

Hot issue publics on Twitter:

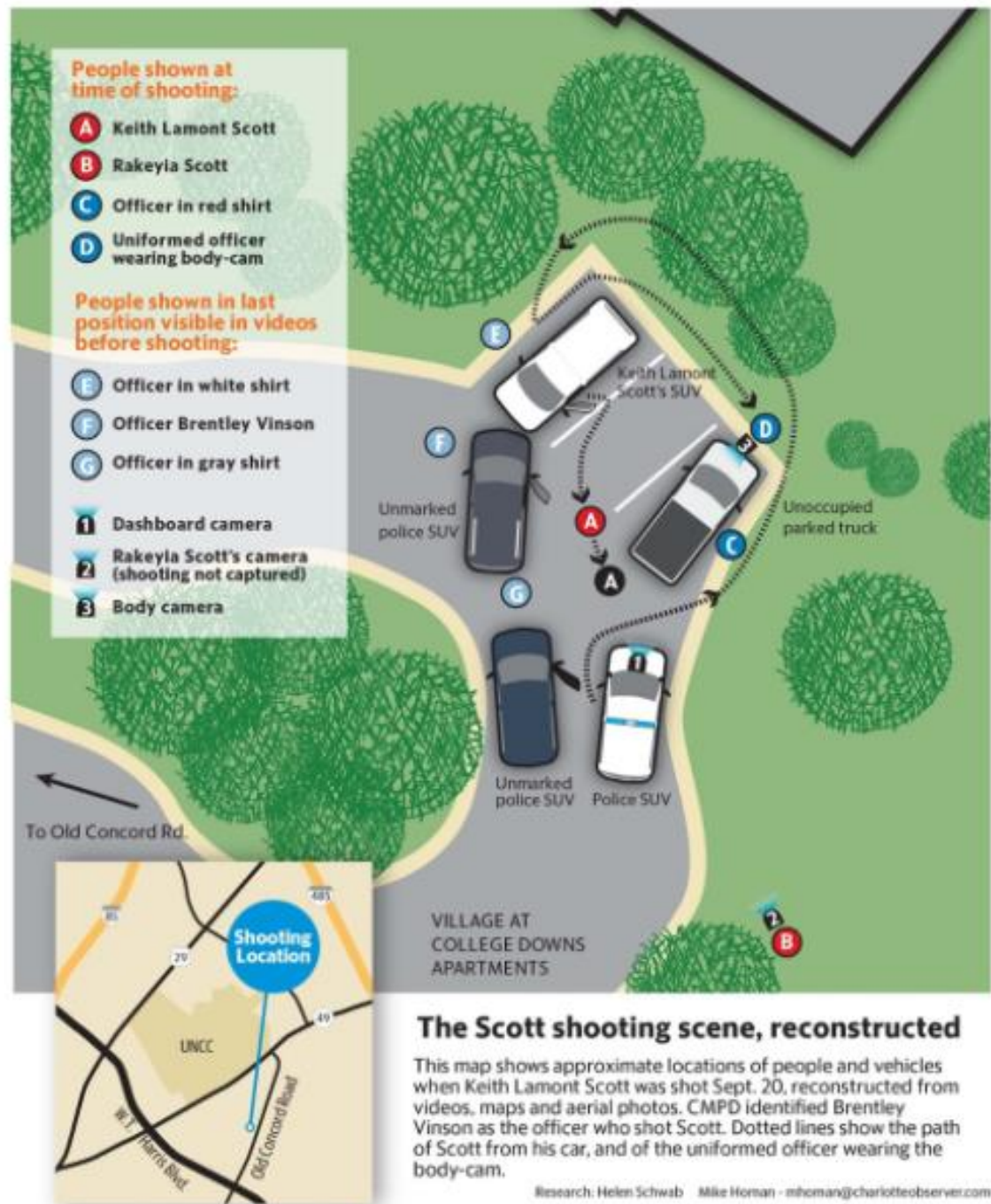
Analyzing Charlotte protests with computer-assisted text analysis



Ryan Wesslen, Tiffany Gallicano,
Jean-Claude Thill & Sagar Nandu
UNC Charlotte
IC²S², July 12, 2017

The Incident

Tuesday, Sept. 20, 2016





By **Catherine E. Shoichet**, CNN

🕒 Updated 4:11 PM ET, Fri September 23, 2016

FACEBOOK: KEIRRAMSPRETTYBABE.SCOTT

Lyric Scott, Scott's daughter, live streamed as she confronted police about the shooting



0:21 / 1:14



Charlotte shooting: Disabled black man shot dead by police 'was reading book while waiting for son's school bus'

Keith Lamont Scott, 43, was shot by police who say he was armed while getting out of a vehicle



Keith Lamont Scott's wife releases video of deadly Charlotte encounter



NEW VIDEO OF SHOOTING SCENE IN CHARLOTTE
DEADLY ENCOUNTER BETWEEN POLICE AND KEITH SCOTT



Charlotte Police Chief Says Dead Man Had a Gun, Not a Book

Keith Lamont Scott was shot and killed by police in North Carolina on 20 September 2016.

