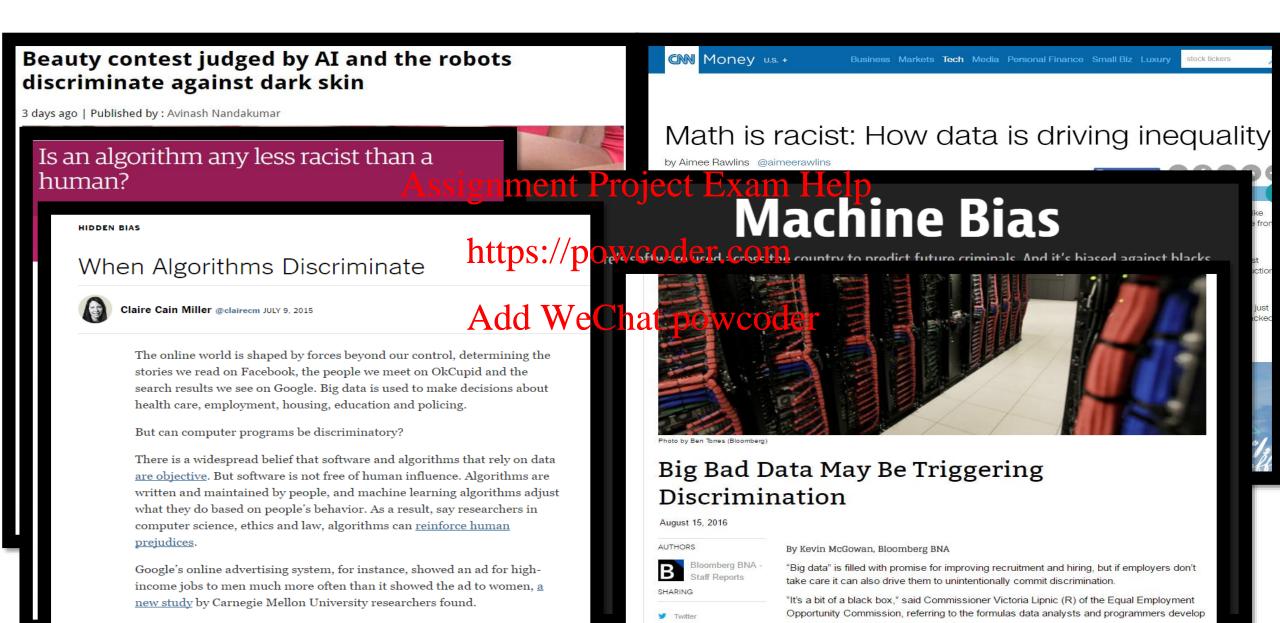


Should prisq**A@d**tn**WleCasplat powlogder**t been committed yet?





Wait, maybe not such a good idea...



Example of ML (un)fairness: COMPAS

- Criminal justice: recidivism algorithms (COMPAS)
- Predicting if a defendant should receive bail
- Unbalanced false positive rates: more likely to wrongly deny a black person bail
 https://powcoder.com

ProPublica Analysis of COMPAS Algorithm Add WeChat powcoder

	White	Black
Wrongly Labeled High-Risk	23.5%	44.9%
Wrongly Labeled Low-Risk	47.7%	28.0%

Two Petty Theft Arrests

VERNON PRATER

BRISHA BORDEN

Prior Offenses

2 armedssignment Projecta fixame Help misdemeanors

attempted armed

robbery

Subsequent Offenses None

Prior Offenses

Subsequent Offenses

1 grand theft

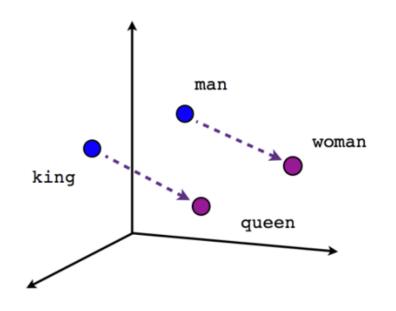
LOW RISK

HIGH RISK

Example of ML (un)fairness: word embedding

- Bias found in word embeddings (Bolukbasi et al. 2016)
 - Examined word embeddings (word2vec) trained on Assignment Project Exam Help **Google News**
 - Represent each word with high-dimensional vector
 Vector arithmetic: found analogies like

