

SciTinyML: Scientific Use of Machine Learning on Low-Power Devices



Responsible AI

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“ Machine intelligence is the
last invention that humanity
will ever need to make ”

Nick Bostrom
Philosopher, University of Oxford

SUSTAINABLE DEVELOPMENT GOALS

17 goals on the United Nations' 2030 Agenda for Sustainable Development:

- Ending poverty and world hunger
- Improving health and education
- Reducing inequality and injustice
- Clean water and sanitation
- ... etc.

Promising Applications of **TinyML**



Industry



Environment



Humans

Microsoft's disastrous Tay experiment shows the hidden dangers of AI

Amazon scraps secret AI recruiting tool that showed bias against women

Predictive policing algorithms are racist. They need to be dismantled.

Google Calls Hidden Microphone in Its Nest Home Security Devices an 'Error'

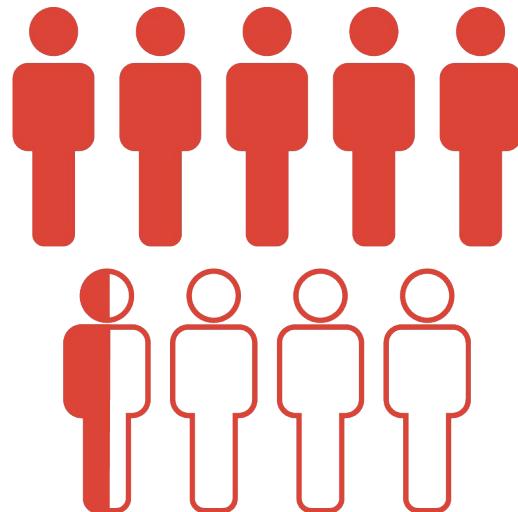
Microsoft's disastrous Tay experiment shows the hidden dangers of AI

Amazon scraps secret tool that showed bias against women

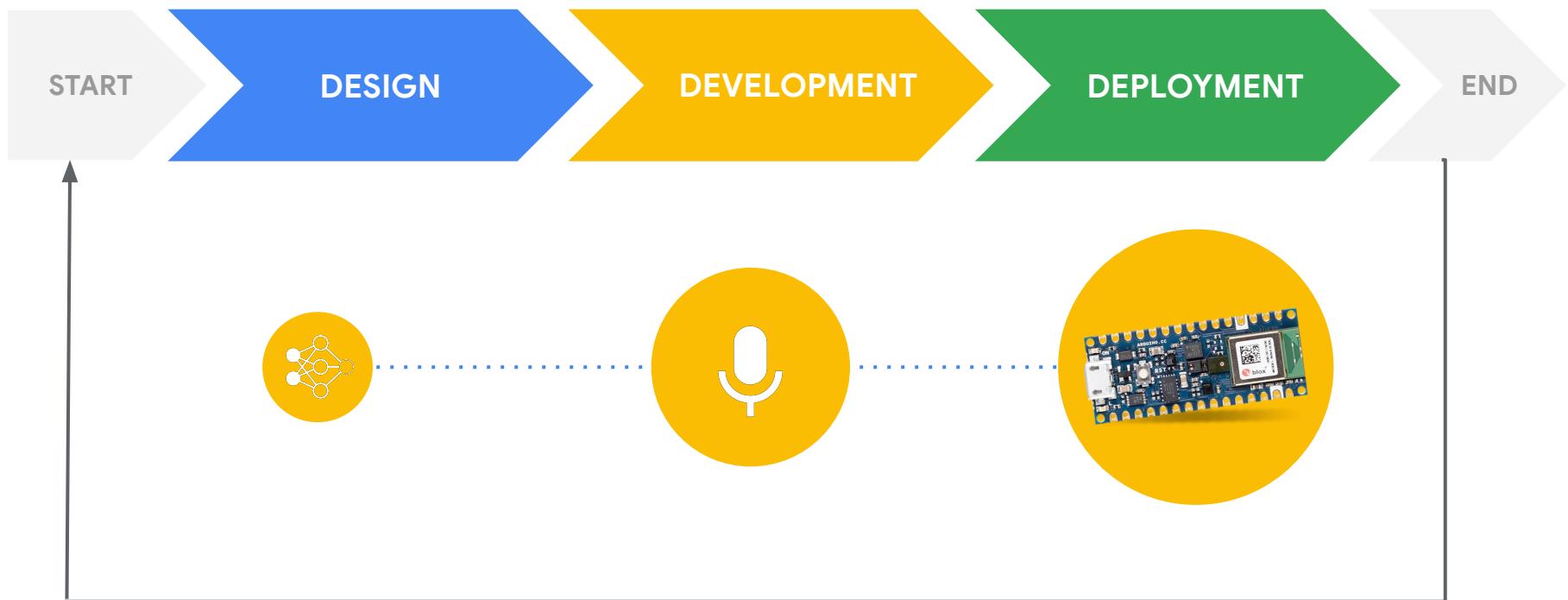
Predictive policing algorithms are racist. They need to be dismantled.

