

Responsible AI: Group Fairness



1. Case study: Male fashion models

The model agency Top Models uses an application (i.e., ADS) that creates a list of potential male fashion models based on their social media profile pictures.

The classification algorithm is trained on images of their current male model pool: <https://www.wilhelmina.com/new-york/men/main/>

Desirable features: short hair, beard, muscular, tall, etc.

Sensitive/protected attribute: Race (For the sake of this example, groups: White and Asian).



2. Train set

Imbalanced --> Asian males are underrepresented in the data --> problematic because the classifier learns the bias; i.e., it is very unlikely that a male model is Asian/it is very likely that a male model is White.

Asian males: High TN and FN

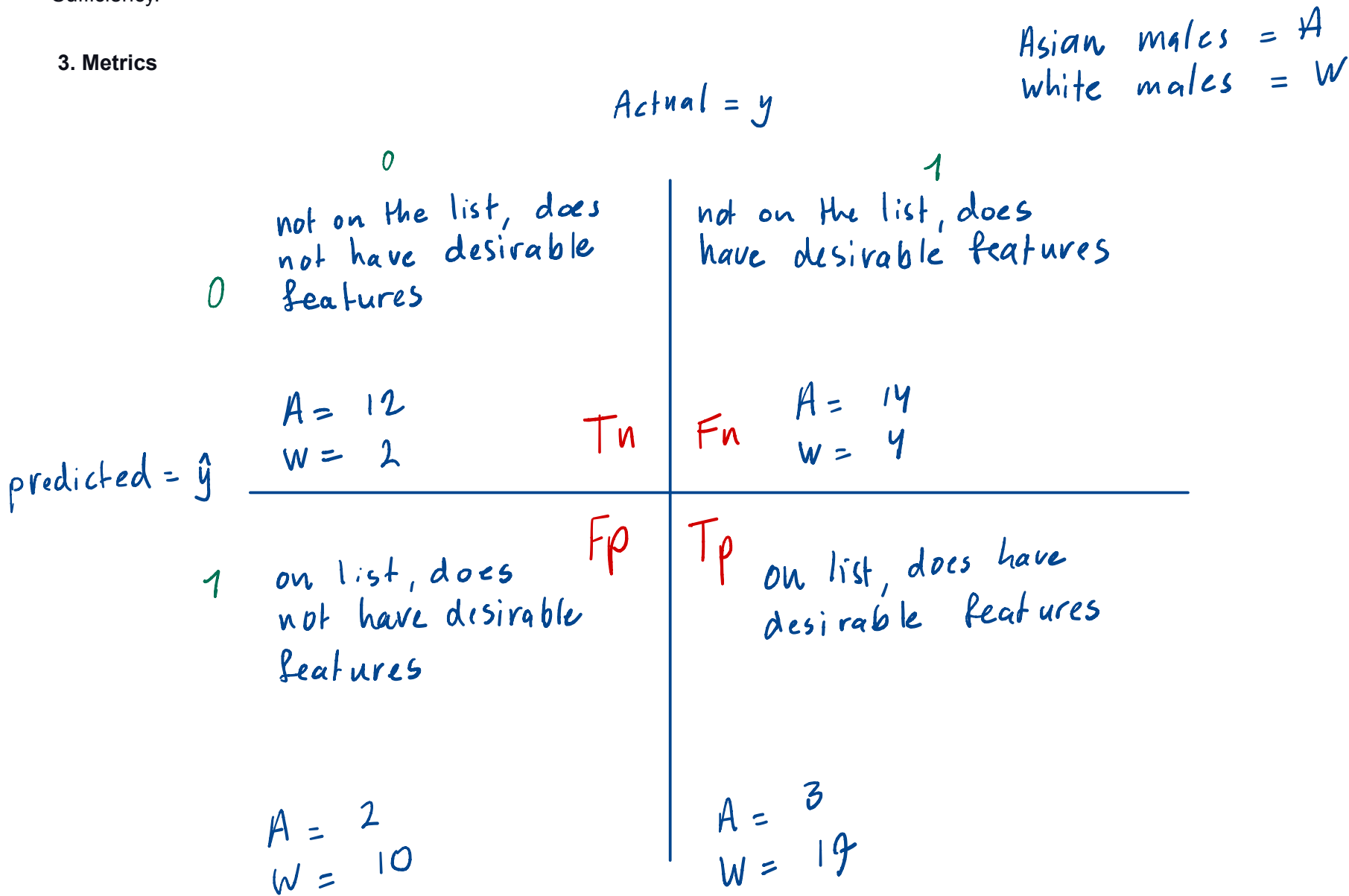
White males: High TP and FP

Representative --> Train data is representative of the population; i.e., there are more White male fashion models than Asian male fashion models.

- Historical bias --> Equal base rates across sensitive/protected attribute groups --> Independence.

Allow for unequal base rates --> investigate if the difference in equal base rates is caused by the sensitive/protected attribute --> Separation or Sufficiency.

