# Tapping Into Advertising Platforms to Monitor ICT Usage and More

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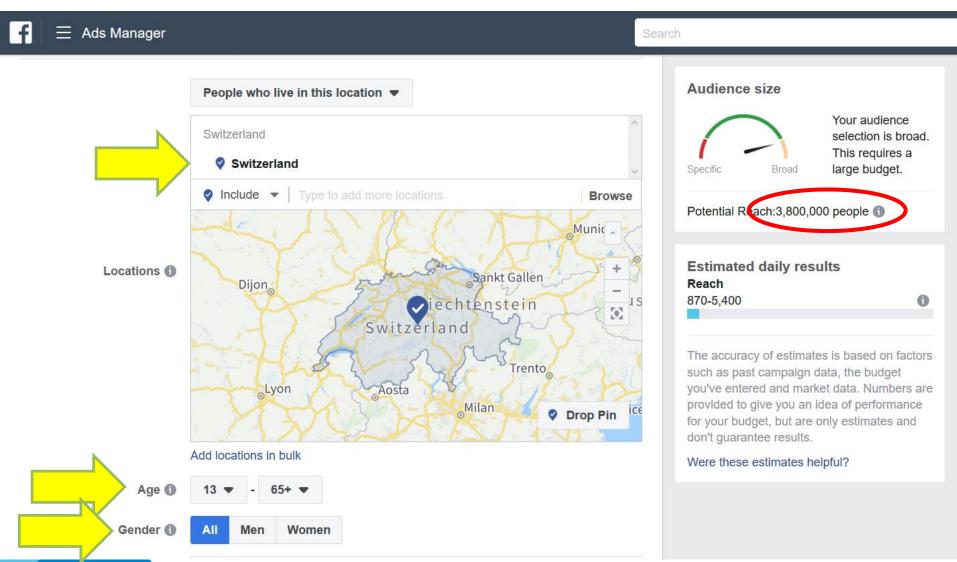


#### tl;dr

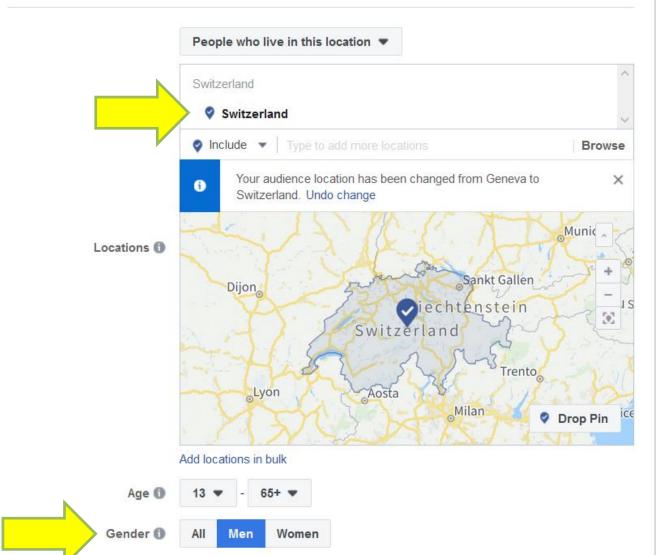
- Online advertising platforms provide information on how many of their users match certain targeting criteria
- How many female Facebook users in Geneva City used to live in France and use an iPhone 8 (Plus) on 4G to access FB? Answer: 1,200.
- This type of information can be used to monitor online gender gaps, income inequalities, migration and more
- Supported by FB, LI, IG, Google, VK, ...

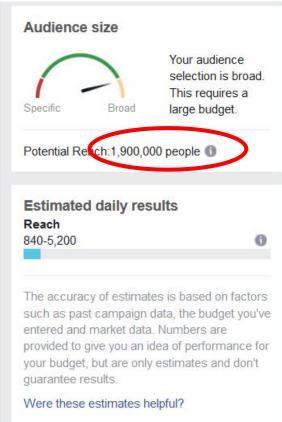


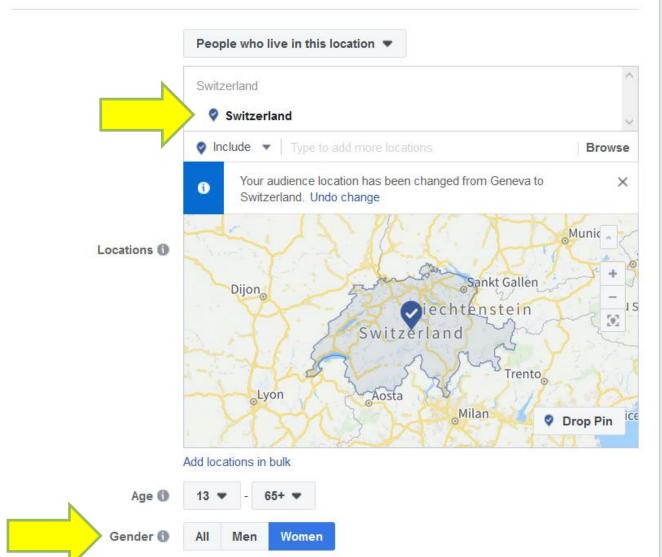
# Facebook's Advertising Platform

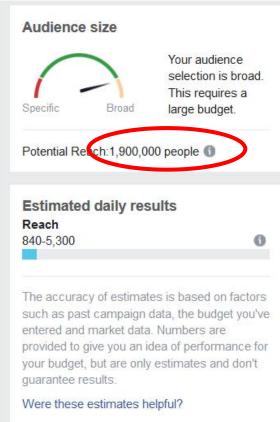


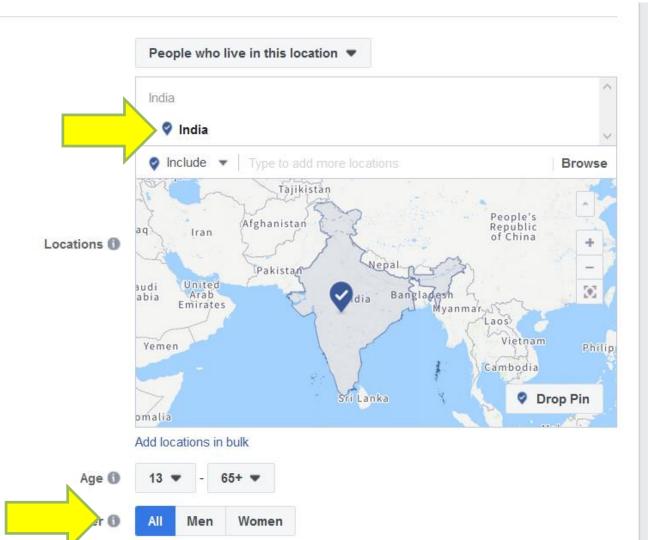
https://www.facebook.com/adsmanager/creation/