By Brooke Binkowski [f](#) [t](#)

Sep 23rd, 2016



Protest Tweet Counts: 48 Hours

Police Deploy Tear Gas



8:30pm: Protester Shot



Looting in Uptown Charlotte



Crowd gathers



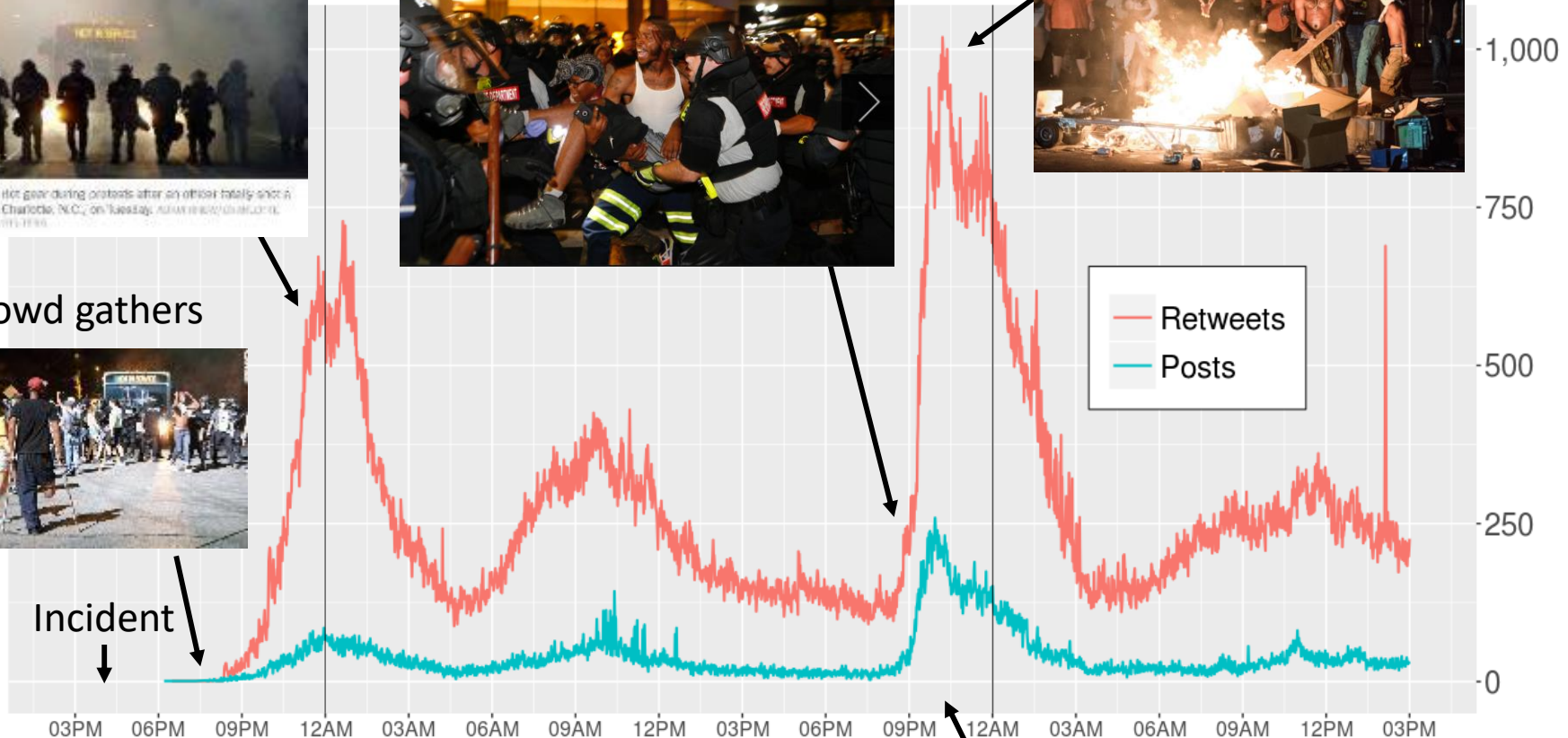
Incident

**Tuesday
Sept. 20th**

**Wednesday
Sept. 21st**

**11pm: State
of Emergency
Declared**

**Thursday
Sept. 22nd**



PR Theory: Hot Issue Publics

What is a public?

People who “begin as **disconnected systems of individuals** experiencing common problems” with the potential to “evolve into **organized and powerful activist groups**” (Grunig, 1997, p. 9)

A group of people that engages in activism on...

...a single issue → **Single Issue Public** (e.g., #BLM)

...many issues → **All Issues Public**

...a single issue that:

- involves nearly everyone in a population
- has received extensive media coverage

→ **Hot Issues Public**



- **Hot issue publics** only become active on an issue when it affects their lives, **often due to locality**.
- Involvement with the issue is more **ego-related** than related to the actual problem itself.

Research Questions

RQ 1. How do **hot issue publics'** Twitter topics differ from **single/all issue publics**?

