# 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

- Initial Sample:
  - Approximately 500,000 emails

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- Ex(1): user "fastow-a" is notably missing
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- Overall, details regarding exclusion criteria have not been made public, and our analyses should be viewed as exploratory and prescriptive

## **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

# 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	$1.\overline{77}$	1	1	1	2	20		

Table 1. Summary Statistics of Sent Mail

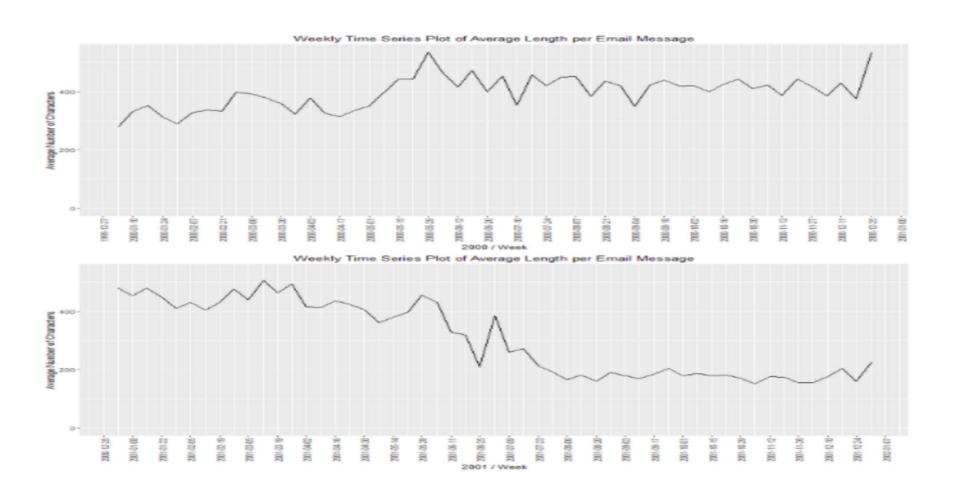
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("cc")								
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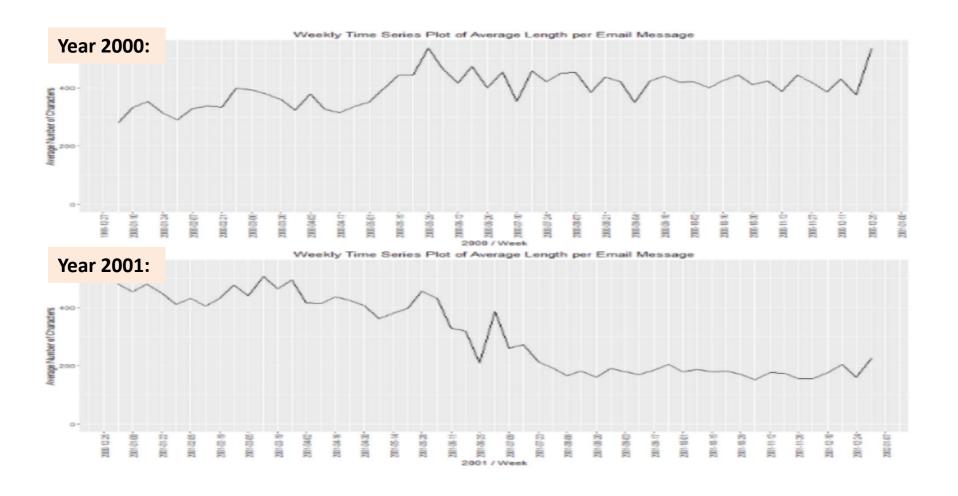
... with an average of 1.77 recipients per sent mail.

