

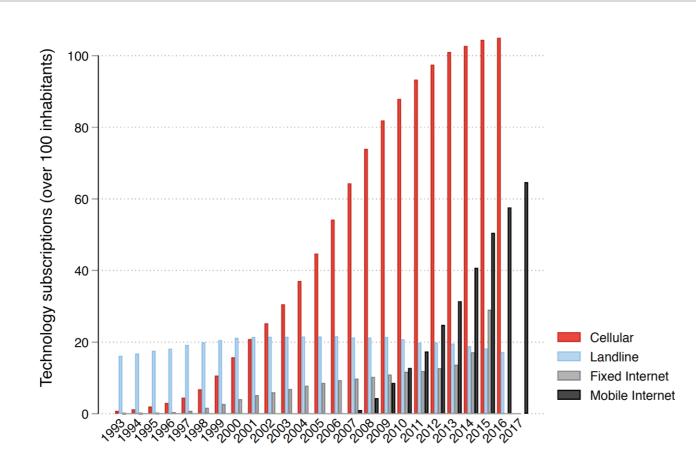
The Digital Revolution and Demography: Perspectives from Digital and Computational Demography

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The Digital Revolution

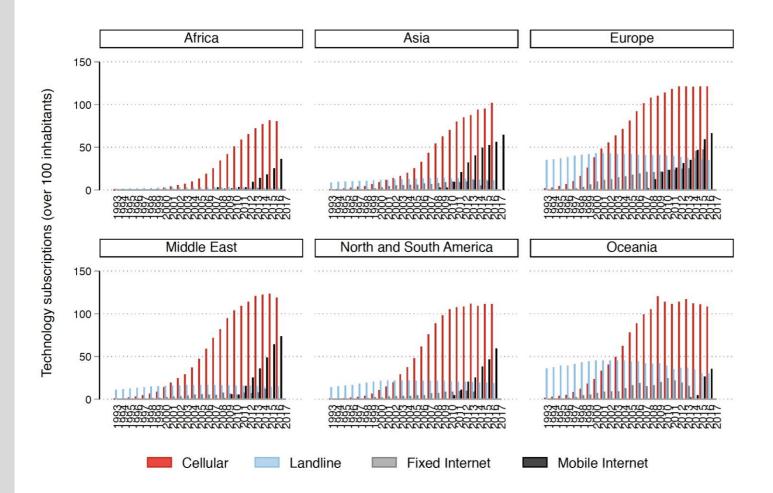


Technology subscriptions per capita, global, 1993-2017. Data from International Telecommunications Union

(ITU)



The Digital Revolution



Technology subscriptions per capita, by region, 1993-2017. Data from International Telecommunications Union (ITU)



• The digital revolution is:

Social revolution:

 \rightarrow How – and for whom – are digital technologies affecting demographic behaviours and outcomes?

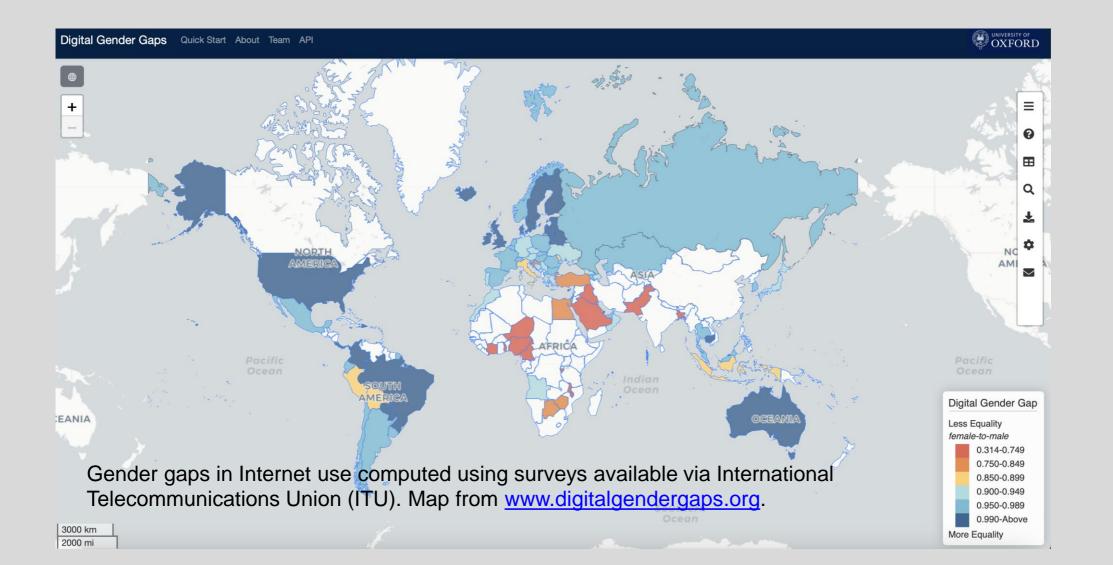
\rightarrow How do technologies (re)shape inequalities?

