

# **Fairness Beyond Disparate Treatment & Disparate Impact: Learning Classification without Disparate Mistreatment (Zafar, Muhammad Bilal, et al.)**

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Github Link: [TianXie1999/Replicate-Paper: Replicate fairness paper \(github.com\)](https://github.com/TianXie1999/Replicate-Paper: Replicate fairness paper (github.com))

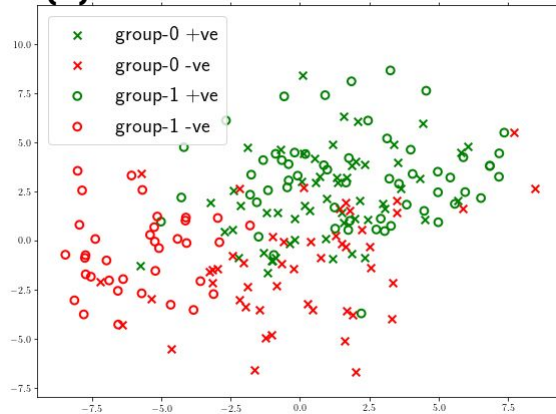
# Why and What?

valuable understandings on trade-off between fairness and accuracy

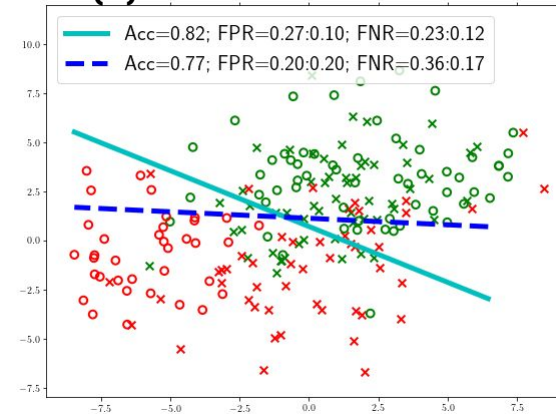
- **a fairness metric derived from "disparate mistreatment"** : with respect to a sensitive attribute, the model would have misclassification rates differ for groups of people having different values of that sensitive attribute (e.g., blacks and whites)
- **Visualize the fairness-accuracy trade-off:** replicate several experiments on synthesized data and real-world data to visualize balance between fairness and accuracy

