# Zero-Revelation RegTech: Detecting Risk through Corporate Emails

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Joint work with: Sanjiv Das (SCU) and Bhushan Kothari (Google Inc.)

# **Big Picture**

- Financials are often delayed indicators of corporate quality
- Internal discussion (e.g., emails) may be used as an early warning system
- An automated platform that parses emails and produces summary statistics would be highly valuable, since...
  - It can analyze vast quantities of textual not amenable to human processing
  - It does not require revelation of individual email content explicitly to monitors/regulators

# **Our Purpose**

- Our purpose is to explore the predictive power of information conveyed by employee emails
- Specifically, we are interested in:
  - The sentiment conveyed by email content
  - The information conveyed by structural characteristics, such as email volume or length
  - Other non-verbal indicators of potential trouble (e.g., shifting email network patterns)

# **Preview of Results**

- We find that the net sentiment conveyed by Enron employee email content is a significant predictor of stock-return performance
- Interestingly, email length was a stronger predictor of subsequent price declines than the net sentiment conveyed by the message body itself.

 We also identify other potential indicators/predictors of escalating risk or malfeasance.

# Data

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- Initial Sample:
  - Approximately 500,000 emails

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- Caveats / Redactions
  - The Enron corpus has been scrubbed over time for legal reasons and to honor requests from affected employees.

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  - Ex(1): user "fastow-a" is notably missing
  - Ex(2): Email chatter surrounding Mr. Skilling's sudden resignation on 8/14/2001 has been expunged.
  - Overall, details regarding exclusion criteria have not been made public, and our analyses should be viewed as exploratory and prescriptive

Data

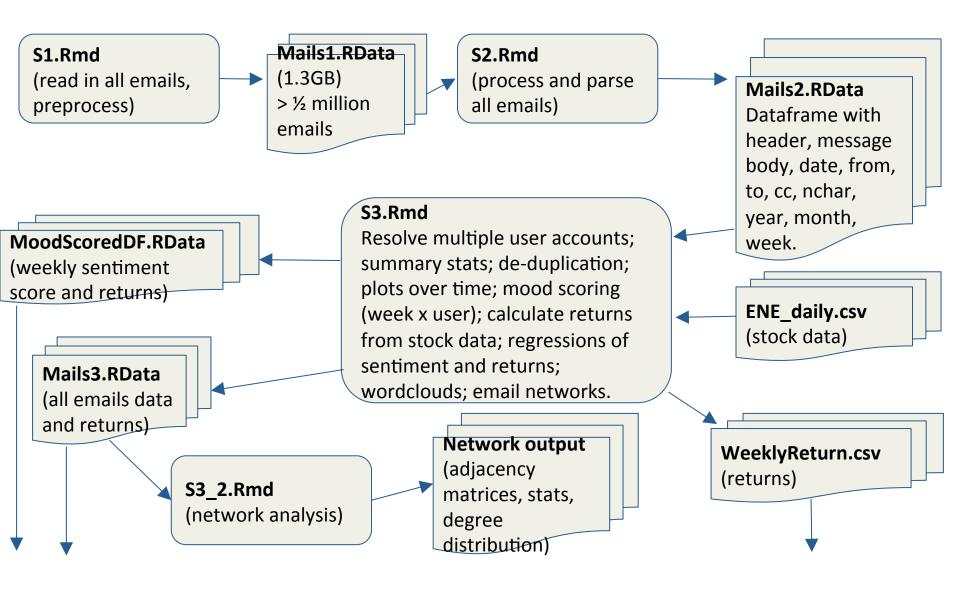
#### Curing the Data

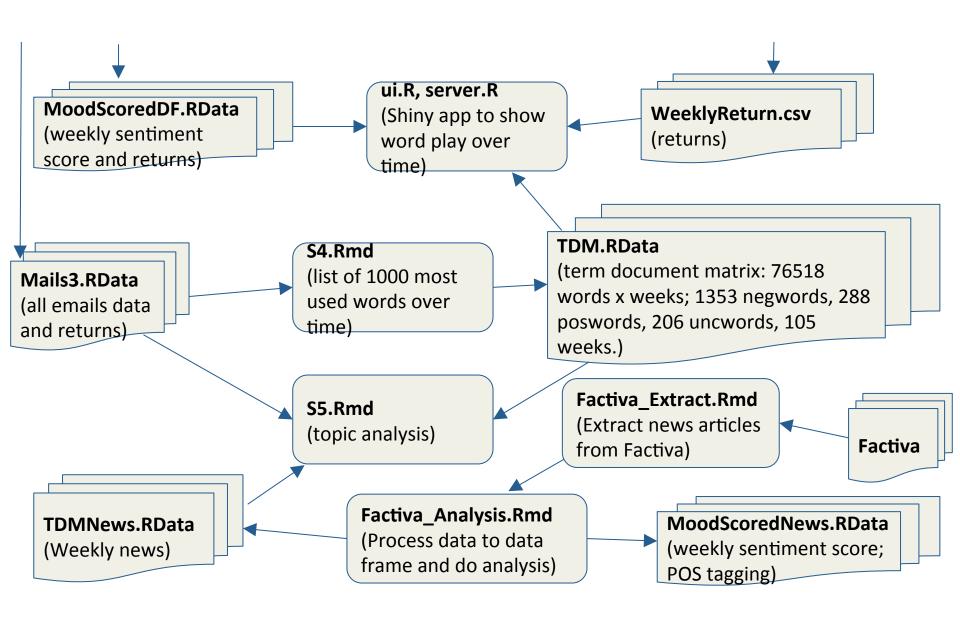
- We focus on "sent" emails (rather than all emails) in order to...
  - Analyze content specifically written by Enron employees
  - Avoid processing the same content more than once
  - i.e., if user "lay-k" sends an email to "skilling-j"
- Other filters applied to remove noisy (junk) mail:
  - Emails greater than 3,000 characters in length
  - Emails sent to more than 20 recipients

# **Our Final Sample**

- Overall, we obtain...
  - The Enron email corpus from the Carnegie Mellon CS site
  - Stock price and stock return information from CRSP
  - News articles from Factiva PR Newswire
  - Sentiment word dictionaries from the Harvard Inquirer and the Loughran and McDonald sentiment word lists
- Final Sample:
  - 144 distinct employees
  - 113,266 sent emails
  - January 2000 through December 2001