Technology has the *theoretical* potential to empower marginalized populations

- SDG 5 on gender equality pledges to "enhance the use of...information and communication technologies to promote the empowerment of women."
- But is this potential being realized?

The Data Gap







• The digital revolution is:

• Social revolution:

How – and for whom – are digital technologies affecting demographic behaviours and outcomes (e.g. health and well-being, gender inequalities, family and fertility, migration)? How do technologies (re)shape inequalities?

• Data revolution:

Our use of the web, social media, and mobile phones generates new data streams that have the potential to provide novel measures of digital adoption and behaviours



→ Advancing understanding of the data and social implications of the digital revolution requires the contributions of digital and computational demography.

- 1. Kashyap, Ridhi., 2021. "Has demography witnessed a data revolution? Promises and pitfalls of a changing data ecosystem" *Population Studies*, *75*(sup1), pp.47-75.
- 2. Kashyap, Ridhi, Gordon Rinderknecht et al. 2023. "Digital and Computational Demography", Handbook of Digital Sociology, ed. Jan Skopek, Edward Elgar Publishing.
- 3. Kashyap, Ridhi, and Emilio Zagheni. 2023. "Leveraging Digital and Computational Demography for Policy Insights." Handbook of Computational Social Science for Policy. Springer International Publishing, 327-344.



- The digital revolution is:
 - Social revolution:

How – and for whom – are digital technologies affecting demographic behaviours and outcomes?

→ Digital expansion and sustainable development, especially SDG 5 on gender equality (The Digital Gender Gaps project, <u>www.digitalgendergaps.org</u>)

• Data revolution:

What are the demographic characteristics of online populations? Can new data streams be used for population-generalizable measurement?



Technology impacts: prior pathways

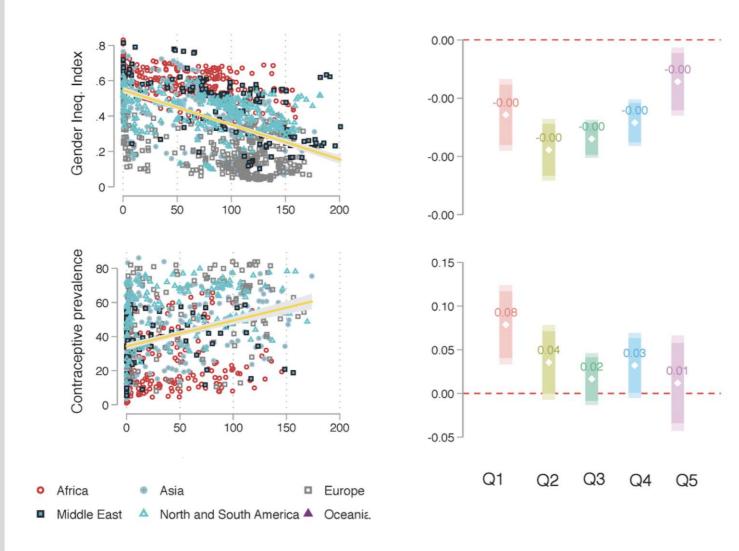
- TV: exposure to new information and behaviours shift in attitudes (e.g. Barber and Axinn 2004, Jensen and Oster 2009, La Ferrara et al 2012)
- Diffusion theories of fertility decline: technology can alter paths for social learning and social interaction

Mobile phones:

- Improved access to (private) information, as well as access to services (e.g. mHealth, mobile money)
- Better connectivity and access to networks
- Exposure to the 'life of others' including more globalized media content

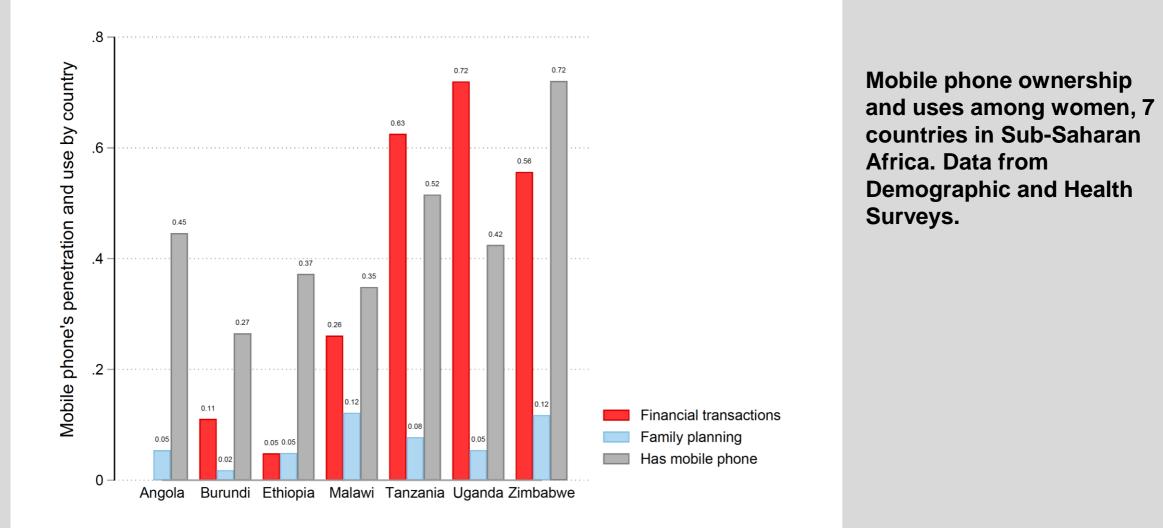
Leveraging mobile phones for sustainable development, *PNAS* (with Valentina Rotondi, Luca Pesando, Simone Spinelli, Francesco Billari)