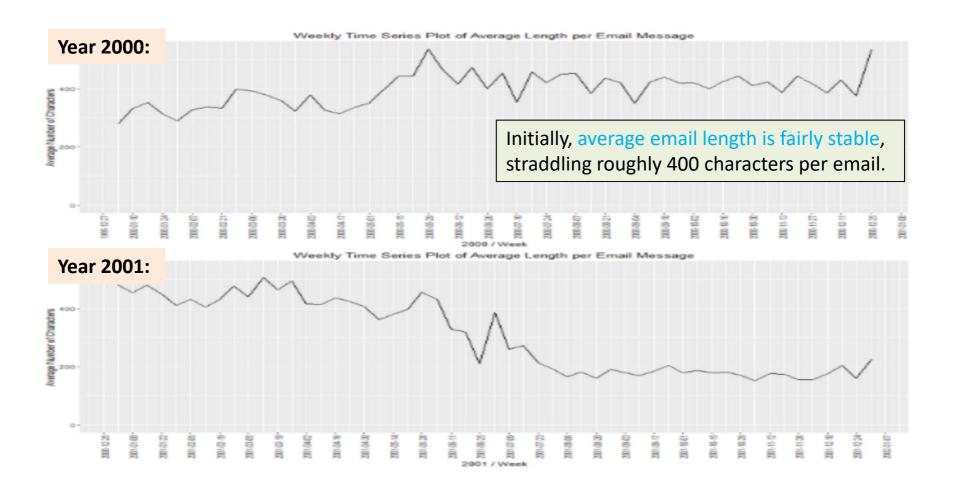
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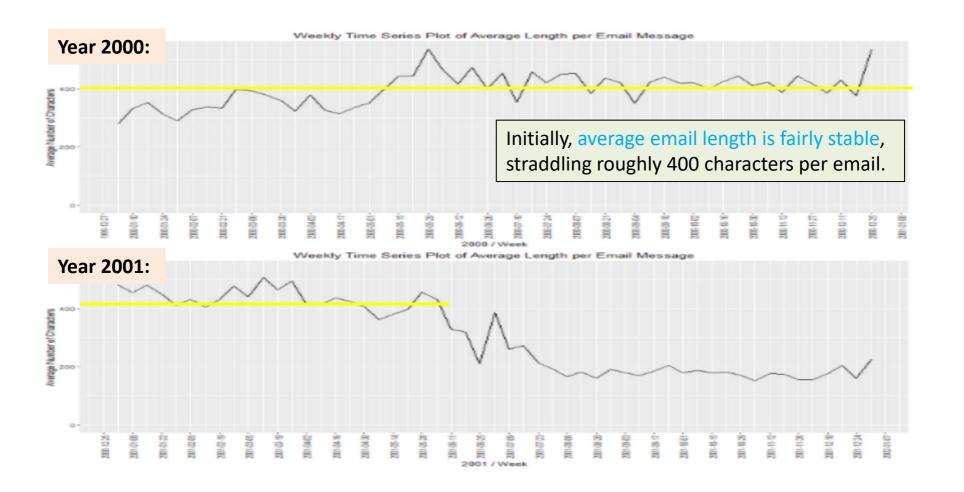
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Indirect Recipients per Email	0.32	Many emails (close to 11%) are simply						
("cc")		forwarded without added text.						
Total Recipients per Email	1.77	1	1	1	2	20		

