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# Fairness in AI: navigating ethical challenges

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**1. Definition**

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# 1. Definition

**Fairness:** the quality of treating people equally or in a way that is right or reasonable.

**Bias:** the action of supporting or opposing a particular person or thing in an unfair way, because of allowing personal opinions to influence your judgment



# Why is this an issue?

Not all biases are unjust, but the term is most often used to indicate an unfair advantage or disadvantage for a certain group of people.





## 2. Criteria

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We define whether an AI system is unfair in terms of its impacts on people:

- Fairness-related harms, including harms involving people's individual experience with AI.



# Examples

AI systems can unfairly allocate opportunities, resources, or information.

Instance 1: AI CV-screening system trained using the CVs of people currently employed in the tech industry, where women are already underrepresented, may conceal employment opportunities from women.

# Examples

AI systems can fail to provide the same quality of service to some people as they do to others.

Instance 1: a facial recognition system that has technical issues recognizing faces with particular skin tones will not provide the same quality service to every user.





# Examples

AI systems can under represent groups of people, or treat them as if they don't exist.

Instance 1: in response to the query “chief executive officer”, a search system that mainly returns images of men in response may underrepresent non-male executive officers.



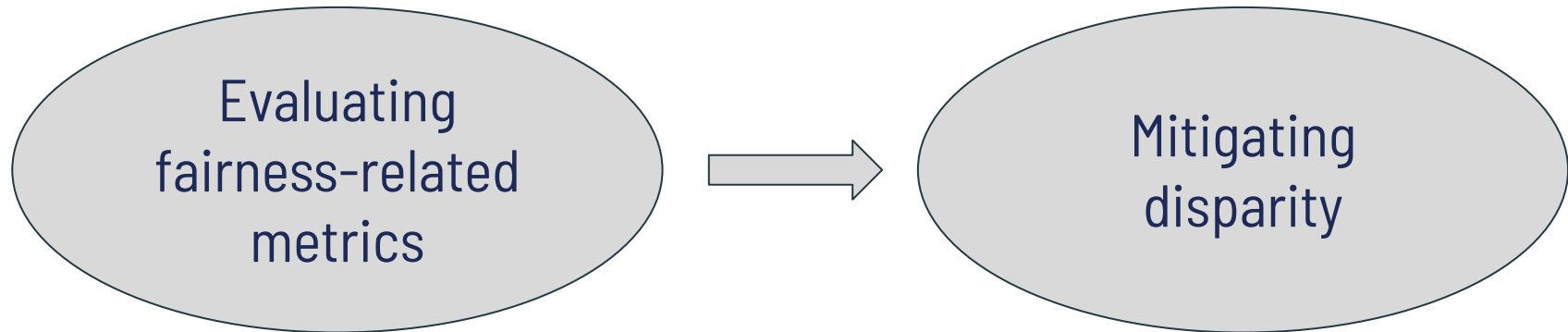


# 3. Practical Application

### 3. Fairlearn

It is an open-source, community-driven project to help data scientists improve fairness of AI systems.