Google Calls Hidden Microsoft Security Devices an 'Error'

Pew Research shows that **65% of Americans** believe that companies “often **fail** to anticipate how their products and services will impact society”



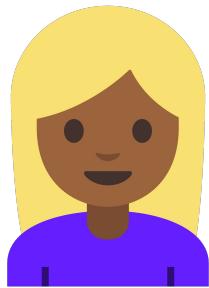
Embedding Ethics Throughout the Workflow



Responsible AI: Design

Stakeholder Analysis

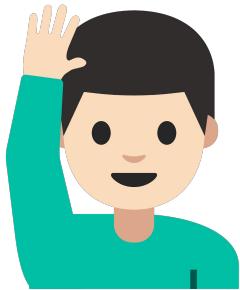
Direct



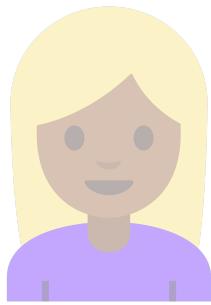
aka the “User(s)”

Stakeholder Analysis

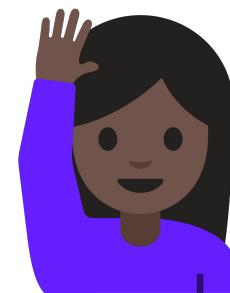
Indirect



Direct

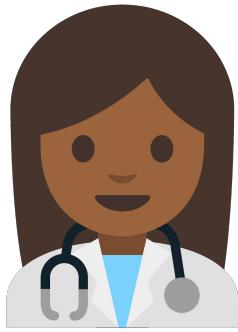


Indirect



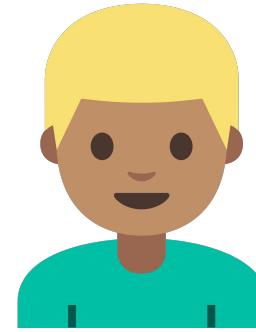
aka the “User(s)”

What do the stakeholders **value**?



Direct (Doctor)

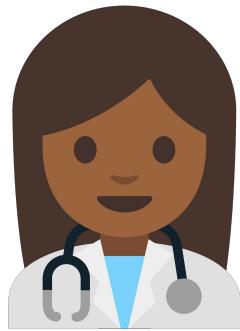
- Accurate diagnosis
- Training/skill set
- Ease of use
- Research advances



Indirect (Patient)

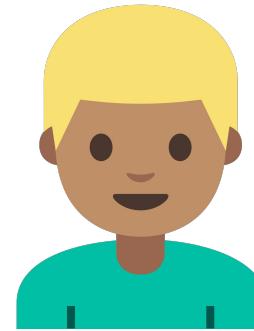
- Personal care
- Being informed / autonomy
- Trust
- Privacy

Do value tensions arise?