RQ 2. How do **local ties** (e.g., city or state level) affect Twitter topics?

RQ 3. What effect does **time** have on Twitter topics?

Structural Topic Modeling (STM)

- **Topic models** are unsupervised text algorithms to detect **word co-occurrence clusters**.
- **Structural Topic Modeling (STM)** incorporates a regression framework to test the **effect of metadata (e.g., covariates)** on topics.

Roberts, Stewart, & Airoldi, 2016.

What and How Effects on Topics

Prevalence Covariates: Effect on Topic Proportions (“what”)

Covariate	Levels	Definition
Publics Type	Hot Issue, All/Single	All/Single = tweeted #BLM 14 days prior to incident Hot Issue = did not tweet #BLM 14 days prior...
Locality	Charlotte, Carolinas, Non-Local or Missing	<ul style="list-style-type: none">• Inferred from profile information.• Supervised classifier to predict missing (if possible).
Time	Hour	Spline to smooth

Content Covariate: Effect on Word Proportions (“how”)

Covariate	Levels	Definition
Locality	Charlotte, Carolinas, Non-Local or Missing	<ul style="list-style-type: none">• Inferred from profile information.• Supervised classifier to predict missing (if possible).

Dataset & Pre-Processing

1.3MM tweets over seven days: Sept. 20-26, 2016.

- Seven hashtags: e.g., #KeithLamontScott, #CharlotteProtest
- Gnip Historical PowerTrack: 100% Twitter firehose

Additional firehose data for feature extraction.

- **Hot Issue:** Did not tweet #BLM 14 days prior to protests.
- Predict **Missing Locality:** Classifier based on 14 days prior tweets.

Exclusions:

- **Remove retweets** (83% of data set) due to “**echo effect**,” problem of attribution, and limited tweet firehose allocation.
- Excluded short, non-English tweets & trending bots (with supervised classifier)

Hypothesis



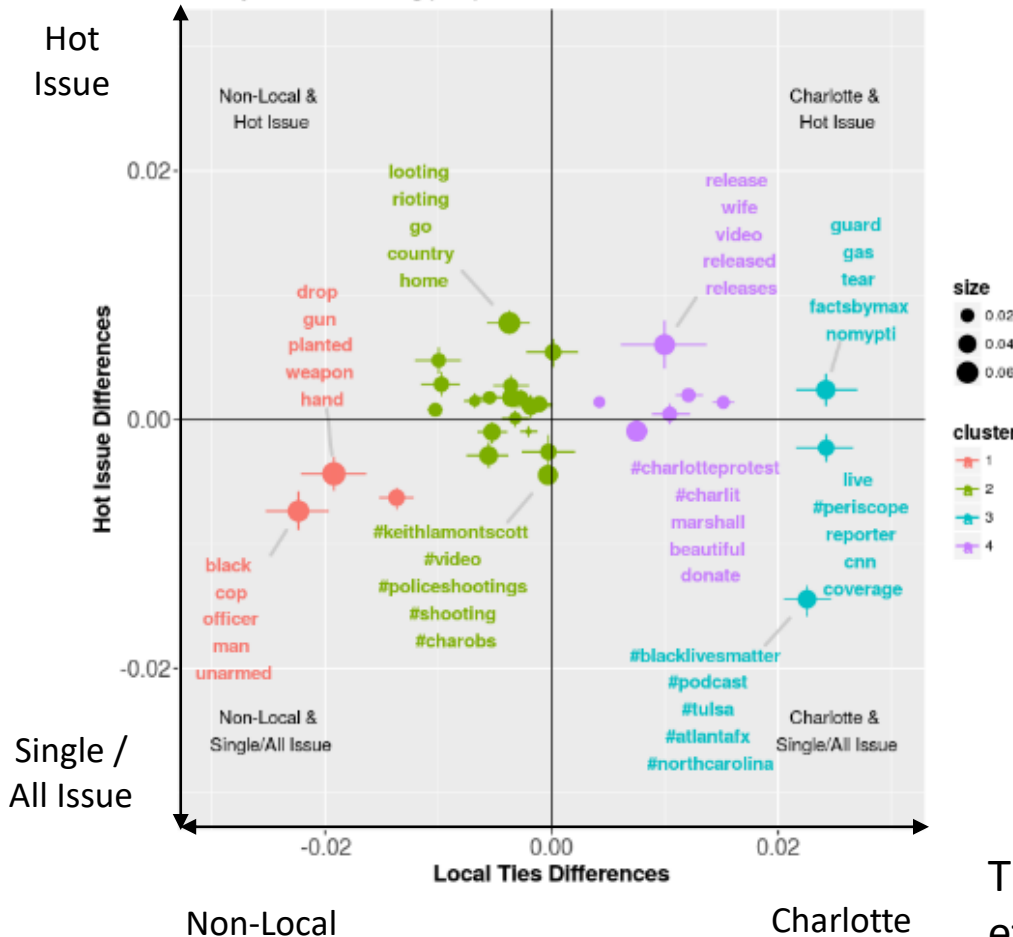
Hot issue public topics will be **short-lived**.

