Social and Ethical Responsibilities of Computing (SERC):

New Activities at

David Kaiser

November 2021

AI in the World



SERC at MIT



New Teams, New Tools



Opportunities

STAT+

MIT News

When AI is the opposite of sinister: An MIT researcher is held up as model of how algorithms can benefit humanity

By Rebecca Robbins Sept. 23, 2020

HEALTH TECH



MIT artificial intelligence researcher Regina Barzilay is the Association for the Advancement of Artificial Intelligence. MIT CSAIL

Faster drug discovery through machine learning

New technique speeds up calculations of drug molecules' binding affinity to proteins.

Daniel Ackerman | MIT News Office March 15, 2021



MIT News ON CAMPUS AND AROUND THE WORLD

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Taming the data deluge

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A National Science Foundation-funded team will use artificial intelligence to speed up discoveries in physics, astronomy, and neuroscience.

Sandi Miller | Department of Physics October 29, 2021



At the crossroads of language, technology, and empathy

With a double major in linguistics and computer science, senior Rujul Gandhi works to surmount language and cultural barriers, globally and on campus.

Alli Armijo | MIT News correspondent October 13, 2021



Challenges

The interface of **artificial intelligence** (AI) + **machine-learning** (ML) techniques with **people** — both individuals and groups — presents **special challenges**.



These challenges are **exacerbated** when AI + ML techniques move **beyond research settings** into **real-world situations**.

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NISTIR 8280

Face Recognition Vendor Test (FRVT) Part 3: Demographic Effects

> Patrick Grother Mei Ngan Kayee Hanaoka

This publication is available free of charge from: https://doi.org/10.6028/NIST.IR.8280

> National Institute of Standards and Technology U.S. Department of Commerce

The **2019 NIST report** (third in a series) analyzed 189 algorithms from 99 (mostly commercial) developers. They assessed performance both for **1:1 matching** (for verification of a known individual) and for **1:***n* **matching** (for attempted matches within a whole dataset). [Main report: 75pp, plus 1200pp technical appendices.]

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Summary: False positive rates highest for people from Africa and East Asia, lowest for people from Eastern Europe (large effect, ~10²x). Across geographical sets, false positives higher for women than men (~ 2 – 5x).

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False match rate in men within E. Europe

The New Hork Times

Another Arrest, and Jail Time, Due to a Bad Facial Recognition Match

A New Jersey man was accused of shoplifting and trying to hit an officer with a car. He is the third known Black man to be wrongfully arrested based on face recognition.

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Nijeer Parks is the third person known to be arrested for a crime he did not commit based on a bad face recognition match. Mohamed Sadek for The New York Times



The US has *thousands* of distinct law-enforcement jurisdictions. Commercial facial-recognition technologies are *already being used* across the country, subject to *no regulation, standardization, or oversight*.



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Several reported cases of Black men being wrongfully arrested due to incorrect FRT matches indicate a *combination of failures*: inadequate technical calibrations *plus* human failures to follow appropriate procedures.

More generally: new technologies are deployed within *existing institutional frameworks*.



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Contrast this with other examples of *forensic science* within the US, such as *fingerprinting*: far from perfect, but subject to *expert review, training,* and *standardization*.



The New York Tin

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One billion surveillance cameras are now in place within 50 countries. Within the US alone, facial images of half the adult US population are already included in databases accessible to law enforement.

By Kashmir Hill Published Dec. 29, 2020 Updated Jan. 6, 2021 The US has *thousands* of distinct law-enforcement jurisdictions. Commercial facial-recognition technologies are *already being used* across the country, subject to *no regulation, standardization, or oversight*.

Several reported cases of Black men being wrongfully arrested due to incorrect FRT matches indicate a *combination of failures*: inadequate technical calibrations *plus* human failures to follow appropriate procedures.

Forbes

Jun 4, 2019, 06:26pm EDT | 2,146 views

Government Watchdog Questions FBI On Its 640-Million-Photo Facial Recognition Database



taff



Officials testify a House Oversight and Reform committee hearing on facial recognition technology in [+] JACQUELYN MARTIN/ASSOCIATED PRESS

As China Tracked Muslims, Alibaba Showed Customers How They Could, Too

The website for the tech titan's cloud business described facial recognition software that could detect members of a minority group whose persecution has drawn international condemnation.