Direct (Doctor)

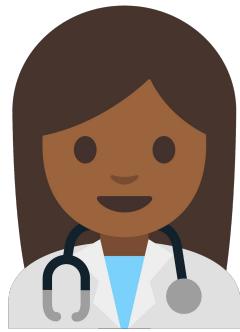
- **Accurate diagnosis**
- Training/skill set
- Ease of use
- Research advances



Indirect (Patient)

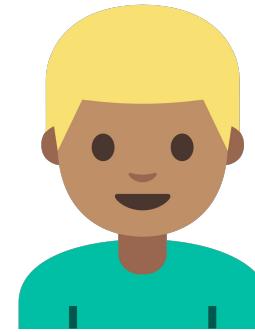
- Personal care
- **Being informed / autonomy**
- Trust
- Privacy

Do value tensions arise?



Direct (Doctor)

- Accuracy
- Training/skill set
- Ease of use
- **Research advances**



Indirect (Patient)

- Personal care
- Being informed / autonomy
- Trust
- **Privacy**

Which type of error is most harmful?

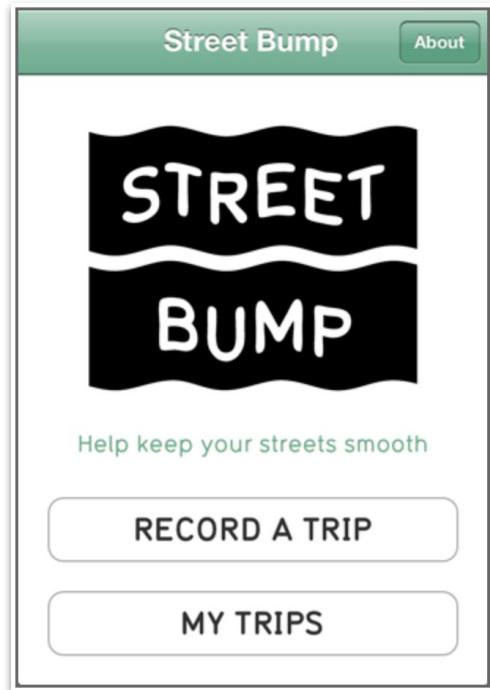
	Actual Disease = Yes	Actual Disease = No
Predicted Disease = Yes	True Positive	False Positive <i>Type 1 Error</i>
Predicted Disease = No	False Negative <i>Type 2 Error</i>	True Negative

Responsible AI: Development

The “garbage in, garbage out” problem



Bias: Sampling the Data



“ ...we need to ask which people are excluded. Which places are less visible? What happens if you live in the shadow of big data sets? ”

Kate Crawford

*Principal Researcher at Microsoft and Professor
at NYU Tandon School of Engineering*

Project Euphonia

Google Research Initiative to
collect data and **refine** speech
recognition algorithms to work
better for individuals with
speech impairments



Open Datasets and Crowdsourcing



Accent

23% United States English, **8%** England English, **5%** India and South Asia, **4%** Australian English, **3%** Canadian English, **2%** Scottish English, **1%** Irish English, **1%** Southern African, **1%** New Zealand English

Age

23% 19–29, **14%** 30–39, **10%** 40–49, **6%** < 19, **4%** 50–59, **4%** 60–69, **1%** 70–79

Industry Solutions: Datasheets for Datasets

Questions for dataset creators to reflect on during the key stages of the dataset lifecycle:

- ***Motivation***
- ***Composition***
- ***Collection Process***
- ***Preprocessing/ labeling***
- ***Uses***
- ***Distribution***
- ***Maintenance***

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Industry Solutions: Data Nutrition Labels

Metadata	
Filename	201612v1-docdollars-product_payments
Format	csv
Url	https://projects.propublica.org/docdollars/
Domain	healthcare
Keywords	Physicians, drugs, medicine, pharmaceutical, transactions
Type	tabular
Rows	500
Columns	18
Missing	5.2%
License	cc
Released	JAN 2017
Range	
From	AUG 2013
To	DEC 2015
Description	This is the data used in ProPublica's Dollars for Docs news application. It is primarily based on CMS's Open Payments data, but we have added a few features. ProPublica has standardized drug, device and manufacturer names, and made a flattened table (product_payments) that allows for easier aggregating payments associated with each drug/device. In [1], one payment record can be attributed to up to five different drugs or medical devices. This table flattens the payments out so that each drug/device related to each payment gets its own line.



