

“Computer-Assisted Reading” and other Discoveries from Quantitative Social Science

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(*Crimson Conversations* Talk, Riverside, CT, 4/12/12)

The Emergence of Quantitative Social Science

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 - We now read a tiny fraction haphazardly, and delude ourselves into thinking we understand all we need

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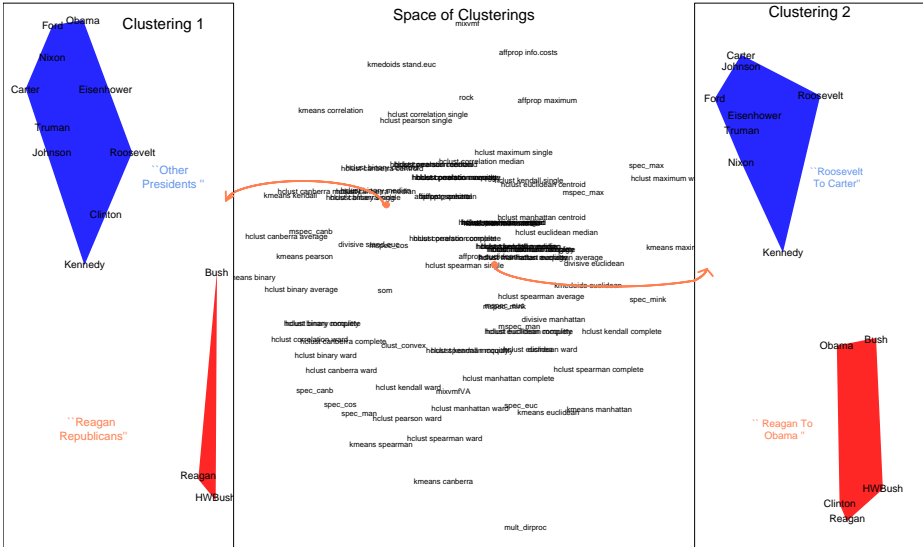
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- **Question: How to organize clusterings so humans can understand?**

Humans Can Zoom in to Read; We Can Zoom Out

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



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 - Both scholars preferred our insight to their's

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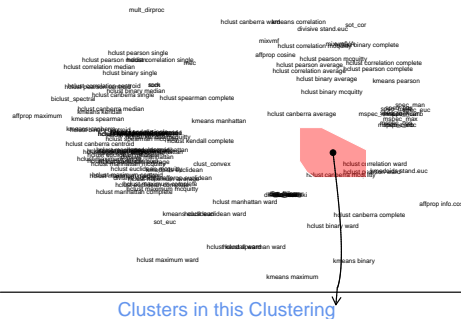
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- Apply our method

Example Discovery



Credit Claiming, Legislation:

“As the Senate begins its recess, Senator Frank Lautenberg today pointed to a string of victories in Congress on his legislative agenda during this work period”



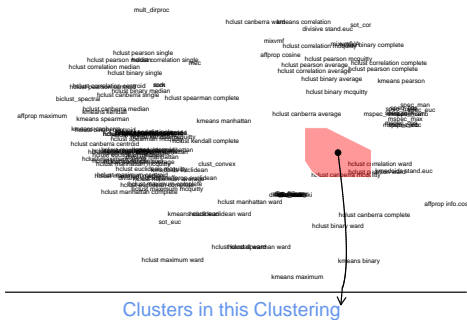
Credit Claiming
Pork



Mayhew Credit Claiming
Legislation

Gary King (Harvard)

Example Discovery



Advertising:

“Senate Adopts Lautenberg/Menendez Resolution Honoring Spelling Bee Champion from New Jersey”



Credit Claiming
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Advertising



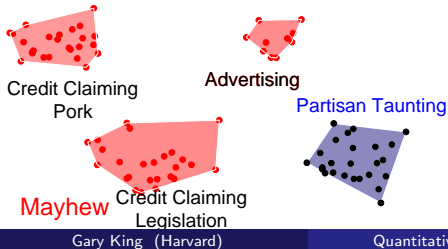
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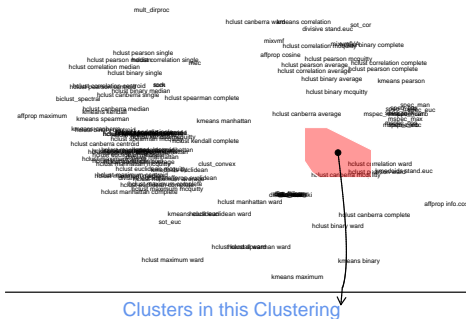
Clusters in this Clustering