# Enron Code Pipeline





# Table 1. Summary Statistics of Sent Mail

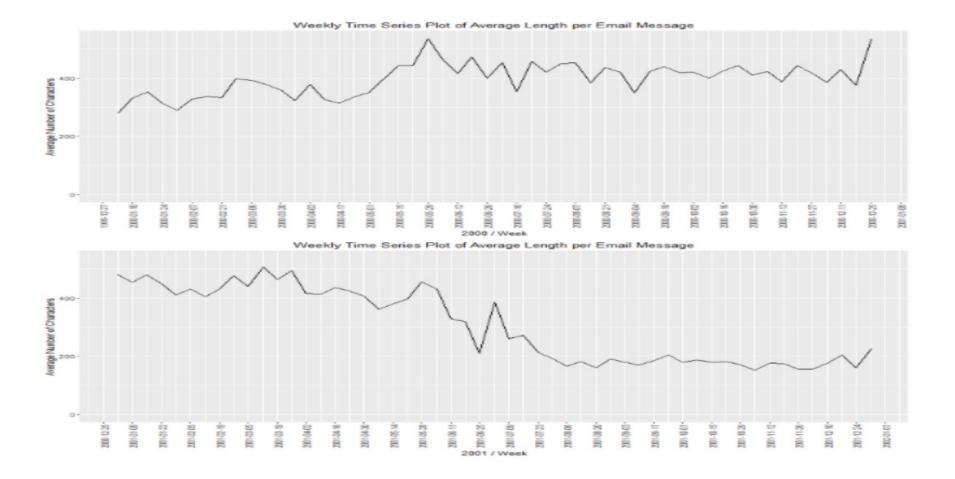
	-					
Panel A. Characteristics by Employee $(N = 144)$						
Variable	Mean	Min	P25	Median	P75	Max
Emails per Person	787	2	105	349	891	8,793
Average "Connectedness"	1.62	1	1.21	1.44	1.76	4.47
Average Length per Person	279.92	19.15	160.45	227.90	338.07	944.23
Panel B. Email Characteristics $(N = 113, 266)$						
Variable	Mean	Min	P25	Median	P75	Max
Length of Email ( $\#$ of characters)	362	0	46	163	466	2,998
Direct Recipients per Email ("to")	1.44	0	1	1	1	20
Indirect Recipients per Email	0. The average email is 362 characters in 19					
( "cc" $)$	length, with a median of 163 characters					
Total Recipients per Email	1.77	1	1	1	2	20

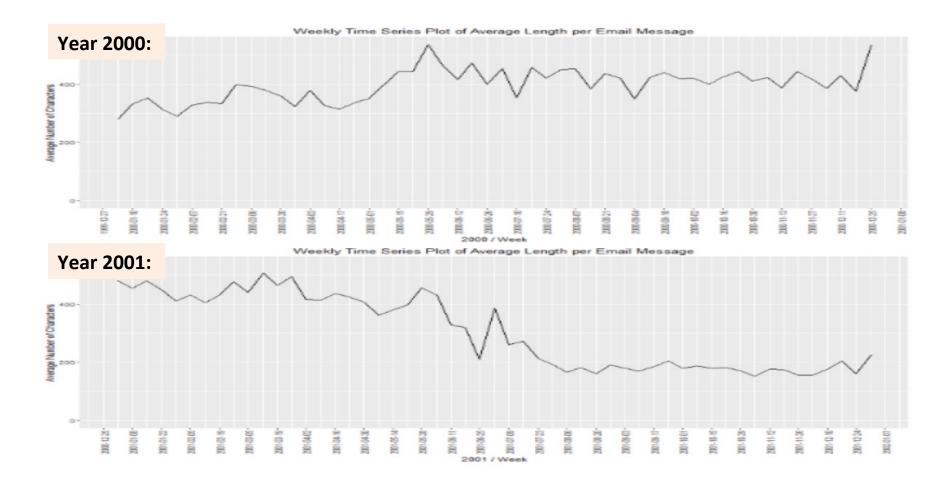
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Indirect Recipients per Email	0.32	0	0	0	0	19
( "cc" $)$						
Total Recipients per Email	1.77	1	1	1	2	20
	with an average of 1.77 recipients per sent mail.					

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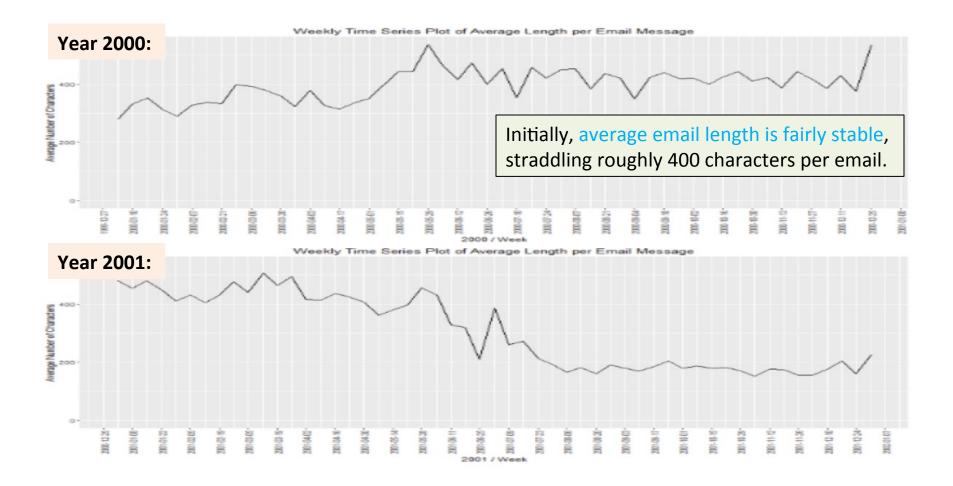
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Indirect Recipients per Email	0.32	Many emails (close to 11%) are simply				
("cc")		forwarded without added text.				
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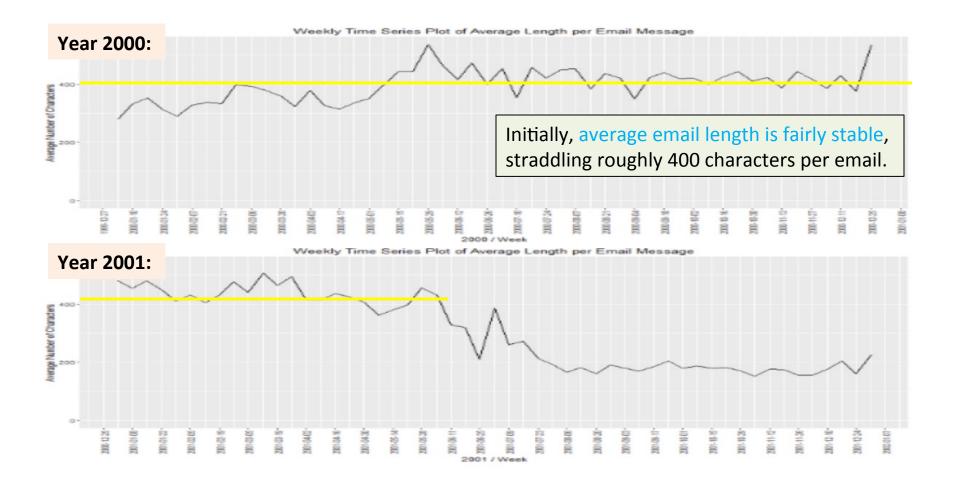
Motivation