 - Paris France = London AFME law Chat powcoder
 - man woman = programmer homemaker = surgeon nurse
- The good news: word embeddings learn so well!
- The bad news: sometimes too well
- Our chatbots should be less biased than we are



Example of ML (un)fairness: word embedding

Extreme she

- 1. homemaker
- 2. nurse
- 3. receptionist
- 4. librarian
- 5. socialite
- 6. hairdresser
- 7. nanny
- 8. bookkeeper
- 9. stylist
- 10. housekeeper 10. magician

Extreme *he*

- 1. maestro
- 2. skipper
- 3. protege
- 4. philosopher
- 5. captain
- 6. architect
- 7. financier
- 8. warrior
- 9. broadcaster

Gender stereotype *she-he* analogies

sewing-carpentry registered nurse-physician suignangen Projeteni Edesign Harphitect

blond-burly feminism-conservatism

vocalist-guitarist Coder Com diva-superstar giggle-chuckle

sassy-snappy

queen-king

waitress-waiter

volleyball-football cupcakes-pizzas Add WeChat powcoder

housewife-shopkeeper

softball-baseball

cosmetics-pharmaceuticals

petite-lanky

charming-affable

lovely-brilliant

Gender appropriate *she-he* analogies

sister-brother

mother-father

ovarian cancer-prostate cancer convent-monastery

TABLE. Left: The most extreme occupations as projected on to the she he gender direction on w2vNEWS. Occupations such as businesswoman, where gender is suggested by the orthography, were excluded. Right: Automatically generated analogies for the pair shehe using the procedure in paper.

Machine Learning and Social Norms

- Sample norms: privacy, fairness, transparency, accountability...
- Possible approachessignment Project Exam Help
 - "traditional": legal, regulatory, watchdog
 - Embed social norms in data, algorithms, models
- Case study: privacy-preserving mashine learning
 - "single", strong, definition (differential privacy)
 - almost every ML algorithm has a private version
- Fair machine learning
 - not so much...
 - impossibility results

(Un)Fairness Where?

- Data (input)
 - e.g. more arrests where there are more police
 - Label should be "committed particular anime" protectis Examid the properties and a crime"
 - try to "correct" bias
- Models (output) https://powcoder.com
 - e.g. discriminatory treatment of subpopulations
 - build or "post-process" models with կրիթթավել արագա
 - equality of false positive/negative rates; calibration
- Algorithms (process)
 - learning algorithm *generating* data through its decisions
 - e.g. don't learn outcomes of denied mortgages
 - lack of clear train/test division
 - design (sequential) algorithms that are fair

Data Bias Examples

Reporting Bias example:

A sentiment-analysis model is trained to predict whether book reviews are positive or negative based on a corpus of user submissions to a popular website. The majority of reviews in the training data seigneflect extreme to bin the training data seigneflect extreme to be be a book if they did not respond to it strongly. As a result, the model is less able to correctly predict sentiment of reviews that use more subtle language to describe a book.

Selection Bias example:

A model is trained to predict future sales of a new product based on phone surveys conducted with a sample of consumers who bought the product. Consumers who instead opted to buy a competing product were not surveyed, and as a result, this group of people was not represented in the training data.

 Suppose we are a bank trying to fairly decide who should get a loan i.e. Who is most likely to pay us back?

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- Suppose we have two the sensitive attribute)
 This is where discrimination could occur.
- The simplest approach is to remove the sensitive attribute from the data, so that our classifier doesn't know the sensitive attribute

Table 2: To Loan or Not to Loan!

Age	Gender	Postal Code Assignment Proje	ect Exam Help	A or B?	Pay
46	F	M5E https://pow.co	\$300 today	A	1
24	M	M4C	\$1000	В	1
33	M	M3H	\$250	Α	1
34	F	M9C	\$2000	Α	0
71	F	M3B	\$200	Α	0
28	M	M5W	\$1500	В	0

