Digital Trace Data for Social Research

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The Digital Revolution

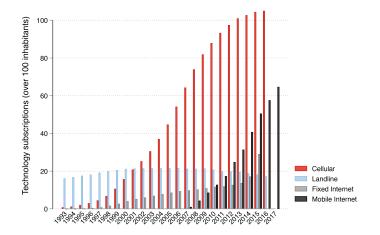


Figure: Technology subscriptions per capita, 1993–2017. Data from International Telecommunications Union (ITU).

The Digital Revolution

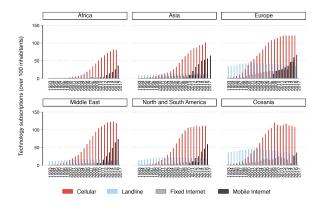


Figure: Technology subscriptions per capita, by global regions, 1993–2017. Data from International Telecommunications Union (ITU).

The Digital Revolution

Social transformation:

Digital technologies permeate social, economic, political life.

Data revolution:

The digitalisation of our lives creates data by-products: digital trace data

A Changing Data Ecosystem

- Digital trace data emerge from two processes –
- Social life is digitally mediated and the adoption of digital technologies and platforms (e.g. social media) generates data
 - Companies that provide these services want to capture these data streams because they are intrinsic to business models, e.g. targeted advertising

A Changing Data Ecosystem

- Digital trace data emerge from two processes –
- Social life is digitally mediated and the adoption of digital technologies and platforms (e.g. social media) generates data
 - Companies that provide these services want to capture these data streams because they are intrinsic to business models, e.g. targeted advertising
- Digitalisation of data has resulted in the storage of diverse types of information – including about non-digital or offline life
 - Information linked to everyday activities, e.g. sensors, consumer transactions, video recordings of cities are digitally stored

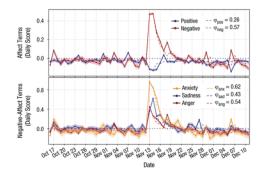
Common Features of Digital Trace Data

- Data that require repurposing because they were not intentionally collected for research
- Non-reactive: provide opportunities to observe without asking and for more dynamic measurement
- Measures of behaviour or activity (contrast with self-reported)
 - Ethics
 - Accessibility
- Digital trace data come in many forms

Digital Trace Data: Examples

Social media sites

Twitter



¹Garcia, David, and Bernard Rime. "Collective emotions and social resilience in the digital traces after a terrorist attack." *Psychological science* (2019): 0956797619831964.

Facebook and Gender Gaps

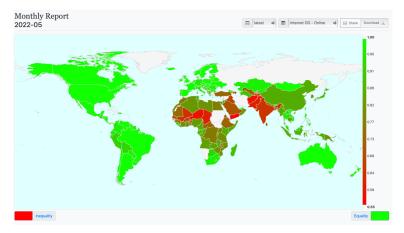


Figure: Gender gaps in internet use computed using data from Facebook (online model) available at www.digitalgendergaps.org

¹Fatehkia, Masoomali, Ridhi Kashyap, and Ingmar Weber. "Using Facebook ad data to track the global digital gender gap." *World Development* 107 (2018): 189-209.

Digital Trace Data: Examples

- Social media sites
- Web search queries

Google Flu

Figure 3 : ILI percentages estimated by our model (black) and provided by the CDC (red) in the mid-Atlantic region, showing data available at four points in the 2007-2008 influenza season.

Data available as of 4 February 2008

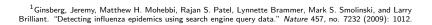
Data available as of 12 May 2008

Week

15 19

From: Detecting influenza epidemics using search engine query data

2.5



Google Search and IPV

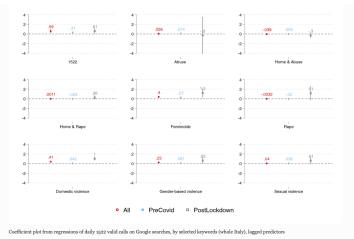


Figure: Relationship between abortion search volume and abortion rates across 37 countries. Marker size indicates the number of restrictions (from 0 to 7) that exist on abortion in each country.

¹Köksal, Selin, Luca Maria Pesando, Valentina Rotondi, and Ebru Şanlıtürk. "Harnessing the Potential of Google Searches for Understanding Dynamics of Intimate Partner Violence Before and After the COVID-19 Outbreak." European Journal of Population (2022): 1-29.