Partisan Taunting:

“Senator Lautenberg’s amendment would change the name of . . . the Republican bill . . . to ‘More Tax Breaks for the Rich and More Debt for Our Grandchildren Deficit Expansion Reconciliation Act of 2006’”

Example Discovery



Definition of Partisan Taunting:
Explicit, public, and negative attacks on another political party or its members



Credit Claiming
Pork



Advertising

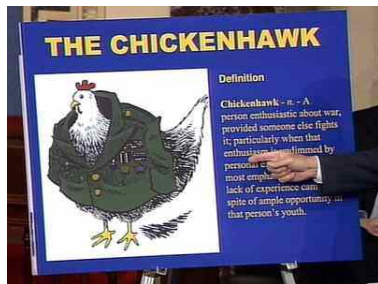
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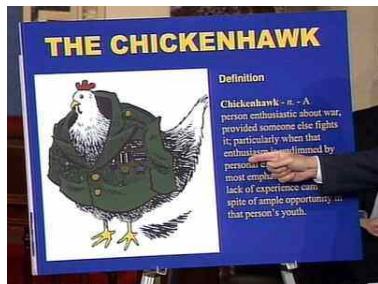
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Sen. Lautenberg
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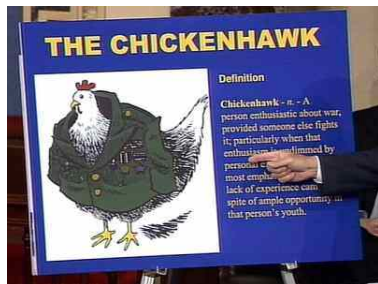
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- "Every day the House Republicans dragged this out was a day that made our communities less safe." [Homeland Security]

Out of Sample Confirmation of Partisan Taunting

- Discovered using 200 press releases; 1 senator.

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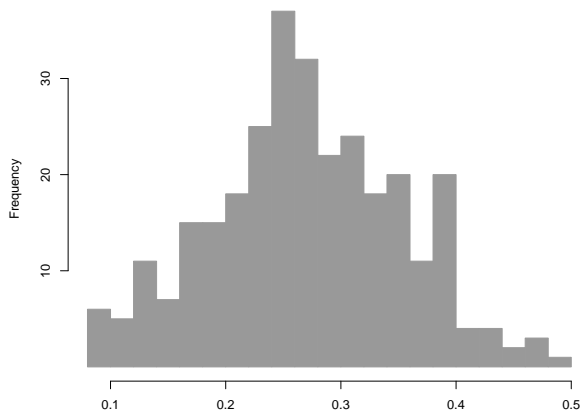
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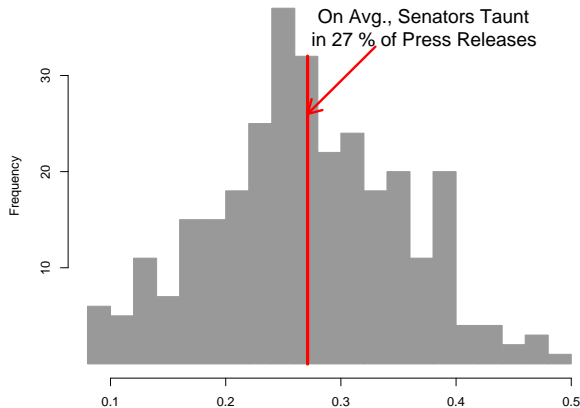
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- 13 **Scholarly data:** the replication movement in academia, led in part by political science, is massively increasing data sharing

For more information



<http://GKing.Harvard.edu>

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(Why Johnny Can't Classify)

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- Now imagine choosing the *optimal* classification scheme by hand!

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 - \rightsquigarrow you can't have both!