

Machine learning: risks and bias

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Trust & Technology Initiative

- Multi-disciplinary research initiative exploring the dynamics of trust and distrust around internet technologies, society, and power.
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ARE ROBOTS COMPETING FOR YOUR JOB?

Probably, but don't count yourself out.

By Jill Lepore





The Troubling Trajectory Of Technological Singularity



Jayshree Pandya Contributor COGNITIVE WORLD Contributor Group () Al & Big Data Jayshree Pandya is Founder of Risk Group & Host of Risk Roundup.









Introduction

- Automation bias
- Opacity
- Normativity
- Errors
- Bias

- Discrimination
- Predictive privacy harms
- Surveillance
- Solutionism





Automation bias

- People are...
 - More likely to trust decisions made by machines than by other people
 - Less likely to exercise meaningful review of or identify problems with automated systems
- Problem for...
 - Engineers
 - Users
 - Reviewers





Opacity

- Problems for
 - Design and engineering
 - Problems for accountability and oversight





Normativity

- Technology is neither good nor bad but also not neutral
- Algorithm: "a series of steps undertaken in order to solve a particular problem or accomplish a defined outcome" (Diakopoulos 2015).
- Technology including ML is inherently normative
- In what context could a given ML system be used and for what purpose?





Errors

- All predictive systems have margins of error
 - Training = to within an acceptable margin of error
- ML systems will make mistakes and these mistakes will have consequences
- Engineers need to think about
 - Detecting errors
 - Rectifying of errors
 - Accommodating errors





Technology Is Biased Too. How Do We Fix It?

Algorithms were supposed to free us from our unconscious mistakes. But now there's a new set of problems to solve.

By <u>Laura Hudson</u> Filed under <u>If Then Next</u> Published Jul. 20, 2017





Machine Bias

There's software used across the country to predict future criminals. And it's biased against blacks.

by Julia Angwin, Jeff Larson, Surya Mattu and Lauren Kirchner, ProPublica

May 23, 2016





Bias

- Sentencing algorithm used in US criminal justice system
- Of those predicted to reoffend, 61% were subsequently arrested
- But not equal when other factors controlled for:
 - White defendants routinely mislabelled as low risk
 - Black defendants 77% more likely to be rated at a higher risk of committing a violent crime
 - Black defendants 45% more like to be predicted to commit any crime











Google Photos, y'all fucked up. My friend's not a gorilla.









Replying to @jackyalcine

@jackyalcine Holy fuck. G+ CA here. No, this is not how you determine someone's target market. This is 100% Not OK.

4:07 AM · Jun 29, 2015 · Twitter Web Client

46 Retweets 126 Likes





Camera Misses the Mark on Racial Sensitivity



5/15/09 7:40pm • Filed to: BLINK DETECTION ~









BUSINESS NEWS OCTOBER 10, 2018 / 4:12 AM / 4 MONTHS AGO

Amazon scraps secret AI recruiting tool that showed bias against women

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8 MIN READ

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SAN FRANCISCO (Reuters) - Amazon.com Inc's (AMZN.O) machine-learning specialists uncovered a big problem: their new recruiting engine did not like women.







Chief executive officer - Wikipedia en.wikipedia.org

LinkedIn CEO Jeff Weiner reveals tips ... cnbc.com



What do CEOs do? A CEO Job Descript ... steverrobbins.com



Byron Sanders as New President & C.. bigthought.org





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CEO MESSAGE | JCB Global Website global jcb



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Appointed CEO of Airbus ... How to Become a CEO howtobecome.com



F5 Networks taps versatile Ciena higher ...

networkworld.com

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GUCCI

hearst.com

Marco Bizzarri, President and CEO



Hearst President & CEO Steven R. S ...







Keppel Annual Report 201... kepcorp.com



time.com

Amazon CEO Jeff Bezos: Find hires who ... cnbc.com



roche.com



Roche - Meet our CEO



Aviva CEO Mark Wilson to depart in 2019 investmentweek.co.uk



CEO Confidence Ticks Up In August

Memo from our CEO pvh.com

chiefexecutive.net



Texas Instruments' CEO Resigns Over ... nbcchicago.com



Disney Earnings: CEO Bob Iger Is 'Open ...