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## ≡ Fairlearn

```
data = fetch_adult()

X = data.data
y_true = (data.target == ">50K") * 1
sex = X["sex"]

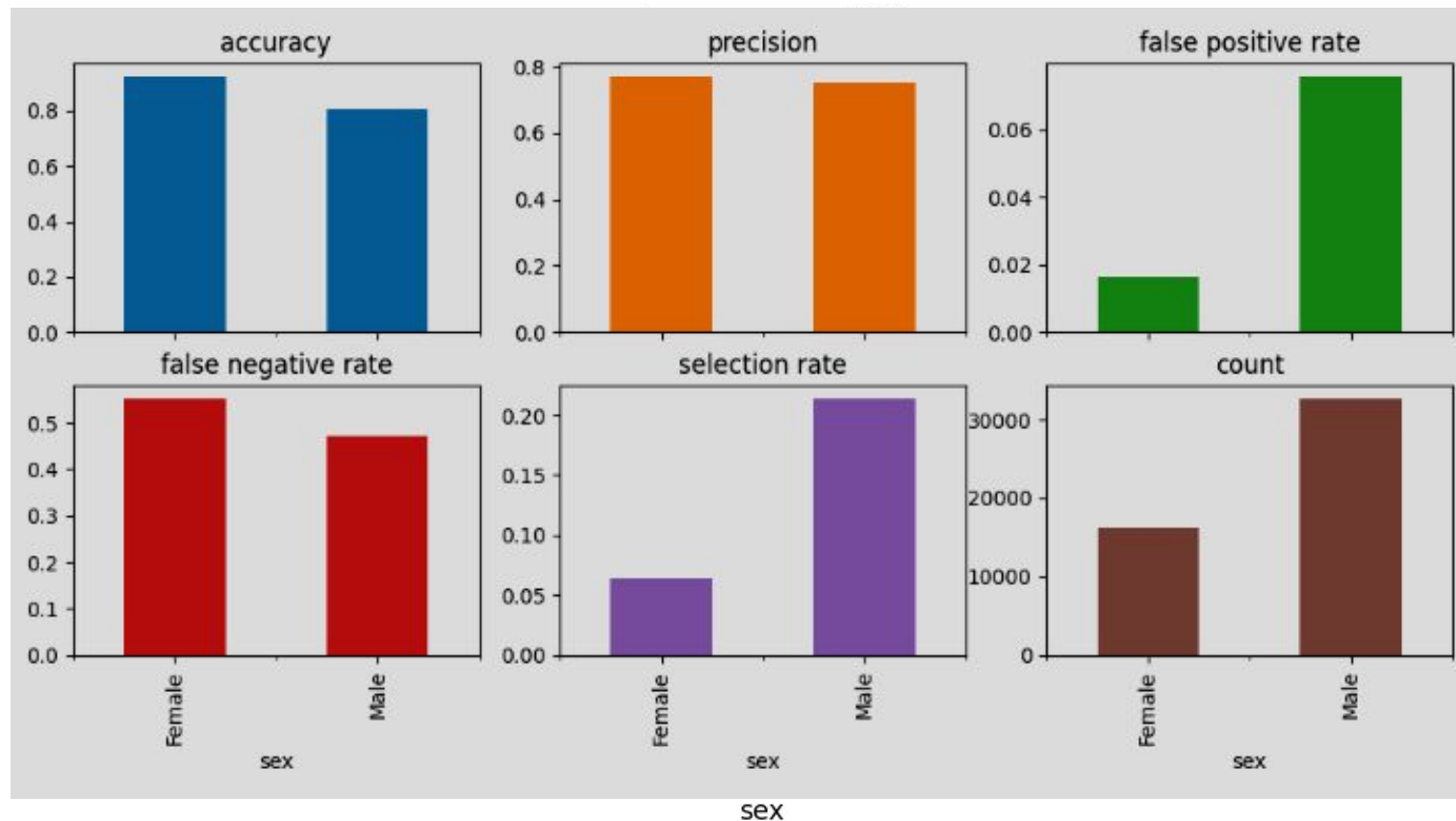
selection_rates = MetricFrame(
    metrics=selection_rate, y_true=y_true, y_pred=y_true, sensitive_features=sex
)

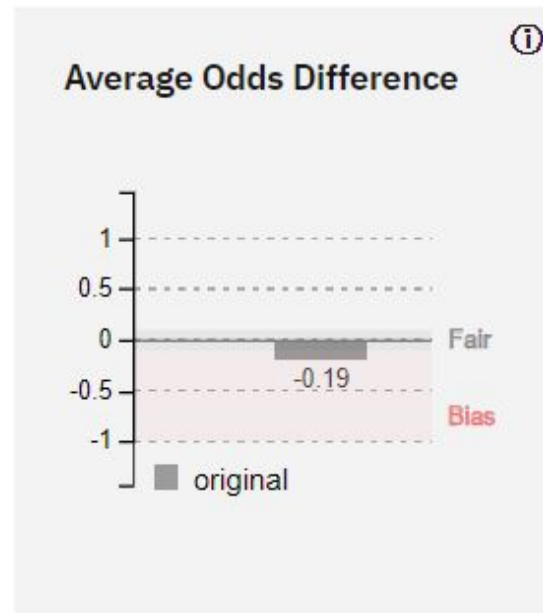
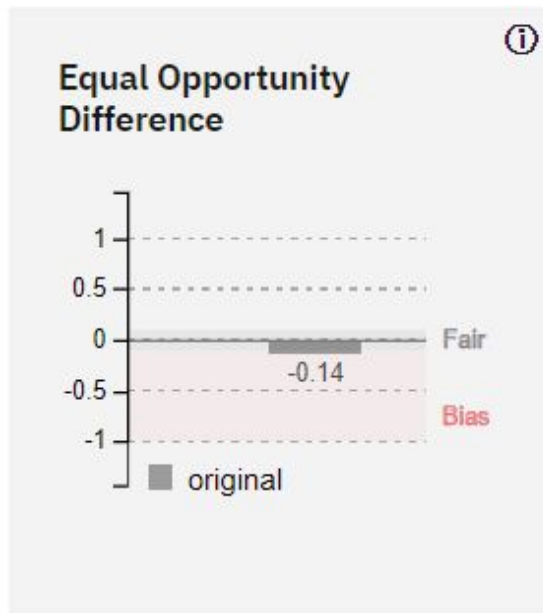
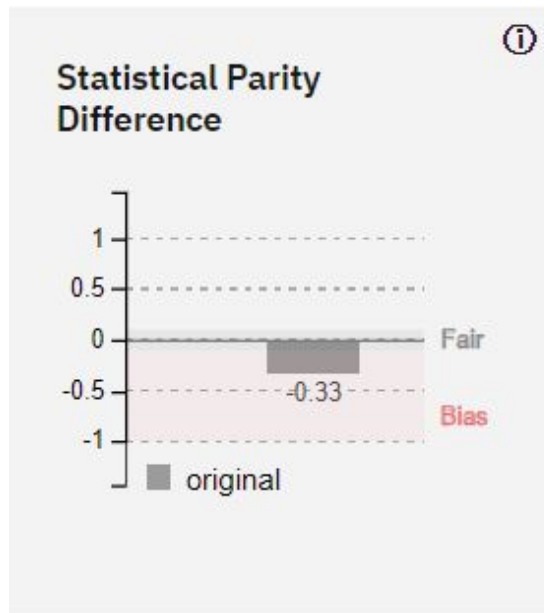
fig = selection_rates.by_group

fig.plot.bar(
    legend=False, rot=0, title="Fraction earning over $50,000"
)

plt.show()
```

