# Machine Learning Fairness

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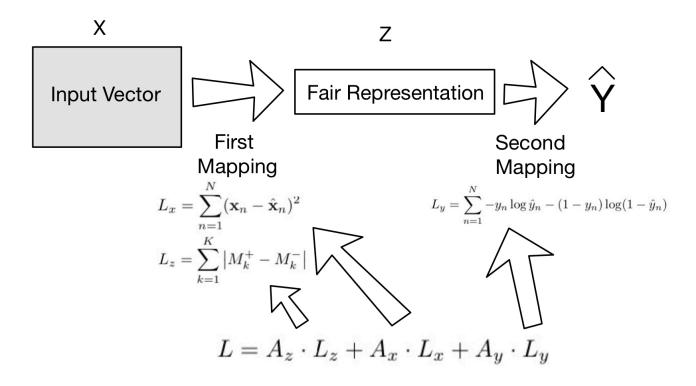
#### **Evaluation Metrics**

- Accuracy (Proportion of Correct Prediction)
- Individual Fairness (Similar Individuals Receive Similar Classification)
- Group Fairness (Similar or Equivalent Classification Distribution)
- Calibration (Difference of Accuracy Between Groups)

## Algorithms

- 1. Learning Fair Representations
- 2. Maximizing Accuracy under Fairness Constraints (C-SVM & C-LR)
- 3. Maximizing Fairness under Accuracy Constraints (gamma & fine-gamma)

#### **Learning Fair Representation**



A here is a hyper-parameter controlling the trade-off between accuracy and fairness.

#### Maximize Accuracy under Fairness Constraint

C-SVM & C-LR

minimize 
$$L(\boldsymbol{\theta})$$
  
subject to  $\frac{1}{N} \sum_{i=1}^{N} (\mathbf{z}_i - \bar{\mathbf{z}}) d_{\boldsymbol{\theta}}(\mathbf{x}_i) \leq \mathbf{c},$   
 $\frac{1}{N} \sum_{i=1}^{N} (\mathbf{z}_i - \bar{\mathbf{z}}) d_{\boldsymbol{\theta}}(\mathbf{x}_i) \geq -\mathbf{c},$ 

Letter c here is a hyper-parameter controlling the trade-off between accuracy and fairness.

## Maximize Fairness under Accuracy Constraint

gamma & fine-gamma

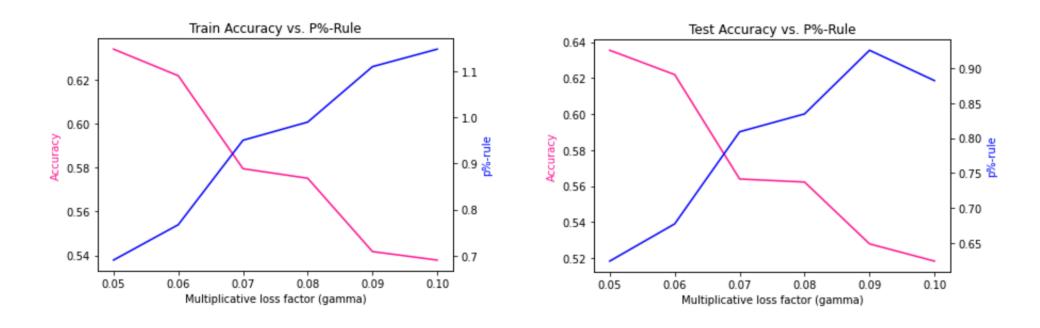
minimize 
$$\left|\frac{1}{N}\sum_{i=1}^{N} (\mathbf{z}_i - \bar{\mathbf{z}}) d_{\boldsymbol{\theta}}(\mathbf{x}_i)\right|$$
 subject to  $L(\boldsymbol{\theta}) \leq (1 + \gamma)L(\boldsymbol{\theta}^*),$ 

Gamma here is a hyper-parameter controlling the trade-off between accuracy and fairness.

#### **Evaluation**

	A1 (LFR)	A2 (C-SVM & C-LR)	A3 (gamma & fine-gamma)
Accuracy	53.66%	66%	56.39%
Calibration	11.77%	4.94%	1.38%
Protected Accuracy	60.69%	63.96%	NA
Non-protected Accuracy	48.91%	68.91%	NA
Individual Fairness	0.8118	NA	NA
p-rule	NA	NA	95%

# Hyper-Tuning Visualization



Negative Correlation Between Fairness and Accuracy

#### Reference

- Muhammad Bilal Zafar, Isabel Valera, Manuel Gomez Rodriguez, and Krishna P. Gummadi. Fairness Constraints: Mechanisms for Fair Classification. https://arxiv.org/abs/1507.05259
- Rjcbard Zemel, Yu Wu, Kevin Swersky, and Toniann Pitassi. Learning Fair Representations. http://proceedings.mlr.press/v28/zemel13.pdf