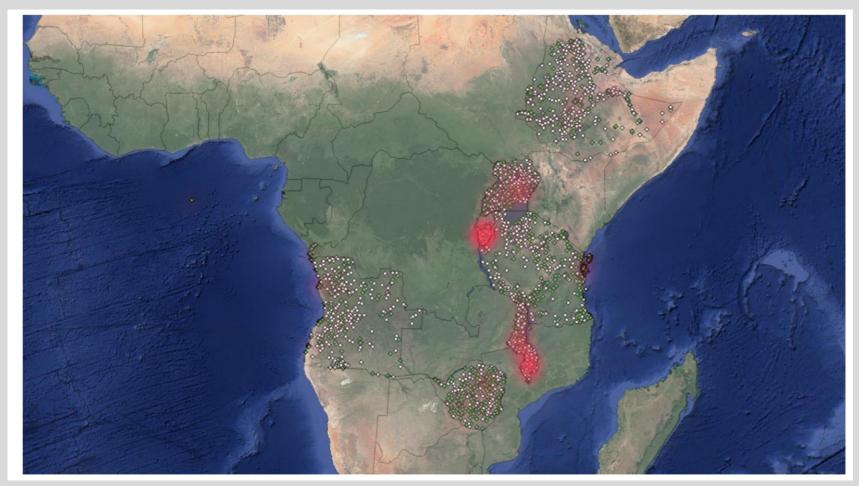
Global correlations between mobile phone diffusion and sustainable development outcomes (left-panel).

Standardized coefficients from models regressing mobile-phone diffusion on sustainable development outcomes by GDP-percapita quintiles (rightpanel). Leveraging mobile phones for sustainable development, *PNAS* (with Valentina Rotondi, Luca Pesando, Simone Spinelli, Francesco Billari)



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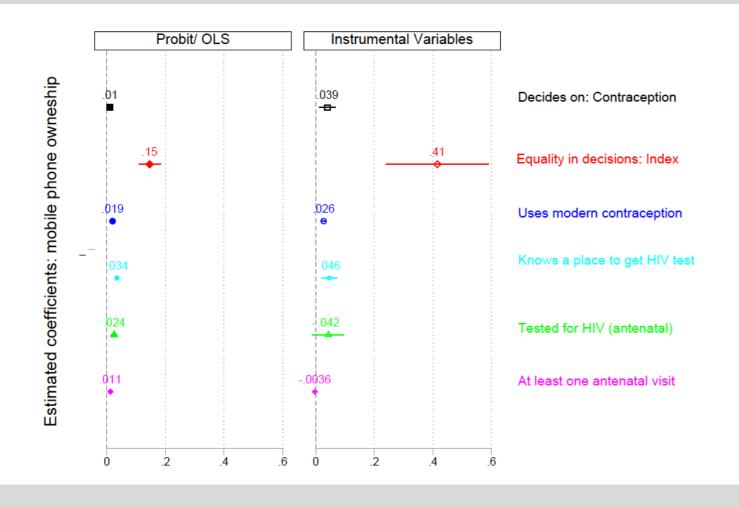




The average number of lightning strikes (red) and mobile ownership (green dots) versus non-ownership (white dots). Source: authors' elaboration from augmented DHS data (2015-16)

Lightning strikes as a quasiexperiment – poorer mobile connectivity in areas with more strikes due to poorer connectivity and broken down towers



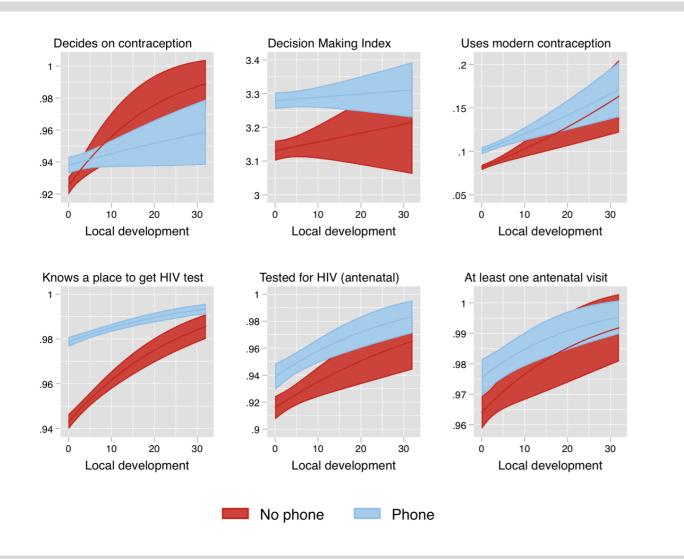


Individual-level effects of mobile phone ownership among women on sustainable development outcomes linked to women's empowerment and sexual and reproductive health.