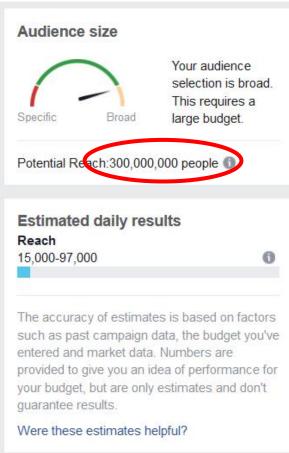


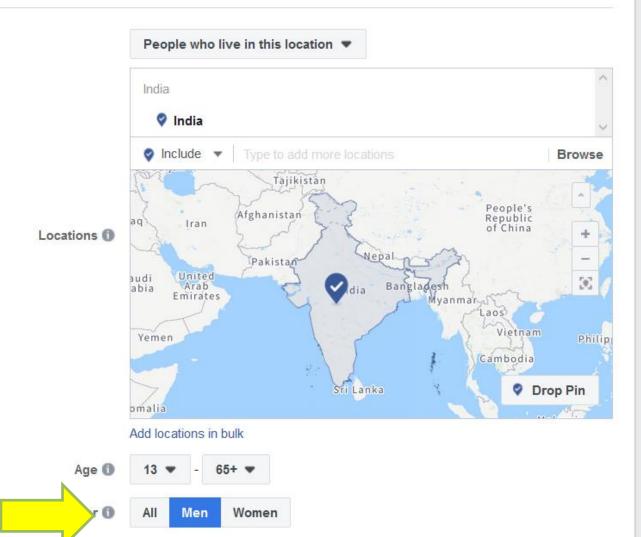


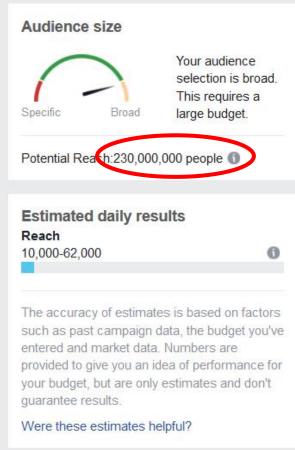


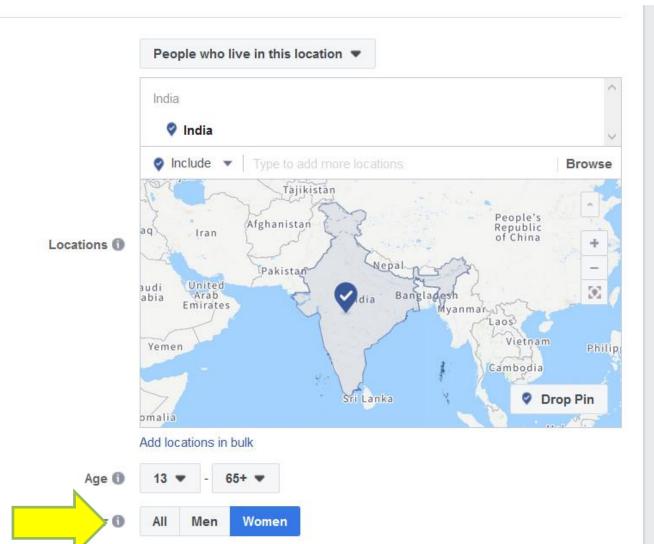


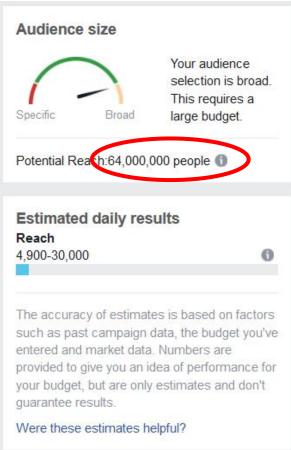






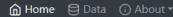






#### Digital Gender Gaps

Measuring digital gender inequalities in real-time



#### Using big data to measure global gender gaps in internet and mobile access

Tracking progress on gender inequalities in internet and mobile access and use is more important than ever to ensure that women benefit from the digital revolution. Data on gender gaps in internet and mobile phone use and access are significantly lacking geographical coverage, comparability, and are slow to be updated.

We show how big data can help close this gender data gap and measure progress towards this important development goal in real-time.

#### Latest indicators

Check out the latest internet and mobile gender gap indicators.

Reports »

#### Project details

See the background of the project, and an overview of how the data is collected and processed.

#### Team

Meet the project members.

Team »

