Using Computer Science to tackle unwanted bias in AI

Daphne Lenders

 \boxtimes daphne.lenders@uantwerpen.be





What is AI?















Save time of tax administrators

Potential for higher accuracy







| | | BI | |
|-------------|--------------------|-------------------|---------------------|
| Nationality | Reported Income | Omitted Income | Potential Fraud? |
| Belgian | 30.000€ | 200€ | No |
| Belgian | 15.000€ | 0€ | No |
| Belgian | 20.000€ | 1000€ | No |
| Non-Belgian | 17.000€ | 500€ | Yes |





| Natio | nality | | Potential Fraud? |
|-------|--------|---|---------------------|
| Belg | jian | | No |
| Belg | jian | | No |
| Belg | jian | | No |
| Non-B | elgian | < | Yes |
| | | | |





• Taking over existing biases





• Taking over existing biases

| • | Amplifying existing biases | | | |
|---|--|--|---------------------|------------------------|
| | Nationality | | Potential Fraud? | AI Fraud Prediction |
| | Belgian | | No | No |
| | | | | |
| | Non-Belgian | | Yes | Yes |





• Taking over existing biases

| • | Amplifying exist | ing biases | | |
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| | | | | |
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| | Non-Belgian | | No | Yes |





- Taking over existing biases
- Amplifying existing biases
- Black Boxes







- Demographic Parity
- Equal Opportunity
- Individual Fairness





- Demographic Parity
 - Comparing base rates of AI predictions
 - 10% of Belgian people are flagged
 - 30% of Non-Belgian people are flagged
 - \rightarrow 20% difference; thus unfair!





- Demographic Parity
 - Comparing base rates of AI predictions
 - 10% of Belgian people are flagged
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 - \rightarrow 20% difference; thus unfair!

What if Non-Belgian people commit more fraud?





- Equal Opportunity
 - Comparing how "correct" the classifier is across groups





Equal Opportunity

Belgian

| Potential Fraud? | AI Fraud Prediction | |
|---------------------|------------------------|--|
| No | No | |
| No | No | |
| No | No | |
| Yes | Yes | |

Of all people who were not flagged by human, **100%** was not flagged by AI

Non-Belgian

| Potential Fraud? | AI Fraud Prediction |
|------------------|------------------------|
| No | No |
| No | Yes |
| Yes | Yes |
| Yes | Yes |

Of all people who were not flagged by human, **50%** was not flagged by AI





Equal Opportunity

| Belgian | | | Non-Belgian | | |
|----------------------|------------------------|---------|------------------|------------------------|--|
| Potential Fraud? | AI Fraud Prediction | Po F | tential raud? | AI Fraud Prediction | |
| No | No | | No | No | |
| No | No | | No | Yes | |
| No | No | | Yes | Yes | |
| Yes | Yes | | Yes | Yes | |

This algorithm does not satisfy equal opportunity!





Equal Opportunity

| Belgian | | Non-Belgian | | |
|---------------------|------------------------|---------------------|------------------------|--|
| Potential Fraud? | AI Fraud Prediction | Potential Fraud? | AI Fraud Prediction | |
| No | No | No | No | |
| No | No | No | No | |
| No | No | Yes | Yes | |
| Yes | Yes | Yes | Yes | |

This algorithm does satisfy equal opportunity

But what about human bias?





- Equal Opportunity
 - Comparing how "correct" the classifier is across groups
 - Makes sure that existing bias is not increased
 - Does not solve the problem of bias in human labels





- Individual Fairness
 - Determine for one individual at a time whether s/he got discriminated

But how to measure this?





Measuring bias – No Silver Bullet

- There's no such thing as a perfect definition of fairness
- How we measure bias may depend on the problem
 - Do we want to fundamentally change a decision process?
 - Are we okay with making mistakes?
 - How high is the risk of preserving existing biases?
- Legal Definitions?





How to tackle bias?

- Many approaches
- Again no silver bullet
- One example: Situation Testing





Situation Testing

- Approach taken from Social Sciences
- Trying to achieve Individual Fairness
- Idea: similar people should be treated alike

| NO FRAUD | FRAUD |
|-------------------------|--------------------------|
| Omitted Income: 500€ | Omitted Income: 500€ |
| Reported Income: 20000€ | Reported Income: 20000€ |
| Nationality: Belgian | Nationality: Non-Belgian |





Situation Testing

- Can't always find "equal" people (that only differ on sensitive attribute)
- Look at "similar" people instead
- How to define similarity?
 - Computer Science can help here!

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Situation Testing

- Still more questions
 - How many people do we have to compare?
 - How similar is similar enough?





Conclusion

- Tackling bias in AI has gained interest but ...
 - No silver bullet
 - Still many open questions
 - Lack of a clear legal framework





It's Time to Combine Forces



