## The Modern Gatekeepers in Mass Media

Patrick Kraft Yanna Krupnikov Kerri Milita John Barry Ryan

74th MPSA Annual Conference, Chicago, II

April 8<sup>th</sup>, 2016







1 / 22

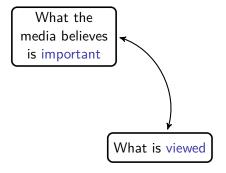
### Introduction

How does social media change the agenda-setting role of traditional news media?

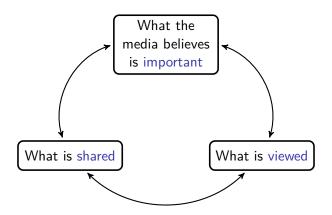


2 / 22

Introduction 0000



0000



Introduction

- Scraped articles from New York Times website between February 18 and April 28, 2015.
- ► Article categories: printed front page, digital front page (top, bottom & opinion section), most viewed, most emailed, most shared on Facebook, most shared on Twitter
- Extracted information: title, author, date, time, URL, full text
- ▶ Total number of unique articles: 5504



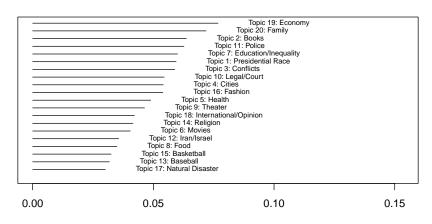
April 8<sup>th</sup>, 2016

Introduction

- ► Summarize content using structural topic model (Roberts et al., 2014; Roberts, Stewart, and Tingley, 2014)
  - ▶ Each document is modeled as a mixture of *K* topics
  - Prevalence of topics in documents can be influenced by a set of covariates
  - ▶ (Word use within each topic can be influenced by a set of covariates)
- ► Number of topics: 20

duction Results Conclusion Appendix Reference

### **Topic Proportions**



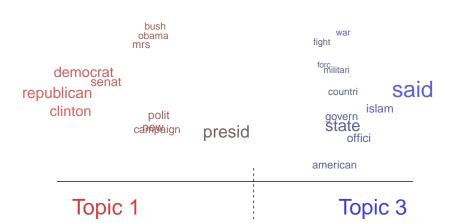
**Expected Topic Proportions** 



 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 0 000000
 000000
 000000

### Example for Topic Perspectives

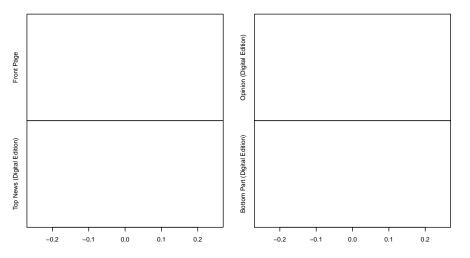




 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 00 ● 00000
 00
 000000

# Changes in Topic Distributions (Newspaper Section)

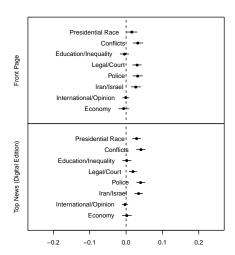


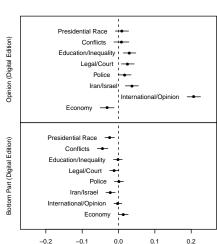


 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 00●0000
 00
 000000

### Changes in Topic Distributions (Newspaper Section)



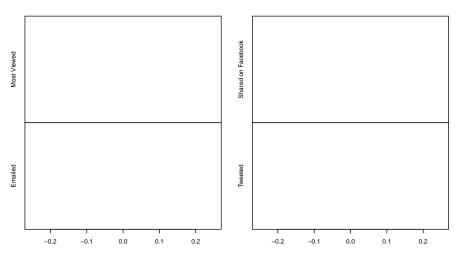




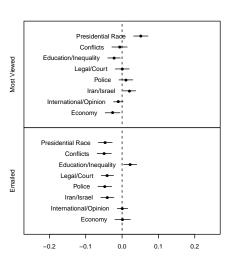
 Introduction
 Results
 Conclusion
 Appendix
 References