Digital Trace Data: Examples

- Social media sites
- Web search queries
- Blogs and internet forums

Reddit

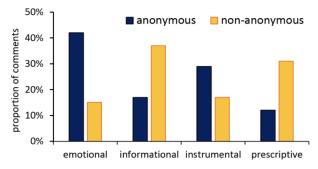


Figure 4. Posts from anonymous and non-anonymous accounts in the light of social support types.

¹De Choudhury, Munmun, and Sushovan De. "Mental health discourse on reddit: Self-disclosure, social support, and anonymity." In Eighth International AAAI Conference on Weblogs and Social Media. 2014.

Digital Trace Data: Examples

- Social media sites
- Web search queries
- Blogs and internet forums
- Call detail records from mobile phones

Mobile Phones

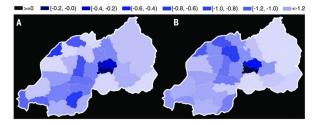


Figure: (A) Predicted composite wealth index (district average), computed from 2009 call data and aggregated by administrative district. (B) Actual composite wealth index (district average), as computed from a 2010 government DHS of 12,792 households.

¹Blumenstock, Joshua, Gabriel Cadamuro, and Robert On. "Predicting poverty and wealth from mobile phone metadata." *Science* 350, no. 6264 (2015): 1073-1076.

Digital Trace Data: Examples

- Social media sites
- Web search queries
- Blogs and internet forums
- Call detail records from mobile phones
- Sensor data

Electricity Smart Meters



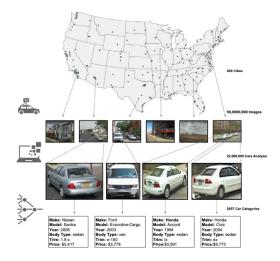


Open in figure viewer PowerPoint

Household load profiles by household composition.

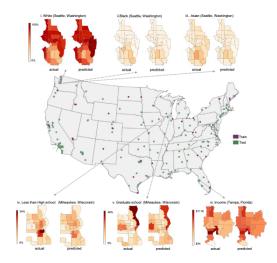
¹Newing, Andy, Ben Anderson, AbuBakr Bahaj, and Patrick James. "The role of digital trace data in supporting the collection of population statistics-The case for smart metered electricity consumption data." *Population, Space and Place* 22, no. 8 (2016): 849-863.

Google Street View



¹Gebru, Timnit, Jonathan Krause, Yilun Wang, Duyun Chen, Jia Deng, Erez Lieberman Aiden, and Li Fei-Fei. "Using deep learning and Google Street View to estimate the demographic makeup of neighborhoods across the United States." *Proceedings of the National Academy of Sciences* 114, no. 50 (2017): 13108-13113.

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Lens for social measurement

- Nowcasting social phenomena
- Analyzing the impacts of events and shocks in real-time
- Measuring tricky, hard-to-capture human behaviours

Digitalized lives and the implications of digitalization

- Understanding digitalized lives and social dynamics in digital spaces
- Examining the implications of digital technologies and platforms for social outcomes

Nowcasting social phenomena

Better temporal and/or geographical resolution, especially in contexts with data gaps in traditional sources.

Example (1): Predicting Migration using Facebook

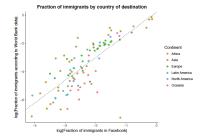
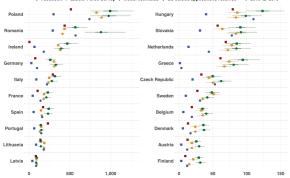


Figure 2: Relationship between stocks of migrants from the Facebook data set, for countries with a least one million Facebook ways are say 2006, and the respective estimates from the World Bank (2015). The data points indicate the fraction of minigrants in the panyl(2016), and ge scale, by country of destimation, color-coded by continent. The dashed line is the OLS regression line through the data.

¹Zagheni, Emilio, Ingmar Weber, and Krishna Gummadi. "Leveraging Facebook's Advertising Platform to Monitor Stocks of Migrants." *Population and Development Review* 43.4 (2017): 721-734.

Example (1): Predicting Migration using Facebook



◆ Facebook ◆ Labour Force Survey ◆ Model estimates ◆ EU settled applications received ● 2018 ■ 2019

¹Rampazzo, Francesco, Jakub Bijak, Agnese Vitali, Ingmar Weber, and Emilio Zagheni. "A framework for estimating migrant stocks using digital traces and survey data: an application in the United Kingdom." Demography 58, no. 6 (2021): 2193-2218.