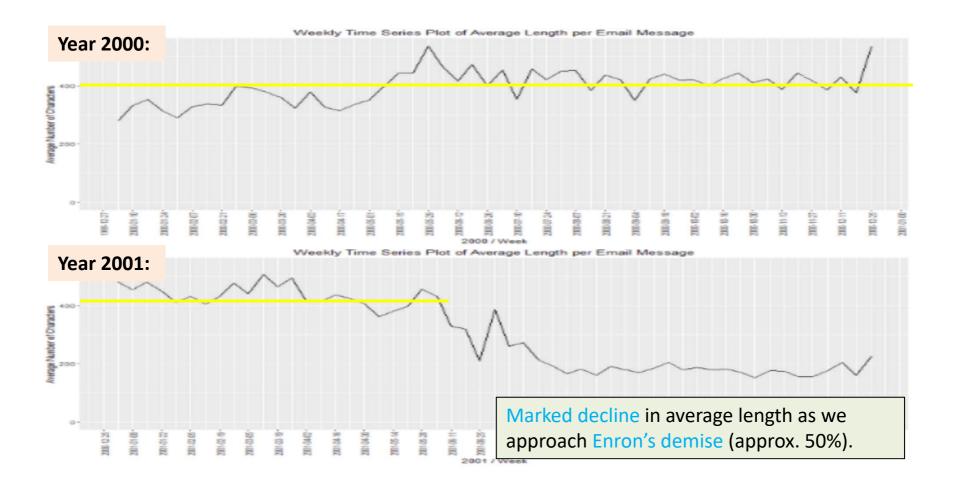
**Analyses** 



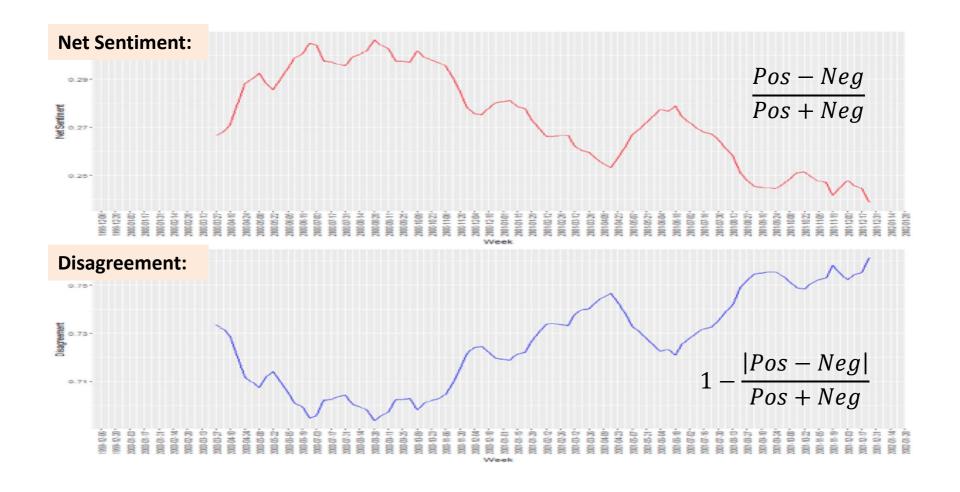




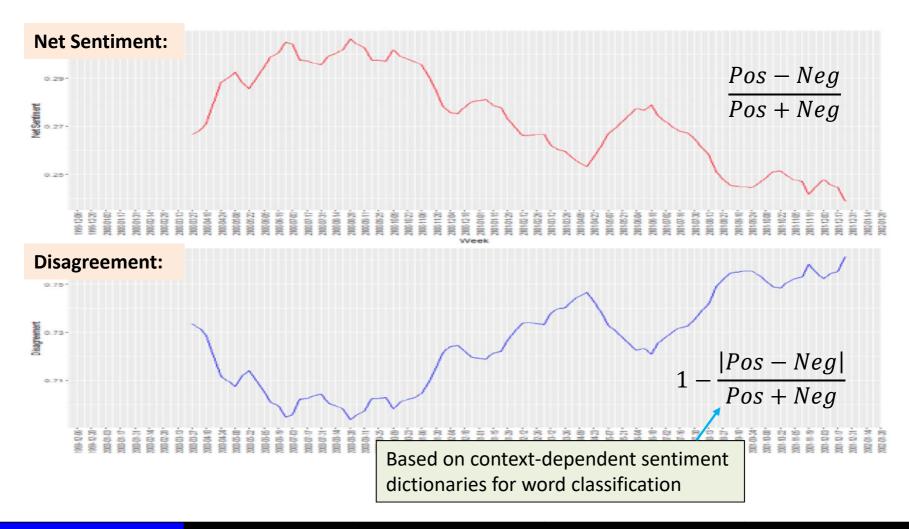




# Figure 2. Email Sentiment and Disagreement over Time

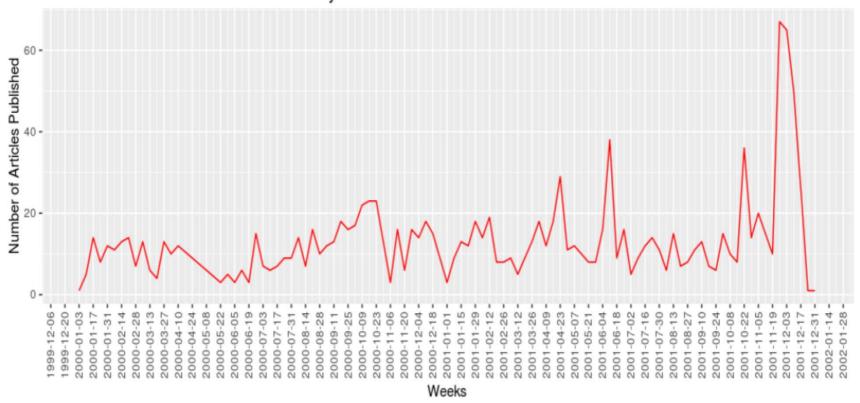


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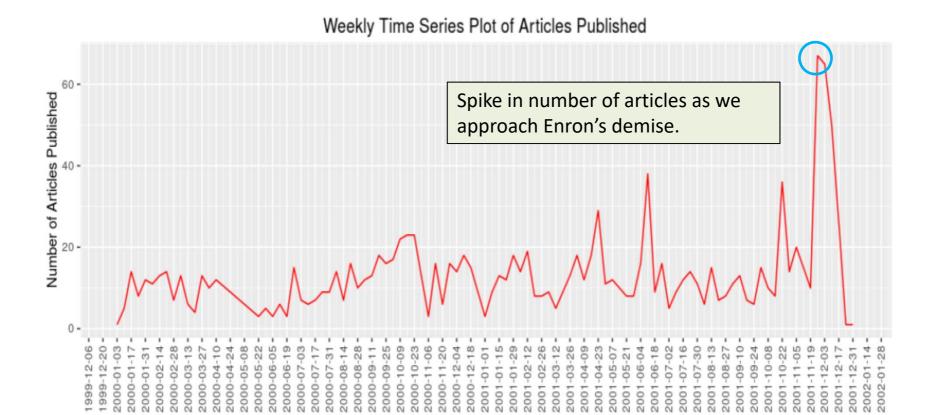