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Alibaba, the leading provider of cloud computing services in China, could be drawn into the global controversy over China's treatment of Uighurs. Aly Song/Reuters



Published Dec. 16, 2020 Updated Jan. 20, 2021



To date, one of the largest and most-used **corpora of email** — critical to algorithmic natural-language processing — comes from employees of **Enron**, the large Texas-based energy company that declared bankruptcy in December 2001.

Ultimately more than **20** executives pleaded guilty or were convicted, including on multiple felony fraud charges.



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In 2003, the U.S. Federal Energy Regulatory Commission released **1.6** *million emails* sent to or from **158** *Enron senior executives* between 2000 – 2002. After minimal processing, the emails were simply made publicly available on a website. Employees were given a *limited opt-out* period.

(Recall 2003: Mark Zuckerberg was still an undergraduate ...)



More than **20k academic studies** have been published that make use of the Enron email corpus (~1k per year).



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Although no legal actions were ever brought against 99.83% of Enron email users, researchers are *still* finding $O(10^4)$ examples of *sensitive personally identifiable information* within the publicly available corpus, including Social Security Numbers, credit card numbers, birthdates, bank account numbers... Not to mention multiple extramarital affairs and other embarrassing episodes.

rXiv.org > cs > arXiv:2001.10374	Search Help A
[Submitted on 24 Jan 2020]	
The Enron Corpus: Where the Email Bodies are Buried?	

David Noever

To probe the largest public-domain email database for indicators of fraud, we apply machine learning and accomplish four investigative tasks. First, we identify persons of interest (POI), using financial records and email, and report a peak accuracy of 95.7%. Secondly, we find any publicly exposed personally identifiable information (PII) and discover 50,000 previously unreported instances. Thirdly, we automatically flag legally responsive emails as scored by human experts in the California electricity blackout lawsuit, and find a peak 99% accuracy. Finally, we track three years of primary topics and sentiment across over 10,000 unique people before, during and after the onset of the corporate crisis. Where possible, we compare accuracy against execution times for 51 algorithms and report human–interpretable business rules that can scale to vast datasets.



More than **20k academic studies** have been published that make use of the Enron email corpus (~1k per year).

Beyond the privacy concerns, the ubiquitous corpus continues to be used as *training data* for various *natural-language-processing algorithms*.

"If you think there might be **significant biases** embedded in emails sent among employees of a Texas oil-and-gas company that collapsed under federal investigation for **fraud** stemming from **systemic, institutionalized unethical culture**, you would be right. The Enron emails are simply **not representative**—not geographically, not socioeconomically, not even in terms of race or gender. Indeed, researchers have used the Enron emails *specifically* to analyze gender bias and power dynamics. *And yet the Enron emails remain a go-to dataset for training AI systems.*"

The Enron email corpus is a canonical example of *"biased, low-friction data."*

Amanda Levendowski, "How copyright law can fix artificial intelligence's implicit bias problem," *Washington Law Review* 93 (2018): 579-630.

Beyond "Algorithmic Bias"

🛱 Summer 2021 👻 🕂 🕂

Published on Aug 10, 2021

Understanding Potential Sources of Harm throughout the Machine Learning Life Cycle

As machine learning (ML) increasingly affects people and society, awareness of its potential unwanted consequences has also grown. To anticipate, prevent, and mitigate undesirable downstream consequences, it is critical that we understand when and how harm might be introduced...



policing and sentencing; healthcare; real estate and finance; hiring ...



Harini Suresh

Department of Electrical Engineering and Computer Science, and Computer Science and Artificial Intelligence Laboratory, MIT



John Guttag

Department of Electrical Engineering and Computer Science, and Computer Science and Artificial Intelligence Laboratory, MIT

