Introduction to mass collaboration

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Schedule:

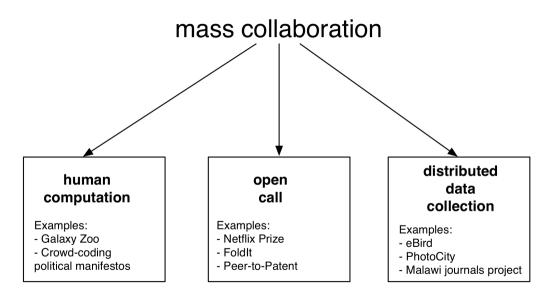
https://compsocialscience.github.io/summer-institute/2019/#schedule

- Observing behavior
- Asking questions
- Running experiments
- ► Creating mass collaboration



Mass collaboration combines ideas from

- crowdsourcing
- citizen science
- collective intelligence



Guiding idea:

Collaborators not cogs (ornithology and astronomy are examples)

▶ Is this really research?

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- ▶ Does this enable new research?

► Is this perfect?

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- ▶ Is this better than we can do without mass collaboration?

▶ Is this impossible?

- ▶ Is this impossible?
- ► Is this possible?

An honest assessment:

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- ► Human computation
- ► Open call
- ► Distributed data collection

► Easy task, big scale

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- ► Increasingly important as we move from numeric survey data to working with text, images, movies, audio, etc.

Crowd-sourced Text Analysis: Reproducible and Agile Production of Political Data

KENNETH BENOIT London School of Economics and Trinity College
DREW CONWAY New York University
BENJAMIN E. LAUDERDALE London School of Economics and Political Science
MICHAEL LAVER New York University
SLAVA MIKHAYLOV University College London

http://dx.doi.org/10.1017/S0003055416000058

Impirical social science often relies on data that are not observed in the field, but are transformed into quantitative variables by expert researchers who analyze and interpret qualitative raw sources. While generally considered the most valid way to produce data, this expert-driven process is inherently difficult to replicate or to assess on grounds of reliability. Using crowd-sourcing to distribute text for reading and interpretation by massive numbers of nonexperts, we generate results comparable to those using experts to read and interpret the same texts, but do so far more quickly and flexibly. Crucially,

http://dx.doi.org/10.1017/S0003055416000058

Here's a piece of the manifesto of the Labor Party in the United Kingdom from 2010:

"Millions of people working in our public services embody the best values of Britain, helping empower people to make the most of their own lives while protecting them from the risks they should not have to bear on their own. Just as we need to be bolder about the role of government in making markets work fairly, we also need to be bold reformers of government."

FIGURE 1. Hierarchical Coding Scheme for Two Policy Domains with Ordinal Positioning

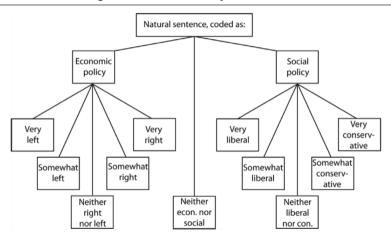
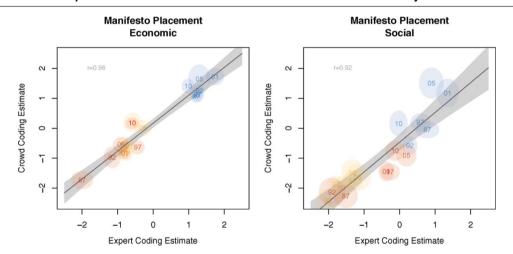


FIGURE 3. Expert and Crowd-sourced Estimates of Economic and Social Policy Positions



What I like about Benoit et al (2016)

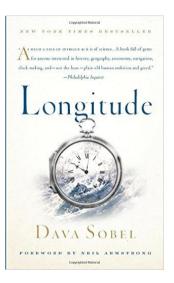
► Better not cheaper

What I like about Benoit et al (2016)

- ► Better not cheaper
- Experts are a bug not a feature

Questions?

- ► Human computation
- ► Open call
- ► Distributed data collection



Solutions are easier to check than to generate

You will participate in an open call in a few moments

Questions?

- ► Human computation
- ► Open call
- ► Distributed data collection

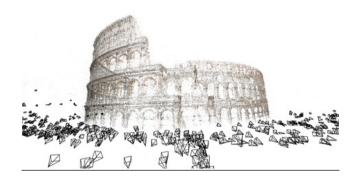
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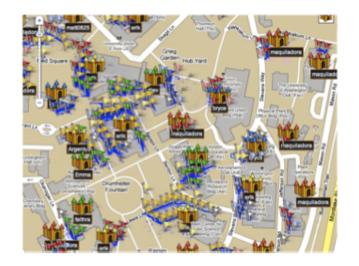
- ▶ people can be where the researchers can't
- scale that researcher cannot match
- sometimes hard to separate from human computation



Tuite et al. (2011) "PhotoCity: Training Experts at Large-scale Image Acquisition Through a Competitive Game" *CHI*: http://dx.doi.org/10.1145/1978942.1979146



Rome in a Day (Agarwal et al., 2009)



Two campuses: University of Washington and Cornell University

Over 2 months, 100,000 photos submitted by 45 players



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- data collection is standardized because of cameras
- verification is automatic by comparison with nearby images
- game points are assigned based on the value of data, trains people to collect more valuable data

Questions about distributed data collection or mass collaboration?