Leveraging growth modeling to examine topic usage over time

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Topic modeling

- **Topic modeling** uses statistical associations of words in a collection of texts (i.e., a corpus) to generate latent topics
 - Uses an algorithm to uncover clusters of co-occurring words that jointly represent higher-order concepts (e.g., novelty or innovation)



Topics over time?

- What if you wanted to exami
- For example:

Three Similarities

- 1. Each is regarding constructs directly or indirectly measured through rhetoric
- 2. Focus is on general rather than specific rhetoric
- Interested in examining changes in generalized rhetoric over time
- How do perceptions of a company's work environment change following the adoption of new policies and procedures?
- How does leader rhetoric regarding social justice issues change in response to social movements like BLM or Time's Up?
- How do corporate values change over time? Are changes to corporate values punctuated by CEO and/or board changes?

Topics over time? – A possible solution

- Companies communicate information to stakeholders about themselves to project a desired image (Elsbach, 1994; 2003, Gioia & Thomas, 1996)
- RQs: How does the IPO process change projected image? How do changes in ownership influence such changes?
- Data
 - 168 IPO firms between 2009-2012
 - "About Us" website information -5 years before to +5 years following IPO
 - Founder/VC ownership collected from IPO prospectuses, proxy statements, and annual reports
- Analysis
 - Topic models & CATA in R using *tm* and *topicmodels* packages
 - Discontinuous growth modeling using *mixed* in Stata

Content analysis code structure overview

- 1. Topic model pre-IPO text data for firm *j* (3 topics)
- 2. Store top 25 words per topic (3 pre-IPO topic word lists)
- 3. Topic model post-IPO data for firm *j* (3 topics)
- 4. Store top 25 words per topic (3 post-IPO topic word lists)
- 5. CATA analyze entire corpus for firm *j* using all 6 word lists
- 6. Repeated steps 1-5 for the remaining 167 IPO firms
- 7. Export to .dta using write.dta (*foreign* library)

		Icar	ipoyear	topic	time	time_ipo	time_post	pretopic_std	posttopic_~d
1	ACOM	2004	2009	1	0	0	0	.118705	.1151079
2	ACOM	2005	2009	1	1	0	0	.2201835	.1100917
3	ACOM	2006	2009	1	2	0	0	.2230769	.0923077
4	ACOM	2007	2009	1	3	0	0	.2610442	.1084337
5	ACOM	2008	2009	1	4	0	0	.2610442	.1084337
6	ACOM	2009	2009	1	5	1	0	.1153846	.025641
7	ACOM	2010	2009	1	6	1	1	.0642202	.0458716
8	ACOM	2011	2009	1	7	1	2	.0566038	.0314465
9	ACOM	2012	2009	1	8	1	3	.0897436	.4102564
10	ACOM	2013	2009	1	9	1	4	.0503597	.1798561
11	ACOM	2004	2009	2	0	0	0	.1978417	.0539568
12	ACOM	2005	2009	2	1	0	0	.1284404	.0642202
13	ACOM	2006	2009	2	2	0	0	.1384615	.0769231
14	ACOM	2007	2009	2	3	0	0	.1084337	.0803213
15	ACOM	2008	2009	2	4	0	0	.1084337	.0803213
16	ACOM	2009	2009	2	5	1	0	.0641026	.1153846
17	ACOM	2010	2009	2	6	1	1	.0366972	.2293578
18	ACOM	2011	2009	2	7	1	2	.0314465	.2012579
19	ACOM	2012	2009	2	8	1	3	.0384615	.1410256
20	ACOM	2013	2009	2	9	1	4	.028777	.0791367
21	ACOM	2004	2009	3	0	0	0	.0863309	.0647482
22	ACOM	2005	2009	3	1	0	0	.2477064	.146789
23	ACOM	2006	2009	3	2	0	0	.2384615	.1307692
24	ACOM	2007	2009	3	3	0	0	.2570281	.124498
25	ACOM	2008	2009	3	4	0	0	.2570281	.124498
26	ACOM	2009	2009	3	5	1	0	.1410256	.1025641
27	ACOM	2010	2009	3	6	1	1	.1100917	.3119266
28	ACOM	2011	2009	3	7	1	2	.1257862	.2767296
29	ACOM	2012	2009	3	8	1	3	.0769231	.0897436
30	ACOM	2013	2009	3	9	1	4	.057554	.0863309

- Other findings
 - Projected image changes more quickly when founders and VCs retain less ownership following IPO
 - Changes to projected image at IPO are positively associated with analyst recommendations
 - Retaining pre-IPO rhetoric is negatively associated with market-to-book ratio

Topics over time?

- Things to consider
 - Why does it matter?
 - Be able to argue why rhetorical change matters; link to important outcomes
 - Text data! A lot of longitudinal text data
 - Sources: Company reports (e.g., annual reports, shareholder letters), social media, Glassdoor, Internet Archive
 - Garbage in, garbage out
 - Some basic R programming skills
 - Familiarize yourself with general procedures and topic modeling libraries
 - Explore different timeframes and topic configurations
 - Topic modeling and growth models are sensitive to cut-offs
 - Try different combinations to better understand rates of change