

Comments slides for Tuesday, November 17:
Privacy and ethics, part 1

Matthew J. Salganik

COS 597E/SOC 555 Limits to prediction
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- ▶ Moral vs adversarial
- ▶ Find something that works for you

Kosinki et al.

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- ▶ Unanticipated secondary use. Are you worried about that for Fragile Families Challenge? How does Kosinki's other work influence how you think about the ethics of this work?

Duhigg and Christl et al.

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- ▶ Can you measure not just tracking but harm?

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