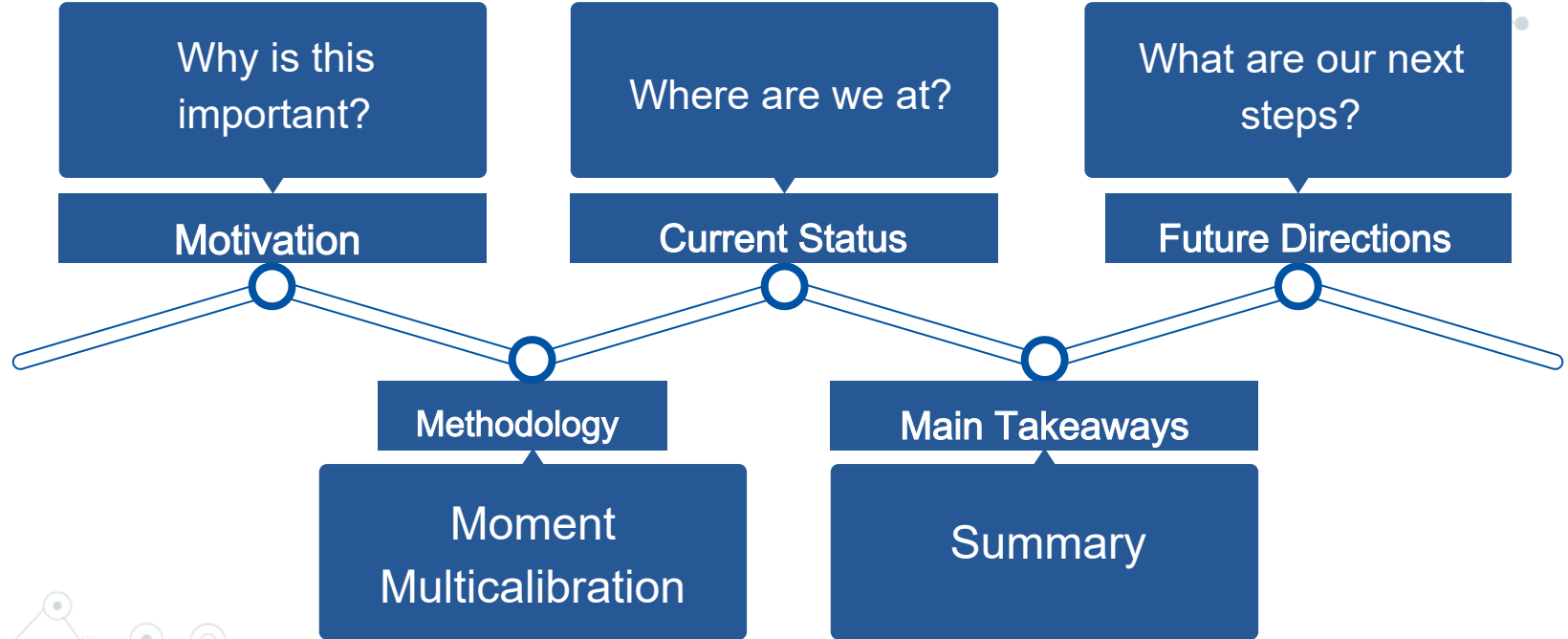




e q u a l m o d e l

A Post-Processing Algorithm for Bias Reduction in Big Data Analytics

Roadmap



Machine Learning is becoming more prevalent but there are **consequences** ...

PROPUBLICA Why America Fails at Gathering Hate Crime Statistics

DOCUMENTING HATE

Why America Fails at Gathering Hate Crime Statistics

The FBI relies on local law enforcement agencies to identify and report crimes

How our data encodes systematic racism

Artificial Technologists must take responsibility for the toxic ideologies that our data sets and algorithms reflect.

Facebook's ad-serving algorithm discriminates by gender and race

Even if an advertiser is well-intentioned, the algorithm still prefers certain groups of people over others.

In 2016, Microsoft's Racist Chatbot Revealed the Dangers of Online Conversation

The bot learned language from people on Twitter—

Racism and discrimination in health care: Providers and patients

POSTED JANUARY 16, 2017, 9:30 AM , UPDATED JULY 09, 2020, 12:34 PM



Monique Tello, MD, MPH

Contributor

The Apple Card Didn't 'See' Gender—and That's the Problem

the way its algorithm determines credit lines makes the risk of bias more acute.

