





### BEYOND PRIVACY Learning Data Ethics

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# Data for Diversity-Aware Technology: Ethical Considerations

### Insights from the project WeNet – The Internet of Us

Supported by

Author: Laura Schelenz, International Center for Ethics in the Sciences and Humanities, Tübingen, Germany

## **Data for Diversity-Aware Technology**



### **Ethical Challenges**

Operationalization of diversity

Diversityawareness in tech development teams

Pattern inference through machine learning



Collection of large amounts of sensitive data

Protection of privacy of data subjects, data minimization

Representation of minorities in the dataset





## WHAT IS DIVERSITY-AWARE TECHNOLOGY?

### **Diversity-aware technology...**



- leverages the diversity of technology users to their advantage
  - Diversity helps achieve a «good» outcome = diversity as **instrumental** value
- can help reduce computer-mediated bias against certain social groups
  - Diversity realizes the goal of inclusion and representation of minorities = diversity as instrumental/**intrinsic** value
- mirrors the diversity of its users  $\rightarrow$  how can we **operationalize** diversity?

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## What kind of diversity?

- WeNet understands diversity in terms of social practices
  - = routine behavior
  - e.g. cooking, riding a bike
- Large amounts of sensitive data: eating habits, transportation, shopping, use of university buildings, student performance

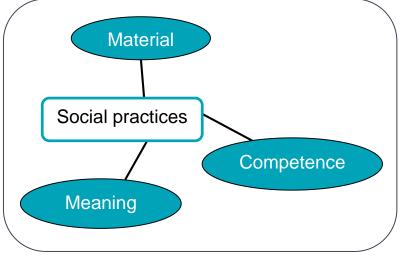
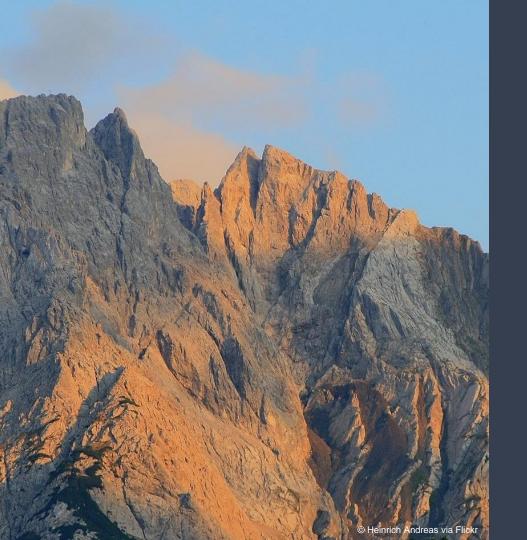


Figure 1: Operationalization of social practices in WeNet

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## ETHICAL CHALLENGES

### **Ethical challenges**



- the need to collect massive amounts of sensitive data
- data minimization, data protection, and privacy rights

how to account for minorities in the dataset
 implicit bias and constraints of category-building

- how to account for minorities in computer models
- →machine learning and statistics

### **Ethical challenge #dataprotection**



### Diversity-aware technology poses risks to data subjects

#### Misuse, loss, hacking of information

particularly risky if sensitive data is involved

#### Easy identification of data subjects

the rarer and more "dispersed" the data points, the easier it will be to trace the information back to the data subjects

#### Discrimination of marginalized groups

the more data that is "out in the open", the more data can be used against a person

#### Easy to circumvent Article 5, data minimization

risk in claiming the need to collect vast amounts of data for diversityaware technology

### **Ethical challenge #equalrepresentation**



Implicit bias = scripts that prevent us from recognizing "full" diversity

#### **Implicit bias**

stems from "schemas" that we internalize at a young age and that are activated subconsciously; schemas help interpret a situation (cf. Sally Haslanger)

#### Example from WeNet: social practice of "working"

contract-based employment vs. reproductive labor

- social practices are coded to ideas of gender
- implicit bias in the operationalization of diversity may result in the marginalization of female users





#### George (user)

- student assistant (5hrs a week) at a local NGO
- has a one-year contract
- wants to gain experience and extra money



#### Aisha (user)

- takes care of her grandma every weekday between 7 and 9 pm (when her mother has to leave for her night shift)
- annoyed because cannot meet friends

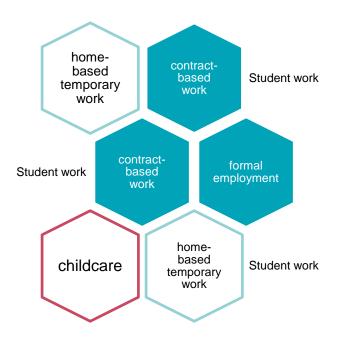
### **Ethical challenge #algorithmicjustice**

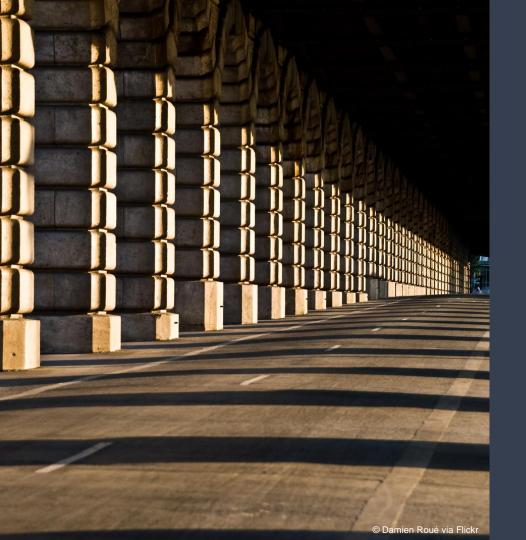


#### Computer models built from data sets must be diversity-aware

# Machine learning and pattern recognition

- Diversity represented in the dataset may be further reduced by machine learning methods
- Algorithms may be optimized for the majority of the population but not minorities







### RECOMMENDATIONS

### Recommendations



- Diversity-aware technology needs interdisciplinary cooperation
- Develop diversity-aware technology that leverages diversity for a "good" outcome and ensures non-discrimination
- Protect data subjects' privacy, explore innovative solutions that help represent diversity by collecting *less* data
- Develop a data collection plan that explicitly seeks to reduce bias in the dataset; answer the question "How do we account for minorities in the dataset in a way that properly represents them?"
- Test how the computer models fare with regard to fairness; answer the question "How do our models affect minorities and is there disparate treatment resulting from our technology?"
- Increase diversity-awareness in tech development teams: provide training to enhance sensitivity to questions of gender, race, and class discrimination

### **GET IN TOUCH**

Website <u>www.internetofus.eu</u> <u>www.izew.uni-tuebingen.de</u>

Email <u>laura.schelenz@uni-tuebingen.de</u>

Twitter
<u>@WeNetProject</u>
<u>@LauraSchelenz</u>



## **THANK YOU!**



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