Data



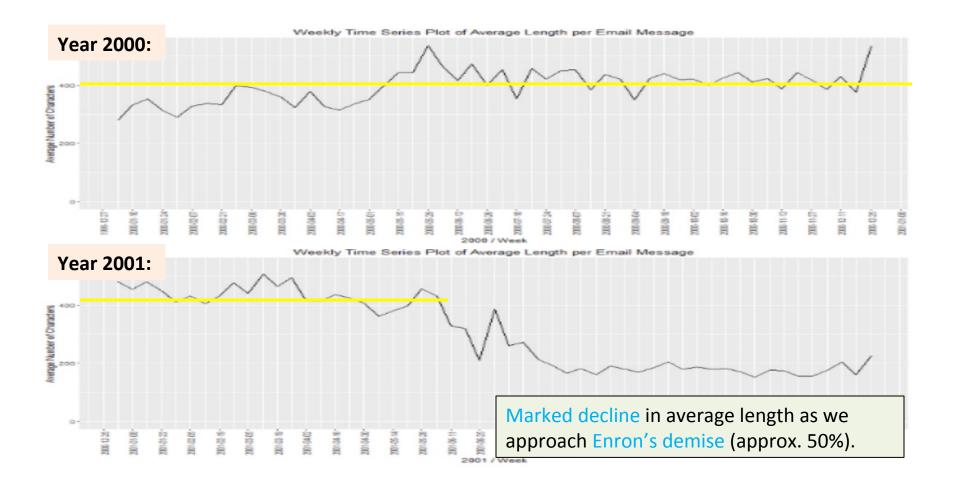
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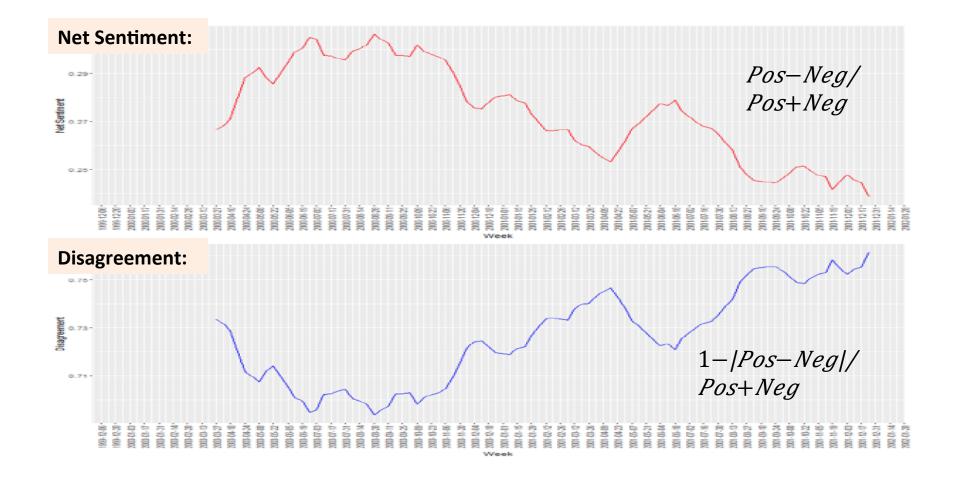
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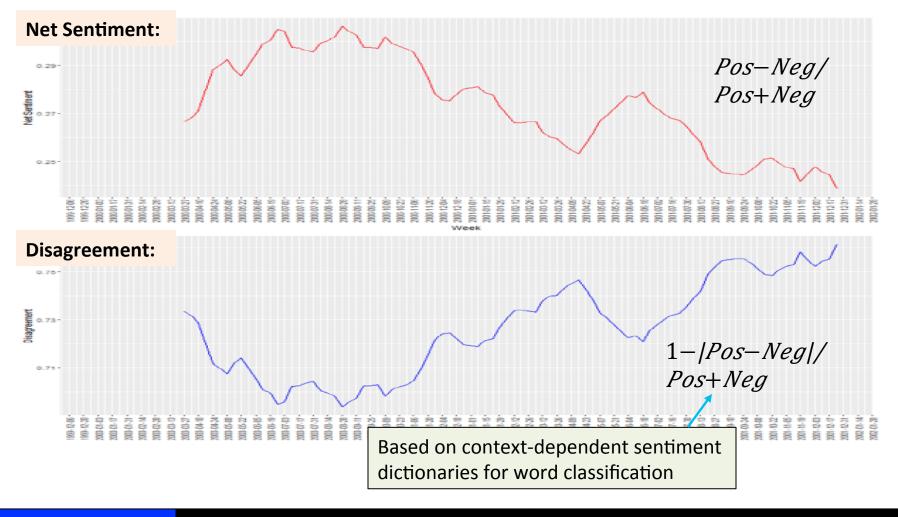
#### Figure 2. Email Sentiment and Disagreement over Time



