## Opening Access to Patient Data – Finding a Balance

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# Privacy and Patient Data



A fundamental expectation



Harms are possible if privacy is not protected



Has tended to be a predominant consideration

# Access to Patient Data for Research



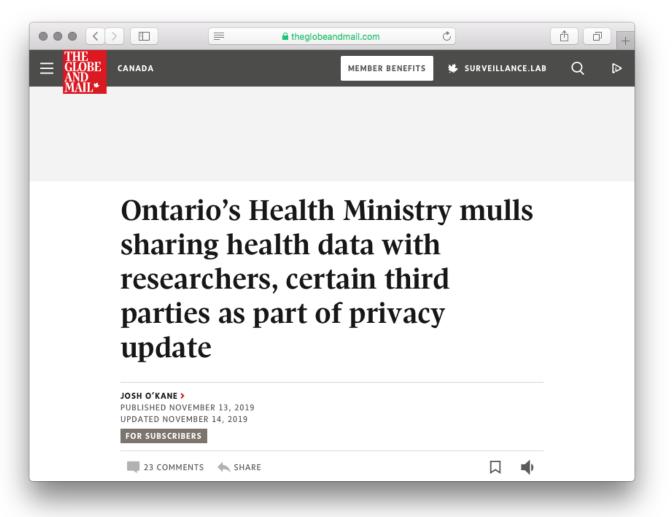
Supports innovation, data are the 'new oil' for a learning health system



Social and economic implications are large



Has tended to be a secondary consideration



## Increasing Demands for Access



Big data are needed for precision medicine



Al requires big computing next to data



Biased or missing data skews analyses and reinforces inequalities

Two **Strategies** for Opening Access and Protecting Privacy



Anonymization



Access control



Removes personally identifiable information (e.g., names, dates, locations, unique variables)

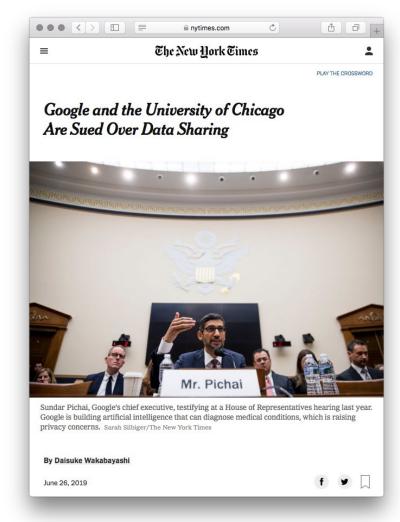
### Anonymization



Trades off information loss against risk of reidentification



There are limits in theory and practice





### **ARTICLE**

https://doi.org/10.1038/s41467-019-10933-3

DPEN

Estimating the success of re-identifications in incomplete datasets using generative models

Luc Rocher 1,2,3, Julien M. Hendrickx & Yves-Alexandre de Montjoye<sup>2,3</sup>

While rich medical, behavioral, and socio-demographic data are key to modern data-driven research, their collection and use raise legitimate privacy concerns. Anonymizing datasets through de-identification and sampling before sharing them has been the main tool used to address those concerns. We here propose a generative copula-based method that can accurately estimate the likelihood of a specific person to be correctly re-identified, even in a heavily incomplete dataset. On 210 populations, our method obtains AUC scores for predicting individual uniqueness ranging from 0.84 to 0.97, with low false-discovery rate. Using our model, we find that 99.98% of Americans would be correctly re-identified in any dataset using 15 demographic attributes. Our results suggest that even heavily sampled anonymized datasets are unlikely to the modern standards for anonymization set forth by GDPR and seriously challeng forget model.

"... we find that 99.98% of Americans would be correctly re-identified in any dataset using 15 demographic attributes."



Requires researchers to access data in secure location

## Access Control



Data cannot be removed, only approved results



Limits linkage, sophisticated analysis

## Example of MUHC Data Warehouse





Requests for access directed by default towards anonymized data



Encrypted data made available in secure cloud environment



Investigators sign data use agreement and additional approvals for external collaborations

## Closing Reflections





Placing more weight on anonymization reinforces focus on certain types of data and health contexts

Value of data for health system improvement is blurring the line between research and business of healthcare



### Google Is Slurping Up Health Data—and It Looks Totally Legal

Tech giants can access all of your personal medical details under existing health privacy laws. The question is how else that data might get used.



PHOTOGRAPH: ALBERTO PEZZALI/GETTY IMAGES

Last week, when Google gobbled up Fitbit in a \$2.1 billion acquisition,

the talk was mostly about what the company would do with all that