Evaluating  
fairness-related  
metrics





# References

- Fairlearn: [A toolkit for assessing and improving fairness in AI, Microsoft](#), Microsoft, version 2020/09/22.
- Fairlearn Homepage: <https://fairlearn.org/>, last updated 2024
- Author: gonzalodom11 et al: [Why we should trust AI?](#), Github, version 2024/04/20
- Ai Fairness 360 Homepage: <https://aif360.res.ibm.com/>, last updated 2024/04

## Recommendations to read:

- Fairlearn: [A toolkit for assessing and improving fairness in AI, Microsoft](#), Microsoft, version 2020/09/22.
- Caltech. (n.d.). Can we trust artificial intelligence? Retrieved from <https://scienceexchange.caltech.edu/topics/artificial-intelligence-research/>
- Ruf, B., & Detyniecki, M. (2021, September 30). Towards the right kind of fairness in ai. arXiv.org. <https://arxiv.org/abs/2102.08453>.



A top-down photograph of a wooden-framed letterboard with a black felt surface. The words "Thank You" are written in white, serif, all-caps letters. The board is placed on a rustic wooden table. To the bottom left is a portion of a vintage orange rotary telephone. To the bottom right is a portion of a vintage typewriter. A green leafy plant is visible at the top edge.

Thank  
You