## An Introduction to Fairness in Machine Learning Using Fairlearn

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## Agenda

- Algorithmic Fairness
- Types of Fairness-related Harms
- Introduction to Tutorial

## Algorithmic Fairness

The idea that algorithmic systems should behave or treat people without unjust or prejudicial treatment on the grounds of sensitive characteristics.

#### Hiring

RETAIL OCTOBER 11, 2018 / 1:04 AM / UPDATED 2 YEARS AGO

#### Amazon scraps secret AI recruiting tool that showed bias against women



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

**Source**: <a href="https://www.reuters.com/article/us-amazon-com-jobs-automation-insight-idUSKCN1MK08G">https://www.reuters.com/article/us-amazon-com-jobs-automation-insight-idUSKCN1MK08G</a>

#### **Fraud Detection**

#### XENOPHOBIC MACHINES

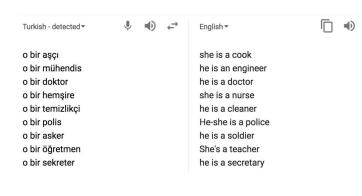
DISCRIMINATION THROUGH UNREGULATED USE OF ALGORITHMS IN THE DUTCH CHILDCARE BENEFITS SCANDAL



#### Source:

https://www.amnesty.nl/content/uploads/2021/10/2021 1014 FINAL Xenophobic-Machines.pdf?x42580

#### **Translation**



**Source**: https://qz.com/1141122/google-translates-gender-bias-pairs-he-with-hardworking-and-she-with-lazy-and-other-examples/

## Types of Harm

Majority of fairness research focuses on these two harms

#### **Allocation**

The system extends or witholds opportunities, resources, or information.

#### **Quality-of-Service**

The system does not work equally well for all groups.

#### Representation

The development/usage of the system overrepresents or underrepresents certain groups.

#### **Stereotyping**

The system reinforces stereotypes.

> Most prevalent in unstructured data

#### **Denigration**

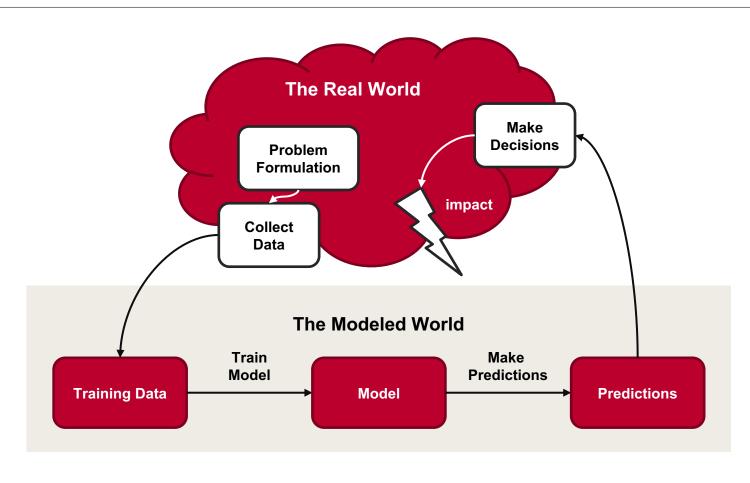
The system is actively derogatory or offensive. The system makes decisions in

a way that violates social norms.

**Procedural** 

Closely related to interpretable machine learning

## What are we optimizing for?



mizing for? What are we Does our problem formulation reflect the real-world context and our (moral) Does the values? Is the data a good system produce me Real World representation of fair outcomes? reality? Make **Decisions Problem Formulation** impact Collect Data Does the model Are we optimizing for make fair the right performance predictions? metric? The Modeled World Train Make Model **Predictions Training Data** Model **Predictions** 

### Conclusion

- Machine learning systems can reproduce, amplify, and introduce unfairness.
- There are different types of fairness-related harms, today we will focus on:
  - allocation harm
  - quality-of-service harm.
- Fairness-related harms can arise due to a mismatch between what we optimize for and what we actually value.



```
if questions:
    try:
        answer()
    except RuntimeError:
        pass
else:
    print("Thank You.")
```

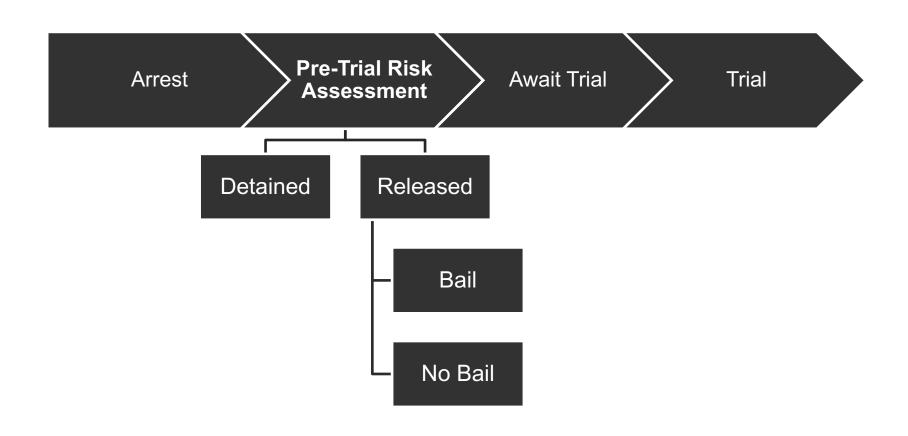
# Tutorial Pre-Trial Risk Assessment

## Learning Objectives

After completing this tutorial, you will be able to:

- apply group fairness metrics in Python;
- explain several **trade-offs** between different group fairness criteria;
- explain how threats to construct validity may impact downstream fairness-related harms;

## **Context** Pre-trial Risk Assessment in the United States



### **Propublica's Analysis of COMPAS**

In 2017, Propublica found that **COMPAS**, a recidivism prediction algorithm used by judges in the United States, failed differently for African-American defendants compared to white Americans.

	White	African- American
False Positive	23.5%	44.9%
False Negative	47.7%	28.0%

**Source:** <a href="https://www.propublica.org/article/machine-bias-risk-assessments-in-criminal-sentencing">https://www.propublica.org/article/machine-bias-risk-assessments-in-criminal-sentencing</a>

