Investigating discrimination bias in predictive modelling

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Background: The Problem

In recent years, many scandalous examples have shown that statistical models trained on large amounts of data can "act" discriminatory. Examples include:

- Adds of high-income jobs being shown less frequently to women, presumable becasue they've been predicted to be less interested or suitable¹
- ▶ Black people's health status being underestimated, leading to inappropriate healtch care measures²
- Black people begin predicted a higher risk for crime recidivism, leading to higher penalties³

¹@datta automated 2015

²@obermeyer_dissecting_2019

³ProPublica (2016)

Project Aims

- ► How can we quantify fairness in order to be able to evaluate algorithmic fairness?
- What methods are available to increase algorithmic fairness? In what type of situations do they apply? (i.e. In what kind of situations can we expect them to be successful?)

Background: Why Discrimination Bias?

- ightharpoonup Correlation between outcome y and protected charateristic x_p
- Correlation between important predictors x_i and protected carachteristic x_p
- Undersampling of groups with protected protected carachteristic x_p

Possible Solutions

Pre-Processing	Training	Prediction
Resampling	Penalty	Threshold
Mapping	Model bias	adjustments
Altering labels	Tuning for fairness	Alter predictions

We've chosen to work with resampling and threshold adjustment.

Possible Goals

Demographic parity

$$P(Y = 1|X = 1) = P(Y = 1|X = 0)$$

Equalized odds

$$P(G = 1|X = 0, Y = 1) = P(G = 1|X = 1, Y = 1)$$

Data: COMPAS

Models

Model	Tuning
Random Forest	Number of predictors sampled at each sp
Artificial neural net	Number of hidden nodes
Logistic ridge regression	Penalisation
K-nearest neighbour (left out)	Number of neighbours
AdaBoost	None

Accuracy and Fairness for the Initial Models

(plot)

Method 1

Method 2

Method 3

Comparison

Final Model Fit

Conclusions

Slide with R Output

summary(cars)

```
speed
                     dist
##
   Min. : 4.0
##
                Min. : 2.00
##
   1st Qu.:12.0 1st Qu.: 26.00
##
   Median: 15.0 Median: 36.00
   Mean :15.4 Mean : 42.98
##
##
   3rd Qu.:19.0
                3rd Qu.: 56.00
   Max. :25.0 Max. :120.00
##
```

Slide with Plot