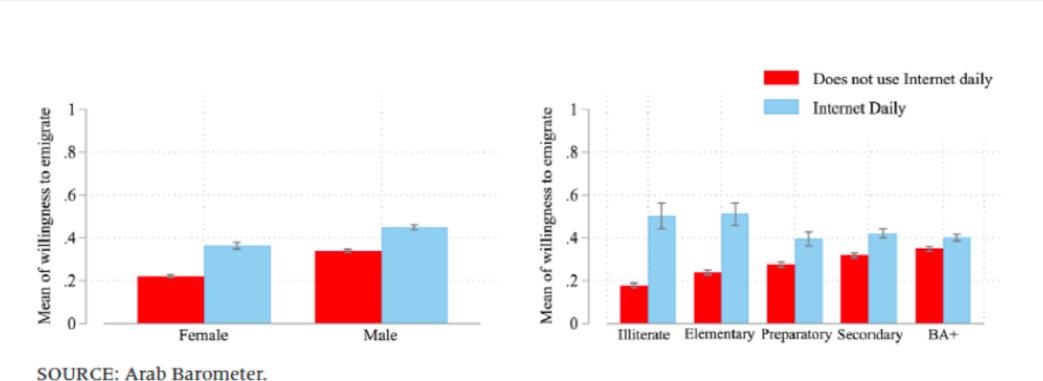
Covariates used in the models: education, age, household size, employment status, radio and tv ownership, urban, local development (nightlights). Country and year fixed effects. Standard errors clustered at the cluster level.

Leveraging mobile phones for sustainable development, *PNAS* (with Valentina Rotondi, Luca Pesando, Simone Spinelli, Francesco Billari)





Marginal effects of the interaction between mobilephone diffusion and local development (as proxied by nightlights). The *Internetization* of international migration, *Population and Development Review* (joint with Luca Pesando, Valentina Rotondi, Manuela Stranges, Francesco Billari)



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booket. mab barometer.

Willingness to migrate by internet use and gender (left panel) and education (right panel)

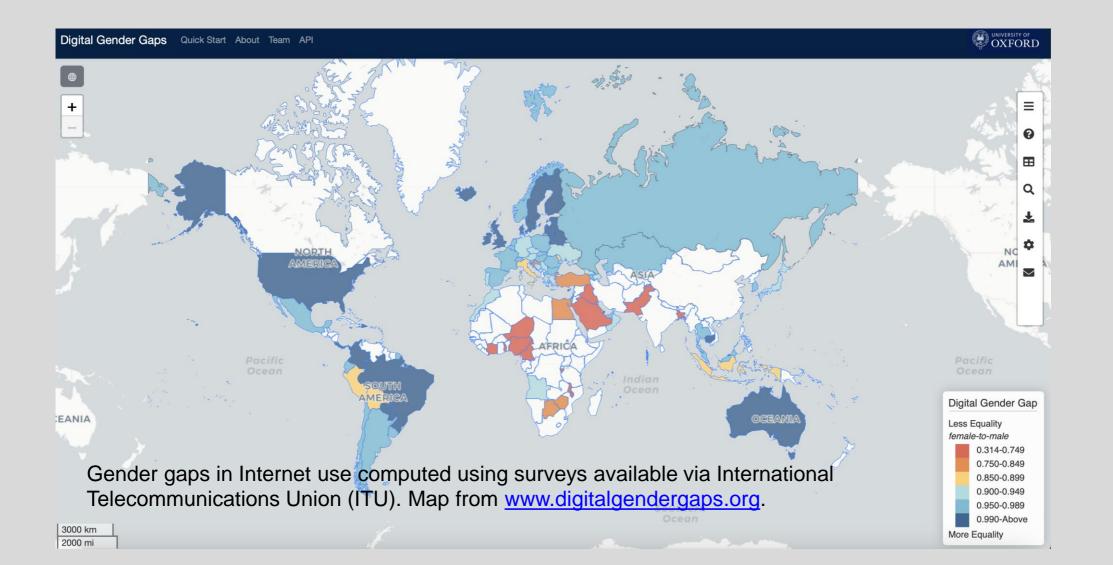


Summary (1)

- Mobile phone ownership and internet adoption linked to improved women's empowerment, reproductive and sexual rights, and health outcomes
- Bigger pay-offs at lower levels of economic development
- Implication: important to track digital connectivity by gender

The Data Gap





The Digital Revolution



- The digital revolution is:
 - Social revolution:

How – and for whom – are digital technologies affecting demographic behaviours and outcomes?