## Figure 3. Factiva News Coverage over Time





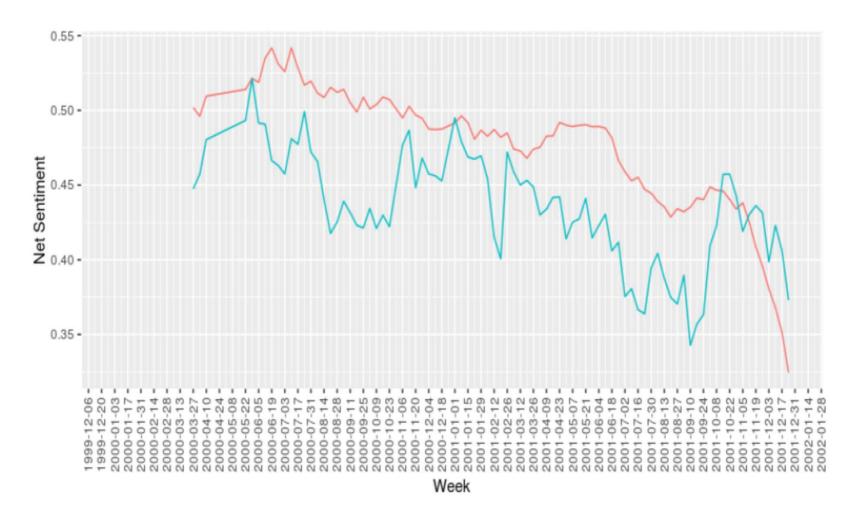
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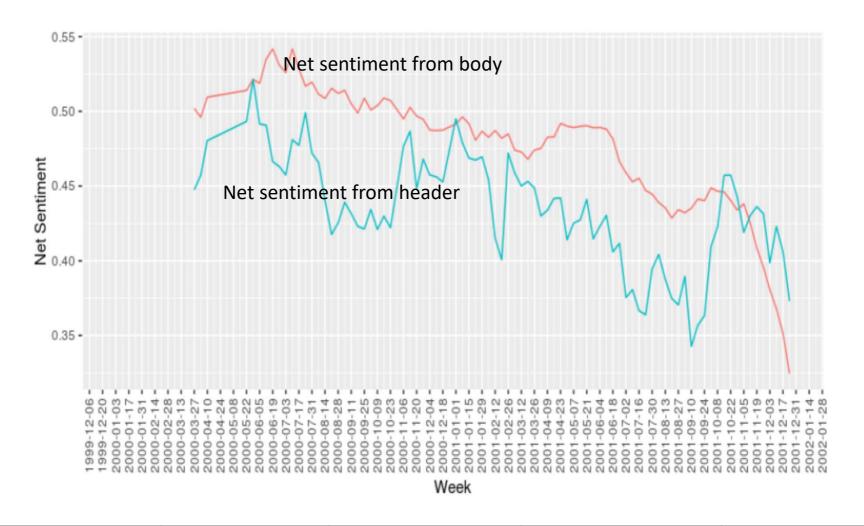
S. Kim / SCU 2017 Motivation Data Analyses Concluding Remarks

Weeks

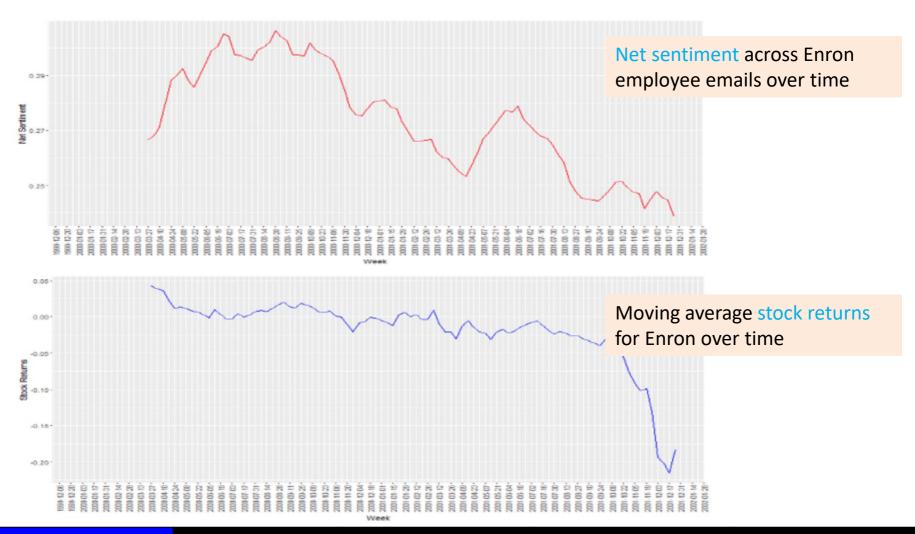
## Figure 4. Factiva News Sentiment over Time