A standard label that highlights the “key ingredients” of a dataset:

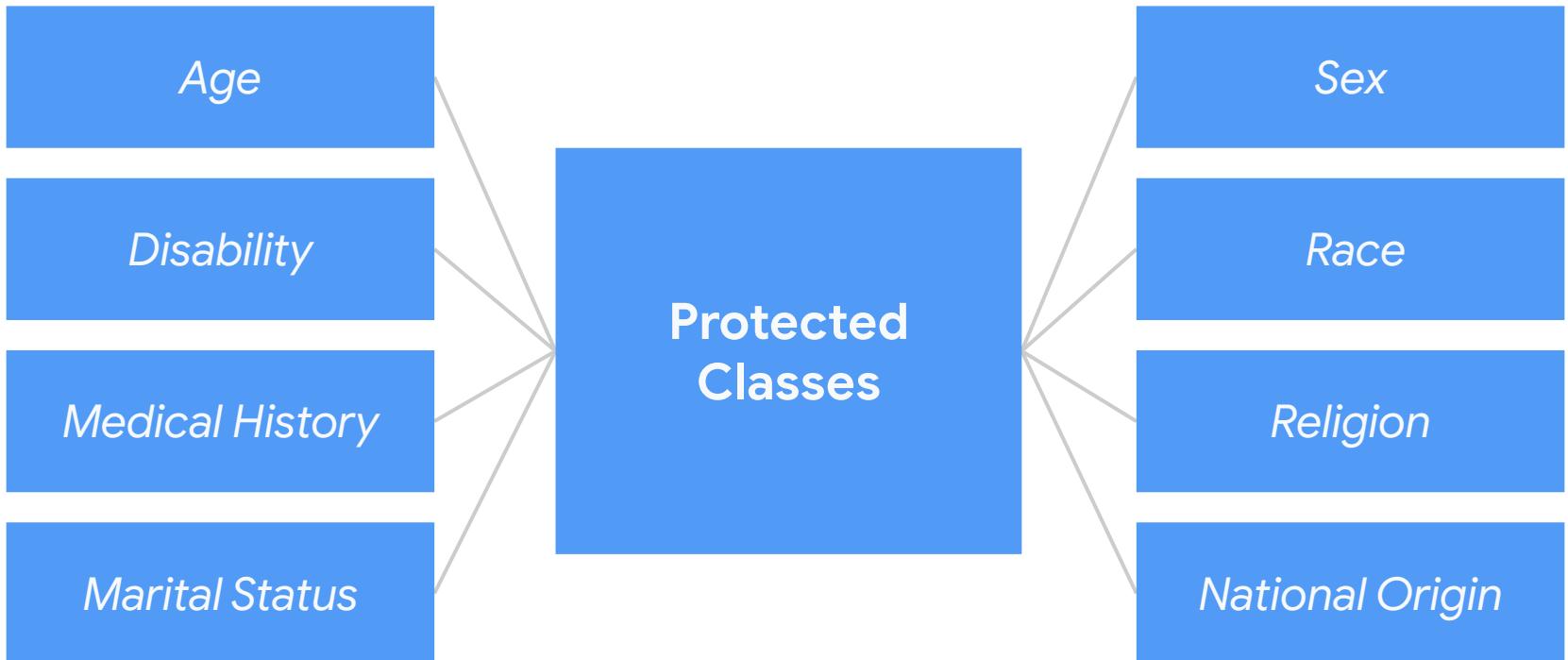
- ***Provenance***
- ***Metadata***
- ***Missing units***
- ***Variables***

Unfairness in ML

Model exhibits **discriminatory biases**, perpetuates **inequality** or performs less well for historically **disadvantaged groups**



- **All ML discriminates** (it just means to recognize a distinction, differentiate)
- Fairness is concerned with **wrongful** discrimination



1. Group Unawareness

Sensitive attributes are **not** included as features of the data (e.g. race, gender)