- 1. **Historical bias** arises when there is a misalignment between world as it is and the values or objectives to be encoded and propagated in a model. It is a normative concern with the state of the world, and exists even given perfect sampling and feature selection.
- 2. **Representation bias** arises while defining and sampling a development population. It occurs when the development population under-represents, and subsequently fails to generalize well, for some part of the use population.
- 3. **Measurement Bias** arises when choosing and measuring features and labels to use; these are often proxies for the desired quantities. The chosen set of features and labels may leave out important factors or introduce groupor input-dependent noise that leads to differential performance.
- 4. Aggregation bias arises during model construction, when distinct populations are inappropriately combined. In many applications, the population of interest is heterogeneous and a single model is unlikely to suit all subgroups.
- 5. Evaluation bias occurs during model iteration and evaluation. It can arise when the testing or external benchmark populations do not equally represent the various parts of the use population. Evaluation bias can also arise from the use of performance metrics that are not appropriate for the way in which the model will be used.
- 6. **Deployment Bias** occurs after model deployment, when a system is used or interpreted in inapppropriate ways.

Social and Ethical Responsibilities of Computing at MIT



SERC Leadership Team



Professor, Department of Aeronautics and Astronautics Director, Interactive Robotics Group, CSAIL Associate Dean, SERC



David Kaiser Germeshausen Professor of the History of Science Professor of Physics Associate Dean, SERC



Participants in **SERC Action Groups** represent **all 5 Schools** at MIT plus the new MIT Schwarzman College of Computing. They have engaged in a sustained fashion within **multidisciplinary groups** for one or more semesters during the past two years.



SERC Teaching

Vision:

Fractal model for embedding SERC material throughout the curriculum, making it *inescapable*.

Develop original pedagogical materials by **multidisciplinary teams** with members from across computing, data sciences, humanities, arts, and social sciences—for use in **each** of these types of classes.



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Each submission is **reviewed by 4-6 senior researchers at MIT**, drawn equally from computing and data sciences and from arts, humanities, and social sciences. Each submission is **also reviewed** by **MIT undergraduate volunteers** for balance and accessibility.

Cases are written by **subject-area experts**, not limited to MIT.

SERC





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The Case of the Nosy Neighbors

by Johanna Gunawan and Woodrow Hartzog

Inspired by companies like Clearview AI, Nextdoor, and Amazon, this case study asks students to assume the role of a high-ranking ethics-focused employee at a (fictional) neighborhood-focused social media company. It



Who Collects the Data? A Tale of Three Maps

by Catherine D'Ignazio and Lauren Klein

Who makes maps and who gets mapped? Using a comparative reading of three maps, this case study introduces the idea that data may be useful, but they are not neutral. Rather, they



The Bias in the Machine: Facial Recognition Technology and Racial Disparities

by Sidney Perkowitz

Facial recognition technology (FRT) appears in uses from providing secure access to



The Dangers of Risk Prediction in the Criminal Justice System

by Julia Dressel and Hany Farid

Courts across the United States are using computer software to predict whether a person will commit a crime, the results of which are



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Hacking Technology, Hacking Communities: Codes of Conduct and Community Standards in Open Source by Christina Dunbar-Hester

Published: Aug 10, 2021

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throughout the Machine Learning Life

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by Harini Suresh and John Guttag

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Research programs conducted with **external partners** that inform the design and use of the tools for **better outcomes**.

Faculty stewards and champions trained to work in **multidisciplinary teams** to assess potential harms as well as benefits from computing research.



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Train crash at Montparnasse station, Paris, 1895



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First study: dataset combinations (including location / mobility data) and the reidentification of sensitive, personal healthcare information.

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New Tools

Ethical Computing Platform: a free and open **online platform** that research groups can use *iteratively* to address a series of questions about **benefits and burdens** of their **proposed research design**, **data collection** and **protection protocols**, and **potential unintended consequences** of a given project.



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coming soon!

STS/Physics, SERC

Executive Director,

MIT IPC



AeroAstro/CSAIL, SERC

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... but like **any new technology**, they carry risks of **unintended consequences** and **real harms** for individuals and groups. Computing and the data sciences are enabling **fantastic progress** and **benefits for society**...



The most **significant challenges** arise when techniques **leave the lab** and impact **real-world situations**. ... but like **any new technology**, they carry risks of **unintended consequences** and **real harms** for individuals and groups.



Identifying potential consequences and rectifying harms requires **input** and **expertise** from **across the fields** of science, engineering, humanities, and the social sciences ...



... as well as working with **partners beyond academia** — including people who are **impacted** by the latest technologies as well as the people who **implement** them.

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