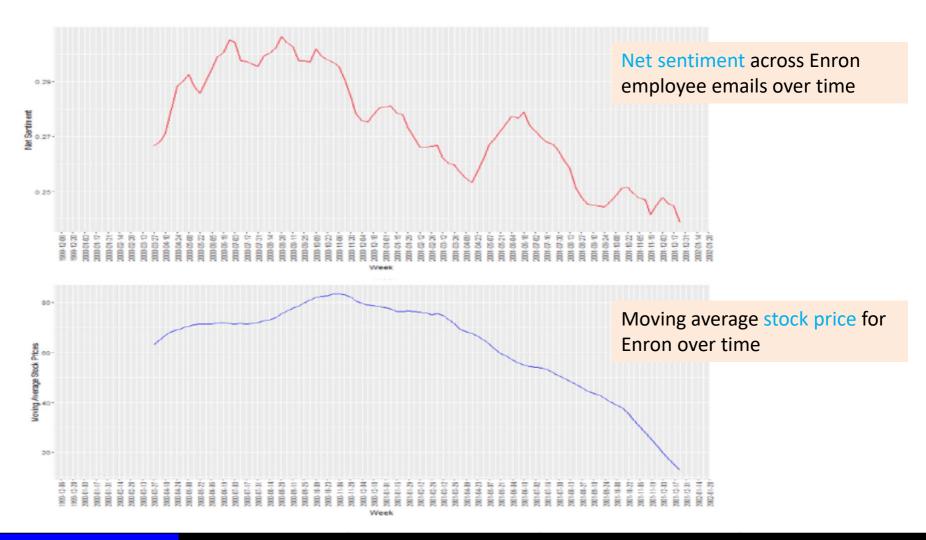
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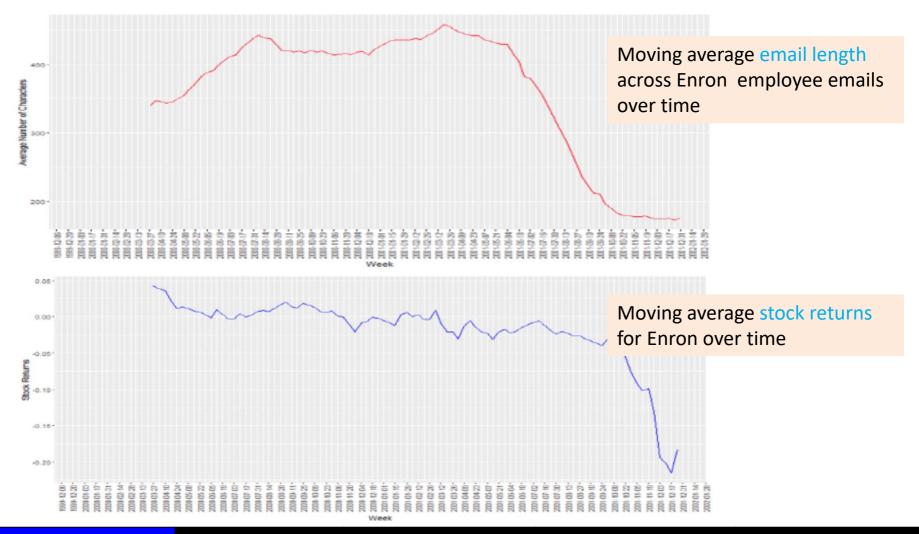
## Figure 5. Stock Returns and Net Sentiment over Time



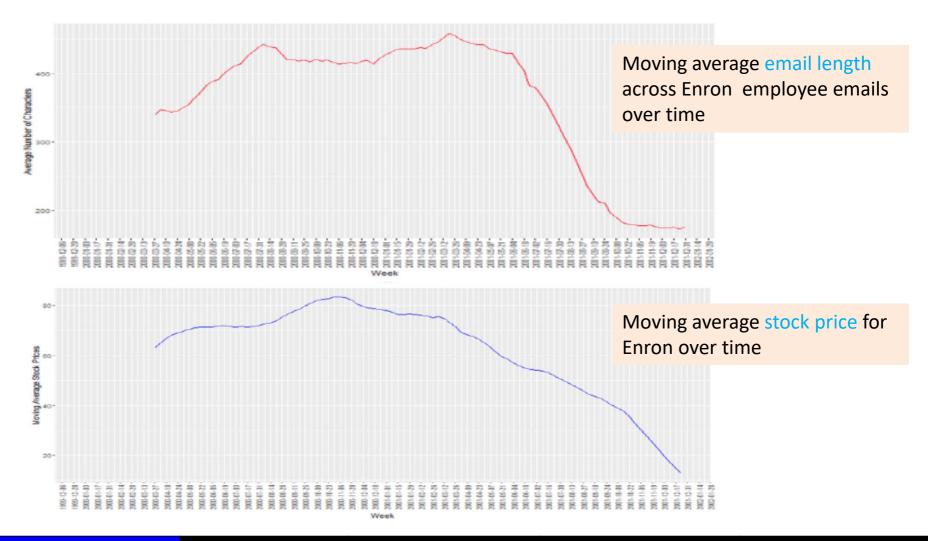
# Figure 6. Stock Prices and Net Sentiment over Time