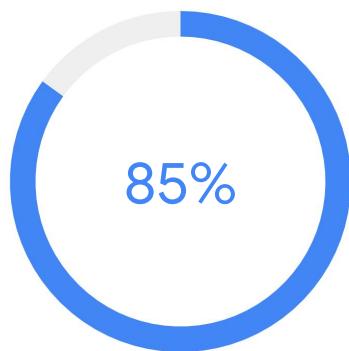


Pro: Avoids disparate treatment

Con: Possibility of highly correlated features that are proxies of the sensitive attribute

2. Group Threshold

Counteract historical biases in data by **adjusting** confidence thresholds *independently* for each group



Group A



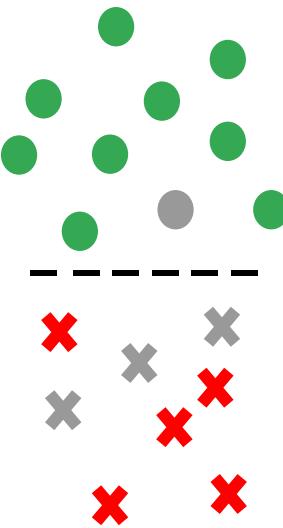
Group B

3. Equal Opportunity

	Actually Healthy = Yes	Actually Healthy = No
Predicted Healthy = Yes	True Positive	False Positive
Predicted Healthy = No	False Negative	True Negative

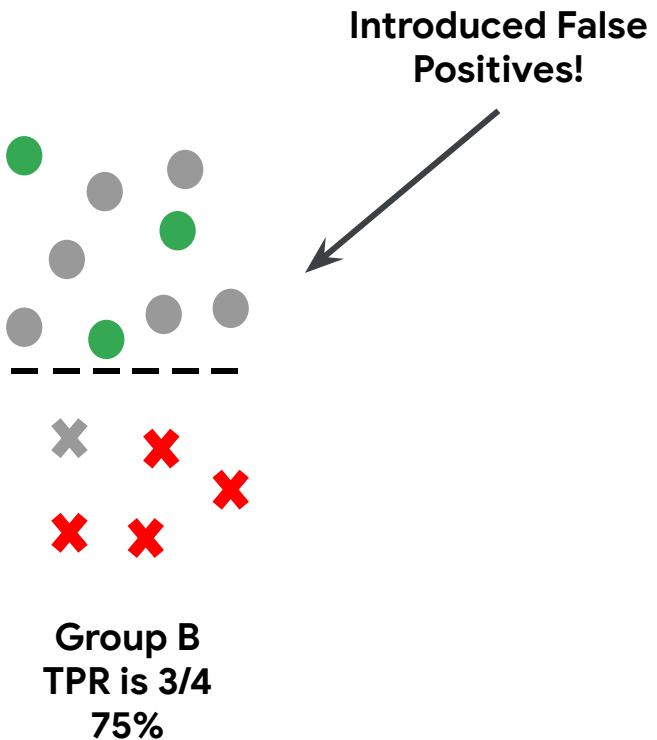
Qualified individuals should have an equal chance of being correctly classified for a desirable outcome.

Problem with Equality of Opportunity



Group A
TPR is $9/12$
75%

- True positive
- False positive
- True negative
- False negative



Group B
TPR is $3/4$
75%

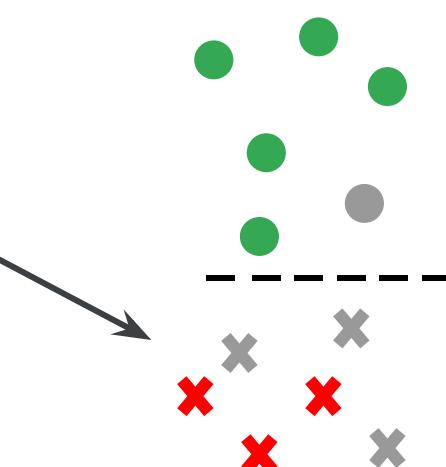
4. Equal Accuracy

	Actually Healthy = Yes	Actually Healthy = No
Predicted Healthy = Yes	True Positive	False Positive
Predicted Healthy = No	False Negative	True Negative