• Data revolution:

Use of the web, social media and mobile phones generates new data streams that have the potential to provide real-time information on digital connectivity and behaviours



Social Media Ad Audience Estimates

| New Campaign > ■ New Ad Set > □ 1 Ad ✓ Edit | • In draft |
|---|---|
| cations | () |
| People living in or recently in this location | Audience definition Your audience selection is fairly broad. |
| Nigeria | Specific Broad |
| Vigeria | Estimated audience size: 12,200,000 - 14,300,000 6 |
| Include Q Search locations Browse NEGAL Bamako BURKINA FASO N'Djamena | Estimates may vary significantly over time based on your targeting selections and available data. |
| GUINEA BENIN NIGEXIA LIBERIA GHANA Lagos CAMEROON Bangui | Estimated daily results Estimated daily results aren't available for this campaign since it has a budget optimized across ad sets. |
| GABON CONSULT OF DEN CC Drop pin NE | |
| e - 65+ | |
| nder men | |

How many
users of 'x'
characteristics
(gender,
location, age,
device type,
etc) are on a
given platform?

Facebook Ads Manager: https://www.facebook.com/adsmanager



Ad Audience Estimates

| | tched audience (option ting options to reach your web | | target accounts. | Select | | ur estimated target audience 00,000+ LinkedIn members |
|--|--|-----------------|------------------|---------------|-----------|---|
| Target by the audie | ence below | ÷ 0 | | | S Neth | erlands |
| What location do | you want to target? (requir | red) | | | A Male | |
| include • Start | typing a country, state, city, or | r town | | See full list | Audier | ce expansion: Enabled |
| | Netherlands × | | | | discrimit | tools may not be used to nate based on personal ristics like gender, age, or actual ved race/ethnicity. Learn more |
| 오 What gender do y All Femal O Male | | | | | × | |
| Select specific targeting | g criteria to zero in on your ide | eal audience: | | | | |
| Company name | Company industry | Company size | Job title | Job function | | |
| Job seniority | Member schools | Fields of study | Degrees | Member skills | | |

LinkedIn Ads Manager



Applications

- Predicting SDG 5 indicators using gender gap indicators computed using social media ad data
 - Internet and mobile access gender gaps (<u>www.digitalgendergaps.org</u>, using Facebook)
 - Professional gender gaps (using LinkedIn)
- Expanding geographical coverage, more timely data on social indicators and for monitoring rapidly evolving situations

Limitations

- No past data; data access constraints
- Black box algorithms



Facebook Gender Gap Index

 $FB \ GGI = \frac{Female \ to \ male \ ratio \ of \ Facebook \ users}{Female \ to \ male \ ratio \ of \ the \ population}$

 Using ITU data, we compute a ground truth internet gender gap indicator derived from survey data:

• Internet $GGI = \frac{\% \text{ of } female \text{ population using the Internet}}{\% \text{ of male population using the Internet}}$

• Similarly, we compute Mobile GGI



Facebook Gender Gap Index

FB GGI \approx Internet GGI

- Correlation of FB GGI 18+ with Internet GGI is 0.81
- Stronger than any other development indicator (including internet penetration)
- Correlation with Mobile GGI is 0.75
- We fit three types of predictive models:
 - online model
 - online-offline model
 - offline model

Using Facebook ad data to track the global digital gender gap. *World Development* (joint work with Masoomali Fatehkia, Ingmar Weber)