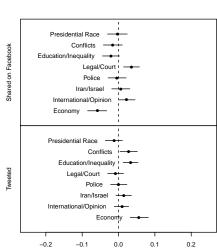
 0000
 000
 00
 000000

# Changes in Topic Distributions (Shared/Viewed)

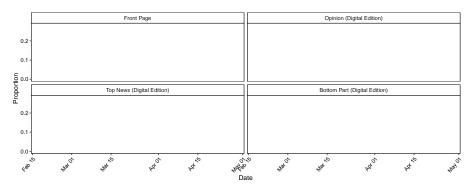


## Changes in Topic Distributions (Shared/Viewed)



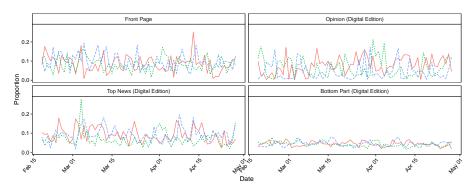


## Topic Proportions over Time (Newspaper Section)



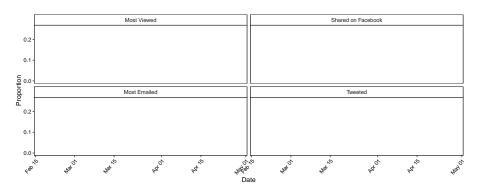
opic - Presidential Race - Legal/Court - Police

## Topic Proportions over Time (Newspaper Section)



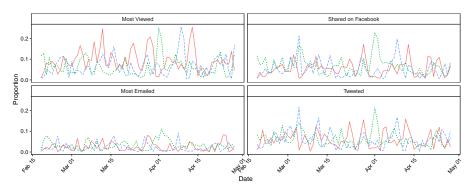
Topic - Presidential Race - Legal/Court - Police

## Topic Proportions over Time (Shared/Viewed)



Topic - Presidential Race - Legal/Court - Police

## Topic Proportions over Time (Shared/Viewed)

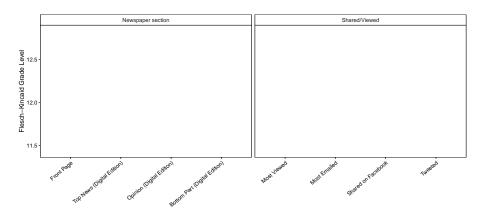


Topic - Presidential Race - Legal/Court - Police

 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 000000 ●0
 00
 000000

### Article Readability by Category

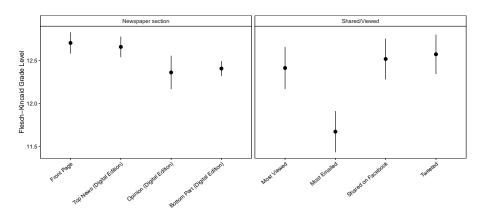




 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 000000 ●0
 00
 000000

## Article Readability by Category

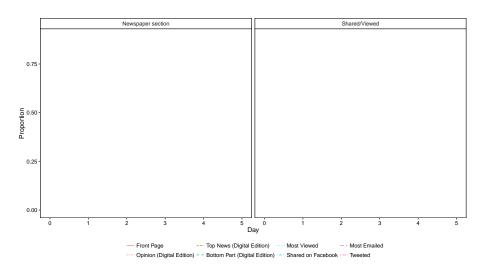




 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 0000000
 00
 000000

### What stays in the loop?

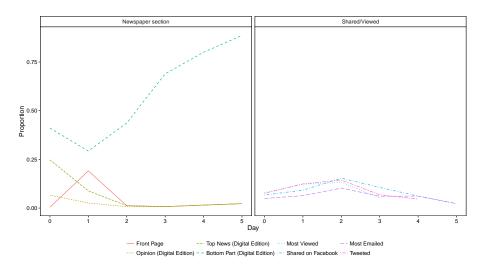




 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 0000000
 00
 000000

### What stays in the loop?





ntroduction Results **Conclusion** Appendix References

### Conclusion

- ► There are differences in types of articles that are shared across media, as well as articles that the New York Times places on its front page
  - ▶ Disagreement on importance of certain news stories
- ▶ Limitations:
  - Reader of the New York Times may be different than readers of other news sources.
  - The editorial team of the New York times may be less influenced by social media than other news outlets.
- Subsequent analyses:
  - Closer analysis of temporal development of articles, track how long articles remain popular in a given cycle
  - Matching individual articles to broader news stories
  - ► Take into account additional meta information (length etc.)



Thank you very much for your attention!

Comments, questions? patrick.kraft@stonybrook.edu



### Topic Words I

# Topic 1:

republican, clinton, democrat, senat, presid, said, new, mrs, polit, campaign, obama, bush, candid, parti, presidenti, will, hous, voter, elect, state

### Topic 2:

book, one, like, work, stori, can, time, write, way, life, think, say, world, make, read, peopl, will, novel, even, seem

### Topic 3:

said, state, islam, offici, govern, presid, american, countri, war, militari, fight, forc, group, attack, year, kill, unit, russian, two, citi

### Topic

new, citi, build, said, park, york, hous, street, year, room, home, open, like, neighborhood, live, hotel, place, art, will, one

#### Table C.

said, patient, studi, health, research, medic, doctor, new, can, use, peopl, drug, year, care, one, found, diseas, hospit, test, scientist

