

LEARNING FROM BULLYING TRACES IN SOCIAL MEDIA

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Snape's Worst Memory

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Victim:
Severus Snape

Defender:
Lily Evans

Reinforcer:
Peter Pettigrew

Bully:
James Potter

Assistant:
Sirius Black

Bystander:
Remus Lupin



<http://harry-potter-spain.deviantart.com/art/Snape-s-Worst-Memory-27310861>

Bullying (Peer Victimization)

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Across a national sample of students in grades 4-12 in the U.S.,
38% reported being bullied by others

32% reported bullying others [Vaillancourt et al., 2010]

More students involved as assistants, reinforcers, bystanders...

Multiple forms:



physical



relational



verbal

Venues: physical world, online (cyber-bullying)

Bullying Hurts

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Symptoms of Victims

Interpersonal problems

Depression, anxiety, loneliness, low self-worth

Absent from school more often and lower grade

Every day, about 160,000 kids stay home from school
because of the fear of being bullied [The U.S. CDC]

Lethal school violence and suicide

Bullying victims are between 2 to 9 times more likely to
consider suicide than non-victims [Kim et al., 2009]

Limitations of the State-of-the-Art

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Traditional social science studies are handicapped by data scarcity

- Small sample size

- Low/no temporal resolution

- Time consuming



Computational study is largely unexplored

- Only a few studies on cyber-bullying, overlooked other bullying episode

Bullying Traces in Social Media

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Bullying trace: social media post talking about actual bullying episode (in physical world or online)

Reporting a bullying episode: *“some tweens got violent on the n train, the one boy got off after blows 2 the chest... Saw him cryin as he walkd away :(bullying not cool”*

Accusing someone as a bully: *“@USERNAME i didnt jump around and act like a monkey T T which of your eye saw that i acted like a monkey :(you’re a bully”*

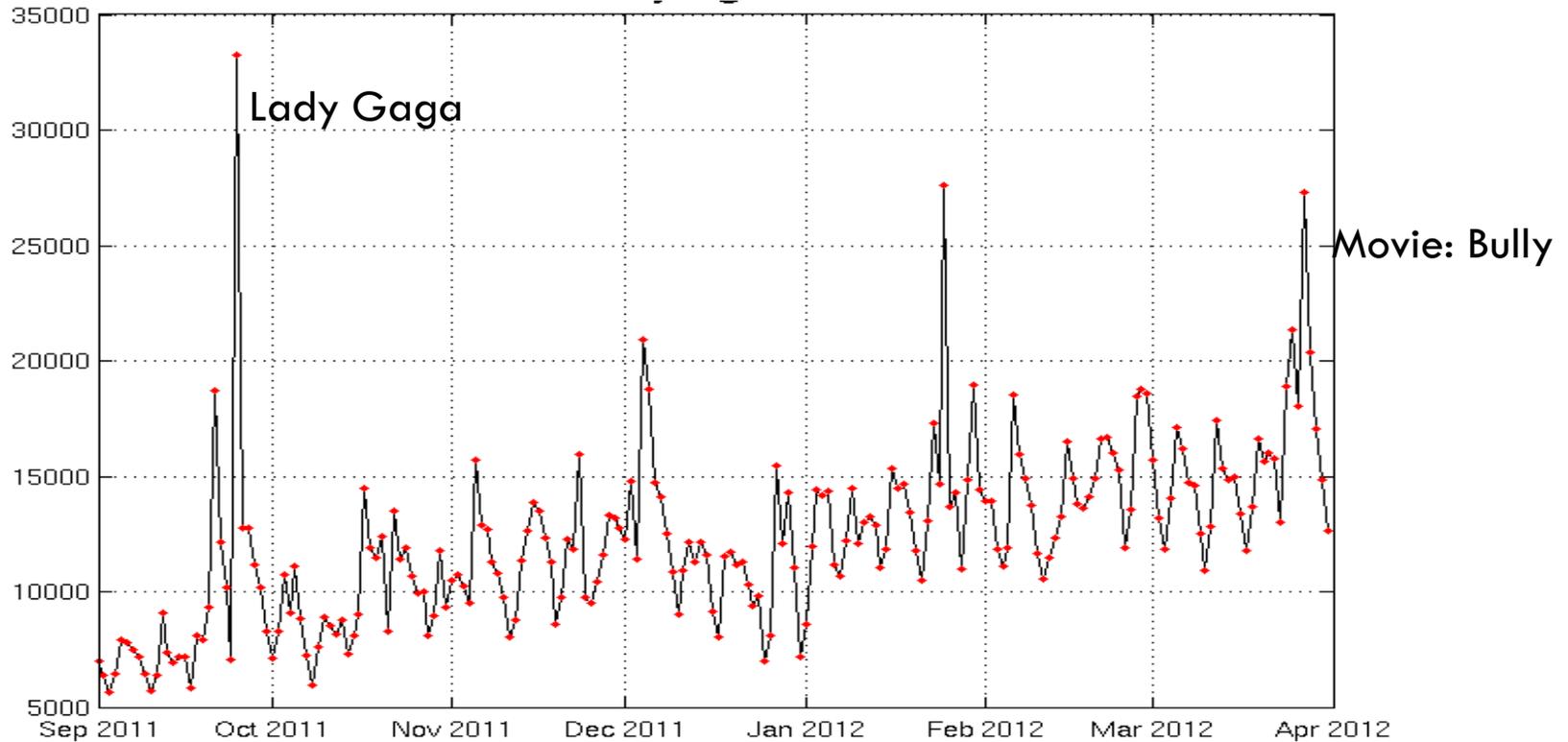
Revealing self as a victim: *“People bullied me for being fat. 7 years later, I was diagnosed with bulimia. Are you happy now?”*