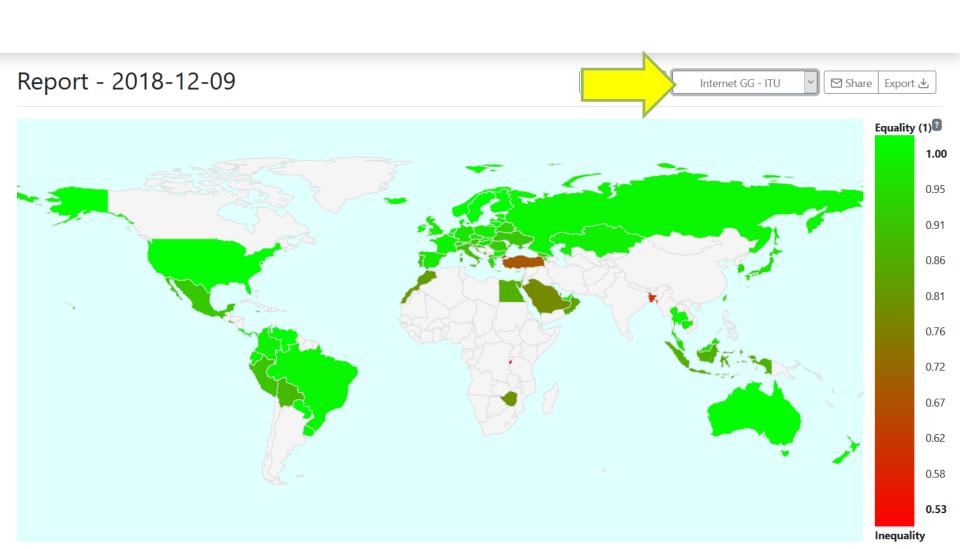


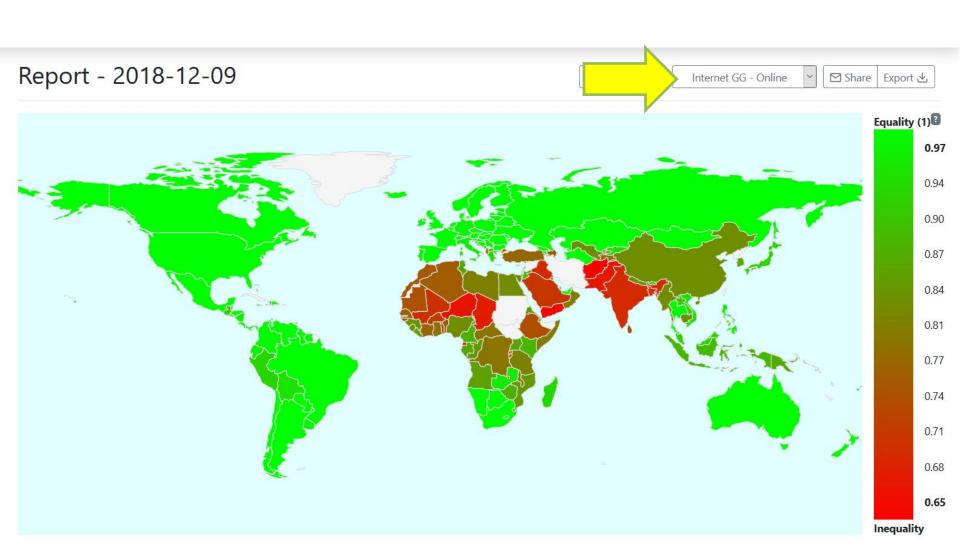


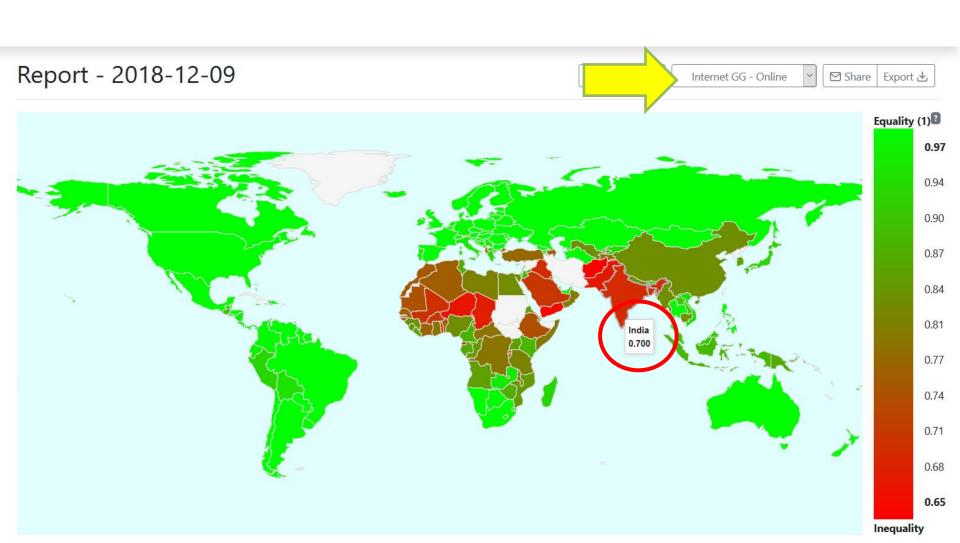


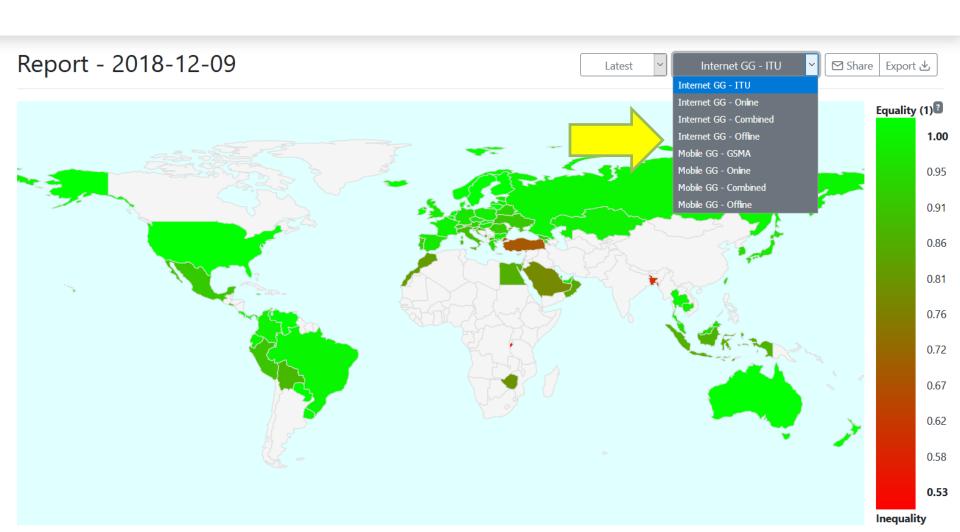












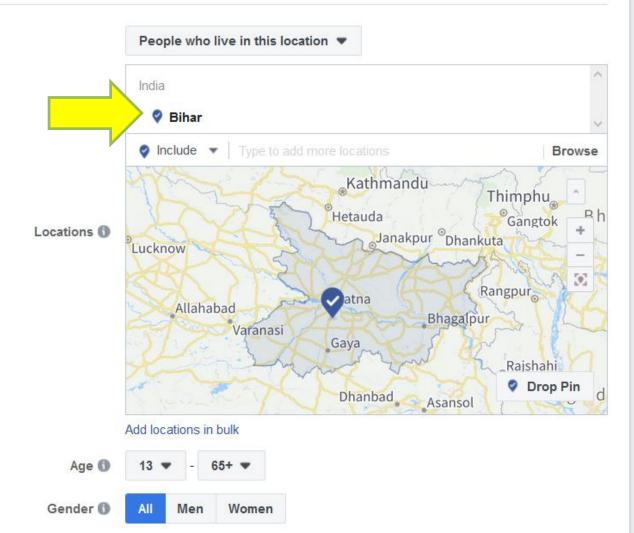
#### Modeling Internet Access Gender Gaps

