## 3. Fairness: Definitions and metrics

### What is fairness?

### Many definitions of fairness:

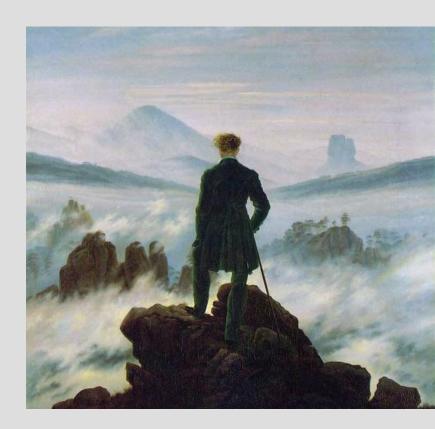
- Individual. seeks for similar individuals to be treated similarly
- Group. split a population into groups defined by protected attributes (e.g. women, black) and seeks for some measure to be equal across groups



## Individual fairness

Am I as an individual being treated as another with similar qualifications?

- Fairness through unawareness.
- Fairness through awareness.
- Counterfactual fairness.



## Fairness through unawareness

• **Definition:** Protected attributes A not explicitly used as features X.

Mathematically: A∩X = Ø

Potential problem: proxy in the data (see practice notebook).

## Fairness through awareness/ Individual Fairness

• **Definition:** Similar predictions to similar individuals (w.r.t similarity metric).

• Mathematically: x,y 2 individuals. Model M must satisfy Lipschitz condition:

$$D(M(x), M(y)) \le d(x, y)$$

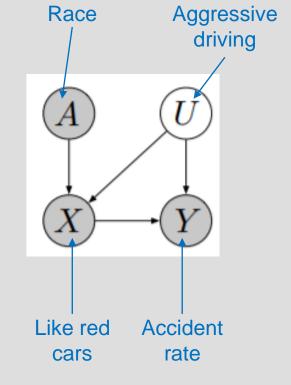
Potential problem: Choice of similarity/distance metric.

## Counterfactual fairness

- **Definition:** Causal graph. Fair if the predicted outcome does not depend on a descendant of the protected attribute
- Mathematically: A protected attribute, X other features, U latent background variables.  $\hat{Y}$  estimator.

$$P(\hat{Y}_{A < -a}(U) = y | X = x, A = a) = P(\hat{Y}_{A < -a}(U) = y | X = x, A = a)$$

Potential problem: determining the causal graph



## Group fairness

Is the minority group to which I belong being treated as the majority group?

- Equality of Outcome. Distribution across groups should be the same (Statistical Parity, Disparate Impact)
- Equality of Opportunity. E.g. Equalized Odds



## **Equality of Outcome**

- Definition: Equal likelihood of positive predicted outcome.
- Mathematically:

$$P(\hat{Y} = 1|A = male) = P(\hat{Y} = 1|A = female)$$

$$SR_{male}$$
 $SR_{female}$ 

- Example metrics:
  - Disparate Impact =  $SR_{female} / SR_{male} \rightarrow$  ideal value of 1
  - Statistical Parity =  $SR_{female} SR_{male} \rightarrow$  ideal value of 0
- Potential problem: Positive discrimination ?

## **Equality of Opportunity**

- **Definition:** This should be equal for all groups:
  - probability of a person in the positive class being correctly assigned a positive outcome
  - · probability of a person in a negative class being incorrectly assigned a positive outcome

• Mathematically: 
$$P(\hat{Y} = 1 | A = m, Y = y) = P(\hat{Y} = 1 | A = f, Y = y), y \in \{0,1\}$$

$$TPR_m \text{ if } y = 1; FPR_m \text{ if } y = 0$$

• **Example metric:** Average Odds Difference

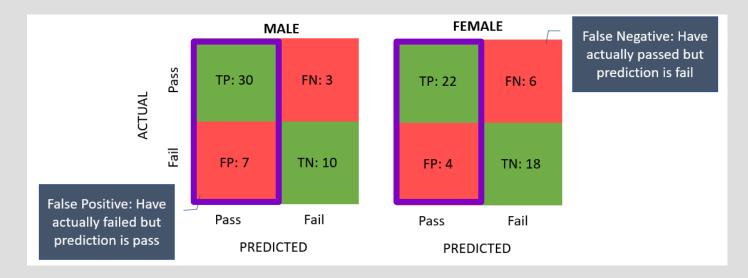
$$AOD = \frac{1}{2}[(FPR_f - FPR_m) + (TPR_f - TPR_m)]$$