Example (2): Gender Data Gaps



Figure: Gender gaps in internet use computed using data from International Telecommunications Union (ITU) available at www.digitalgendergaps.org

Example (2): Predicting Internet Gender Gaps

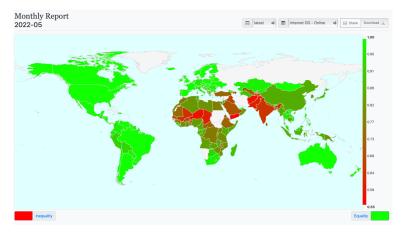


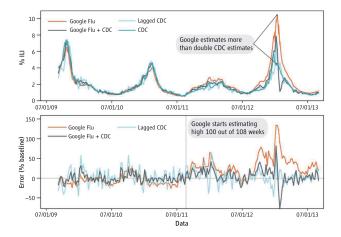
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 $^{^{1}\}mathsf{F}atehkia,$ Kashyap, Weber. 2018. "Using Facebook ad data to track the global digital gender gap" World Development

Nowcasting social phenomena

- Digital trace data are predictors within a model trained to predict 'ground truth' measure of interest
- Think carefully about how to operationalize measures from the digital trace source.
 - Algorithmic bias
 - Population drift

Cautionary Tale: Google Flu



¹Lazer, David, Ryan Kennedy, Gary King, and Alessandro Vespignani. "The parable of Google Flu: traps in big data analysis." Science 343, no. 6176 (2014): 1203-1205.

Nowcasting social phenomena

- Construct generated using digital trace source is a predictor within a model trained to predict 'ground truth'
- Think carefully about how to operationalize measures from the digital trace source.
 - Algorithmic bias
 - Drift: population drift, usage drift, system drift
- Digital trace often better as complements: combine across multiple sources of data, and consider using multiple platforms.

Lens for social measurement

- Nowcasting social phenomena
- Analyzing the impacts of events and shocks in real-time
- Measuring tricky, hard-to-capture human behaviours

Digitalized lives and the implications of digitalization

- Understanding digitalized lives and social dynamics in digital spaces
- Examining the implications of digital technologies and platforms for social outcomes

Example: Law and Public Sentiment

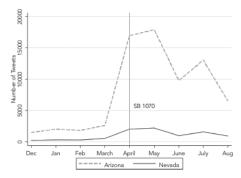


FIG. 2.—Number of Twitter messages related to immigrants per month in Arizona and Nevada (December 2010–August 2011). The vertical line on April 2010 indicates when the Arizona governor approved SB 1070.

¹Flores, René D. "Do anti-immigrant laws shape public sentiment? A study of Arizona's SB 1070 using Twitter data." *American Journal of Sociology* 123, no. 2 (2017): 333-384.

- Analyzing the impacts of events
- What is the appropriate counterfactual trend?

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Example: Online Dating

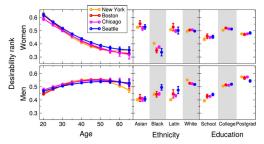


Fig. 2 Desirability, quantified using the measures defined here, as a function of demographic variables of the user population.

¹Bruch, Elizabeth E., and M. E. J. Newman. "Aspirational pursuit of mates in online dating markets." *Science Advances* 4, no. 8 (2018): eaap9815.

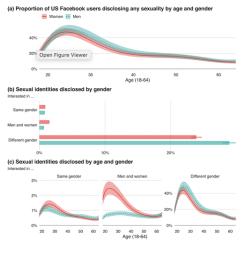
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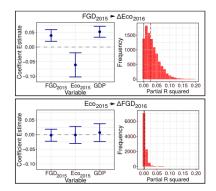
- Understanding digitalized lives and social dynamics in digital spaces
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Example: Sexuality Disclosure on Social Media



¹Gilroy, Connor, and Ridhi Kashyap. "Digital traces of sexualities: understanding the salience of sexual identity through disclosure on social media." *Socius* 7 (2021): 23780231211029499.

Example: Social Media Gender Gaps



¹Garcia, David, Yonas Mitike Kassa, Angel Cuevas, Manuel Cebrian, Esteban Moro, Iyad Rahwan, and Ruben Cuevas. "Analyzing gender inequality through large-scale Facebook advertising data." Proceedings of the National Academy of Sciences 115, no. 27 (2018): 6958-6963.

Considerations when Using Digital Trace Data

- What is your research question? How can digital trace data help you address it?
- Is it possible to use the data in an ethical manner?
- Are the data accessible?
- What is your theoretical or measurement framework?
- Who do your data include and who do they exclude?
- Can you compare or validate your measures against other sources of data?