Motivation

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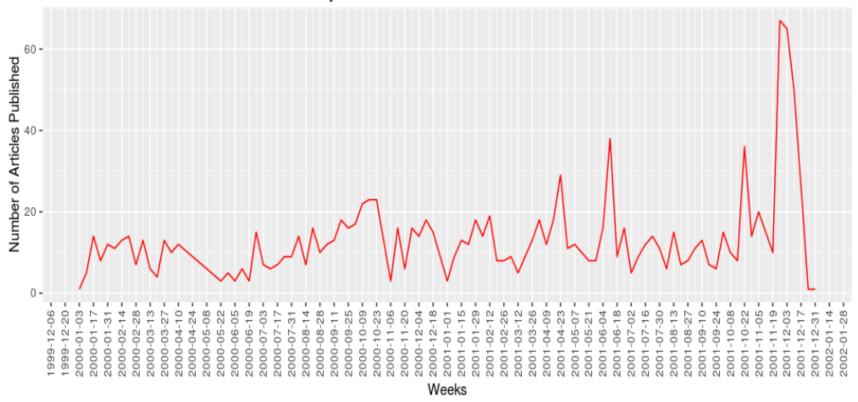
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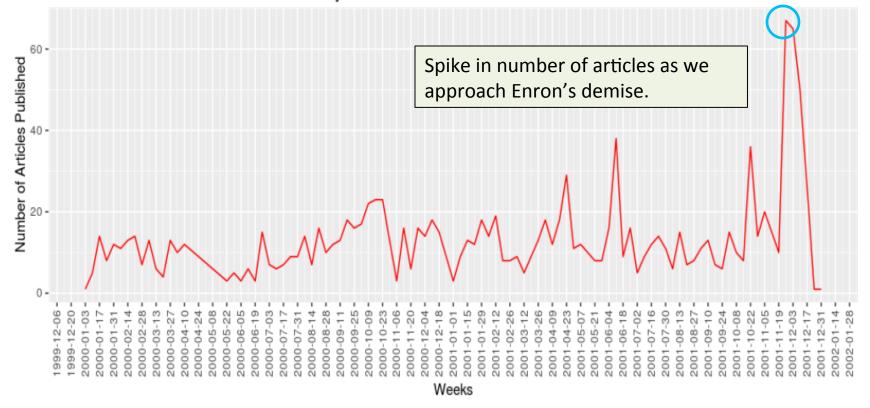
#### Figure 3. Factiva News Coverage over Time

Weekly Time Series Plot of Articles Published

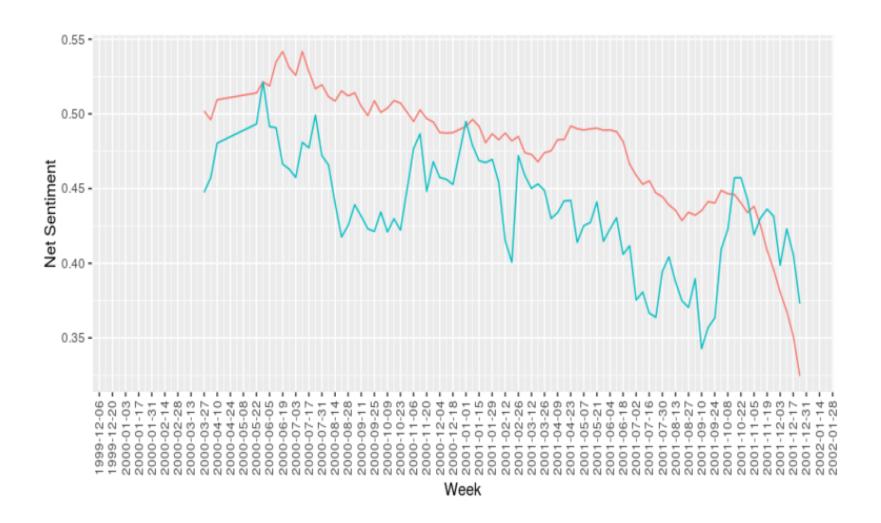


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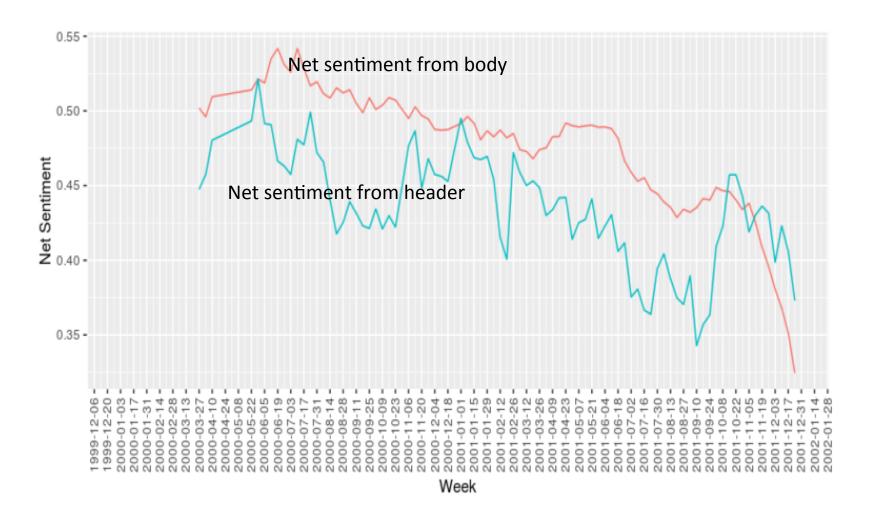