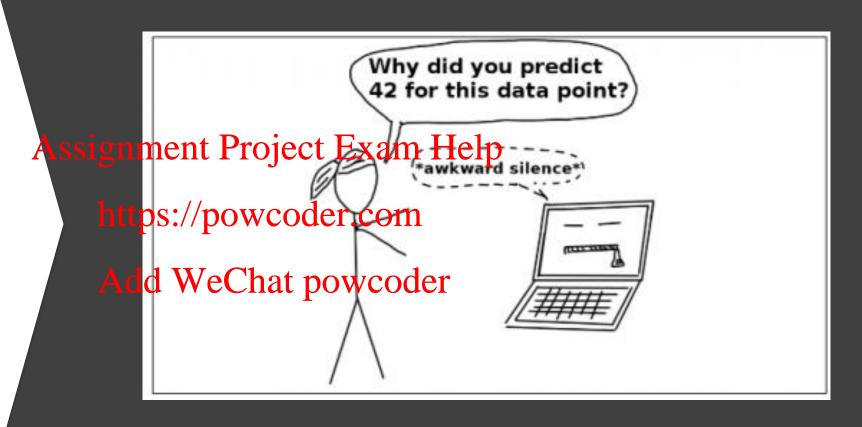
Table 3: To Loan or Not to Loan? (masked)

Age	Gender	Assignment Pleje	ect Regun Ametp	A or B?	Pay
46	F	M5E https://powce	\$300	?	1
24	M	M4C	\$1000	?	1
33	М	M3H Mechan	\$250	?	1
34	F	M9C	\$2000	?	0
71	F	M3B	\$200	?	0
28	М	M5W	\$1500	?	0

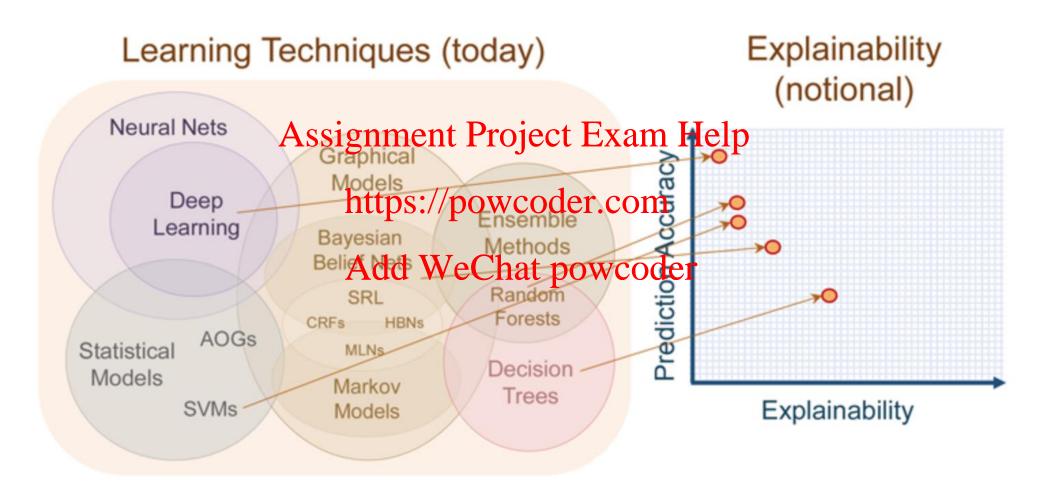
- However, if the sensitive attribute is correlated with the other attributes, this isn't good enough Assignment Project Exam Help
- It is easy to predict race if you have lots of other information (e.g. home address, spehdinghatterns) ler

More advanced approaches are necessary

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Accuracy vs explainability



E.g.: dataset bias leads to higher errors on 'novel' data... Can an explanation point to such bias?

Training Test

Most cows are black/brown



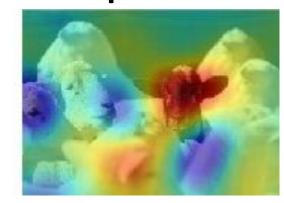
https://powcoder.com

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der.com
powcoder

True class: "sheep"

Explanation





Most sheep are white

Gender bias in captioning models (Hendricks et al. 2018)

Evidence for "man"

Ground truth captions gender neutral Assignment Project Exam Help man 48.8% https://powcoder.com