3. Metrics



Calculate (positive) base rates for Asian and White males:

$$\text{Base rate} = \frac{F_n + T_p}{F_n + T_p + F_p + T_n}$$

Asian males = $\frac{14 + 3}{14 + 3 + 2 + 12} = 0.54 \rightarrow$ of all the outcomes 54% was positive

White males = $\frac{4 + 17}{4 + 17 + 10 + 2} = 0.63$

Separation: 'Gotta catch 'em all' --> recall --> positive outcome is favorable (e.g., hiring example).

What proportion is truly positive (i.e., correctly classified as positive); A male that is on the list, and has the desirable features.

$$T_{PR} = \frac{T_p}{T_p + F_n}$$

Asian males = $\frac{3}{3 + 14} = 0.17 \rightarrow$ of the positives only 17% was predicted correctly
↳ decrease F_n , increase F_p

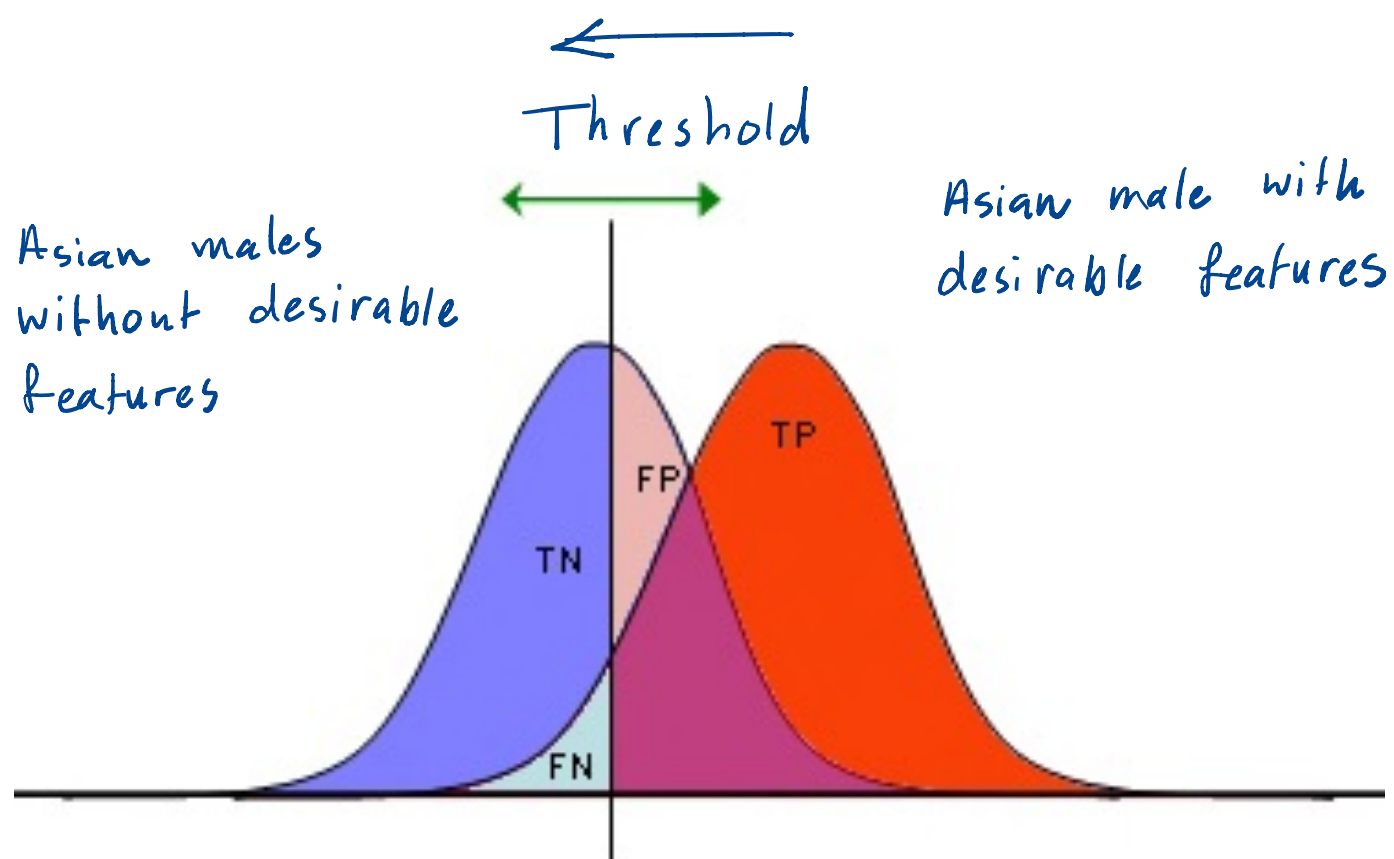
White males = $\frac{17}{17 + 4} = 0.80$
↳ increase F_n , decrease F_p

Equalized Opportunities: Equal chance to get a correct prediction, i.e., be on the list generated by the application, when they have the desirable features.

False positive.

High	unfair for individuals: some Asian males favoured without having desired features	
low	fair	unfair for individuals: some white males disfavoured while having the desired features
	Low	High

False negatives



for white males the threshold goes the other way.