The percentage of correct classifications should be the same for all individuals

Problem with Equal Accuracy

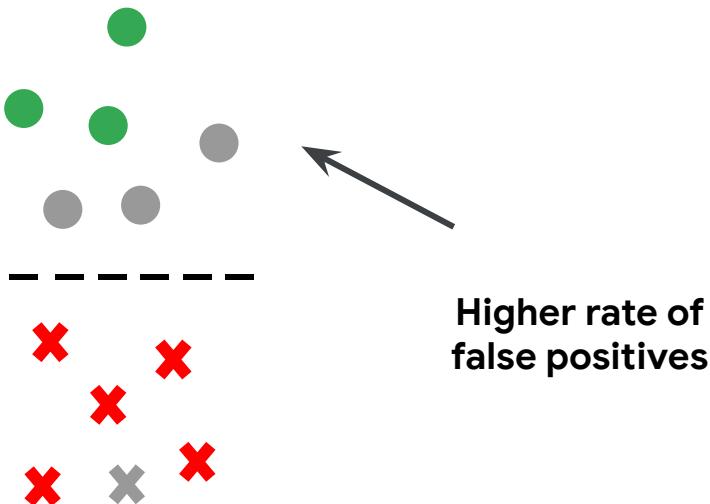
Higher rate of
false negatives



- True positive
- False positive
- True negative
- False negative

Group A
Accuracy is
75%

Higher rate of
false positives



Group B
Accuracy is
75%

Impossibility Theorem

We cannot satisfy all fairness metrics
at the same time!



For example:

- **Group Unawareness** is incompatible with **Group Threshold**
- **Equal Opportunity** is incompatible with **Equal Accuracy**