- Expect they will be high in peaks (esp. 2nd night) but should **quickly subside** in days that follow.
- Interaction with locality => Local & Hot Issue
- Alternatively, Non-local & Single/All Issue Topics should be the **most persistent**.

Hot Issue & Local Topics

Difference Topic Proportions with 99% Confidence Intervals

Color by K-Means Clustering (k = 4)



- Cluster topics into four categories (k-means).
- Expect **Non-Local & Single/All Issue** (red) will be the **most persistent**.
- **Charlotte/Hot Issue** (purple) expected to be the **least persistent**.

This example only considers the “Charlotte” effect. See Appendix for “Carolinas”.

Top 3 topics by hour

Protest Periods

City Curfews

LEGEND

- 1st Highest
- 2nd Highest
- 3rd Highest

Top Topics by Top Five Words

reading + book + disabled + son + sick

black + cop + officer + man + unarmed

drop + gun + planted + weapon + hand

#keithlamontscott + #video + #policeshootings

riots + telling + erupt + calm + also

violence + stop + act + congress + must

looting + rioting + go + country + home

#charlotteprotest + #charlit + marshall

guard + gas + tear + factsbymax + nomypti

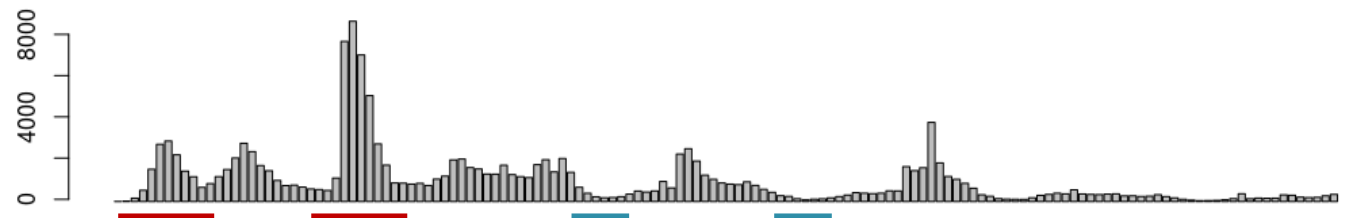
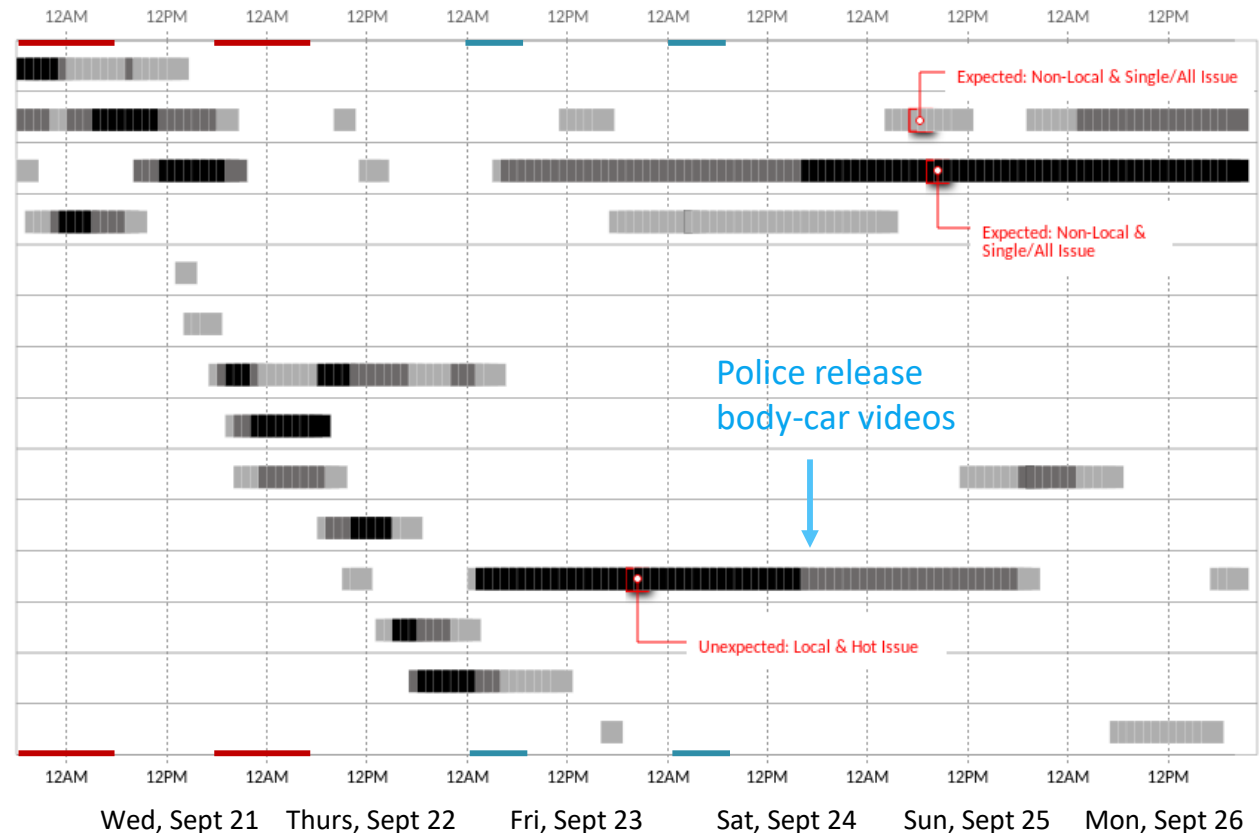
obama + thugs + rioters + obamas + blm