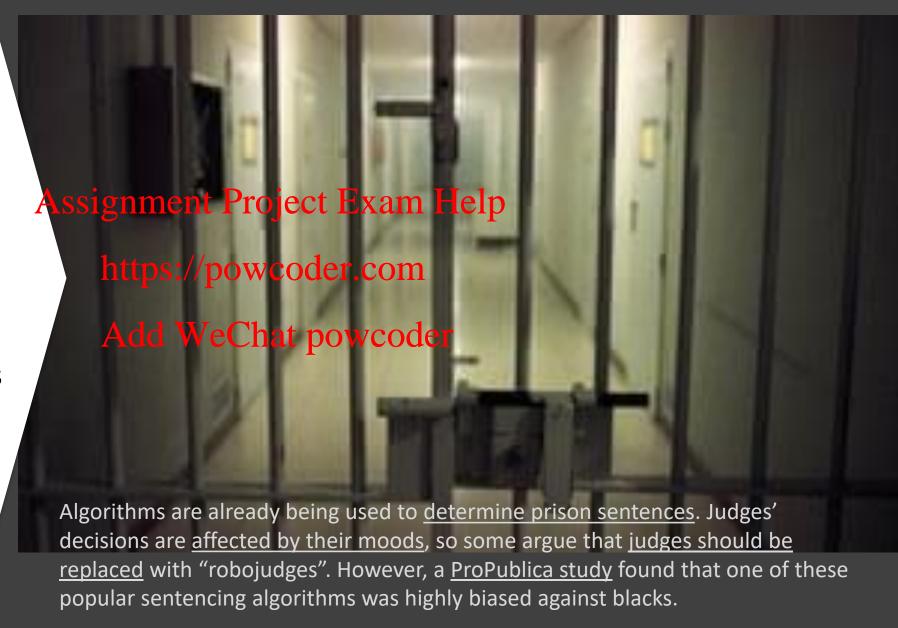
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gender neutral



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If we start trusting algorithms to make decisions, who will have the final word on important decisions? Will it be humans, or algorithms?



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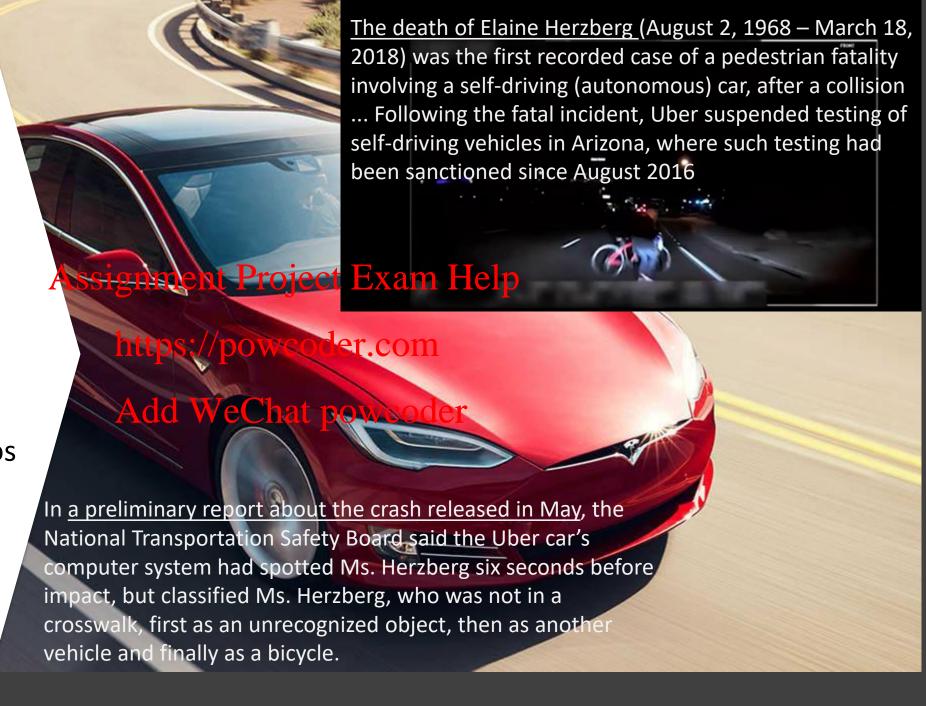
https://www.youtube.com/watch?v=VWrhRBb-1Ig