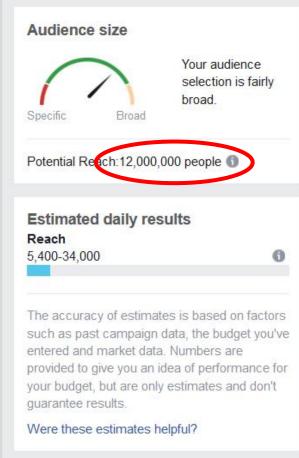
	Online Model	OnlOffl. Model	Offline Model
Intercept	0.933*** (0.006)	0.932*** (0.005)	0.933*** (0.007)
FB GG (age 18+)	0.071***	0.093***	(0.007)
log(GDP per capita)	(0.011)	(0.017) 0.018*	
GGGR – Literacy		(0.008) -0.018	
GGGR – Education		(0.016) -0.019	
Internet Penetration		(0.019)	0.040***
GGGR – Tertiary Educ.			(0.009) 0.032
GGGR – Economy			(0.021) 0.043**
GGGR Score			(0.014) -0.024
			(0.012)
Adjusted R-squared	0.691	0.791	0.615
# predicted countries <sup>a</sup>	152	127	132

\*\*\* p < 0.001, \*\* p < 0.01, \* p < 0.05.

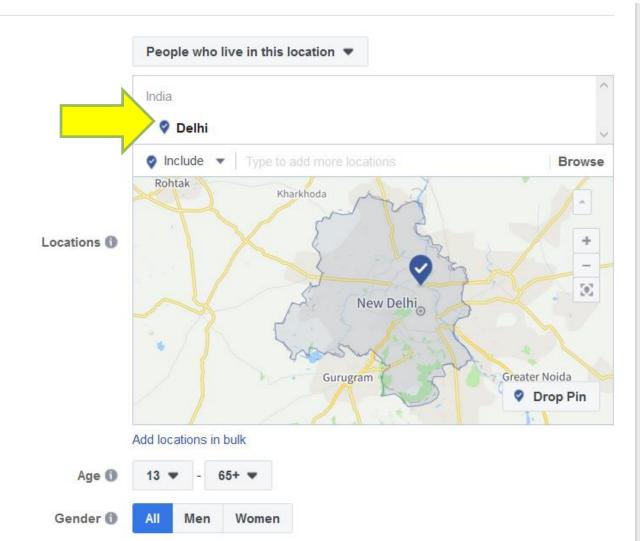
معهد قطر لبحوث الحوسبة "Using Facebook ad data to track the global digital gender gap", M. Fatehkia, R. Kashyap, I. Weber, World Development, 2018

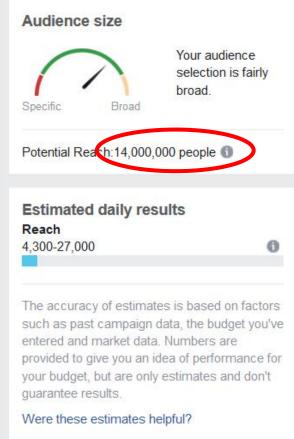
#### Device Type as Proxy for Income

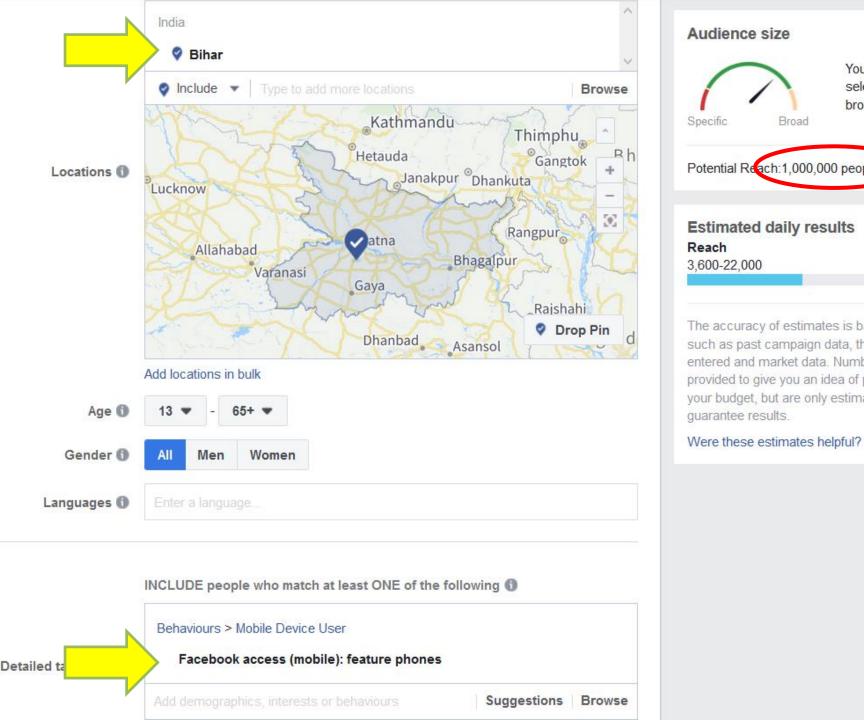




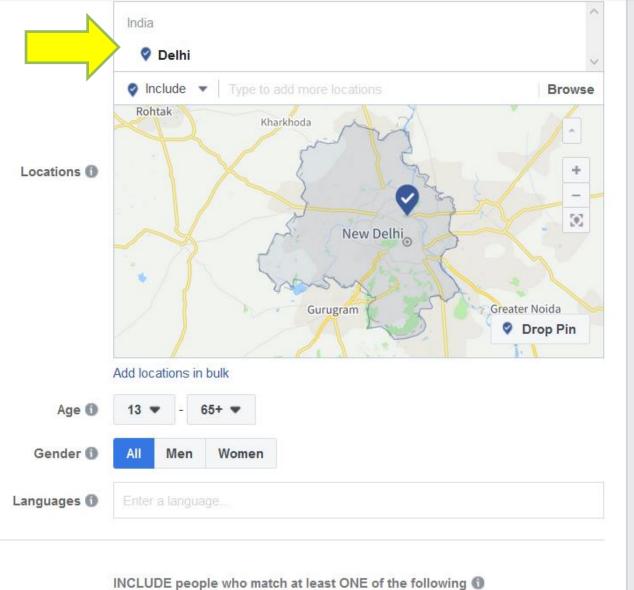
#### Device Type as Proxy for Income







#### Audience size Your audience selection is fairly broad. Specific Broad Potential Reach:1,000,000 people @ Estimated daily results Reach 3,600-22,000 The accuracy of estimates is based on factors such as past campaign data, the budget you've entered and market data. Numbers are provided to give you an idea of performance for your budget, but are only estimates and don't guarantee results.