release + wife + video + released + releases

#charlotteriots + #msnbc + organization + group

curfew + midnight + #hillary + #charlotteprotests

last + hide + died + bullets + rubber




Timeline created by
Microsoft's Timelines Storyteller
<https://timelinestoryteller.com/>

Results and Discussion

The **most persistent topics** tended to be by Non-Locals and Single/All Issue Publics (i.e., previously engaged in #BlackLivesMatter).

However, we find an unexpected persistence of “release the video” topic, which was **Local & Hot Issue**.



Explore the role local and national organizations (e.g., Color of Change, ACLU) to influence topics with “cluster tweets” (Gallicano, Wesslen and Thill, 2017).



Future Work

Need for additional studies

- “Natural Experiment”: Compare 2014 Charlotte incident vs 2016.
- Other #BlackLivesMatters and non-protests events.

Problems with topic modeling/STM:

- Zero-sum problem of topic models (Fong & Grimmer, 2016)
- Effect of retweets: weighted STM?

Social Network & Information Diffusion

- Can hot issue relate to periphery vs. core? (Barberá et al., 2015)

Ryan Wesslen

rwesslen@uncc.edu

wesslen.github.io

@ryanwesslen

Questions?

Thanks to Tiffany Gallicano, Jean-Claude Thill, Sagar Nandu, Wenwen Dou and Ali Karduni.



UNC CHARLOTTE

Project Mosaic

Data Science Initiative

Charlotte Visualization Center

References

Grunig, J. E. (1997). A situational theory of publics: Conceptual history, recent challenges and new research. In D. Moss, T. MacManus, & D. Vercic (Eds.), *Public relations research: An international perspective* (pp. 3-48). London: Thomson.

Roberts, M. E., Stewart, B. M., and Airoidi, E. M. (2016). A model of text for experimentation in the social sciences. *Journal of the American Statistical Association* 111.515: 988-1003.


Fong, C., & Grimmer, J. (2016). Discovery of Treatments from Text Corpora. In *ACL (1)*.

Gallicano, T. D., Wesslen, R., and Thill, J.C. (2017). From cluster tweets to retweets: A big data, rhetorical exploration of digital social advocacy in the context of the Charlotte protests on Twitter. In 2017 International Public Relations Research Conference.

Barberá, P., Wang, N., Bonneau, R., Jost, J. T., Nagler, J., Tucker, J., & González-Bailón, S. (2015). The critical periphery in the growth of social protests. *PloS one*, 10(11), e0143611.

Feature Extraction

Local ties extraction on Twitter profile location

1. Use a list of North and South Carolina terms.
2. Location entity recognition & geolocation.  Google Maps APIs
3. Build supervised classifier (CV Ridge Regression) on out-of-sample tweets (10% two weeks prior) to predict labels for missing; remaining are “Missing.”

Hot Issue vs. Single/All Issue Publics

- Single/All Issue = Tweeted #BLM two weeks before protest
- Hot Issue = Did not tweet #BLM two weeks before protest

Count of Tweets	Charlotte	Carolinas	Non-Local	Missing	Total
Hot Issue	7,860	6,024	81,423	13,404	108,711
Single/All Issue	739	1,914	37,787	1,888	42,328
Total	8,599	7,938	118,210	15,292	151,039

Local Ties Effect on Content

Topic 7: drop + gun + planted + weapon + hand + marijuana + @thejusticedept

Carolinas cmpd putney #wral chief investigation #cmpd sbi evidence #releasethetape kerr marijuana gun case drop
Charlotte putney investigation chief sbi says kerr marijuana case possession handgun evidence information original gun
Missing hear drop ankle holster didnt planted day empty weapon knew wearing cmpd commands hand
Not Local gun drop planted evidence holster transparency weapon @thejusticedept putney chief hand scene marijuana planting