Cyber-bullying direct attack: *“Lauren is a fat cow MOO BITCH”*

Bullying Traces in Social Media

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Daily number of bullying traces collected



Pros: Large sample size, High temporal resolution, Easy to collect
Cons: Unknown biases

Questions NLP Can Help with

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Is the post a bullying trace or not?

Text Categorization

Who are the participants? What are their roles?

Role Labeling

How do they feel during the episode?

Sentiment Analysis

What are people saying about bullying?

Latent Topic Modeling

.....

Manually Labeled Training Data

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Collected from Twitter Streaming API

Each has keyword such as “bully”, “bullied”, or “bullying”

Re-tweets are removed

Annotated by experienced experts

Is it a bullying trace or not ($\kappa = .72$)

Bullying roles of author and person mentions ($\kappa = .61$)

Is the bullying trace written jokingly ($\kappa = .81$)

Task A: Text Categorization

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Classify bullying or not

684 out of 1762 posts in enriched dataset were bullying traces

“Is Google + the new BULLY PLAYGROUND?” (no episode)

“You know what? I’ve never seen a Bully beat a kid up for their lunch money lol.” (the episode did not happen)

Standard procedure

Tokenization (emoticon, hash-tag, user mention, url)

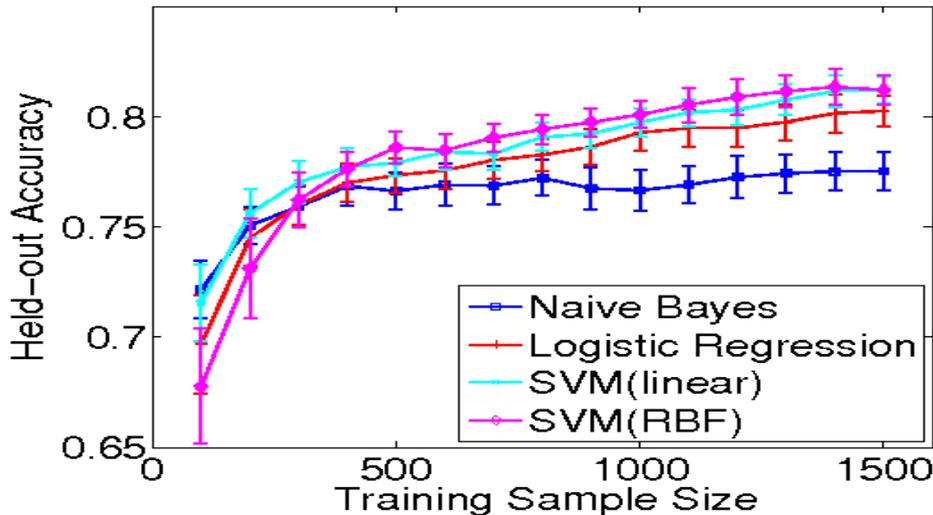
Bag-of-words representation (unigrams and bigrams)