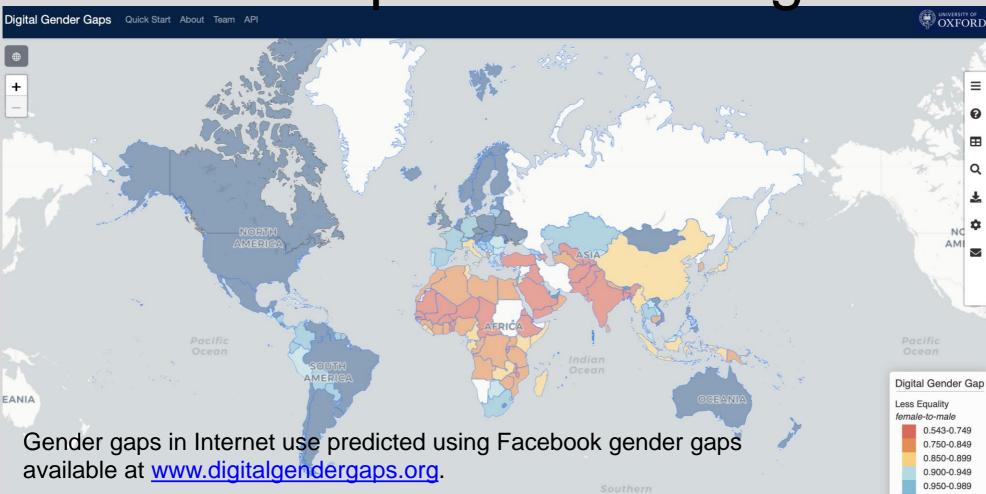
Internet GGI predicted using FB GGI

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0.990-Above

More Equality

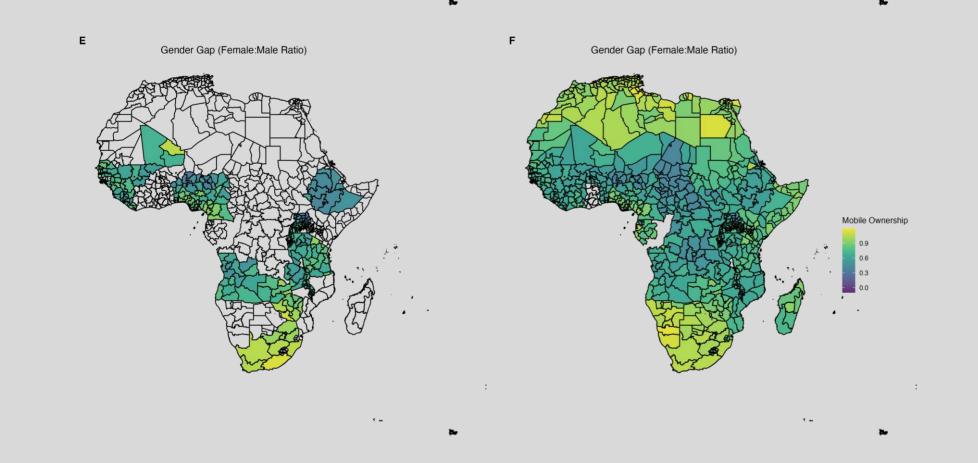
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3000 km 2000 mi

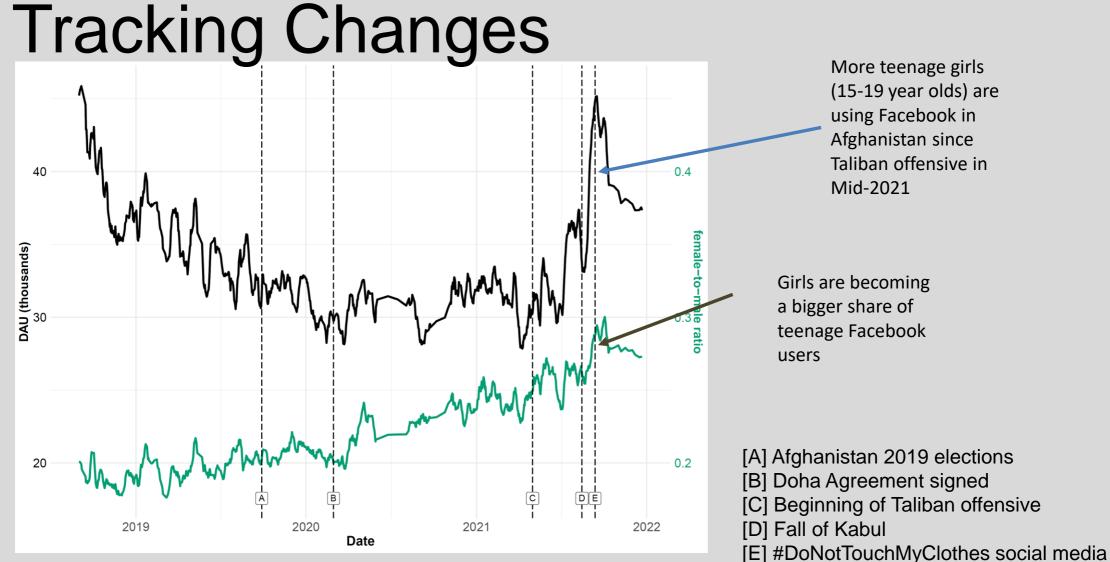


Next Output: Subnational Maps of Digital Gender Gaps (Africa)



Leveraging online advertising data for measuring the Sustainable Development Goals: applications for gender gaps and SDG5 (with Reham Al Tamime, Masoomali Fatehkia and Ingmar Weber)





campaign

Selection and Bias

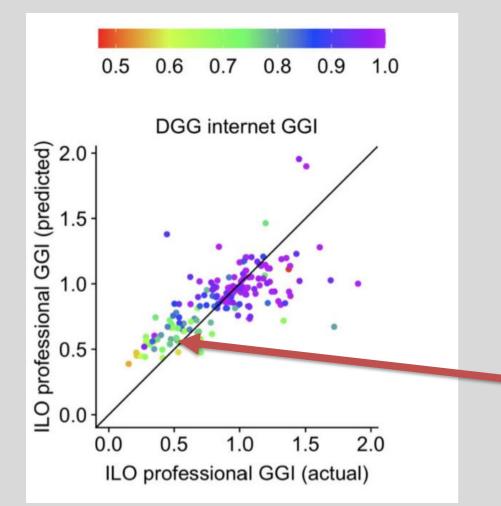


- Digital gender gaps are important to track themselves
- But they can also help illuminate biases in analyses using online samples/populations

Analysing global professional gender gaps using LinkedIn advertising data, *EPJ Data Science* (with Florianne Verkroost)



Selection and Bias



In countries where women are less likely to be online, LinkedIn Gender Gaps over-predict professional gender equality

➔ Women who are online are more selective



Summary

- The digital revolution has both **social** and **data** implications.
- Understanding differential **impacts of digitalization** is vital to understanding demographic and sustainable development dynamics.
- Harnessing the data revolution requires both 'old' and 'new' data sources, and their linkage, within a diversifying data ecosystem
- Demographic analyses of digital populations can help understand biases and generalizability but important for illuminating population inequalities in new ways.
- Digital and computational demography has a vital role to play in advancing an global vision of computational social science.