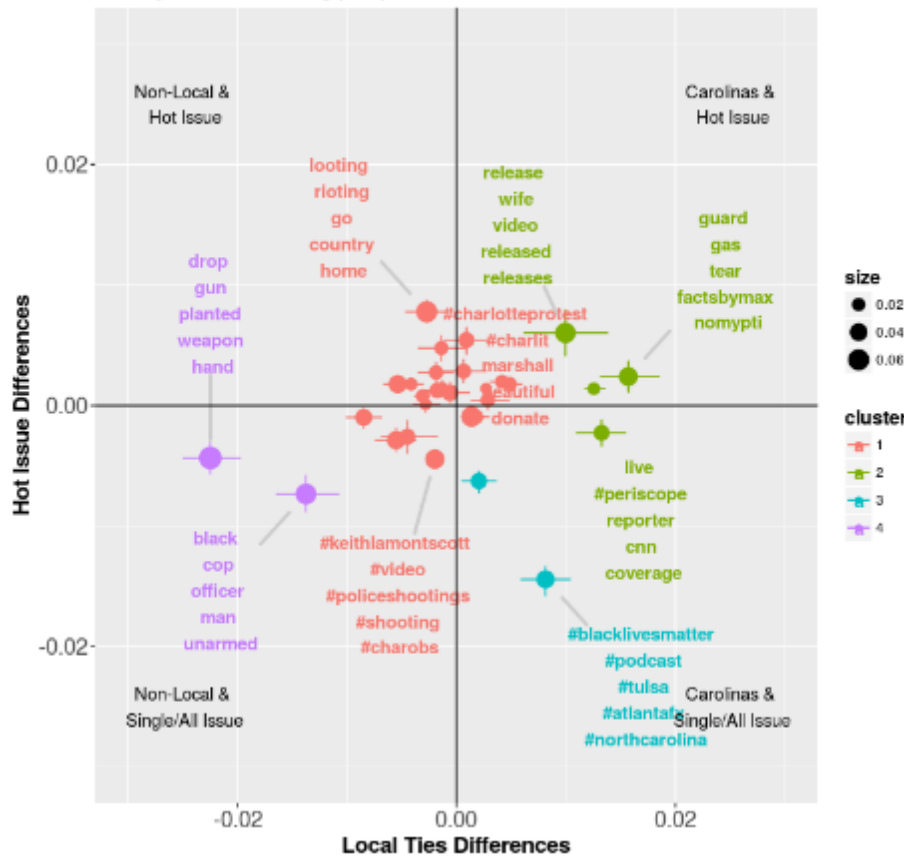
Topic 15: looting + rioting + country + home + protesting + way

Carolinas home rioting looting work protesting people close go youre way really bring country destroying
Charlotte city home together work people come way looting go close protesting rioting isnt leave
Missing rioting looting bunch start city country isnt movement cause everyone go way respect acting
Not Local looting rioting go country protesting home way change destroying like property youre work animals

Non-Local vs Carolinas/Missing

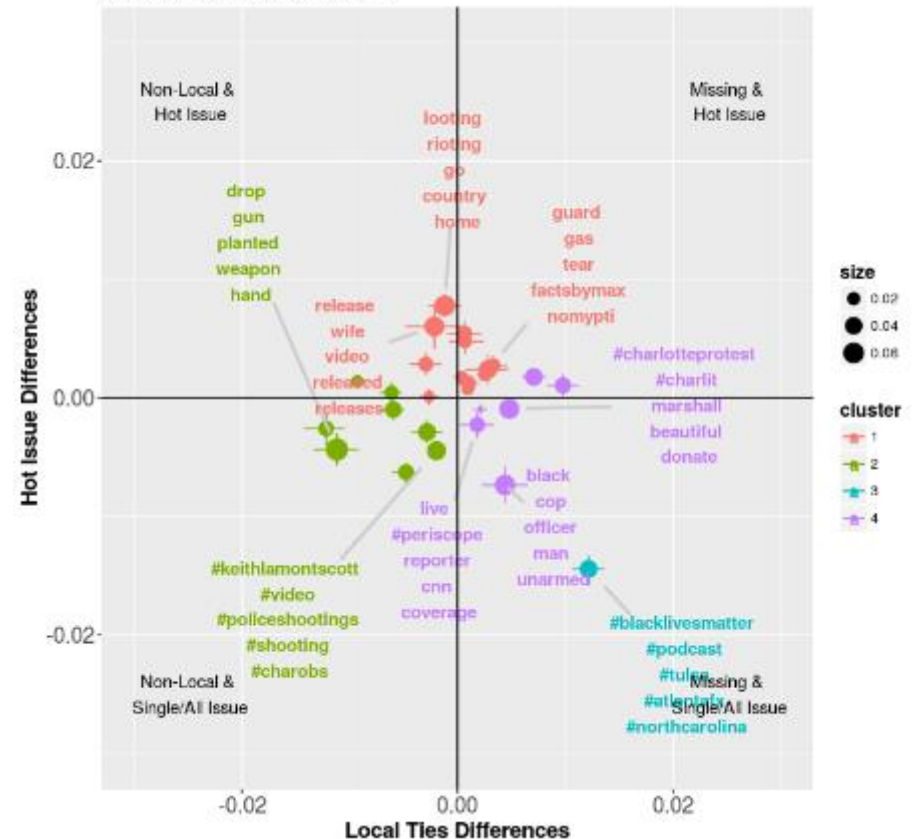
Difference Topic Proportions with 99% Confidence Intervals

Color by K-Means Clustering ($k = 4$)



Difference Topic Proportions with 99% Confidence Intervals

Color by K-Means Clustering ($k = 4$)



Predicting Missing Locality

- Used 5-fold Ridge Regression to predict users with missing profile location.
- Built two models: Charlotte and Carolinas. Used the “known” labels and held out 20% for testing.
- Achieved 90%+ accuracy but low precision/recall.

