Fortune 500 Electoral Influence

1990-2012

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About the data

Sunlight Foundation State and Federal Campaign Contributions
1990 - 2012

(Two Year Increments Corresponding to Cycle)

Federal law requires that campaigns and PACs disclose information about the identities of individuals, corporate or private, who donate over \$200 to a in a given year or cycle.

Name
Residence
Employer/Organization
Recipient Information
Date

15.1 GB

Or 50.7 Million transactions for all Federal and State level campaigns

Top individual contributors tend to be lesser known

Contributions Individual

1 \$729,005,408.00 Mr. Lee L Mercer Jr.

2 \$54,756,566.00 Stephen L. Bing

3 \$53,822,872.00 Linda McMahon

4 \$39,830,810.31 Bob J. Perry

Contributions Individual

5 \$28,215,750.00 Thomas F. Steyer

11 \$26,798,373.20 James & Virginia Stower

13 \$19,516,336.04 James E. Pederson

15 \$17,889,650.00 John R. Raese







Top organizations are well known heavy-weights

Rank	Contributions	Organization
1	\$1,348,150,439.38	National Republican Congressional Committee
2	\$991,885,410.28	Republican National Committee
3	\$777,393,014.00	DNC Services Corps
4	\$683,447,745.15	Democratic Congressional Campaign Committee
5	\$631,654,092.89	Democratic Senatorial Campaign Committee
11	\$137,855,333.64	Service Employees International Union
13	\$128,364,947.88	California Teacher's Association
15	\$97,952,177.43	Philip Morris Company
21	\$84,236,959.00	National Rifle Association
23	\$81,540,958.00	National Association of Realtors

There is substantial variation between the causes to whom Fortune 500 corporations donate and those to whom their employees donate.

Resolve entities with substantial spelling variation and the occasional unicode character.

Sample of Goldman Sachs Variations

GOLDMAN SAKS GOLDRAN SACHS

GOLDMAN, SACHS & CO GOLDMAN SACHS & CO.

GOLDMAN- SACHS & CO. GOLDMAN SACH & COMPANY

GOLMAN SACHS GOLDMAN SACHS & COMPANY

GOLDMAN SACHS GROUP GOLDMAN SACHS ASIA LLC

Entity resolution algorithm

Read in each entry and thoroughly clean all unicode

Loop through F500 authoritative list and match

Output JSON for later use

GOLDMAN SAKS —— Prudential
Financial

Humana, Inc.

The Goldman Sachs
Group, Inc.

*[{ 'GOLDMAN SAKS':
 'The Goldman Sachs
 Group, Inc.' }]

- 1. Remove stop words
- 2. Strip misc. Chars
- Simple edit distance score using fuzzywuzzy

Tesoro Corporation

Liberty Mutual Holding Company

Runtime.

Runtime issues, even with parallelization

Entity resolution algorithm

200,000 records:

37 minutes 10 seconds on on local 512.729 records:

5x the work per node AWS 20 nodes about 30 - 40 minutes

Data aggregation algorithm

- l. Remove unicode encoding
- 2. Strip characters
- Look-up organization namein JSON, held in memory on AWS using configure options()
- 4. Determine whether corporate or individual contribution
- 5. Export to CSV

Achieved constant run time and could process the full dataset in 30 minutes on 40 nodes

Entity reduction results from 1 percent of total data

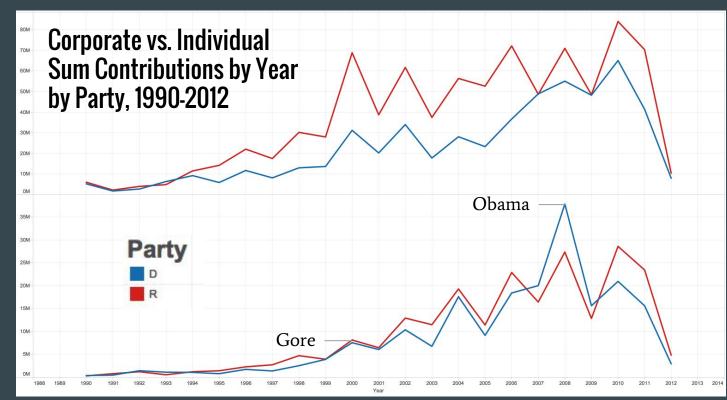
28 Aliases 26 Aliases 18 Aliases 17 Aliases

Confirmed.

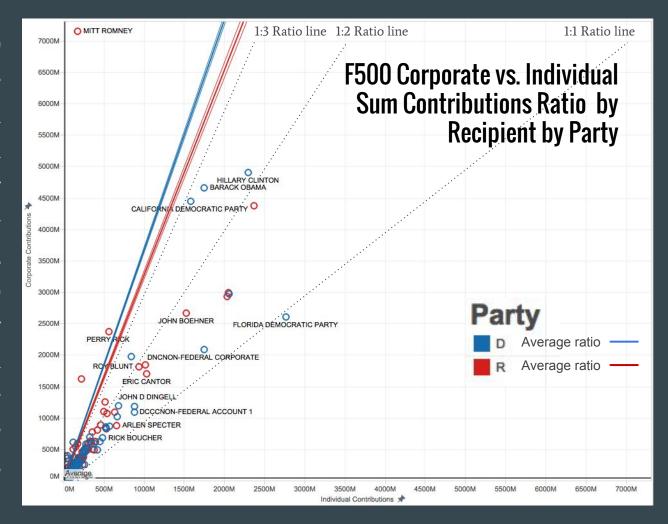
F500 Corporations don't always agree with employees

Corporate Contributions

Individual Contributions



While overall F500 corporations donate more to Republican candidates, when compared to their employees at a dollarto-dollar ratio, corporations tend to slightly outweigh their employees on Democratic candidate and committee contributions

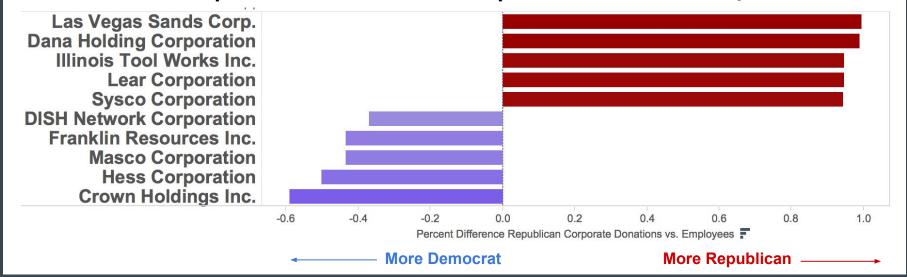


Quantifying Disconnect





Difference Percent Republican Contributions for Corporations vs. Individual, by Co.



Stuff We Want to Do

More specifically, we ran out of time to do this stuff because of AWS issues

Analysis of Donor Networks

Extraction of all pairs and loading into a network analysis package, where edges represent money transfers, to study relationship of centrality to electoral success.

Full Dataset Entity Resolution

Because of time constraints and issues with MRJob on AWS, we were constrained to Fortune 500 companies and their employees.

Things We Learned

What we would tell ourselves

MRJob data passing protocols are difficult to understand, but useful

Reading in data from any source must deal with encoding issues by default

Entity reduction is computationally very expensive

AWS costs can balloon quickly

Thanks!

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