# Synthesized Data

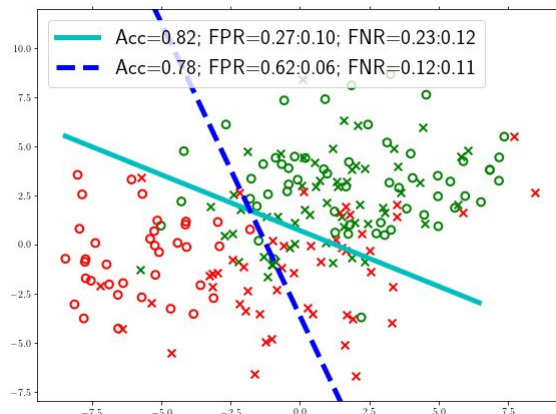
(a) Data Set Visualization



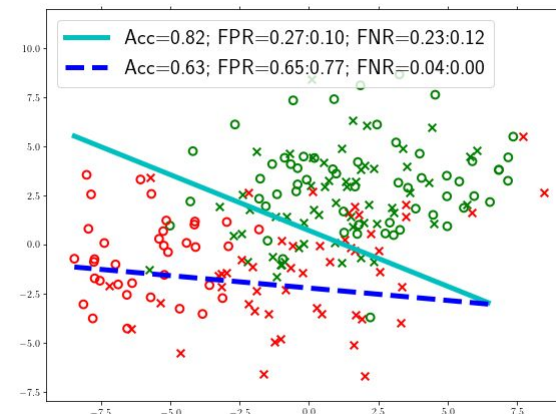
(b) Constraint on FPR



(c) Constraint on FNR

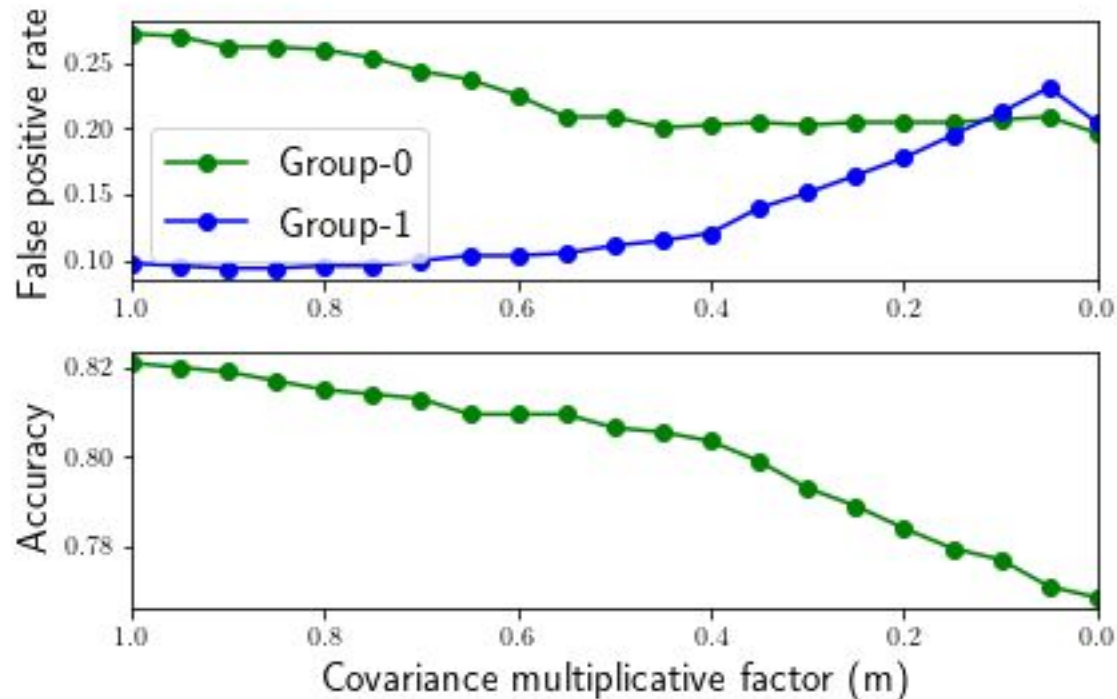


(d) Constraint on FPR&FNR



multivariate normal distribution: 2 non-sensitive features and 1 sensitive feature

# Accuracy-Fairness Trade-off While relaxing the Threshold



Measure fairness using the **covariance** between the users sensitive attributes and the signed distance between the feature vectors of misclassified users and the classifier decision boundary

# Real-world data:

## ProPublica COMPAS data set

- Crime offenders for different demographic groups
- Logistic Regression
- **Sensitive attribute:** race
- **Y:** whether the individual recidivated

### == Unconstrained (original) classifier ==

Accuracy: 0.671

|| s || FPR. || FNR. ||

|| 0 || 0.35 || 0.32 ||

|| 1 || 0.15 || 0.59 ||

### == Constraints on FPR ==

Accuracy: 0.653

|| s || FPR. || FNR. ||

|| 0 || 0.28 || 0.41 ||

|| 1 || 0.24 || 0.51 ||

## Credit Card data from Kaggle

- Credit status of client, personal and applicant information  
->predict future defaults and credit card borrowing
- Logistic Regression
- **Sensitive attribute:** gender
- **Y:** whether the client has good credit (paid off that month)

### == Unconstrained (original) classifier ==

Accuracy: 0.515

|| s || FPR. || FNR. ||

|| 0 || 0.38 || 0.58 ||

|| 1 || 0.36 || 0.64 ||

### == Constraints on FPR ==

Accuracy: 0.514

|| s || FPR. || FNR. ||

|| 0 || 0.39 || 0.57 ||

|| 1 || 0.35 || 0.65 ||