Carolinas	FALSE	TRUE
FALSE	7,325	167
TRUE	295	81

Charlotte	FALSE	TRUE
FALSE	6,948	411
TRUE	348	161

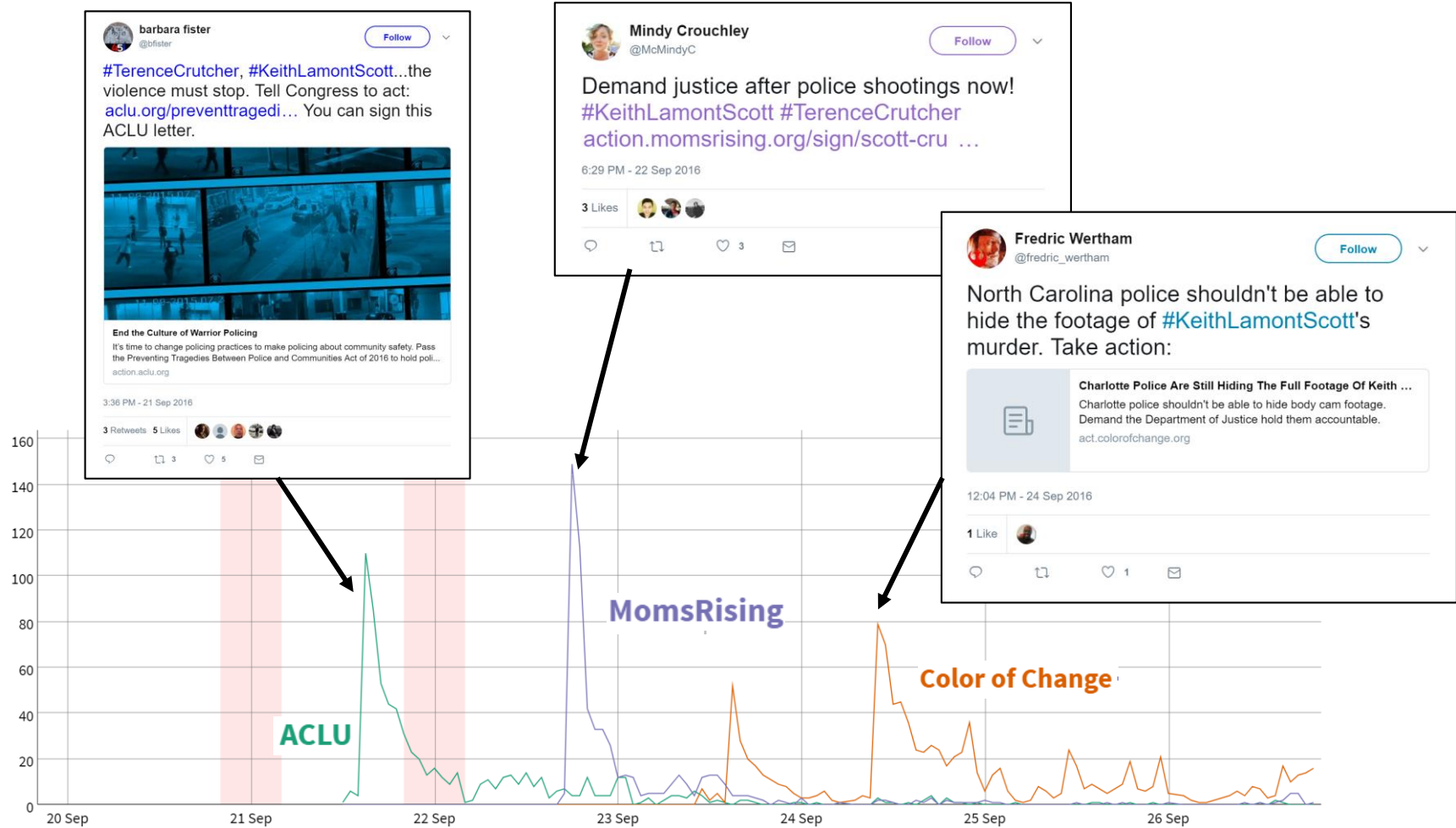
Carolinas Top Features

Rank	Coeff	word
1	0.600672	clt
2	0.589524	raleigh
3	0.438203	mccrory
4	0.436338	durham
5	0.410346	@coliestudent
6	0.400823	#repealhb2
7	0.393499	#asheville
8	0.390386	uptown
9	0.381046	@wbvt_news
10	0.379816	ecu
11	0.372657	shrimp
12	0.352878	@roycoopernc
13	0.347727	tackling
14	0.343472	triad
15	0.326004	trippin
16	0.319209	unc
17	0.312744	brackeen
18	0.307389	@cameronnewton
19	0.300898	@snapchat
20	0.298132	charlotte