 Potential problem: Where does ground-truth come from, is it a valid unbiased measure?

# Group fairness - Illustration through an example

### Equality of outcome →

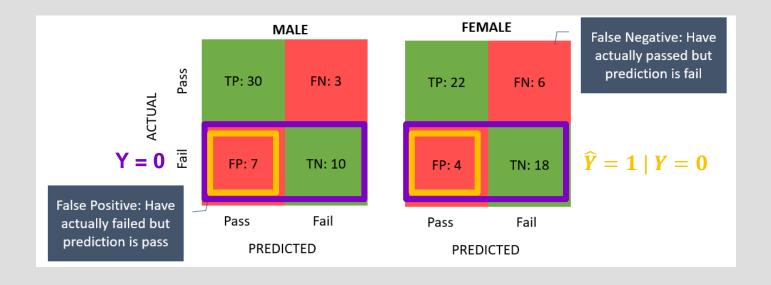
$$P(\widehat{Y} = 1|A = male) = P(\widehat{Y} = 1|A = female)$$



$$SR_{male} = 37/50 = 0.74$$
 $SR_{female} = \frac{26}{50} = 0.52$ 
 $SR_{female} = 0.52 = 0.52$ 
 $SR_{female} = 0.52 = 0.52$ 
 $SR_{female} = 0.52 = 0.52$ 

### Equality of opportunity >

$$P(\widehat{Y} = 1 | A = male, Y = 0) = P(\widehat{Y} = 1 | A = female, Y = 0)$$

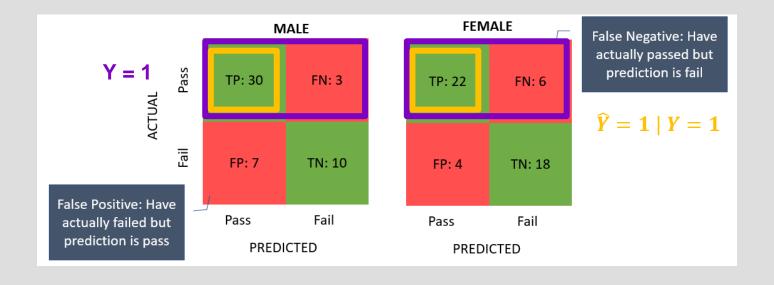


$$P(\hat{Y} = 1 | A = m, Y = 0) = FPR_{male} = \frac{FP}{FP + TN} = 7/17 = 0.41$$

$$P(\hat{Y} = 1 | A = f, Y = 0) = 4/22 = 0.18$$

### Equality of opportunity $\rightarrow$

$$P(\hat{Y} = 1 | A = male, Y = 1) = P(\hat{Y} = 1 | A = female, Y = 1)$$



$$P(\hat{Y} = 1 | A = m, Y = 1) = TPR_{male} = \frac{TP}{TP + FN} = 30/33 = 0.91$$

$$P(\hat{Y} = 1 | A = f, Y = 1) = 22/28 = 0.79$$

### **Average Odds Difference**

$$AOD = \frac{1}{2}[(0.18 - 0.41) + (0.79 - 0.91)] = -0.175$$

## Reading/References

### **Supporting Notebook**

### **Definitions of fairness**

(Verma & Rubin, 2018) <u>Fairness Definitions Explained</u> (Mehrabi et al., 2021). <u>A survey on bias and fairness in machine learning</u>

#### **Metrics**

https://www.holisticai.com/open-source