Discovery

Gary King Institute for Quantitative Social Science Harvard University

covering joint work with

Justin Grimmer (Harvard) and Eleanor Powell (Harvard → Yale)

(talk @ Institute for Qualitative and Multimethod Research, Syracuse University, 5/26/09)

0

Reading

- Gary King and Eleanor Neff Powell. 2008. "How Not to Lie Without Statistics"
- Justin Grimmer and Gary King. 2009. "Quantitative Discovery of Qualitative Information: A General Purpose Document Clustering Methodology"

http://GKing.harvard.edu

Is Qualitative or Quantitative Research Better?

This question has an answer (hard to know from the qualitative methods literature!)

If insufficient information is quantified, qualitative judgment wins

- The vast majority of human decisions are qualitative
- If not, we wouldn't have survived as a species

If sufficient information is quantified, statistics wins

- The information required: shockingly little
- 6 crude variables beat 83 law professors (Martin et al., 2004)
- Simple quantitative models forecast elections better than expert pundits (Gelman and King, 1993; Campbell, 2005)
- 284 experts forecast the political future with "less skill than simple extrapolation" (Tetlock, 2005)
- statistical model out-performs physicians in determining cause-specific mortality rates (King and Lu, 2008)
- Hundreds of head-to-head contests, mostly with the same conclusion (Meehl, 1954; Grove, 2005)
- The march of quantification across fields of academia, professions, commerce, sports, etc. (Moneyball, SuperCrunchers, Numerati)

Discovery

3 Steps

- Conceptualization (e.g., a categorization scheme)
- Measurement (e.g., classifying objects into categories)
- Verification (e.g., testing a hypothesis that could be wrong)

Styles & Approaches (the cause of many misunderstandings)

- Quantitative: focus on measurement and verification → assumes & underplays conceptualization
- Qualitative: focus on (& iterate between) conceptualization and measurement → ignores or underplays verification

Proposed Solutions

- Do both qualitative & quantitative analysis, separately
- Atheoretical unsupervised learning algorithms
- $\bullet \leadsto \underline{\text{Integrated}} \text{ quant/qual: Computer-assisted qualitative analysis}$

The Problem: Discovery from Unstructured Text

- Examples: scholarly literature, news stories, medical information, blog posts, comments, product reviews, emails, social media updates, audio-to-text summaries, speeches, press releases, legal decisions, etc.
- 10 minutes of worldwide email = 1 LOC equivalent
- An essential part of discovery is classification: "one of the most central and generic of all our conceptual exercises. . . . the foundation not only for conceptualization, language, and speech, but also for mathematics, statistics, and data analysis. . . . Without classification, there could be no advanced conceptualization, reasoning, language, data analysis or, for that matter, social science research." (Bailey, 1994).
- We focus on cluster analysis: discovery through (1) classification and (2) simultaneously inventing a classification scheme
- (We analyze text; our methods apply more generally)

Why Johnny Can't Classify (Optimally)

- Bell(n) = number of ways of partitioning n objects
- Bell(2) = 2 (AB, A B)
- Bell(3) = 5 (ABC, AB C, A BC, AC B, A B C)
- Bell(5) = 52
- \bullet Bell(100) $\approx 10^{28} \times \text{Number of elementary particles in the universe}$
- Now imagine choosing the optimal classification scheme by hand
- Qualitative-only approaches are hopeless
- That we think of all this as astonishing . . . is astonishing

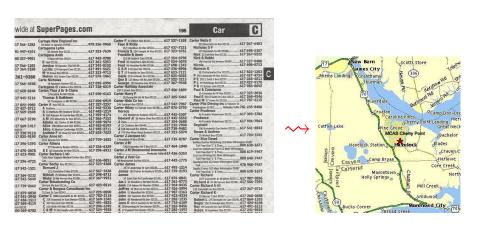
Why HAL Can't Classify Either

- The Goal an optimal application-independent cluster analysis method — is mathematically impossible:
 - No free lunch theorem: every possible clustering method performs equally well on average over all possible substantive applications
- Existing methods:
 - Many choices: model-based, subspace, spectral, grid-based, graphbased, fuzzy k-modes, affinity propogation, self-organizing maps,...
 - Well-defined statistical, data analytic, or machine learning foundations
 - How to add substantive knowledge: With few exceptions, who knows?!
 - The literature: little guidance on when methods apply
 - Deriving such guidance: difficult or impossible
 - (Perhaps true by definition in unsupervised learning: If we knew the DGP, we wouldn't be at the discovery stage.)

If Ex Ante doesn't work, try Ex Post

- Methods and substance must be connected (no free lunch theorem)
- The usual approach fails: can't do it by understanding the model
- We do it ex post (by qualitative choice)
 - For discovery (our goal): No problem
 - For estimation & confirmation: more difficult or biased
- Complicated concepts are easier to define ex post:
 - "I know it when I see it" (Justice Stewart's definition of obscenity)
 - Anchoring Vignettes (on defining concepts by example)
- But how to choose from an enormous list of clusterings?