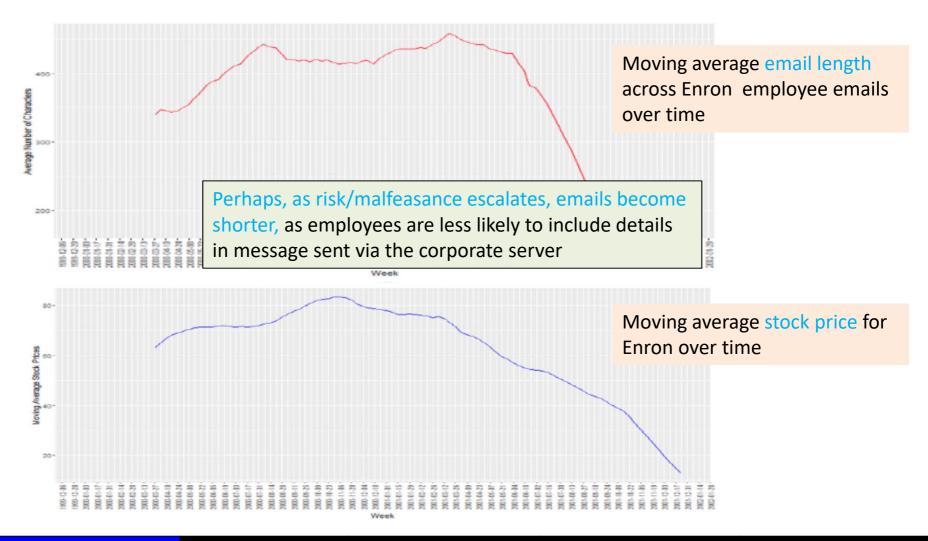
# Figure 7. Stock Returns and Email Length over Time



# Figure 8. Stock Prices and Email Length over Time



# Figure 8. Stock Prices and Email Length over Time



## 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)
$\texttt{MA Email Length}_t$		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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	asso	ociated with	a 4.5% declin	e in stock retu	rns
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Overall, 20-character decline in moving average email length is associated with a 1.17% decline in stock returns.

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Panel B.	News 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
	(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

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Panel A	. News Body	Sentiment	and Retui	ns	
MA Body Sentiment <sub>t</sub>	1.410***	1.501**	0.657	1.505**	-0.827
	(3.95)	(2.49)	(0.87)	(2.48)	(-0.92)
$\texttt{MA Email Sentiment}_t$		-0.245	0.377	-0.284	-1.293
		(-0.19)	(-0.29)	(-0.22)	(-1.02)
$\texttt{MA Email Length}_t$			0.486*		1.380***
			(1.81)		(3.34)
MA Total 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)
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MA Email Langth					
$\texttt{MA Email Length}_t$		On th	e other ha	nd, email cor	ntent contains l
		inforr	mation tha	n content fro	m the news bo
MA Total Emails $_t$				0.000	0.10.1444
THE TOTAL EMBILIS $_t$		(could	d this be di	ue to redaction	ons on the Enro
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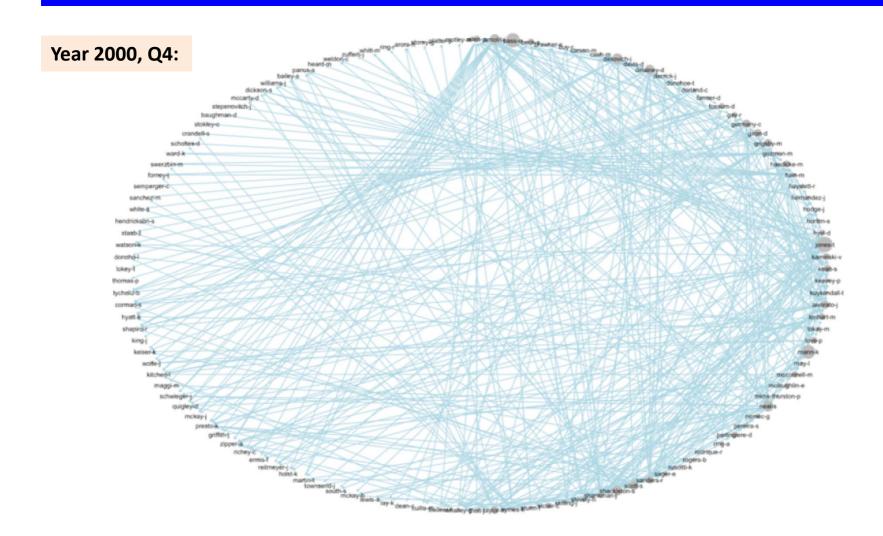
## **Summary and Implications**