#### Figure 4. Factiva News Sentiment over Time



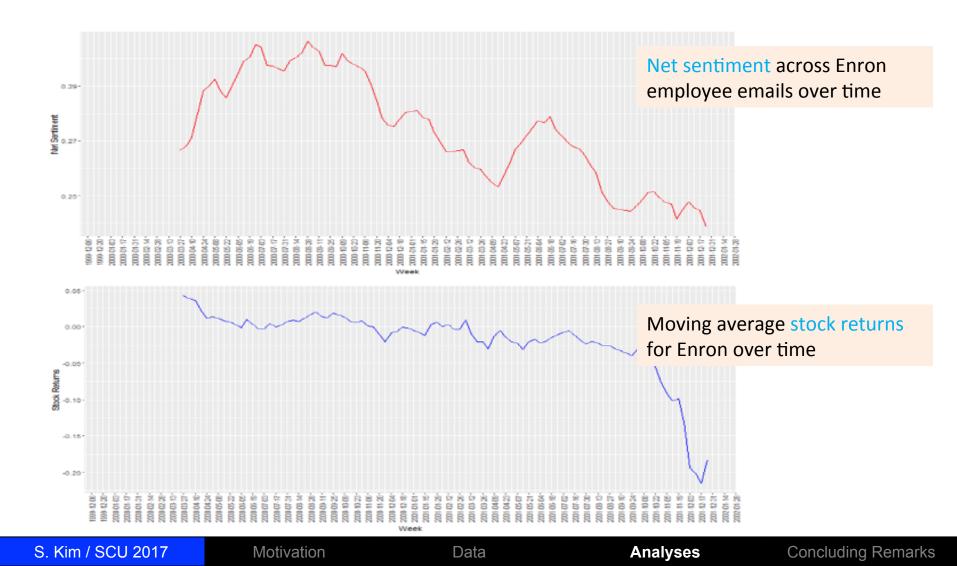
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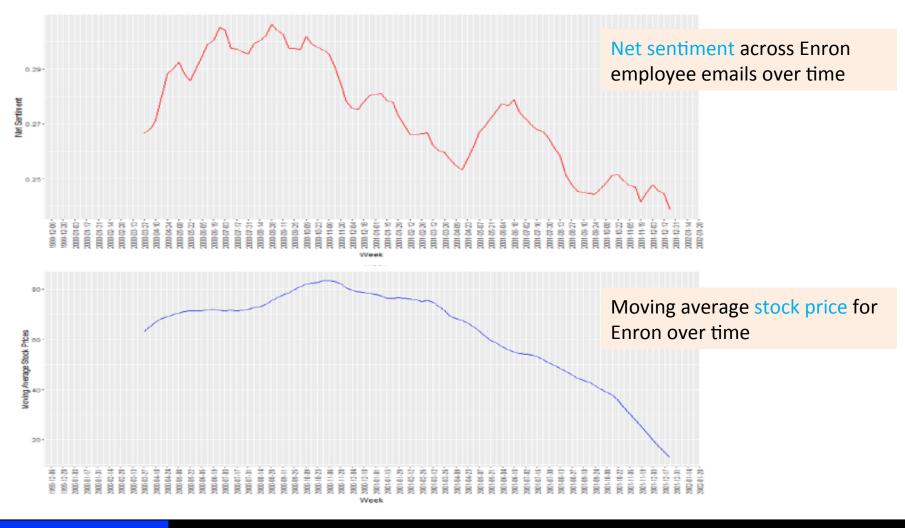


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#### Figure 5. Stock Returns and Net Sentiment over Time