Our Idea: Meaning Through Geography



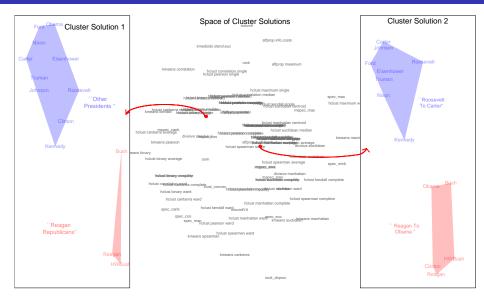
→ We develop a (conceptual) geography of clusterings

A New Strategy

- Code text as numbers (in one or more of several ways)
- Apply all existing clustering methods (that have been used by at least one person other than the author) to the data — each representing different substantive assumptions (<15 mins)</p>
- Develop an application-independent distance metric between clusterings
- Oreate a metric space of clusterings, and a 2D projection
- Introduce the local cluster ensemble to summarize any point, including points with no existing clustering
- Propose a new animated visualization: use the local cluster ensemble to explore the space of clusterings (smoothly morphing from one into others)
- → meaning revealed through a geography of clusterings

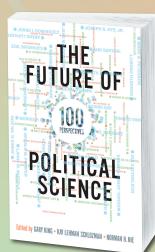
Many Thousands of Clusterings, Sorted & Organized

You choose one (or more), based on insight, discovery, useful information,...



Application-Independent Distance Metric: Axioms

- Clusterings with more pairwise document agreements are closer (we prove: pairwise agreements encompass triples, quadruples, etc.)
- Invariance: Distance is invariant to the number of documents (for any fixed number of clusters)
- Scale: the maximum distance is set to log(num clusters)
- → Only one measure satisfies all three (the "variation of information")



Available March 2009: 304pp Pb: 978-0-415-99701-0: **\$24.95** www.routledge.com/politics

THE FUTURE OF POLITICAL SCIENCE

100 Perspectives

Edited by Gary King, Harvard University, Kay Lehman Schlozman, Boston College and Norman H. Nie, Stanford University

"The list of authors in *The Future of Political Science* is a 'who's who' of political science. As I was reading it, I came to think of it as a platter of tasty hors d'oeuvres. It hooked me thoroughly."

—Peter Kingstone, University of Connecticut

"In this one-of-a-kind collection, an eclectic set of contributors offer short but forceful forecasts about the future of the discipline. The resulting assortment is captivating, consistently thought-provoking, often intriguing, and sure to spur discussion and debate."

-Wendy K. Tam Cho, University of Illinois at Urbana-Champaign

"King, Schlozman, and Nie have created a visionary and stimulating volume. The organization of the essays strikes me as nothing less than brilliant. . . It is truly a joy to read."

—Lawrence C. Dodd, Manning J. Dauer Eminent Scholar in Political Science, University of Florida



Evaluators' Rate Machine Choices Better Than Their Own

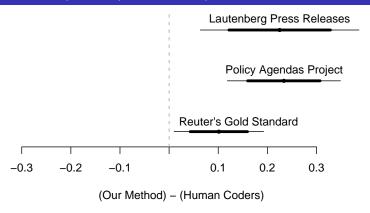
- Scale: (1) unrelated, (2) loosely related, or (3) closely related
- Table reports: mean(scale)

Pairs from	Overall Mean	Evaluator 1	Evaluator 2
Random Selection	1.38	1.16	1.60
Hand-Coded Clusters	1.58	1.48	1.68
Hand-Coding	2.06	1.88	2.24
Machine	2.24	2.08	2.40

p.s. The hand-coders did the evaluation!

Cluster Quality Experiments

Scale: mean(within clusters) — mean(between clusters)



Lautenberg: 200 Senate Press Releases (appropriations, economy, education, tax, veterans, ...)

Policy Agendas: 213 quasi-sentences from Bush's State of the Union (agriculture, banking & commerce, civil rights/liberties, defense, . . .)

Ceuter's: financial news (trade earnings copper gold coffee): "gold 15/1

What do Members of Congress Do?

Substantive example of a finding, using our approach

- David Mayhew's (1974) famous typology
 - Advertising
 - 2 Credit Claiming
 - Osition Taking
- We find one more: Partisan Taunting
 - "Senator Lautenberg Blasts Republicans as 'Chicken Hawks' "
 [Government Oversight]
 - "The scopes trial took place in 1925. Sadly, President Bush's veto today shows that we haven't progressed much since then." [Healthcare]
 - "John Kerry had enough conviction to sign up for the military during wartime, unlike the Vice President, who had a deep conviction to avoid military service" [Government Oversight]
 - \rightsquigarrow Is this what it means to be a member of a political party?

More Informative Discoveries

- Found 2 scholars analyzing lots of textual data for their work
- For each: created 2 clusterings from each of 3 methods, including ours
- Created info packet on each clustering (for each cluster: exemplar document, automated content summary)
- Asked for $\binom{6}{2}$ =15 pairwise comparisons
- Both cases a Condorcet winner:

```
"Immigration":
```

```
\underline{\text{Our Method 1}} \rightarrow \text{vMF 1} \rightarrow \text{vMF 2} \rightarrow \underline{\text{Our Method 2}} \rightarrow \text{K-Means 1} \rightarrow \text{K-Means 2}
```

```
"Genetic testing":
```

 $\underline{\mathsf{Our}\;\mathsf{Method}\;1} \to \{\underline{\mathsf{Our}\;\mathsf{Method}\;2},\;\mathsf{K}\text{-}\mathsf{Means}\;1,\;\mathsf{K}\text{-}\mathsf{means}\;2\} \to \mathsf{Dir}\;\mathsf{Proc}.\;1 \to \mathsf{Dir}\;\mathsf{Proc}.\;2$

Intended contributions

- An encompassing cluster analytic approach for discovery
- A new approach to evaluating results in unsupervised learning
- Especially useful for the ongoing spectacular increase in the production and availability of unstructured text
- Multiple approaches: Integrated, not separate

For more information:

http://GKing.Harvard.edu