# Mapping subnational gaps in internet and mobile adoption using social media data

Digital Demography

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July 5, 2024

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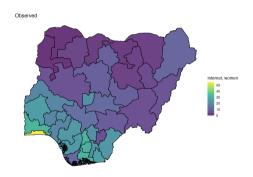
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- ▶ Yet large **inequality** in who has access to digital technology...

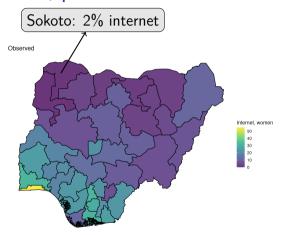


# Adoption of digital technology varies geographically



Source: Nigeria, Demographic and Health Survey

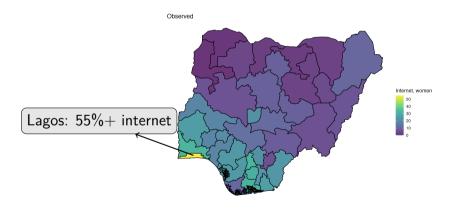
# Women using internet, past 12 months





Introduction

# Women using internet, past 12 months



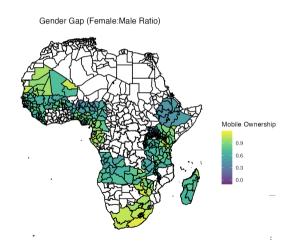


# Develop subnational estimates of adoption

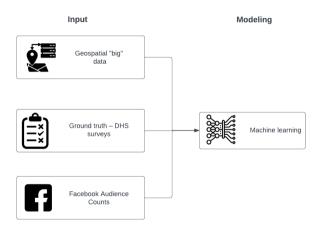
▶ Goal: Develop estimates of internet and mobile adoption by gender and digital gender gaps

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- First subnational level
  - ▶ 52 countries, 874 subnational units



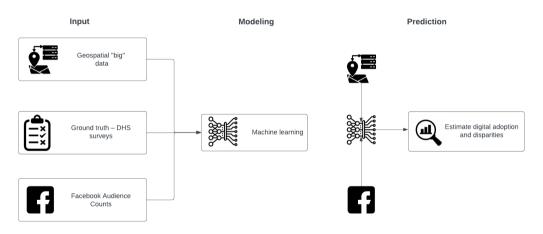
# Overview of approach



24 countries, with ground truth

Introduction

# Overview of approach



19 countries, with ground truth

Data + methods

52 countries, with and without ground truth

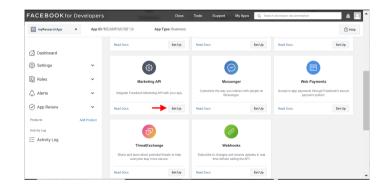
References

# Ground truth – Demographic and Health Surveys (DHS)

- Household surveys representative at the first subnational level
  - Standardized sample design, questionnaire, implementation, etc.
  - Questions on individual-level internet use and mobile phone use (wave 7 onwards)
- ► Focus on 24 different DHS surveys, 2016-2020

#### Facebook audience counts

- Collected through public marketing API
- Specify geographic region (FB template or custom region)
- Disaggregated counts by gender, age, device type, etc.



# Big geospatial and population data

- Include 'offline' predictors that are uniformly available and consistent across subnational units
  - Satellite-derived nighlights data
  - Population density
  - Relative wealth index (Meta)
  - Subnational education index, income index, human development index (HDI), gender development index (GDI)

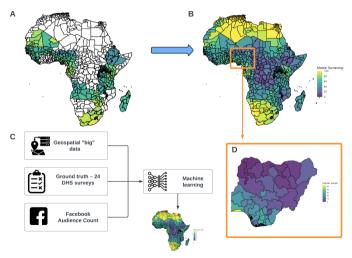
# Outcomes of interest (from DHS)

Indicators	Women	Men	Gender Gap
Mobile Phone Ownership	<b>√</b>	<b>√</b>	<b>√</b>
Internet Use, Past 12 Mo	$\checkmark$	$\checkmark$	$\checkmark$

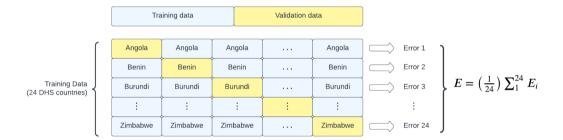
# Modeling approach - machine learning

Algorithm	Description
glmnet (Lasso)	Lasso Regression
glmnet (Ridge)	Ridge Regression
glmnet (Elastic Net)	Elastic Net with 50% L1 Ratio
polspline	Polynomial Spline
ranger	Random Forest with 100 Trees
gbm	Gradient Boosted Machine
glm	Generalized Linear Model
xgboost	Extreme Gradient Boosting
SuperLearner	Ensemble method combining multiple learning algorithms

# Greatly expanded coverage



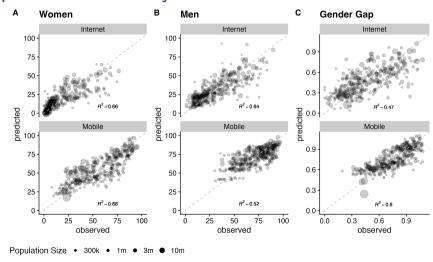
#### Leave-one-country-out cross validation



Introduction O Background 0000 Data + methods

Results 00000 Conclusion

### Overall predictive accuracy



Introduction

Background 0000 Data + methods

Results

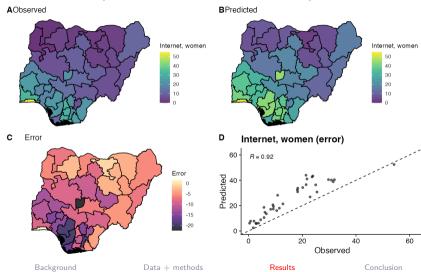
Conclusion

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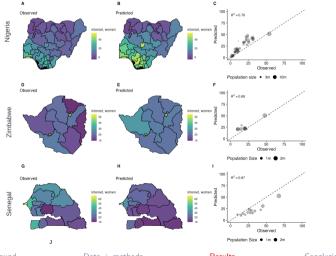
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# Results for Nigeria (Leave-one-country-out)

Introduction



# Error by country

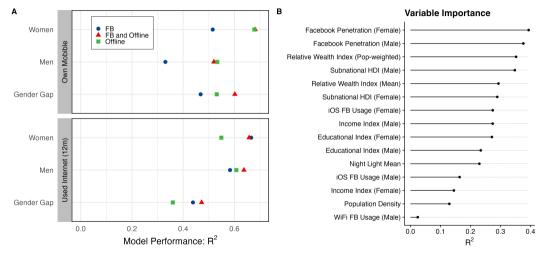


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Conclusion

### Most important predictors



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Data + methods

Results

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References

# Next steps and future opportunities

- Expanding to all 135 LMIC countries (ongoing...)
- Quantifying uncertainty

Results



# Summary

► Using Facebook audience counts **greatly expands** our ability to accurately predict internet adoption in countries with no ground truth

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- Huge disparities in access to mobile and internet technologies between and within countries

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- Using Facebook audience counts greatly expands our ability to accurately predict internet adoption in countries with no ground truth
- Huge disparities in access to mobile and internet technologies between and within countries
- New opportunities to study population-level impacts of digital technology using these subnational estimates

#### Thank You

#### **Funders:**

- Bill and Melinda Gates Foundation (INV-045370)
- ► Leverhulme Trust (Grant RC-2018-003) for the Leverhulme Centre for Demographic Science

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digitalgendergaps.org

Results



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