Industry Solutions: Bias Testing Toolkits

IBM Research Trusted AI

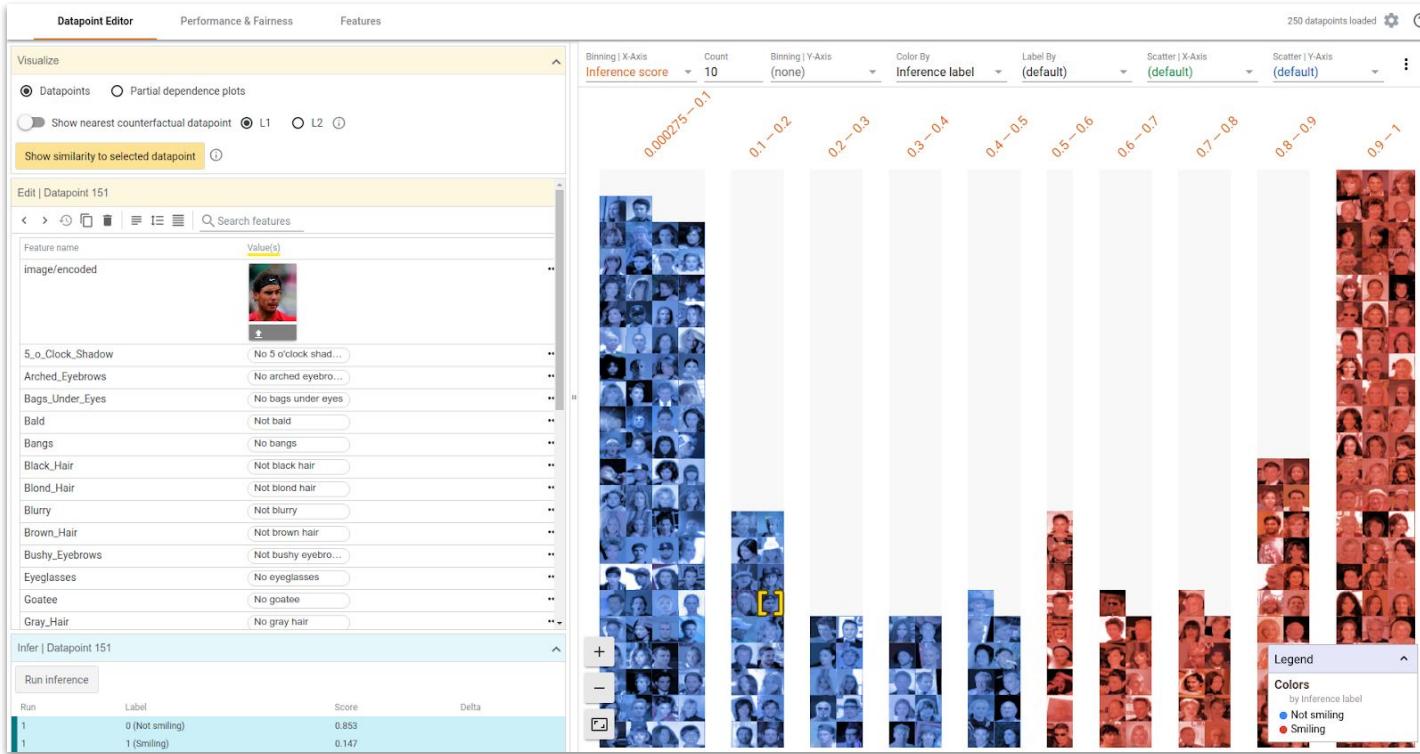


AI Fairness 360

This extensible open source toolkit can help you examine, report, and mitigate discrimination and bias in machine learning models throughout the AI application lifecycle. We invite you to use and improve it.

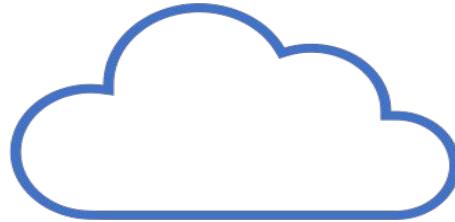


Google's What-If Tool



Responsible AI: Deployment

Privacy preserving?



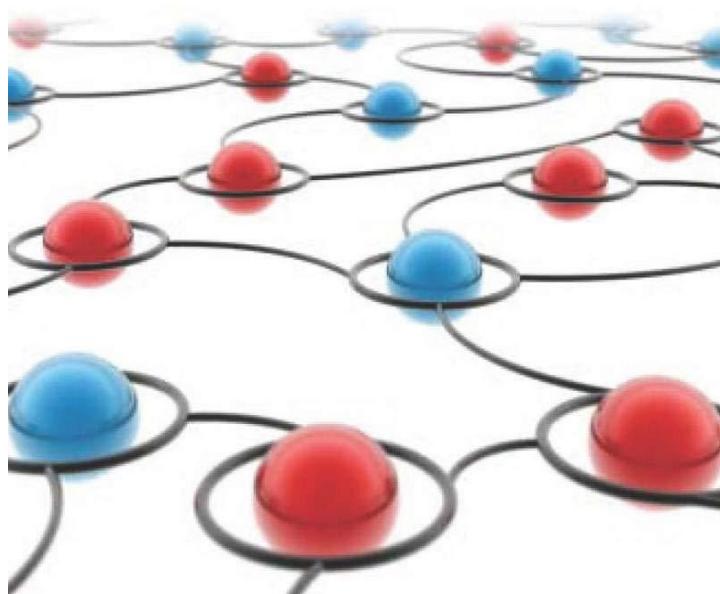
Privacy in Context

- Context shapes our expectations of privacy
- Privacy is a right to the appropriate flow of personal information (contextual integrity)
- Privacy can either be preserved or violated by the introduction of new technologies

PRIVACY IN CONTEXT

Technology, Policy, and the Integrity of Social Life

HELEN NISSENBAUM



Context-Relative Informational Norms



Context: What is the prevailing context?



Actors: Who are the subjects, senders and recipients of information?



Attributes: What is the type or nature of information?



Transmission principle: What are the constraints on the flow of information?



