



Cornell University  
ILR School

Cornell University ILR School  
**DigitalCommons@ILR**

---

Labor Dynamics Institute

Centers, Institutes, Programs

---

8-2018

## The U.S. Census Bureau Adopts Differential Privacy

John M. Abowd

*U.S. Census Bureau*, [john.maron.abowd@census.gov](mailto:john.maron.abowd@census.gov)

Follow this and additional works at: <https://digitalcommons.ilr.cornell.edu/ldi>



Part of the [Labor Economics Commons](#), [Social Statistics Commons](#), and the [Statistical Methodology Commons](#)

**Thank you for downloading an article from DigitalCommons@ILR.**

**Support this valuable resource today!**

---

This Article is brought to you for free and open access by the Centers, Institutes, Programs at DigitalCommons@ILR. It has been accepted for inclusion in Labor Dynamics Institute by an authorized administrator of DigitalCommons@ILR. For more information, please contact [catherwood-dig@cornell.edu](mailto:catherwood-dig@cornell.edu).

If you have a disability and are having trouble accessing information on this website or need materials in an alternate format, contact [web-accessibility@cornell.edu](mailto:web-accessibility@cornell.edu) for assistance.

---

# The U.S. Census Bureau Adopts Differential Privacy

## Abstract

The U.S. Census Bureau announced, via its Scientific Advisory Committee, that it would protect the publications of the 2018 End-to-End Census Test (E2E) using differential privacy. The E2E test is a dress rehearsal for the 2020 Census, the constitutionally mandated enumeration of the population used to reapportion the House of Representatives and redraw every legislative district in the country. Systems that perform successfully in the E2E test are then used in the production of the 2020 Census.

Motivation: The Census Bureau conducted internal research that confirmed that the statistical disclosure limitation systems used for the 2000 and 2010 Censuses had serious vulnerabilities that were exposed by the Dinur and Nissim (2003) database reconstruction theorem. We designed a differentially private publication system that directly addressed these vulnerabilities while preserving the fitness for use of the core statistical products.

Problem statement: Designing and engineering production differential privacy systems requires two primary components: (1) inventing and constructing algorithms that deliver maximum accuracy for a given privacy-loss budget and (2) insuring that the privacy-loss budget can be directly controlled by the policy-makers who must choose an appropriate point on the accuracy-privacy-loss tradeoff. The first problem lies in the domain of computer science. The second lies in the domain of economics.

Approach: The algorithms under development for the 2020 Census focus on the data used to draw legislative districts and to enforce the 1965 Voting Rights Act (VRA). These algorithms efficiently distribute the noise injected by differential privacy. The Data Stewardship Executive Policy Committee selects the privacy-loss parameter after reviewing accuracy-privacy-loss graphs.

## Keywords

differential privacy, Economics of privacy

## Disciplines

Labor Economics | Social Statistics | Statistical Methodology

## Comments

Citation: Abowd, John M. "The U.S. Census Bureau Adopts Differential Privacy," KDD '18 Proceedings of the 24th ACM SIGKDD International Conference on Knowledge Discovery & Data Mining, London, UK (August 2018): 2867, DOI: [10.1145/3219819.3226070](https://doi.org/10.1145/3219819.3226070).

The presentation files are available at <https://hdl.handle.net/1813/60392>.

# The U.S. Census Bureau Adopts Differential Privacy

John M. Abowd  
 United States Census Bureau  
 Washington, DC, USA  
[john.maron.abowd@census.gov](mailto:john.maron.abowd@census.gov)

## ABSTRACT

The U.S. Census Bureau announced, via its [Scientific Advisory Committee](#), that it would protect the publications of the 2018 End-to-End Census Test (E2E) using differential privacy. The E2E test is a dress rehearsal for the 2020 Census, the constitutionally mandated enumeration of the population used to reapportion the House of Representatives and redraw every legislative district in the country. Systems that perform successfully in the E2E test are then used in the production of the 2020 Census.

**Motivation:** The Census Bureau conducted internal research that confirmed that the statistical disclosure limitation systems used for the 2000 and 2010 Censuses had serious vulnerabilities that were exposed by the Dinur and Nissim (2003) database reconstruction theorem. We designed a differentially private publication system that directly addressed these vulnerabilities while preserving the fitness for use of the core statistical products.

**Problem statement:** Designing and engineering production differential privacy systems requires two primary components: (1) inventing and constructing algorithms that deliver maximum accuracy for a given privacy-loss budget and (2) insuring that the privacy-loss budget can be directly controlled by the policy-makers who must choose an appropriate point on the accuracy-privacy-loss tradeoff. The first problem lies in the domain of computer science. The second lies in the domain of economics.

**Approach:** The algorithms under development for the 2020 Census focus on the data used to draw legislative districts and to enforce the 1965 Voting Rights Act (VRA). These algorithms efficiently distribute the noise injected by differential privacy. The Data Stewardship Executive Policy Committee selects the privacy-loss parameter after reviewing accuracy-privacy-loss graphs.

## CCS Concepts/ACM Classifiers

Privacy-preserving protocols; Privacy protections

## Author Keywords

Differential privacy; Economics of privacy

## BIOGRAPHY

John M. Abowd is Chief Scientist and Associate Director for Research and Methodology at the U.S. Census Bureau and the Edmund Ezra Day Professor of Economics, Professor of Statistics and Information Science at Cornell University. He is also Research Associate at the National Bureau of Economic Research, Research Affiliate at the Centre de Recherche en Economie et

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.

KDD 2018, August 19–23, 2018, London, United Kingdom.

© 2018 Copyright is held by the owner/author(s).

ACM ISBN 978-1-4503-5552-0/18/08.

DOI: <https://doi.org/10.1145/3219819.3226070>

Statistique (CREST, Paris, France), Research Fellow at the Institute for Labor Economics (IZA, Bonn, Germany), and Research Fellow at IAB (Institut für Arbeitsmarkt-und Berufsforschung, Nürnberg, Germany). He is the past President (2014-2015) and Fellow of the Society of Labor Economists; Fellow of the American Statistical Association; elected member of the International Statistical Institute; and a fellow of the Econometric Society. He served as Distinguished Senior Research Fellow at the United States Census Bureau from 1998 to 2016, and on the National Academies' Committee on National Statistics (CNSTAT 2010-2016). He currently serves on the American Economic Association's Committee on Economic Statistics (2013-2018). He was the scientific lead on the team that implemented the first formally private production disclosure limitation system worldwide: [OnTheMap](#) (see Machanavajjhala et al. 2008). <https://johnabowd.com>



## REFERENCES

1. Dinur, I. and K. Nissim (2003) "Revealing Information while Preserving Privacy" PODS 2003 3 ACM 1-58113-670-6/03/06.
2. Machanavajjhala, A., D. Kifer, J. Abowd, J. Gehrke, and L. Vilhuber (2008) "Privacy: Theory Meets Practice on the Map," ICDE, doi:10.1109/ICDE.2008.4497436