Let's first ground our discussion...

How sure are you?

But I am part of a demographic representing less than 5% of the population

Given features x , your dosage for a drug is $f(x)$.

The variance conditional on my estimate is $g(x)$.

For Asian Americans under the age of 50, the confidence interval is $[a,b]$

For women with a family history of diabetes, the confidence interval is $[c,d]$



Key Observation

The dosage prediction is averaged over the population, *not* an individual, so the dose might not be accurate for an individual.

Problem

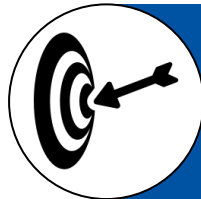


Implicit bias against underrepresented populations in the systems we rely on



Ethical standards of both fairness and privacy are breached

Goal



- 1) Make a tool that can supplement any existing algorithm, making it more fair
- 2) Reduce implicit bias

Solution: Multicalibration

- ◎ **Calibration** assures that our predictions are accurate overall
 - Fails to make the same guarantee for subpopulations
 - E.g. 90% accuracy for the total population does not guarantee 90% accuracy for a subpopulation
- ◎ **Multicalibration** offers the same assurance across all possible subgroups

Project Components & Resources



Software

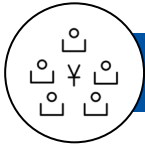


PyTorch



Papers

- © Moment Multicalibration for Uncertainty Estimation (Jung, Lee, Pai, Roth, Vohra)
- © Multiaccuracy: BlackBox PostProcessing for Fairness in Classification (Kim, Ghorbani, Zou)



Stakeholders

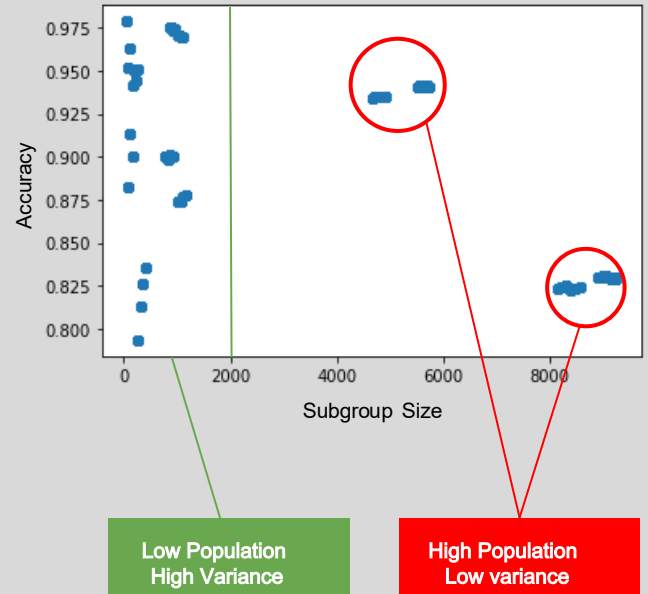
- © Cary Coglianese, Edward B. Shils Professor of Law and Professor of Political Science
- © The Defender Association of Philadelphia

The Problem with ML in Criminal Justice

- ◎ **Data Description** : Data combines socioeconomic data, law enforcement data, and crime data
- ◎ **Goal** : predict violent crime number
- ◎ **Problem**: (1) High variance in accuracies for underrepresented people. (2) Models not calibrated to underrepresented people will only cause further harm



An Imbalanced Dataset



Algorithm Overview

Auditor

- Select subgroups
 - list of predefined subgroups
 - learning oracle algorithm
- Decide whether the subgroup prediction is calibrated

Fixer

- Adjust predictions for the chosen subgroup
- Return updated result to the Auditor