### Topic 6:

film, said, movi, music, year, new, show, will, also, includ, award, watch, like, time, televis, record, song, star, band, play

### Topic 7:

percent, year, school, student, will, tax, peopl, rate, educ, work, job, like, colleg, new, program, worker, mani, pay, polici, univers

### Topic 8:

water, food, like, use, can, wine, restaur, cook, fish, one, make, chef, also, eat, new, chicken, anim, sandwich, will, recip

#### onic Q

show, play, new, music, theater, like, charact, seri, perform, time, open, broadway, star, york, one, critic, direct, actor, product, also

#### nic 10:

said, state, law, court, feder, rule, case, justic, legal, right, conserv, wildlif, vote, bill, republican, member, occup, say, refug, govern

### Topic Words II

### Topic 11:

said, offic, polic, new, report, depart, citi, investig, charg, offici, case, vear, arrest, vork, prison, kill, death, two, last, time

#### Topio 13

iran, said, state, israel, nuclear, deal, unit, presid, obama, american, negoti, will, offici, agreement, netanyahu, iranian, isra, minist, saudi, cuntri

### Topic 13:

said, player, game, team, year, sport, one, play, yanke, last, season, leagu, first run, will basebal met time like two

### Topic 14:

women, said, univers, year, black, student, nation, white, church, men, one, school, mani, peopl, first, war, die, histori, right, presid

game, team, said, season, player, coach, play, state, basketbal, point, iowa, year, two, one, second, first, made, tournament, win, kentucki

### Topic 16:

show, new, fashion, art, design, collect, week, said, one, look, work, like, vork, made, dress, artist, wear, model, way, see

### Topic 17:

said, crash, two, pilot, year, flight, day, ebola, one, plane, gas, peopl, mile, report, ship, last, offici, airlin, passeng, earthquak

### Topic 18:

american, govern, new, will, russia, countri, support, agreement, allow, european, histori, strategi, exist, approv, depend, doesnt, regim, hope, scandal, least

### Topic 19:

compani, said, bank, year, busi, will, market, percent, servic, like, execut, new, million, billion, industri, also, last, investor, now, use

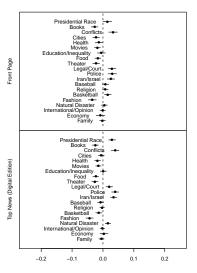
#### ic 20:

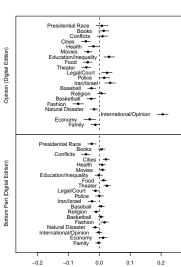
said, one, like, peopl, day, just, want, time, get, famili, friend, back, home,



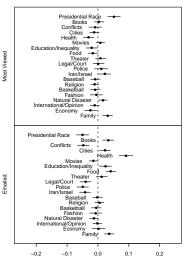
ntroduction Results Conclusion Appendix References

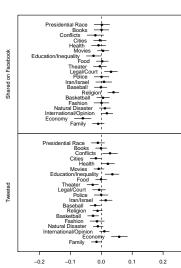
# Changes in Topic Distributions (Newspaper Section)





# Changes in Topic Distributions (Shared/Viewed)

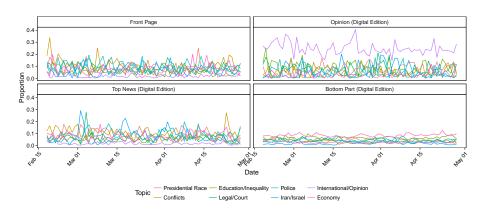




 Introduction
 Results
 Conclusion
 Appendix
 References

 0000
 0000000
 00
 0000●0

# Topic Proportions over Time (Newspaper Section)



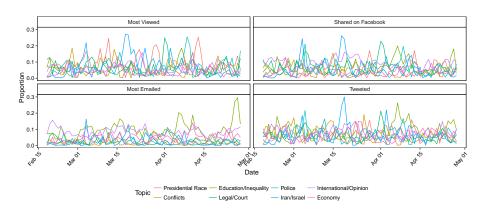


20 / 22

Introduction Results Conclusion Appendix References

0000 00000000 00 00000●

## Topic Proportions over Time (Shared/Viewed)





ntroduction Results Conclusion Appendix References

### References

Roberts, Margaret E, Brandon M Stewart, and Dustin Tingley. 2014. "stm: R package for structural topic models." *Journal of Statistical Software* 1: 1–49.

Roberts, Margaret E., Brandon M. Stewart, Dustin Tingley, Christopher Lucas, Jetson Leder-Luis, Shana Kushner Gadarian, Bethany Albertson, and David G. Rand. 2014. "Structural Topic Models for Open-Ended Survey Responses." *American Journal of Political Science* 58 (4): 1064–1082.