# Audience size Your audience selection is fairly broad. Potential Reach: 440,000 people Estimated daily results Reach 2,700-17,000 The accuracy of estimates is based on factors such as past campaign data, the budget you've

The accuracy of estimates is based on factors such as past campaign data, the budget you've entered and market data. Numbers are provided to give you an idea of performance for your budget, but are only estimates and don't guarantee results.

Were these estimates helpful?

#### Examples of Targetable Devices

INCLUDE people who match at least ONE of the following

Behaviours > Mobile Device User > All Mobile Devices by Brand > Apple

Facebook access (mobile): iPhone 8

Facebook access (mobile): iPhone 8 Plus

Facebook access (mobile): iPhone X

Behaviours > Mobile Device User > All Mobile Devices by Brand > Samsung

Owns: Galaxy S8

Owns: Galaxy S8+

Owns: Galaxy S9

Owns: Galaxy S9+

Add demographics, interests or behaviours

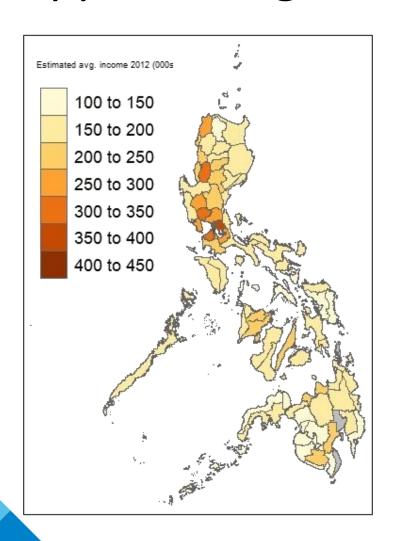
Suggestions

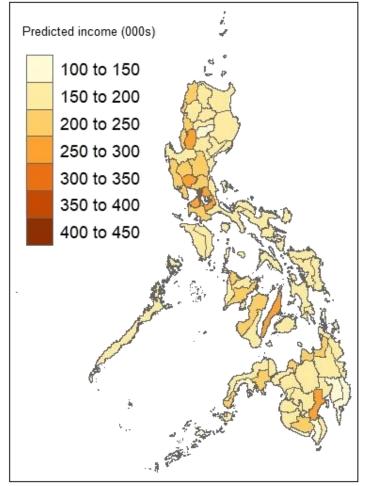
**Browse** 





#### Philippines: Regional Income Inequalities



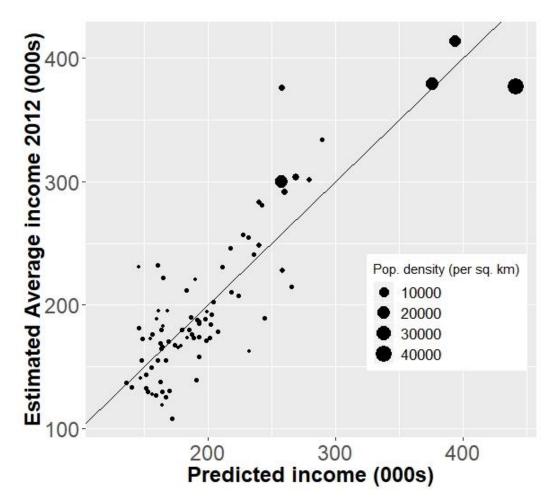




#### Philippines: Regional Income Inequalities

Most predictive features: %-age high end devices (r=.83) FB penetration (r=.81)

