

MIT Open Access Articles

Optimal and Differentially Private Data Acquisition: Central and Local Mechanisms

The MIT Faculty has made this article openly available. *Please share* how this access benefits you. Your story matters.

Citation: Fallah, Alireza, Makhdoumi, Ali, Malekian, Azarakhsh and Ozdaglar, Asuman. 2022. "Optimal and Differentially Private Data Acquisition: Central and Local Mechanisms."

As Published: https://doi.org/10.1145/3490486.3538329

Publisher: ACM|Proceedings of the 23rd ACM Conference on Economics and Computation

Persistent URL: https://hdl.handle.net/1721.1/146417

Version: Final published version: final published article, as it appeared in a journal, conference proceedings, or other formally published context

Terms of Use: Article is made available in accordance with the publisher's policy and may be subject to US copyright law. Please refer to the publisher's site for terms of use.



Optimal and Differentially Private Data Acquisition: Central and Local Mechanisms

ALIREZA FALLAH, Massachusetts Institute of Technology, United States ALI MAKHDOUMI, Duke University, United States AZARAKHSH MALEKIAN, University of Toronto, Canada ASUMAN OZDAGLAR, Massachusetts Institute of Technology, United States

We consider a platform's problem of collecting data from privacy sensitive users to estimate an underlying parameter of interest. We formulate this question as a Bayesian-optimal mechanism design problem, in which an individual can share her (verifiable) data in exchange for a monetary reward or services, but at the same time has a (private) heterogeneous privacy cost which we quantify using differential privacy. We consider two popular differential privacy settings for providing privacy guarantees for the users: central and local. In both settings, we establish minimax lower bounds for the estimation error and derive (near) optimal estimators for given heterogeneous privacy loss levels for users. Building on this characterization, we pose the mechanism design problem as the optimal selection of an estimator and payments that will elicit truthful reporting of users' privacy sensitivities. Under a regularity condition on the distribution of privacy sensitivities we develop efficient algorithmic mechanisms to solve this problem in both privacy settings. Our mechanism in the central setting can be implemented in time $O(n \log n)$ where n is the number of users and our mechanism in the local setting admits a Polynomial Time Approximation Scheme (PTAS).

The full paper is available at: https://arxiv.org/abs/2201.03968

 $\label{eq:CCS} \mbox{Concepts:} \bullet \mbox{Theory of computation} \rightarrow \mbox{Algorithmic game theory and mechanism design;} \bullet \mbox{Security and privacy} \rightarrow \mbox{Privacy-preserving protocols}.$

Additional Key Words and Phrases: differential privacy, minimax lower bounds, data markets

ACM Reference Format:

Alireza Fallah, Ali Makhdoumi, Azarakhsh Malekian, and Asuman Ozdaglar. 2022. Optimal and Differentially Private Data Acquisition: Central and Local Mechanisms. In *Proceedings of the 23rd ACM Conference on Economics and Computation (EC '22), July 11–15, 2022, Boulder, CO, USA*. ACM, New York, NY, USA, 1 page. https://doi.org/10.1145/3490486.3538329

Permission to make digital or hard copies of part or all of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage and that copies bear this notice and the full citation on the first page. Copyrights for third-party components of this work must be honored. For all other uses, contact the owner/author(s).

EC '22, July 11-15, 2022, Boulder, CO, USA

© 2022 Copyright held by the owner/author(s).

ACM ISBN 978-1-4503-9150-4/22/07.

https://doi.org/10.1145/3490486.3538329