- 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.

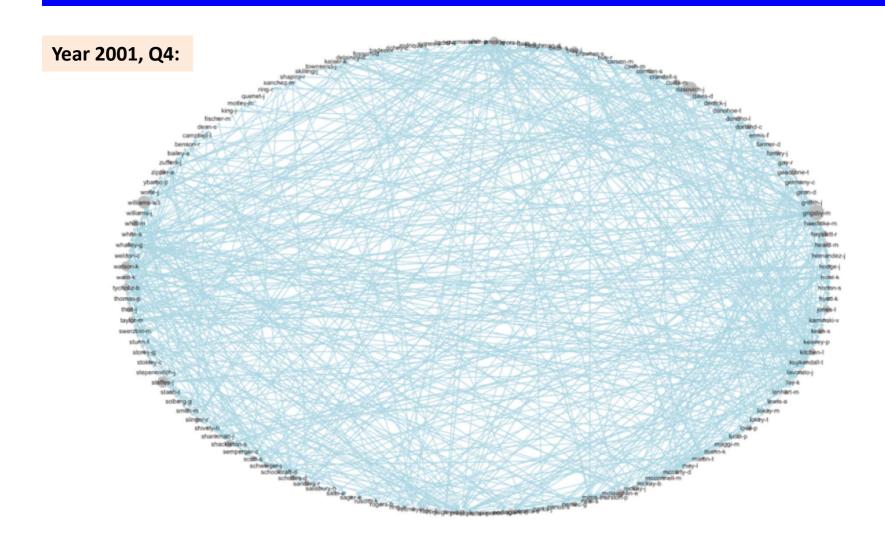
# Additional Explorations

Other dimensions ripe for investigation....

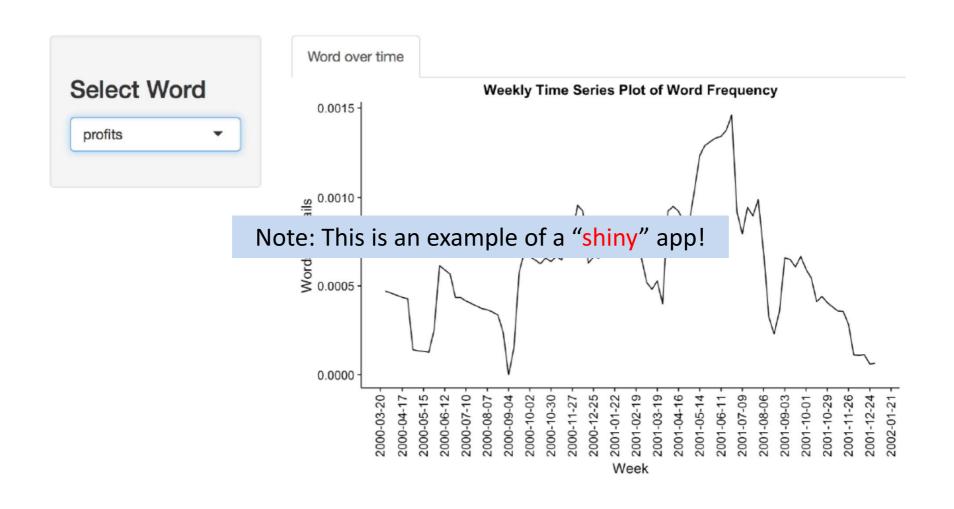
# Figure 11. Email Networks



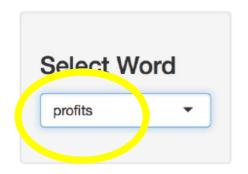
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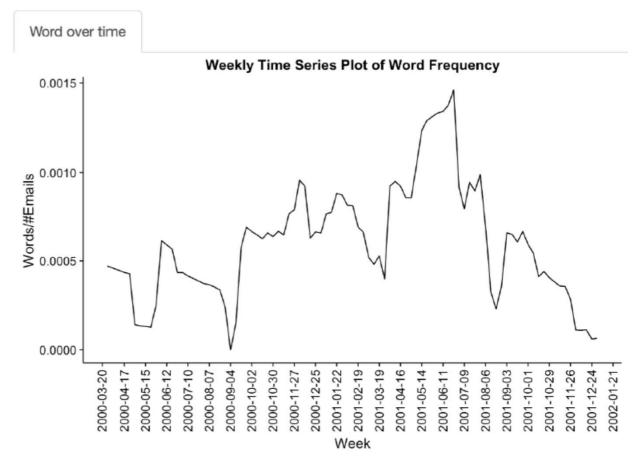