Adj.  $R^2 = .72$ 





#### Brazil: State-Level Ground Truth (2016)

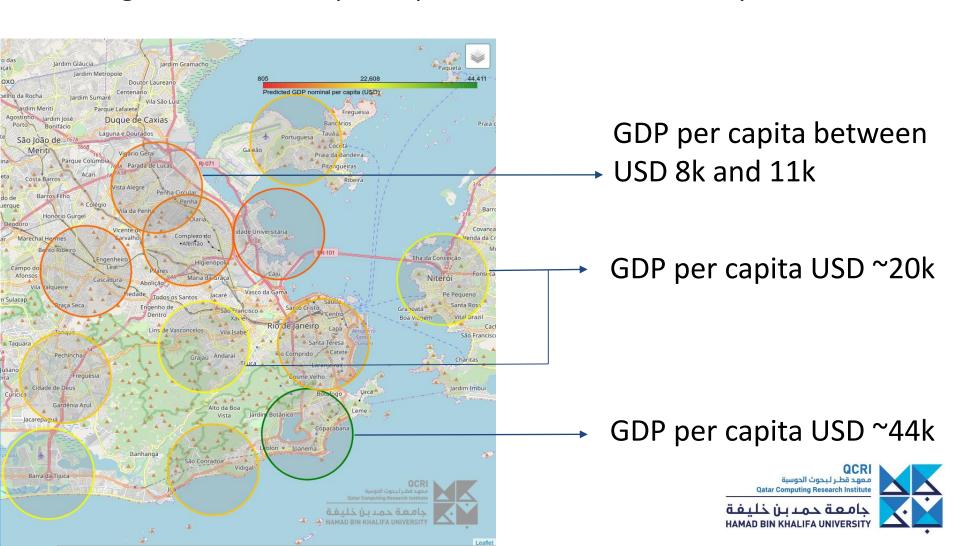


#### **Brazil: State-Level Predictions**

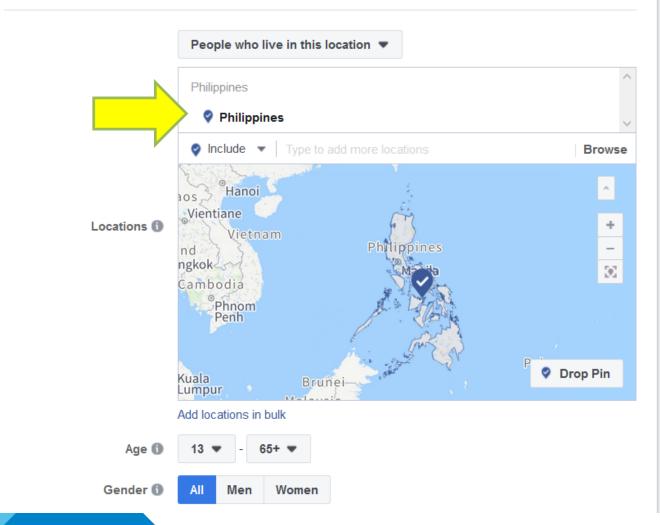


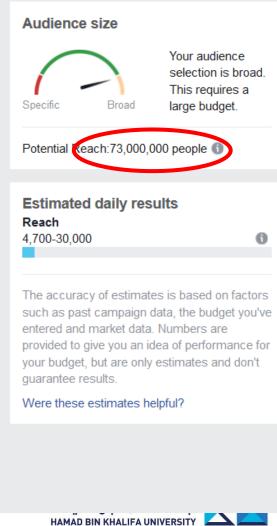
#### **Sub-City Income Monitoring**

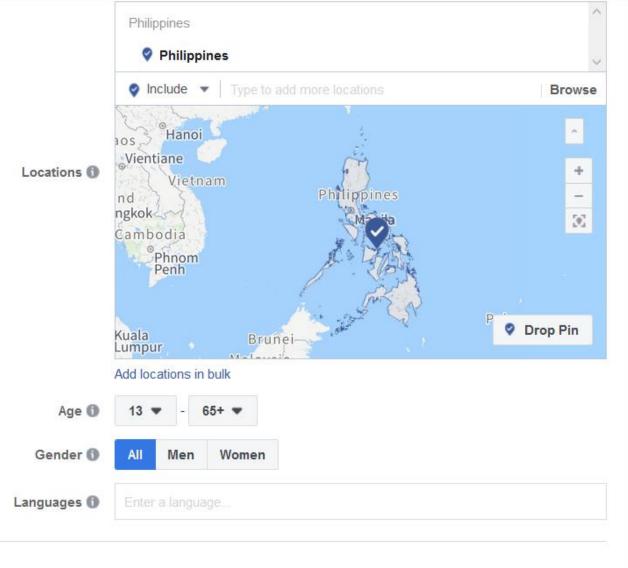
Average GDP nominal per capita for Rio de Janeiro city: ~USD 13K

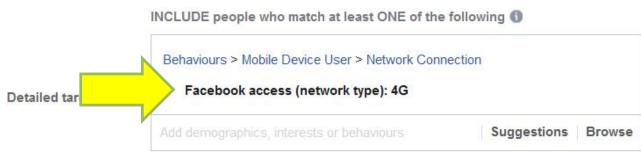


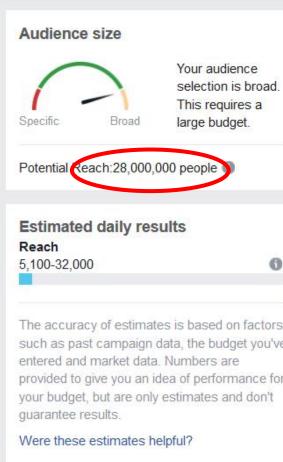
#### Access Type as Proxy for Development











#### Targetable Access Types

INCLUDE people who match at least ONE of the following 1

Behaviours > Mobile Device User > Network Connection

Facebook access (network type): 2G

Facebook access (network type): 3G

Facebook access (network type): 4G

Facebook access (network type): Wi-Fi

Suggestions

**Browse** 

X

Exclude people or Narrow audience

**Size:** 997,556,534

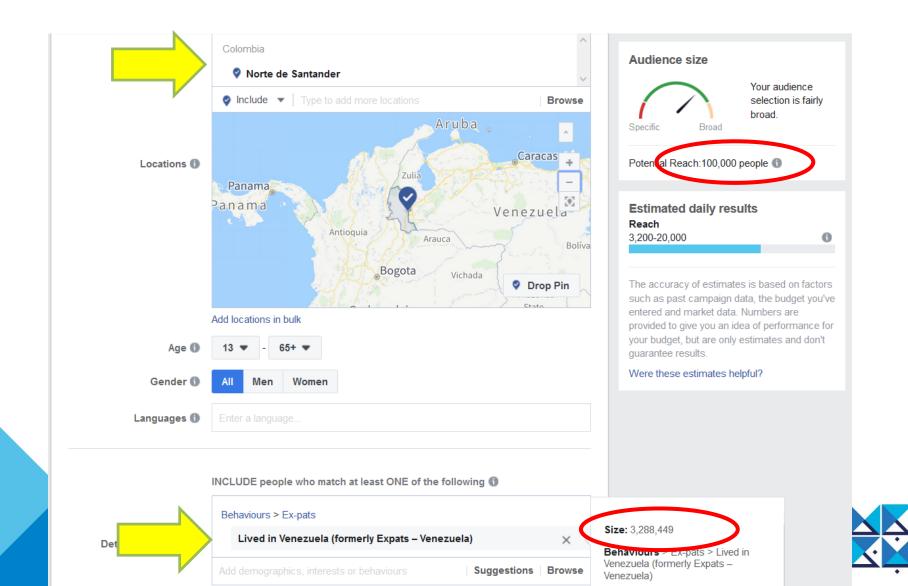
Behaviours > Mobile Device User > Network Connection > Facebook access (network type): 4G

