On Valuing and Procuring Personal Data



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Part 1: Background (manifesto)

- What does it mean to own personal data?
- How might people and firms value personal data?

Part 2: Research on valuing and procuring data efficiently

- Active-learning based (with Jake Abernethy, Yiling Chen, C.J. Ho)
- Prediction-market based (with Raf Frongillo; Jake Abernethy; Justin Harris)

Part 3: Discussion (screed)

1 What does it mean to **own**, **buy**, **sell** personal data? *for physical objects: ownership* ≈ *possession*

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Research #1: An Active-Learning Approach

Low-Cost Learning via Active Data Procurement. Abernethy, Chen, Ho, **Waggoner**. EC 2015.

Problem: How to model and achieve procurement of personal data?



Related approaches: Purchase data to estimate population statistics, especially

- Roth, Schoenebeck. EC 2012.
- Chen, Immorlica, Lucier, Syrgkanis, Ziani. EC 2018. extends RS12
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Drawbacks: specialized to statistics; not data efficient

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Key idea: base prices on value of data to the learning algorithm

Results: regret bounds $T\sqrt{\frac{\gamma}{B}}$ (online setting) and generalization bounds $\sqrt{\frac{\gamma}{B}}$ (i.i.d. data)

Research #2: A Markets-Based Approach

A Market Framework for Eliciting Private Data. Waggoner, Frongillo, Abernethy. NeurIPS 2015.

Goal: use a "market" to procure data privately and with good incentives!

See also:

- A Collaborative Mechanism for Crowdsourcing Prediction Problems (NeurIPS 2011). Abernethy, Frongillo.
- The Possibilities and Limitations of Private Prediction Markets (EC 2016). Cummings, Pennock, Wortman Vaughan.
- An Axiomatic Study of Scoring Rule Markets (ITCS 2018). Frongillo, Waggoner.
- Bounded-Loss Private Prediction Markets (NeurIPS 2018). Frongillo, Waggoner.

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Key ideas:

- Aligned incentives, bounded budget
- Opt-in for users
- Pay only for useful data
- Can add differential privacy

Part 3: Discussion

(Apologies in advance)

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Technical approaches to control; freedom-respecting software principle: software should respect and empower its users ... not exploit them! in particular: control over what information is revealed about you

Summary

- "Ownership:" data rights
- Value for data: "willingness to sell", loss function proxy
- Research: active-learning style, prediction-markets style
- Why markets? economic role of data
- Technical solutions? libre software



Thank you!