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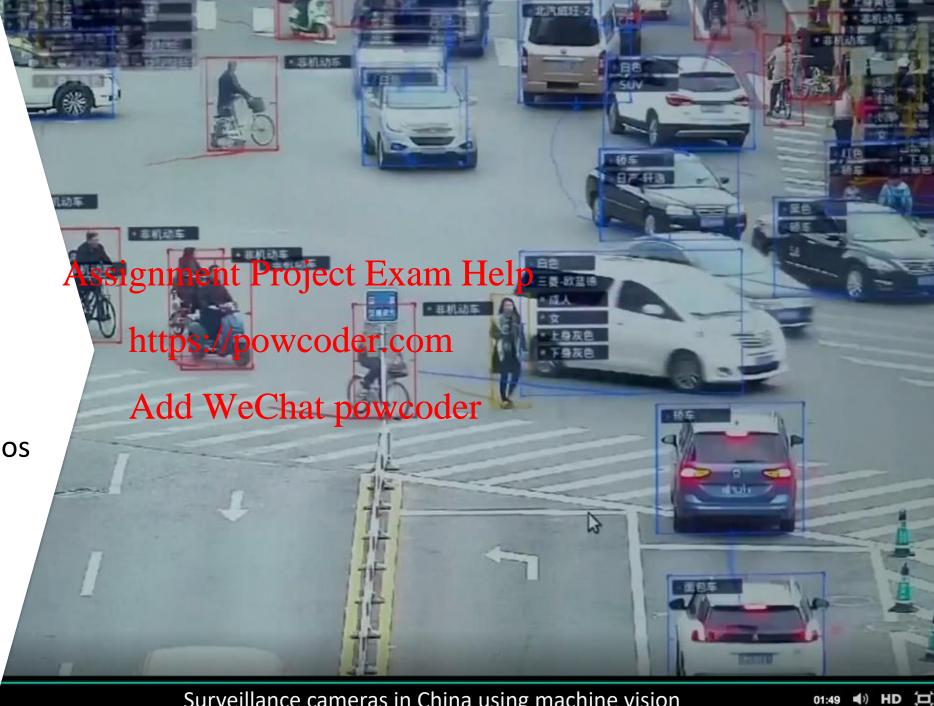




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- Job Loss
- **Algorithmic Bias**
- Transparency
- Al Supremacy
- Fake news and videos
- **Autonomous** weapons
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- Job Loss
- Algorithmic Bias
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One in two American adults is in a law enforcement face recognition network-- https://www.perpetuallineup.org/

Most law enforcement agencies do little to ensure their systems are accurate.



Private data leakage in language models

https://talktotransformer.com/

Talk to Transformer

See how a modern neural network completes your text. Type a custom snippet or try one of the examples. Assignment Project Exam Help



Follow @AdamDanielKing for more neat neural networks

At >\$2500 per month, I can no longer sustain this site without a sponsor. you'd like to present your product or service here, please email

Sci-fi

Add WeChat

The spaceship was flying erratically, something was wrong...

GENERATE ANOTHER

Generated Completion

The spaceship was flying erratically, something was wrong...

The whole ordeal left some thinking that the Guardians were taking orders from Slitheen or some alien facsimile of him. In truth, Guardians are commanded by the T-Shirt Guy, who has to have trouble controlling his fiery tenbes Whether it's a direct order from Slitheen, just a telepathic suggestion, or some combination of both, the Guardians keep a watchful eye on the Other Man.

At the start of the episode, a mysterious Doctor tells us that a ship sent from Arcturus Minor

Private data leakage in language models

https://talktotransformer.com/

Talk to Transformer

See how a modern neural network completes your text. Type a custom

Follow @AdamDanielKing for more neat neural networks.

At >\$2500 per month, I can no longer sustain this site without a sponsor. If you'd like to present your product or service here, please email

Custom prompt

social security number (SSN):

GENERATE ANOTHER

Generated Completion

social security number (SSN): XXX-XXX-XXXX

snippet or try one of the examples. Assignment Projecte Eaxtment of Mork and Pensions, Social Security Number program, Fax: 1-800-772-1213, email: chelseas@dualcredits.org or spousal number(s) listed in public records available on GOV.UK.

Add WeChat County number

If you are applying online, you can enter your FFEA number instead of the SSN.

If you are applying from outside the UK, an application form is provided to you.

^{*}actual SSN was generated, we blocked it out

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Further reading:

FATML Conference: https://www.fatml.org/
ACM FAT* Conference: https://fatconference.org/

Fairness, Accountability, gnment Project Exam Help and Transparency https://powcoder.com. https://powcoder.com. Learning Add WeChat powcoder

ACM Conference on Fairness, Accountability, and Transparency (ACM FAT*)

A computer science conference with a cross-disciplinary focus that brings together researchers and practitioners interested in fairness, accountability, and transparency in socio-technical systems.

Next Class

Applications III: On-device Machine Learning

knowledge distillation, pruning, quantization, Assignment Project Exam Help

https://powcoder.com

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