**Description:** People who primarily access Facebook using a 4G network

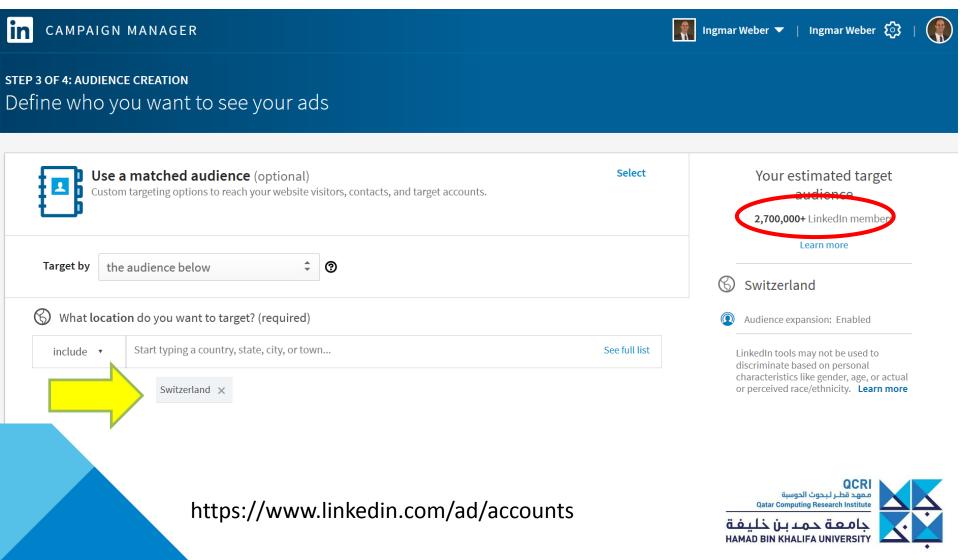
Report this as inappropriate



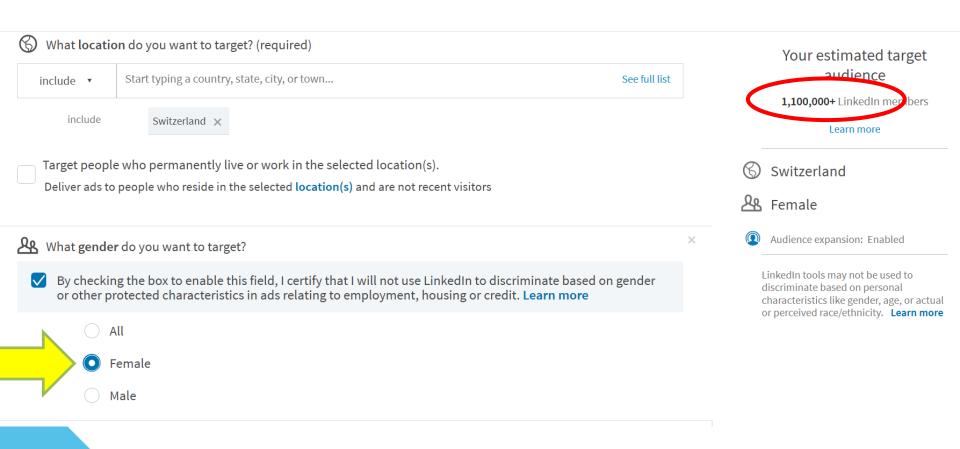
#### Migration Monitoring



#### LinkedIn's Advertising Platform

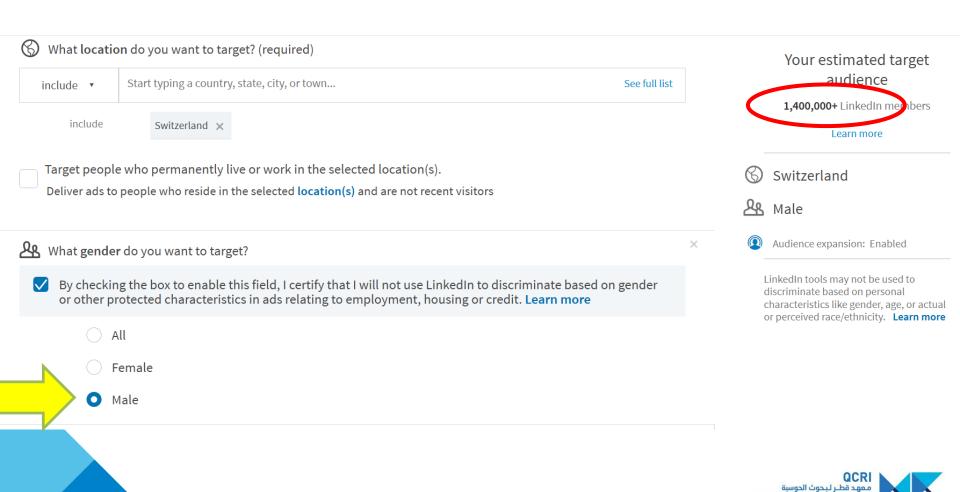


## Gender Gaps on LinkedIn



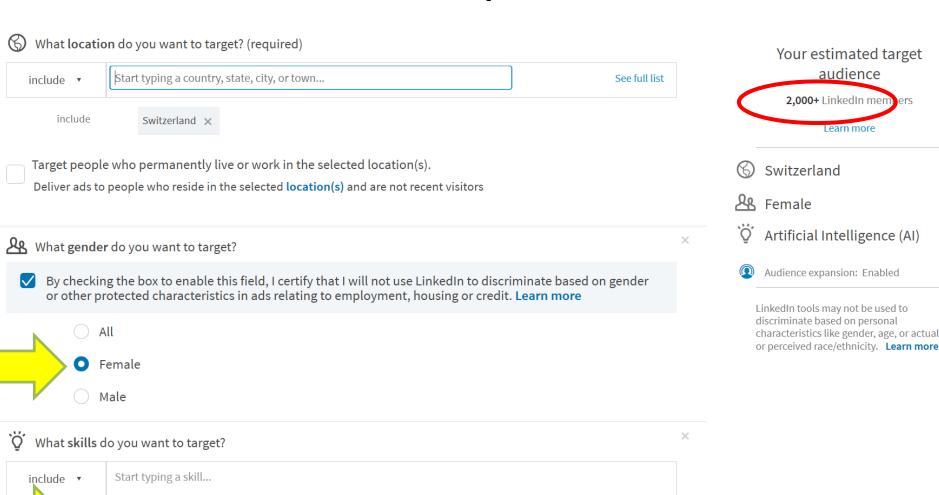


## Gender Gaps on LinkedIn



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#### Gender Skill Gaps on LinkedIn

