Leveraging Facebook's Advertising Platform to Estimate Migration Stocks

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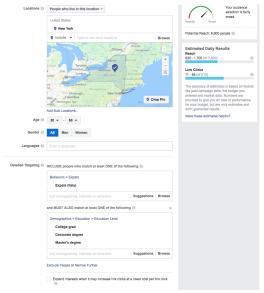
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IUSSP International Population Conference 2017 Cape Town, South Africa

Online (targeted) advertising: A booming trend



Targeting a demographic group on Facebook

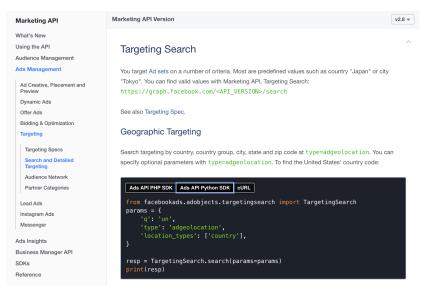


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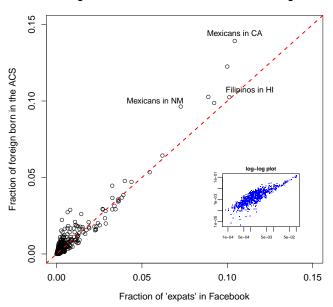
Your audience selection is fairly broad.
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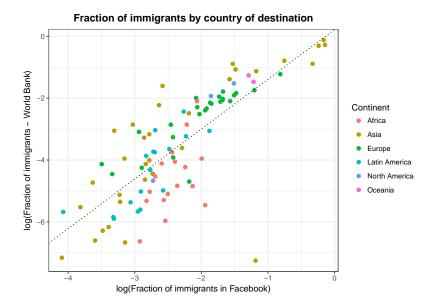
http://www.facebook.com/business

We can access the data in a programmatic way



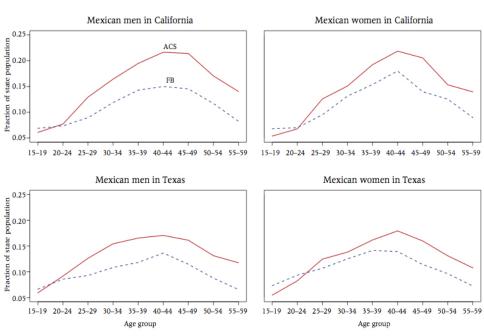
Migrants to US states for different countries of origin





Profiles by age and sex

FIGURE 3 Facebook and ACS profiles of stocks of migrants by age and sex for Mexicans in California and in Texas



Evaluating patterns in the bias

$$\begin{split} log(\text{ACS foreign born pop}_{ij}^z) &= \beta_0 + \beta_1 log(\text{Facebook expats}_{ij}^z) + \\ &+ \beta_2 \mathbb{1}(\text{Origin 1}) + \dots + \beta_{30} \mathbb{1}(\text{Origin 29}) + \\ &+ \beta_{31} \mathbb{1}(\text{Age group 1}) + \dots + \beta_{38} \mathbb{1}(\text{Age group 8}) + \\ &+ \epsilon_{ij}^z \end{split}$$

	$\log(\text{Foreign-born Population - ACS})$
og(FB Expats Population)	0.744*** (0.005)
ustria	0.420*** (0.083)
l'anada	0.200*** (0.051)
'hina	1.132*** (0.050)
rance	0.013 (0.056)
Germany	0.879*** (0.050)
Freece	1.443*** (0.064)
Iungary	0.264*** (0.078)
ndia	0.648*** (0.051)
ndonesia	-0.223****(0.065)
reland	0.193** (0.064)
srael	0.077 (0.063)
taly	0.051 (0.057)
apan	0.538*** (0.052)
falaysia	0.159* (0.068)
lexico .	0.540*** (0.052)
Iepal	-0.018 (0.062)
hilippines	0.098 (0.051)
oland	0.526*** (0.060)
ortugal	0.479*** (0.067)
uerto Rico	0.136* (0.053)
tomania	0.174** (0.059)
lussia	1.069*** (0.052)
ingapore	0.367*** (0.075)
outh Korea	0.811*** (0.051)
pain	0.041 (0.060)
urkev	0.044 (0.060)
JAE	0.376*** (0.099)
TK	-0.634*** (0.055)
lietnam	0.301*** (0.052)
ige group (20-24)	-0.483*** (0.032)
ige group (25-24) ige group (25-29)	-0.483 (0.032) -0.291*** (0.032)
ige group (20-23) ige group (30-34)	-0.231 (0.032) -0.010 (0.031)
ige group (35-34)	0.094** (0.031)
ige group (30-39) ige group (40-44)	0.301*** (0.031)
	0.309*** (0.031)
ge group (45-49)	0.460*** (0.031)
ge group (50-54)	
ge group (55-59)	0.519*** (0.031)
Constant	1.374*** (0.052)
/ 2	13,328
	0.698
djusted R ²	0.697
tesidual Std. Error	0.813 (df = 13289)
Statistic Statistic	807.060*** (df = 38; 13289)

*p < .05; **p < .01; ***p < .001

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Age pattern

Age group $(20-24)$	$-0.483^{***} (0.032)$
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Predictive capacity

▶ Goal: Predict the total number of foreign born from country *i* living in US state *j* (e.g, what is the stock of Mexicans in CA, Italians in NY?)

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- ► Split the data into training set (80% of US states) and test set (remaining 20% of US states)
- ► Estimate a model with no age & country of origin disaggregation and a model with disaggregation

Predictive capacity (continued)

The average out-of-sample Mean Absolute Percentage Error (MAPE) for total number of foreign born from country i living in state j:

- ▶ MAPE with no disaggregation by age-origin = 56%
- ▶ MAPE with disaggregation by age-origin = 37%

Predictive capacity (continued)

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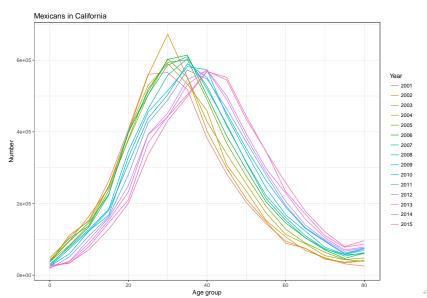
- ▶ MAPE with no disaggregation by age-origin = 56%
- ▶ MAPE with disaggregation by age-origin = 37%
- ⇒ Indication that accounting for biases for subgroups of the population, in a way analogous to post-stratification, helps with predictions

Next steps

With K. Polimis, M. Alexander, I. Weber and F. Billari

- Combining Facebook data with time series analysis of ACS data
 - Facebook provides timely, but biased data, with no constraints on age patterns
 - A Lee-Carter type of model for time series of migration stocks adds constraints on demographic patterns

ACS time series



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- ▶ In the meanwhile, getting there will be a fun process...

Thank you for your attention