If the new practice results in any changes to these features, the practice is **flagged** as violating privacy



Context: Establish the prevailing context



Actors: Establish key actors



Attributes: Ascertain what attributes are affected



Transmission principle: Establish changes in transmission principles

Second Chances



Practices that are flagged as violating privacy may still be desirable all things considered

- Does the new practice provide better support for contextual values?
- Does it promote autonomy?
- Does it improve power relations?
- Does it create a fair distribution of costs and benefits



Justified?

Thousands of Amazon Workers Listen to Alexa
Users' Conversations

How can privacy be preserved?

- **Minimize**
 - Avoid collecting unnecessary data, and dispose or delete data periodically
- **Protect**
 - Use encryption techniques to protect data
- **Informed consent**
 - Be transparent with users about how their data is being collected and used
- **Map the flow of information**
 - Context, the type of information, and who has access, etc.

Responsible AI: Post-Deployment

Sustainability of TinyML

	Microprocessor	vs	Microcontroller
Platform		>	
Power	30W–100W	~1000X	150µW–23.5mW

Environmental Impact

Operational (Recurring)

- Product use
- Operational energy consumption
- e.g., training, inference

Capital (one-time)

- Supply chain for raw materials
- Chip manufacturing
- e.g., hardware production, transport, end-of-life processing

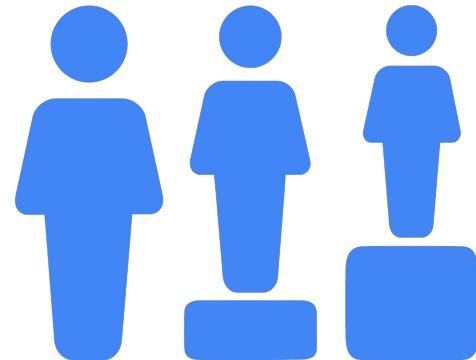
“ Development that meets the needs
of the **present** without
compromising the ability of future
generations to meet their **own** needs ”

World Commission on Environment and Development
Brundtland Report 1987

Equitable Resource Distribution

Equity

Fair distribution of burdens,
benefits, resources, etc.



- *Intragenerational justice*
Within a generation
- *Intergenerational justice*
Between generations

Sustainability Pledges



Carbon neutral since 2007, carbon free by 2030



100% renewable energy by 2025, carbon neutral by 2040



Microsoft

Carbon negative by 2030,
remove historical carbon emissions by 2050

Sustainability Calculator

Microsoft | Power BI Microsoft Sustainability... ▾

File Export Share ...

Contoso (Demo)

Microsoft carbon emissions from my company cloud usage

Most recent data available: Dec 16, 2020

Enrollment ID: All Subscription Name: All Azure Service: All Region: All Year: All Month: All

Carbon emissions (MTCO2e) for scopes 1, 2 and 3	Carbon emissions (MTCO2e) for scopes 1 and 2	Projected end-of-year carbon emissions (MTCO2e)	Carbon intensity (MTCO2e/usage)
34.42	22.46	3.91	0.1188086

▼ 22% compared to previous year

Microsoft carbon emissions from my company cloud usage: scope 1, 2 and 3 (MTCO2e)

Total emissions Remaining year projections

Year	Emissions (MTCO2e)
2017	15.86
2018	10.20
2019	5.00
2020	8.36

My company Microsoft cloud usage

Year	Usage (Tb)
2017	20.25
2018	80.21
2019	184.03
2020	344.4462

My company carbon intensity (MTCO2e/usage)

Quarter	Intensity (MTCO2e/usage)
Q4 17	3.444462
Q1 18	2.655800
Q2 18	1.125867
Q3 18	0.926316
Q4 18	0.741333
Q1 19	0.624000
Q2 19	0.506218
Q3 19	0.066098
Q4 19	0.055489
Q1 20	0.009965
Q2 20	0.000000
Q3 20	0.000000
Q4 20	0.000000

Dashboard

- Emissions details
- Emissions savings
- Preparation report
- Calculation methodology
- Learn more

AI is a science and an art form

There is no substitute for
critical thinking!

Embedded Ethics