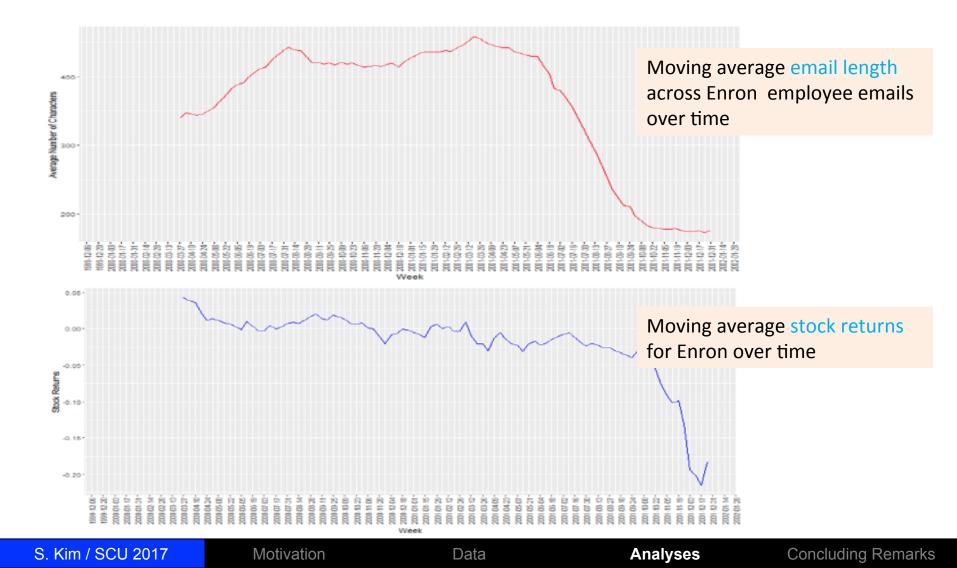


#### Figure 6. Stock Prices and Net Sentiment over Time

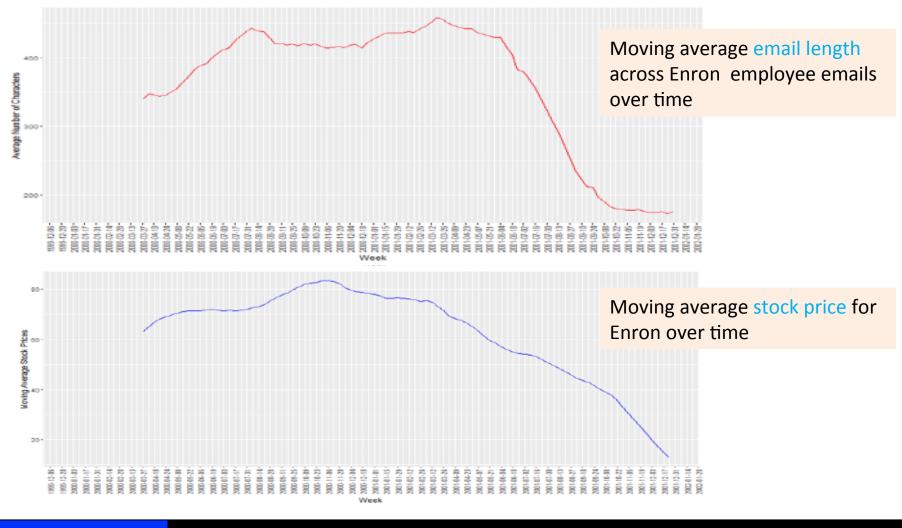


Motivation

#### Figure 7. Stock Returns and Email Length over Time



#### Figure 8. Stock Prices and Email Length over Time

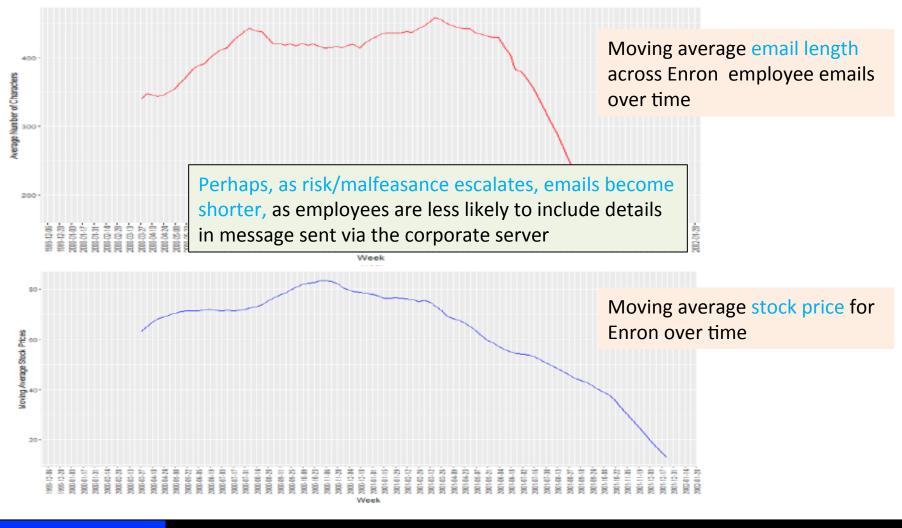


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Motivation

Data

#### Figure 8. Stock Prices and Email Length over Time



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Motivation

Data

## Table 2. Email Content and Stock Returns

Variable	Coef	ficient Esti	mate $(t-stat)$	istic)
	(1)	(2)	(3)	(4)
MA Email Sentiment $_t$	$2.347^{***}$	0.575	$2.330^{***}$	-1.397
	(3.27)	(0.63)	(3.14)	(-1.25)
MA Email Length <sub>t</sub>		0.584***		1.046***
		(2.97)		(4.19)
MA Total $Emails_t$			-0.004	-0.131***
			(-0.10)	(-2.83)
Intercept	-0.680***	-0.406*	-0.671***	0.117
	(-3.45)	(-1.93)	(-3.08)	(0.43)
Adjusted $R^2$	0.10	0.18	0.09	0.24
No. of observations	88	88	88	88

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	S. Kim / SCU 2017	Motivation	Data	Analyses	Concluding Remarks
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S. KIII / SCO 2017 Molivation Data Analyses Concluding Reman	S. Kim / SCU 2017	Motivation	Data	Analyses	Concluding Remark
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Panel B. N	Vews Heade	er Sentiment	and Retu	rns	
MA Header Sentiment $_t$	-0.795	-1.136*	-0.772	-1.210**	-0.893
	(-1.31)	(-1.96)	(-1.34)	(-2.03)	(-1.61)
MA Email Sentiment $_t$		2.628***	0.705	2.566***	-1.254
		(3.30)	(0.66)	(3.18)	(-1.03)
MA Email Length $_t$			0.560**		1.026***
			(2.59)		(3.93)
MA Total Emails $_t$				-0.024	-0.138***
				(-0.59)	(-2.91)
Intercept	0.307	-0.256	-0.096	-0.178	0.485
I	(1.15)	(-0.84)	(0.75)	(-0.54)	(1.39)
Adjusted $R^2$	0.01	0.12	0.18	0.11	0.25
No. of observations	81	81	81	81	81

8. Kim / SCU 2017	Motivation	Data	Analyses	Concluding Remarks
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Panel B.	News Head	er Sentiment	and Retu	rns	
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Kim / SCU 2017 Motiva	tion	Data		Analyses	Concluding Ren

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-			But neither email lengt	•	t when accounting
Intercept	0.307			•	t when accounting 0.485
Intercept	$     \begin{array}{l}       0.307 \\       (1.15)     \end{array} $	for	email lengt	h.	
Intercept Adjusted $R^2$		-0.256	email lengt -0.096	h. -0.178	0.485
	(1.15)	-0.256 (-0.84)	email lengt -0.096 (0.75)	-0.178 (-0.54)	0.485 (1.39)

#### Dependent Variable = *Stock Returns*<sub>t</sub>

S

Pan	el A. News Body	Sentiment	and Retu	rns	
MA Body Sentiment $_t$	$1.410^{***}$	$1.501^{**}$	0.657	$1.505^{**}$	-0.827
	(3.95)	(2.49)	(0.87)	(2.48)	(-0.92)
MA Email Sentiment $_t$		-0.245	0.377	-0.284	-1.293
		(-0.19)	(-0.29)	(-0.22)	(-1.02)
MA Email Length $_t$			$0.486^{*}$		1.380***
			(1.81)		(3.34)
MA Total $\textsc{Emails}_t$				-0.009	-0.164***
				(-0.24)	(-2.77)
Intercept	-0.711***	-0.688***	-0.426*	-0.668***	0.399
	(-4.18)	(-3.27)	(-1.69)	(-2.94)	(1.04)
Adjusted $\mathbb{R}^2$	0.15	0.14	0.17	0.13	0.23
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6. Kim / SCU 2017 Mo	tivation	Data		Analyses	Conclu

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MA Email Sentiment $_t$		-0.245 (-0.19)	$0.377 \\ (-0.29)$	-0.284 (-0.22)	-1.293 (-1.02)		
MA Email Length $_t$		On the other hand, email content contains less information than content from the news					
MA Total Emails $_t$				ue to redactio	ons on the Enron		
Intercept	-0.711***	-0.688***	-0.426*	-0.668***	0.399		
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S. Kim / SCU 2017	lotivation	Data		Analyses	Concluding Rem		

