
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:** $A \cap X = \emptyset$
- **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:

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

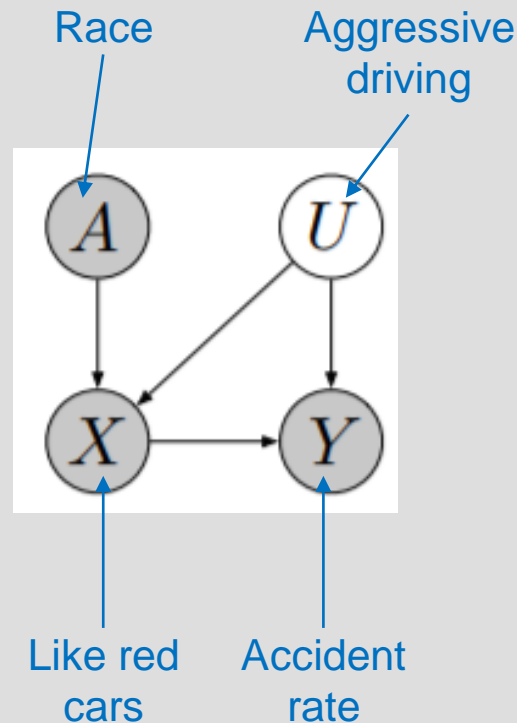
- **Potential problem:** Choice of similarity/distance metric.

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.

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

- **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:**

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

- **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:**
$$\underbrace{P(\hat{Y} = 1 | A = m, Y = y)}_{\text{TPR}_m \text{ if } y = 1 ; \text{FPR}_m \text{ if } y = 0} = P(\hat{Y} = 1 | A = f, Y = y), y \in \{0,1\}$$
- **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 →

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

MALE

ACTUAL	Pass	TP: 30	FN: 3
	Fail	FP: 7	TN: 10
		Pass	Fail
		PREDICTED	

False Positive: Have actually failed but prediction is pass

FEMALE

ACTUAL	Pass	TP: 22	FN: 6
	Fail	FP: 4	TN: 18
		Pass	Fail
		PREDICTED	

False Negative: Have actually passed but prediction is fail

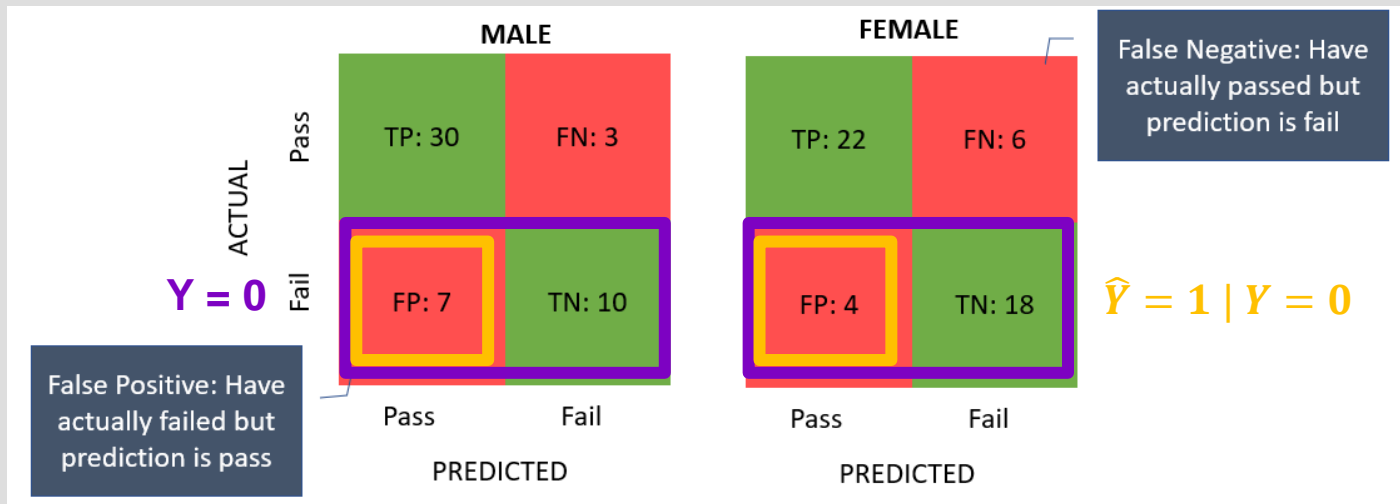
$$\begin{cases} SR_{male} = 37/50 = 0.74 \\ SR_{female} = \frac{26}{50} = 0.52 \end{cases}$$