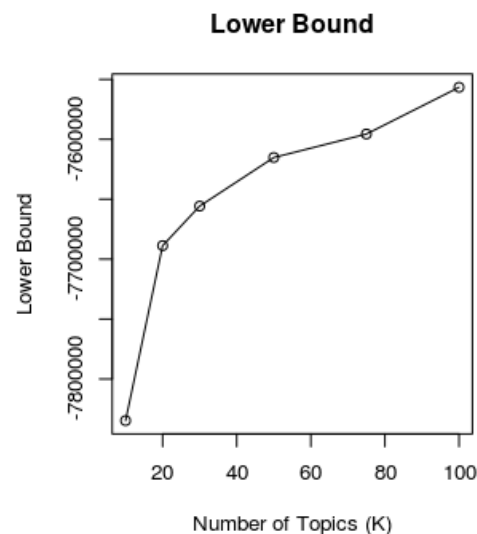
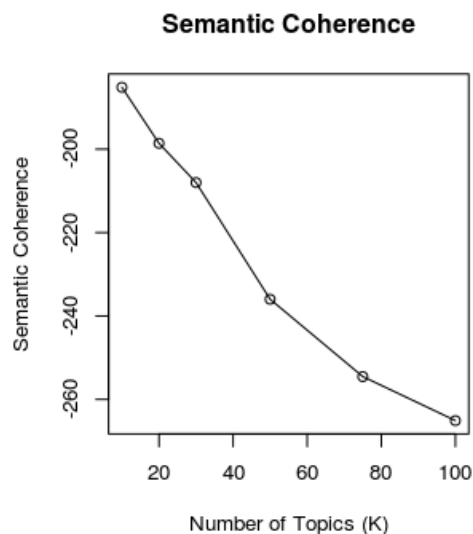
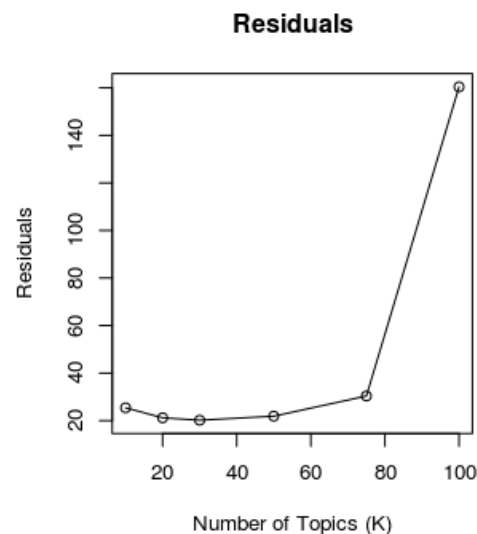
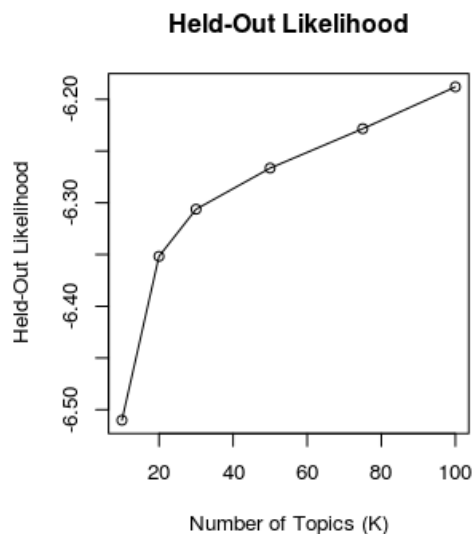
Charlotte Top Features

Rank	Coeff	word
1	1.125938	clt
2	0.920002	uptown
3	0.711054	@coliestudent
4	0.708484	@wbvt_news
5	0.694427	mod
6	0.679505	parkway
7	0.559287	charlotte
8	0.548697	gemini
9	0.53564	sized
10	0.529665	hannah
11	0.524252	tt
12	0.522303	@cameronnewton
13	0.520000	debuted
14	0.517259	reservation
15	0.515416	menace
16	0.513435	#charobs
17	0.50302	ashes
18	0.499604	plugs
19	0.491436	oak
20	0.489624	tailgate

Cluster Tweets



Number of Topics Evaluation



Racial Diversity

Green: African-Americans

Blue: Caucasians

Source:

2010 US Census

Racial Dot Plot Map
University of Virginia
Weldon Cooper Center



Charlotte ranks last in upward mobility study

The study ranked Charlotte dead last among the country's 50 largest cities for upward mobility, meaning it is harder to climb out of poverty here than in any other major U.S. metro area. The odds of moving up in Charlotte - just 4.4%.

Study: Chetty, R., & Hendren, N. (2017). The effects of neighborhoods on intergenerational mobility II: County-level estimates. Retrieved from <http://www.equality-of-opportunity.org/documents/>