No. of observations	81	81	81	81	81		
Adjusted $\mathbb{R}^2$	0.15	0.14	0.17	0.13	0.23		
	(-4.18)	(-3.27)	(-1.69)	(-2.94)	(1.04)		
Intercept		-0.688***		-0.668***	0.399		
MA Total $\text{Emails}_t$		But, again, neither is significant when accounting for email length.					
			(1.81)	]	(3.34)		
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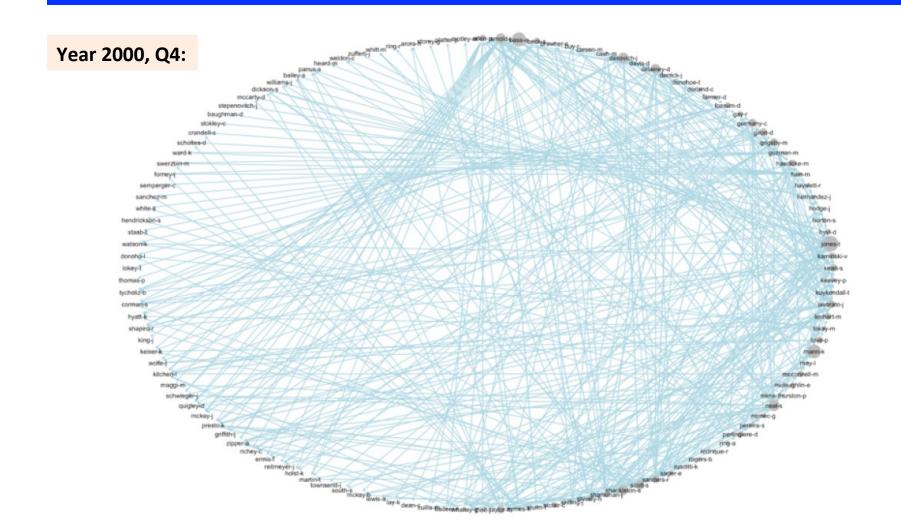
## Summary and Implications

- Thus far, we have shown that the net sentiment conveyed by employee sent mails is a significant predictor of stock-return performance
- Interestingly, email length was a stronger predictor of subsequent price declines than the net sentiment conveyed by the message body itself.
- Overall, email content may be controlled or manipulated
  - Thus, we are also (and perhaps even more!) interested in the non-verbal, interaction- or network-based indicators of potential trouble.

## **Additional Explorations**

# Other dimensions ripe for investigation....

## Figure 11. Email Networks

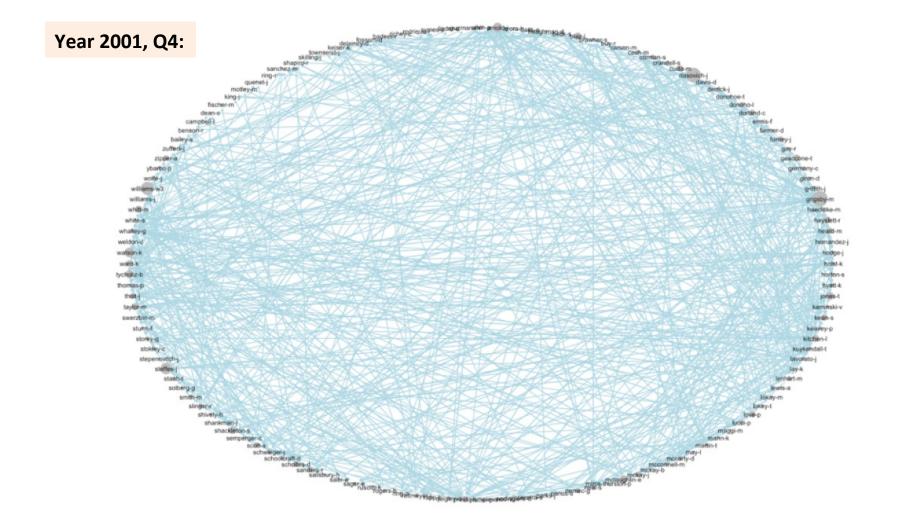


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Motivation

Data

### Figure 11. Email Networks



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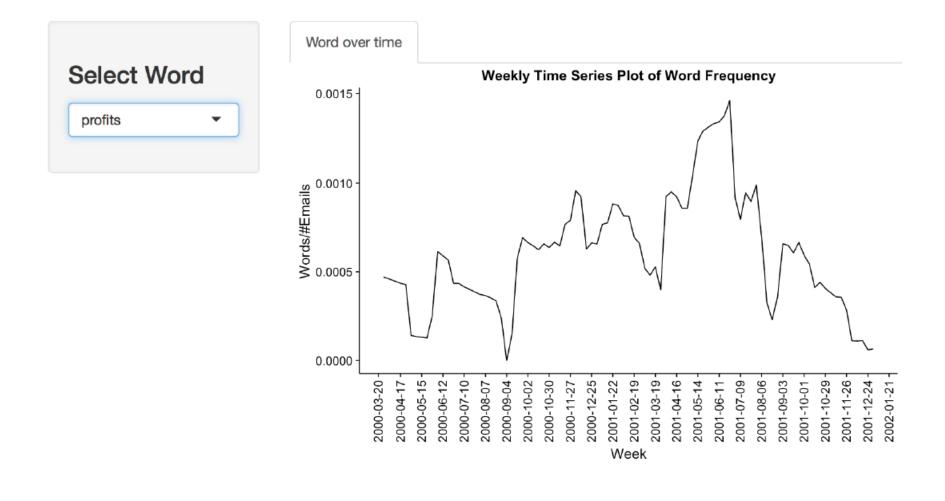
Motivation