$$\begin{cases} SP = 0.52 - 0.74 = -0.22 \\ DI = 0.52/0.74 = 0.7 \end{cases}$$

Equality of opportunity →

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

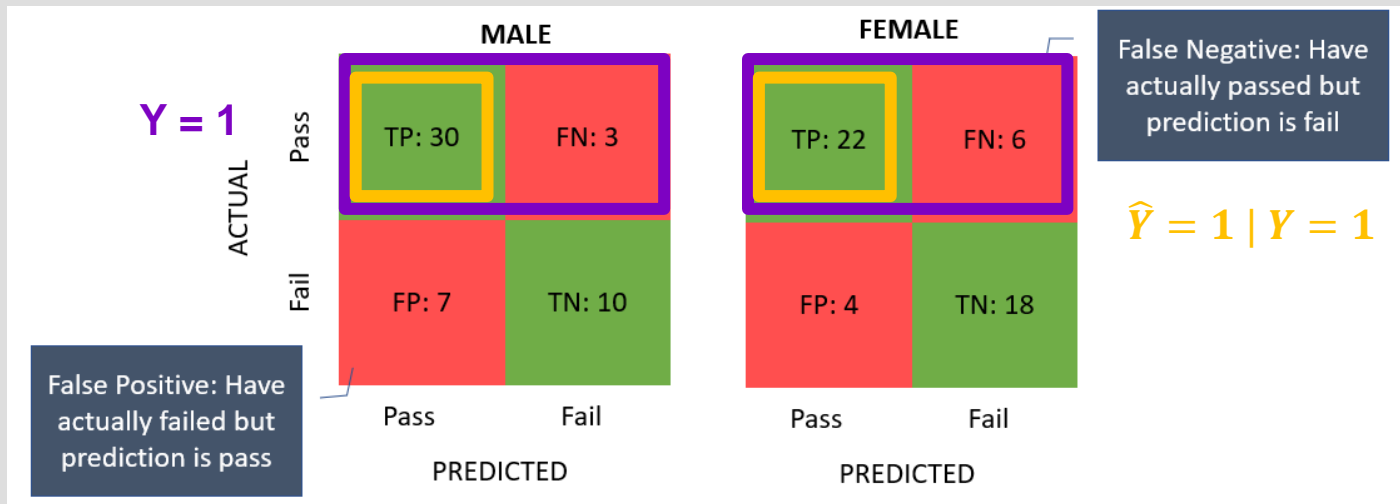


→

$$\begin{cases} P(\hat{Y} = 1 | A = m, Y = 0) = FPR_{male} = \frac{FP}{FP + TN} = 7/17 = 0.41 \\ P(\hat{Y} = 1 | A = f, Y = 0) = 4/22 = 0.18 \end{cases}$$

Equality of opportunity →

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



→

$$\begin{cases} P(\hat{Y} = 1 | A = m, Y = 1) = TPR_{male} = \frac{TP}{TP + FN} = 30/33 = 0.91 \\ P(\hat{Y} = 1 | A = f, Y = 1) = 22/28 = 0.79 \end{cases}$$

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) Fairness Definitions Explained

(Mehrabi et al., 2021). A survey on bias and fairness in machine learning

Metrics

<https://aif360.mybluemix.net/>