## Figure 13. Vocabulary Trends

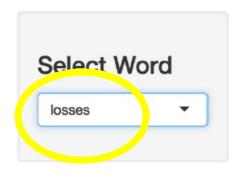


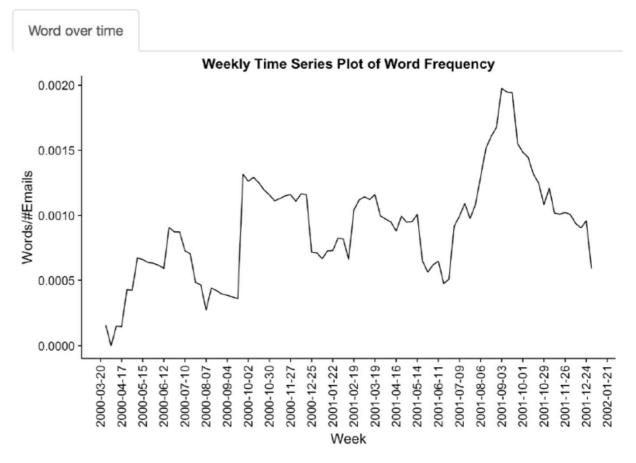
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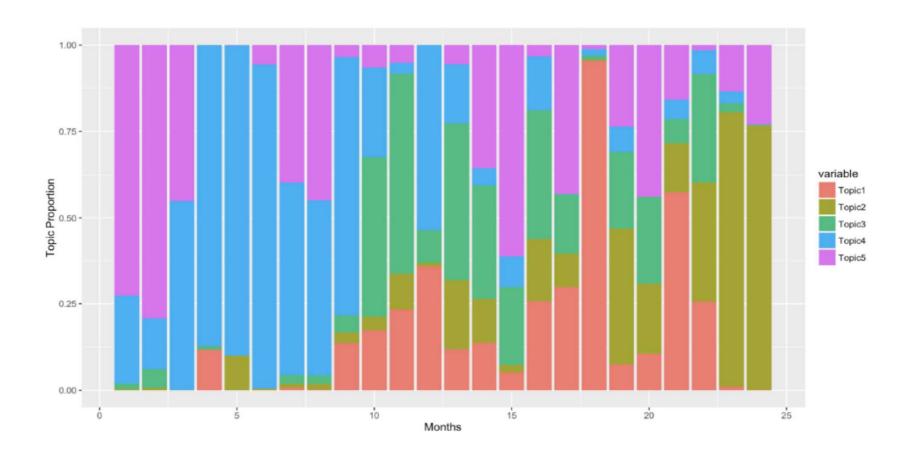


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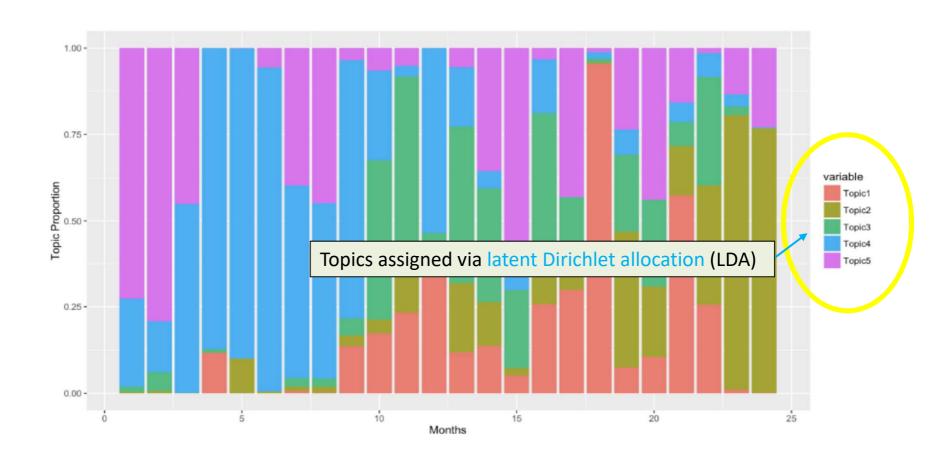




# Figure 14. Topic Analysis over Time



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## **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.

Thank you.