Thank you!

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- Kashyap, R., Fatehkia, M., Al Tamime, R., & Weber, I. (2020). Monitoring global digital gender inequality using the online populations of Facebook and Google. *Demographic Research*, *43*, 779-816.
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Thank you!

Learn more at: www.digitalgendergaps.org

Contact: <u>ridhi.kashyap@nuffield.ox.ac.uk</u>

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Social Media Ad Data Streams

- Social media ad data serve as **digital censuses** that provide aggregated demographic data on user populations
 - Expanding geographical coverage, filling data gaps
 - Finer temporal and spatial resolution
 - Valuable for rapidly evolving situations or for monitoring social indicators
 - Good complement to survey data, but not substitute

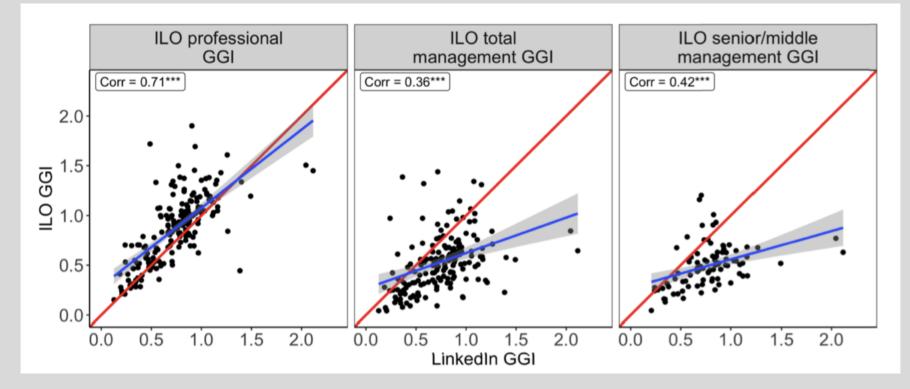
Limitations

- No past data
- Black box algorithms

Analysing global professional gender gaps using LinkedIn advertising data, *EPJ Data Science* (with Florianne Verkroost)



Professional Gender Gaps

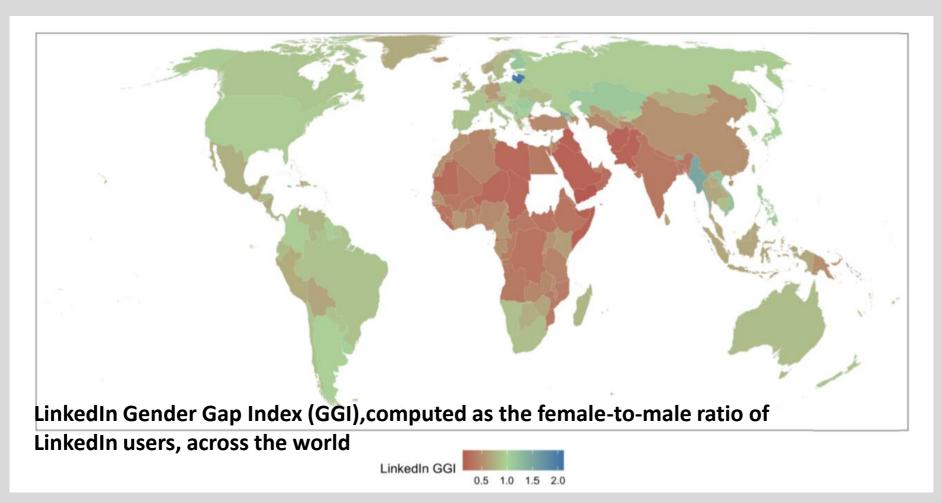


LinkedIn Gender Gap Index (GGI) and its correlations with different International Labour Organization (ILO) Gender Gaps

Analysing global professional gender gaps using LinkedIn advertising data, *EPJ Data Science* (with Florianne Verkroost)



LinkedIn Gender Gap Index



Predictive Fit



Table 3

Summary of results for three regression models predicting ITU internet Gender Gap Index using (i) a single online Facebook variable; (ii) online and offline variables; (iii) offline variables. Bootstrap estimates of the coefficient standard errors are reported in parentheses.