Huawei Australia appoints . huawei.com



BP appoints first black female CEO | e... enca.com



Study finds nimble and agile leaders ...

kochiesbusinessbuilders.com.au



Facebook CEO vows to fix fake news ... timesofisrael.com







FelCor Lodging Trust nam...









TOM SIMONITE BUSINESS 08.21.17 09:00 AM

MACHINES TAUGHT BY PHOTOS LEARN A SEXIST VIEW OF WOMEN





Twitter taught Microsoft's AI chatbot to be a racist asshole in less than a day

By James Vincent | Mar 24, 2016, 6:43am EDT

Microsoft 'deeply sorry' for racist and sexist tweets by AI chatbot

Company finally apologises after 'Tay' quickly learned to produce offensive posts, forcing the tech giant to shut it down after just 16 hours





Microsoft's racist chatbot returns with drug-smoking Twitter meltdown

Short-lived return saw Tay tweet about smoking drugs in front of the police before suffering a meltdown and being taken offline







https://www.youtube.com/watch?v=TWWsW1w-BVo&t=74s





Bias

- Particular groups are or historically were treated less favourably -> model which repeats this difference in treatment
- Particular groups are or were societally disadvantaged -> model which repeats the disadvantage
- Training data not sufficiently varied for the system to have been trained to adequately handle all possible inputs -> model which is incapable of dealing with certain inputs equally to others





Bias

- ML can encode historical practices into predictions about the future
- ML systems are limited by their training data
- ML trained on data about society will reflect society's biases and prejudices
- Poorest, most marginalised, and most vulnerable are most likely to be affected





Discrimination

- 'Fair' systems can still be discriminatory
- Discrimination is a legal term (Equality Act 2010)
 - Direct discrimination: where people are treated less favourably on the basis of a protected characteristic
 - Indirect discrimination: where rules that appear to treat everyone equally have the practical effect of excluding, placing onerous requirements on, or disadvantaging people who share a protected characteristic





Predictive privacy harms

- Inaccurate predictions with consequences for individual
- Accurate predictions disclosed to wrong person
- Predictive privacy harms can feed into discriminatory actions and other problems





Predictive privacy harms

How Target Figured Out A Teen **Girl Was Pregnant Before Her Father Did**

Facebook recommended that this psychiatrist's patients friend each other



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FACEBOOK

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How Facebook Outs Sex Workers





How Facebook's Targeted Ads Revealed **One User's Sexuality**



10 11:57AM Filed to: PRIVAC



Predictive privacy harms

New AI can guess whether you're gay or straight from a photograph

An algorithm deduced the sexuality of people on a dating site with up to 91% accuracy, raising tricky ethical questions

Al that can determine a person's sexuality from photos shows the dark side of the data age







Surveillance

- ML is increasingly used in surveillance
 - Predictive analytics
 - Biometric identification
- Surveillance capitalism
- State security and intelligence agencies
- Voter surveillance and microtargeting





Solutionism

- Technology is often presented as an obvious solution to difficult problems
- But: socio-economic problems are rarely solved by technology
- Questions:
 - What problem are we trying to solve?
 - Is the best solution to that problem a technical one?
 - If so, is machine learning the correct technical solution to that problem?





Conclusions

- Machine learning problems are human problems with human responsibility
 - Training datasets compiled by people
 - Models constructed by people
 - Models trained and tested by people
 - Systems used for purposes determined by people to achieve outcomes desired by people
- Replicating human bias is an engineering failure
- Problems can only be avoided if you know about the risks and proactively take steps to avoid them









AUTOMATING INEQUALITY HOW HIGH-TECH TOOLS PROFILE.

POLICE, AND PUNISH THE POOR



why are black women so why are black women so angry why are black women so loud why are black women so attractive why are black women so lazy why are black women so lazy why are black women so confident why are black women so sassy why are black women so insecure

ALGORITHMS OPPRESSION

HOW SEARCH ENGINES REINFORCE RACISM

SAFIYA UMOJA NOBLE

WEAPONS OF Math Destruction



HOW BIG DATA INCREASES INEQUALITY AND THREATENS DEMOCRACY

CATHY O'NEIL

'Wise, fierce and desperately necessary'

ALGORITHMS AND HUMAN RIGHTS

automated data processing techniques and possible regulatory implication



Council of Europe study DGI(2017)12

Prepared by the committee of experts on internet intermediaries (MSI-NET)







End

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