Data

Analyses

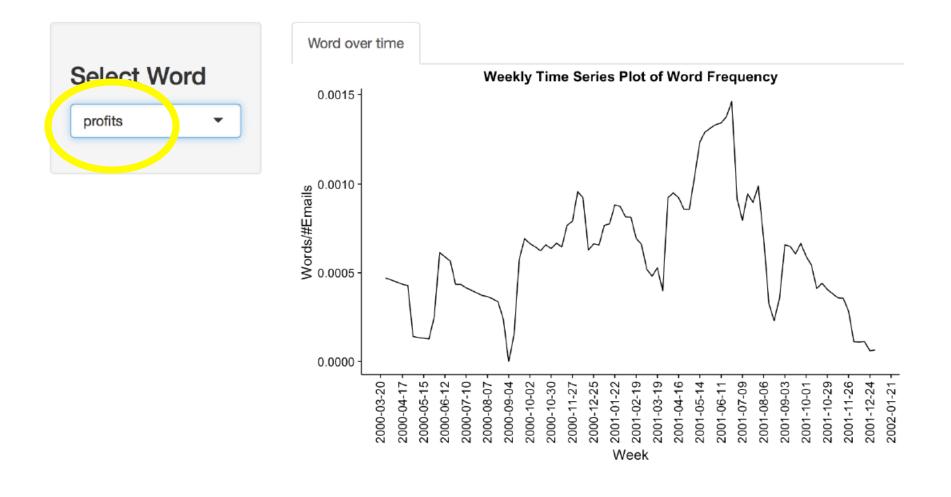
**Concluding Remarks** 

#### Figure 13. Vocabulary Trends

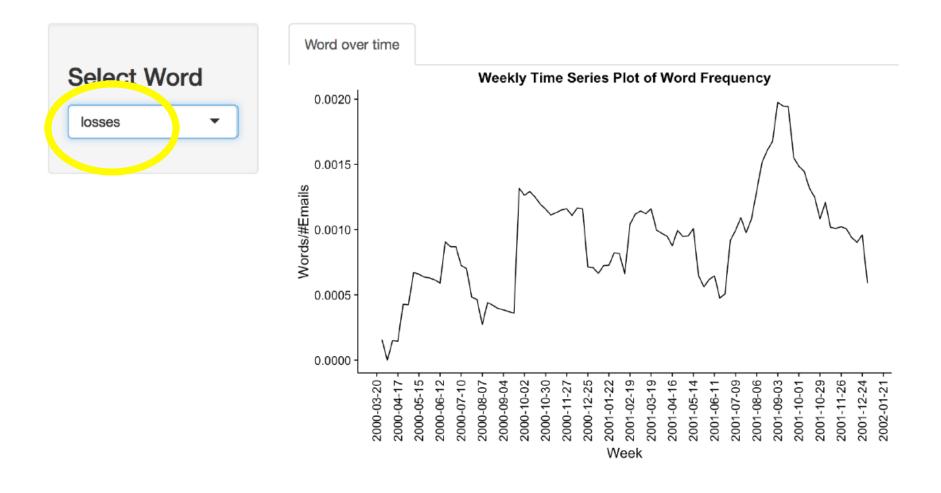


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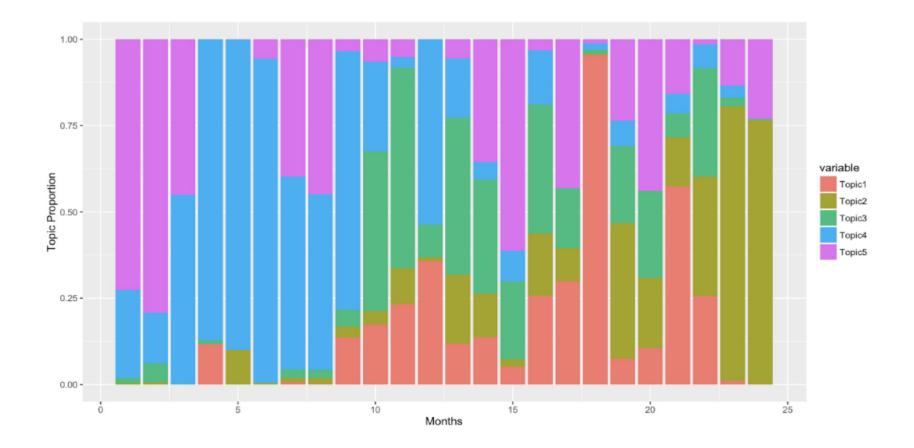


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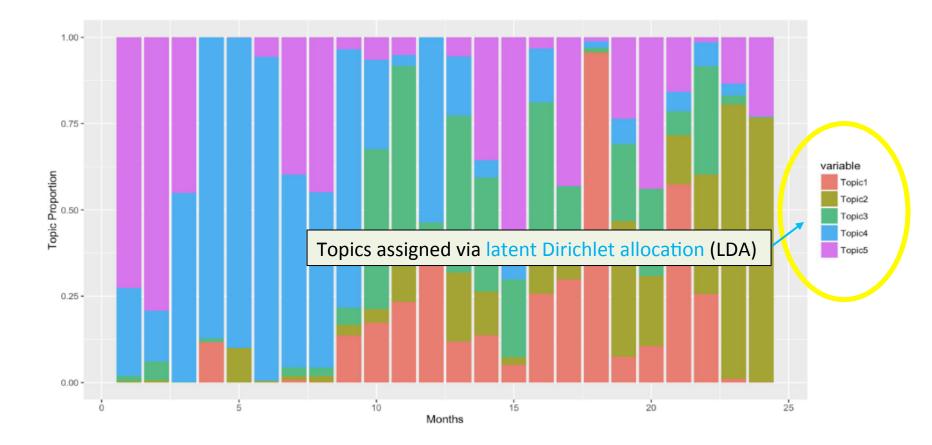
## Figure 14. Topic Analysis over Time



Motivation

Data

### Figure 14. Topic Analysis over Time



Motivation

Data

## **Concluding Remarks**

- We introduce an automated platform to parse corporate email content, and we find that the net sentiment conveyed by employee sent mails is a timely indicator of stock-return performance.
- Non-verbal indicators, such as email length and network structure, are particularly promising avenues to explore.
- Overall, we suggest the promise of a regulatory technology (RegTech) approach by which to systematically parse email content and network structure to detect indicators of risk or malfeasance on an ongoing and more timely basis.