Repeat until
Multicalibrated



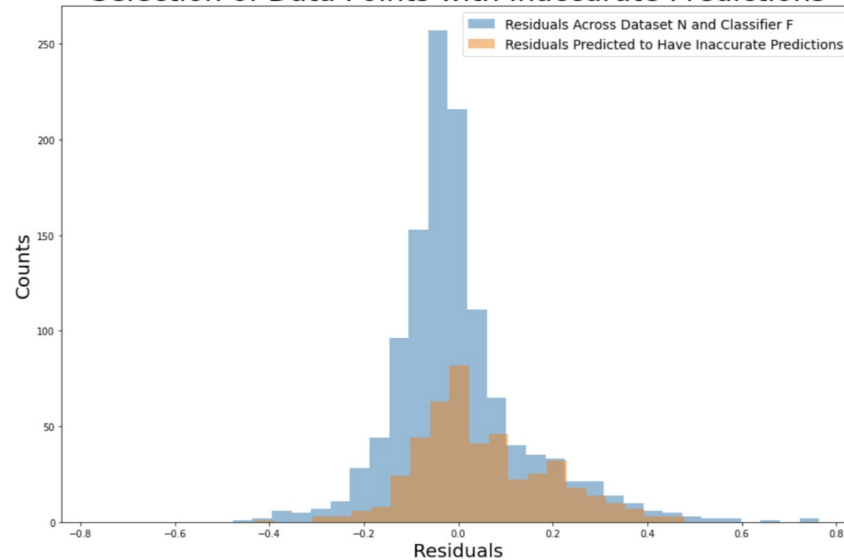
Auditor Visualization



Key Result

This algorithm creates a classifier to predict points in a dataset that will likely have inaccurate predictions

Selection of Data Points with Inaccurate Predictions



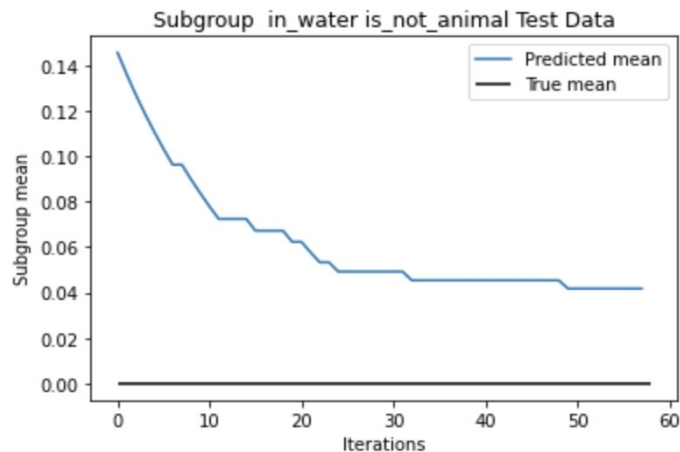
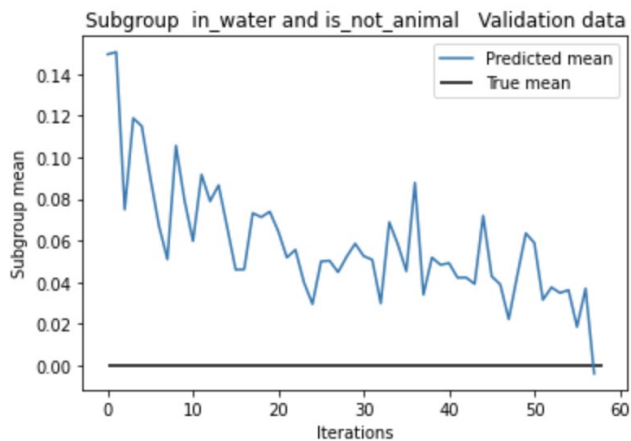
residuals ($n.$) The difference between the prediction and true label

Fixer Visualization



Key Result

After *T* iterations of post-processing, predicted mean is closer to the true mean



Mean predictions adjusted during post-processing for validation data



Mean predictions adjusted during post-processing for test data

Main Takeaways

1. **Implemented** mean multicalibration based on the algorithm in the paper by Jung, Lee, Pai, Roth, Vohra
2. **Tested** the auditor and fixer on two different datasets
3. **Demonstrated** the results showing that both components work

Future Steps

1. **Testing:** *evaluate* the algorithm on a variety of datasets
2. **Application:** *run* algorithm on specific use cases such as housing and medication
3. **Publishing** : *put* our code on a publicly accessible site like IBM AI Fairness 360 Package for ML developers to utilize

Contact Us @

eseseniordesign20202lteam06@gmail.com



Hyewon Lee



Trish la
Pokharna



Brian Handen



Tashweena
Heeramun



Margaret Ji