| | Online Model | OnlOffl. Model | Offline Model |
|------------------------------------|--------------|----------------|---------------|
| Intercept | 0.933*** | 0.932*** | 0.933*** |
| - | (0.006) | (0.005) | (0.007) |
| FB GG (age 18+) | 0.071*** | 0.093*** | |
| | (0.011) | (0.017) | |
| log(GDP per capita) | | 0.018* | |
| | | (0.008) | |
| GGGR – Literacy | | -0.018 | |
| | | (0.016) | |
| GGGR – Education | | -0.019 | |
| | | (0.019) | |
| Internet Penetration | | | 0.040*** |
| | | | (0.009) |
| GGGR – Tertiary Educ. | | | 0.032 |
| | | | (0.021) |
| GGGR – Economy | | | 0.043** |
| | | | (0.014) |
| GGGR Score | | | -0.024 |
| | | | (0.012) |
| Adjusted R-squared | 0.691 | 0.791 | 0.615 |
| Mean Abs. Error | 0.0325 | 0.0288 | 0.037 |
| SMAPE | 3.92% | 3.90% | 4.97% |
| F-statistics | 169 | 67.38 | 29.79 |
| df | 74 | 66 | 68 |
| Ν | 76 | 71 | 73 |
| # predicted countries ^a | 152 | 127 | 132 |
| ** p < 0.001, ** p < 0.01, * p | o < 0.05. | | |

First stage regression



| | (1) | (2) | (3) |
|------------------------|-------------|----------|----------------|
| | Flash rates | Coverage | Flash+Coverage |
| Gray Flash rates (log) | -0.013*** | | -0.013*** |
| | (0.002) | | (0.002) |
| Cell phone coverage | | 0.017*** | 0.014*** |
| | | (0.004) | (0.004) |
| F-stat | 19.96*** | 7.99*** | 18.15*** |
| Hansen J statistic | | | 0.484 |

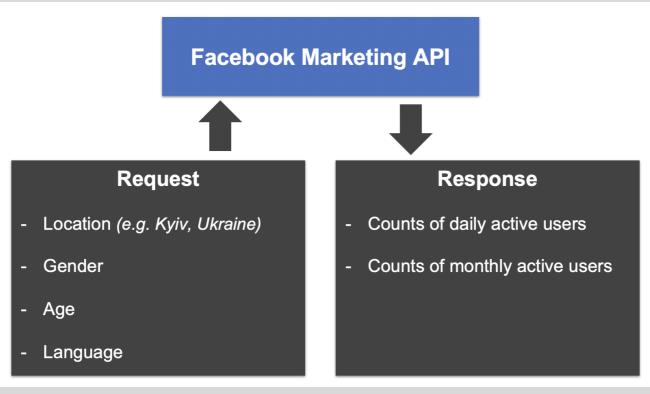
Note: First stage. Instruments: Flash rates (columns 1, 3), cellphone network (column 2,3). * p<0.10, ** p<0.05, *** p<0.01



Further applications

- Monitoring population displacement during Ukraine war through the same social media marketing database

- Help fill critical data gaps to inform humanitarian response



"Nowcasting daily population displacement in Ukraine through social media advertising data", *Population and Development Review* (joint work with Douglas Leasure, Francesco Rampazzo, et al)



From FB estimates to Population Counts

Population Estimation

30-34 year old women in Kyiv before the conflict

Baseline population = 100

Baseline Facebook users = 25

Baseline Facebook penetration = 25%



"Nowcasting daily population displacement in Ukraine through social media advertising data", *Population and Development Review* (joint work with Douglas Leasure, Francesco Rampazzo, et al)



From FB estimates to Population Counts

Population Estimation

How many 30-34 year old women are in Kyiv **today?**

Facebook users = 15

Baseline Facebook penetration = 25%

Current population = 15 / 0.25 = 60

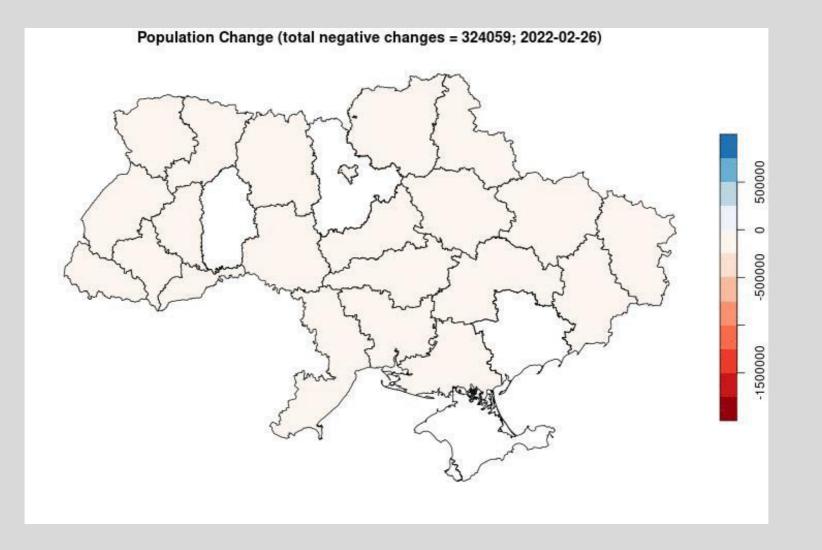


POPULATION = USERS / PENETRATION

"Nowcasting daily population displacement in Ukraine through social media advertising data", *Population and Development Review* (joint work with Douglas Leasure, Francesco Rampazzo, et al)



Daily net population change per Oblast



"Nowcasting daily population displacement in Ukraine through social media advertising data", *Population and Development Review* (with Douglas Leasure, Francesco Rampazzo, et al)