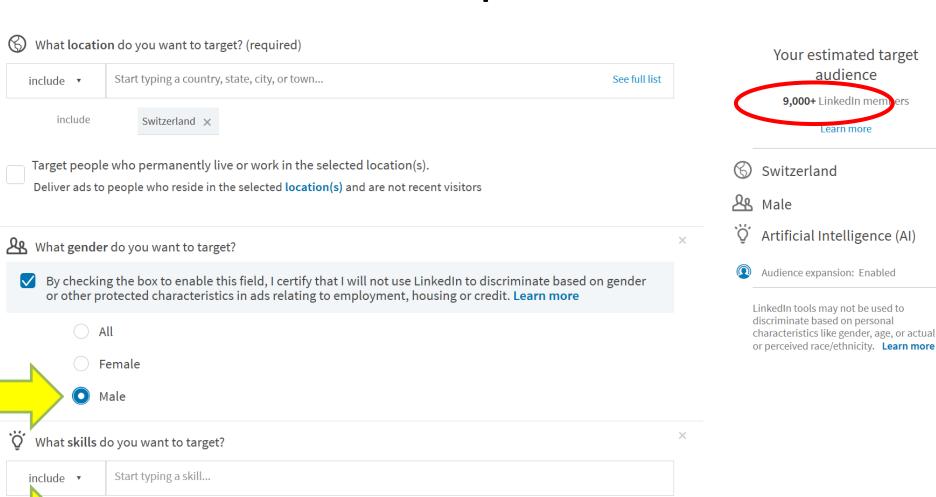


clude

Artificial Intelligence (AI) ×



#### Gender Skill Gaps on LinkedIn

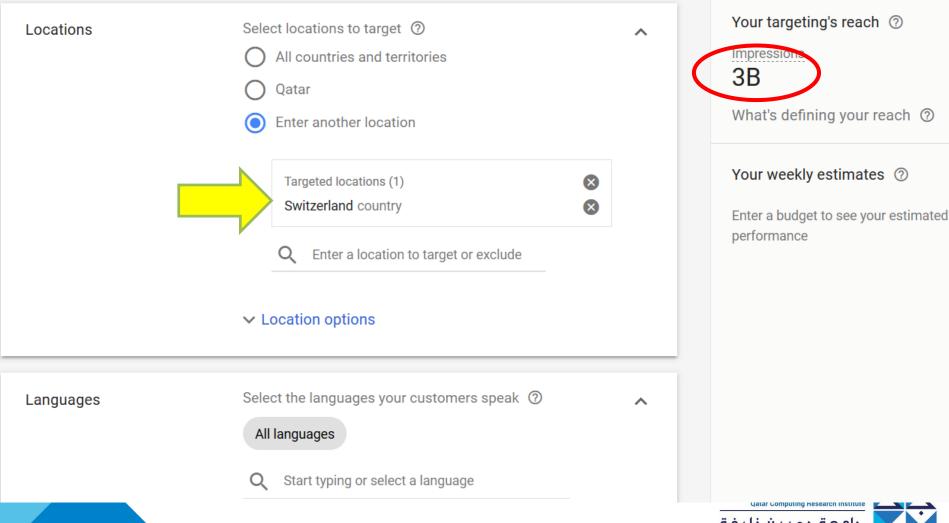


clude

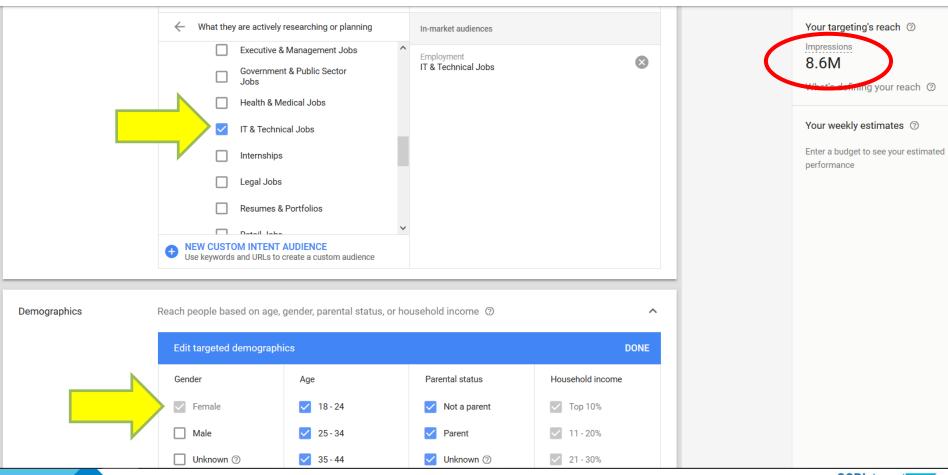
Artificial Intelligence (AI) ×



#### Google's Advertising Platform

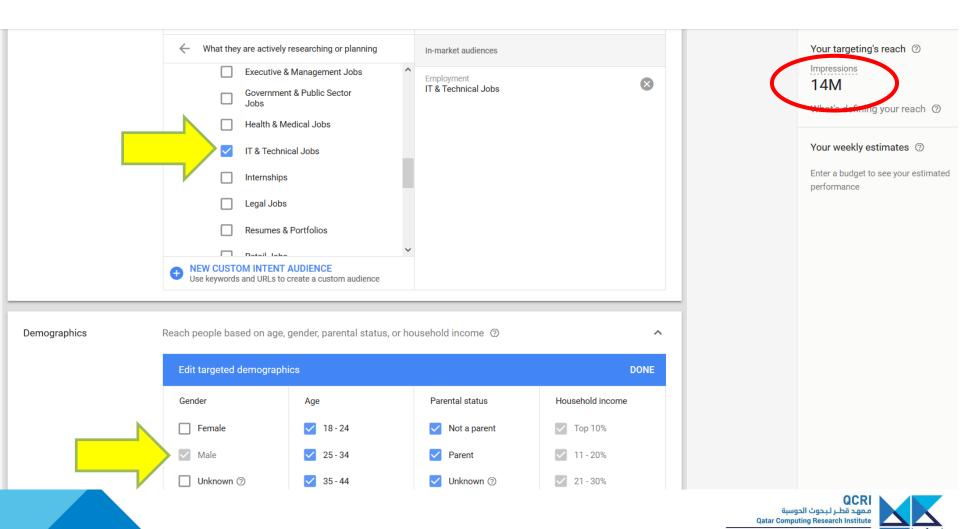


#### Gender Employment Interests





#### Gender Employment Interests



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#### Summary: Advertising Audience Estimates

- + Global reach with over 2 billion users
- + FB, LinkedIn, Google, Twitter, Snapchat, Instagram, vKontakte, ...
- + Real-time estimates
- + Uses anonymous and aggregate data
- + Disaggregation by gender, age, location, country of origin, ...
- + Includes non-traditional attributes such as interests and skills
- Proprietary black box on how attributes are inferred
- Needs modeling for bias correction, don't use at face value
- Needs periodic adjustments of the models as usage patterns change



#### **Key Partners**

















#### **Amazing Collaborators!**

Francesco Billari, Antoine Dubois, Masoomali Fatehkia, Harsh Gandhi, Kiran Garimella, Krishna Gummadi, Karri Haranko, Ridhi Kashyap, Yelena Mejova, Joao Palotti, Tejas Rafaliya, Francesco Rampazzo, Vatsala Singh, Reham Tamime, Agnese Vitali, Emilio Zagheni, ...



#### Thanks!

# Want to collaborate? Get in touch: iweber@hbku.edu.qa

References and full texts: <a href="https://ingmarweber.de/publications/">https://ingmarweber.de/publications/</a>