SVM (RBF kernel)

Tune parameters with 10-fold cross-validation

Task A: Text Categorization (cont.)

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Majority class baseline
Accuracy 61%

Future Work:

Apply the trained model to other social media stream

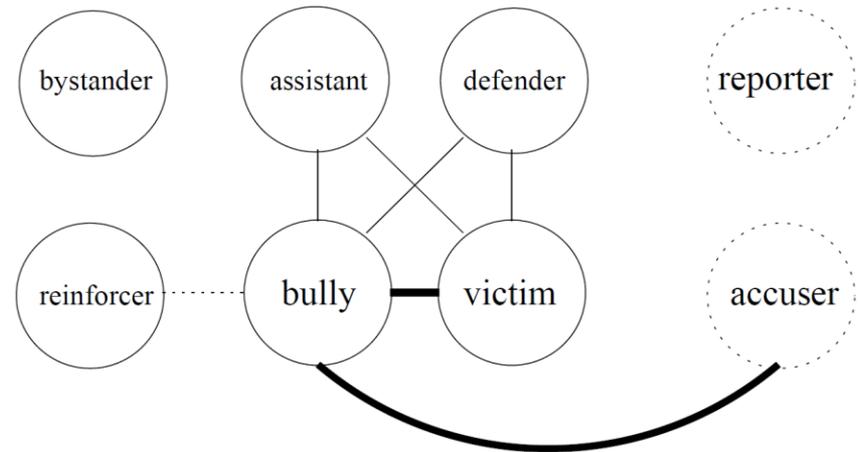
Keyword filtering -> Other forms

Twitter -> Facebook, google+,...

English -> Other language, Weibo

Task B: Role Labeling

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AUTHOR(R): “**We(R)** visited **my cousin(V)** today & #Itreallymakesmemad that **he(V)** barely eats bec **he(V)** was bullied . :(**I(R)** wanna kick the crap out of those **mean kids(B)**.”

Task B-1: Author's Role Labeling

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Multi-class text categorization task

	predicted as				
	A	B	R	V	O
A	33	3	39	10	1
B	5	25	57	11	0
R	15	5	249	27	0
V	1	4	48	109	0
O	1	1	37	3	0

A: Accuser

B: Bully

R: Reporter

V: Victim

O: Other

Accuracy: 61%

Baseline acc: 43%

Future work:

Take advantage of self-mention

Jointly classify many tweets authored by the same person

Task B-2: Person-Mention's Role Labeling

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Sequential tagging

	Accuracy	Precision	Recall	F-1
CRF	0.87	0.53	0.42	0.47
SVM	0.85	0.42	0.31	0.36

A: Accuser B: Bully
R: Reporter V: Victim
O: Other N: Not a person

	predicted as					
	A	B	R	V	O	N
A	0	4	5	10	0	4
B	0	406	13	125	103	302
R	0	28	31	67	0	13
V	0	142	28	380	43	202
O	0	112	4	42	156	86
N	0	78	4	41	16	9306

CRF

Future work:

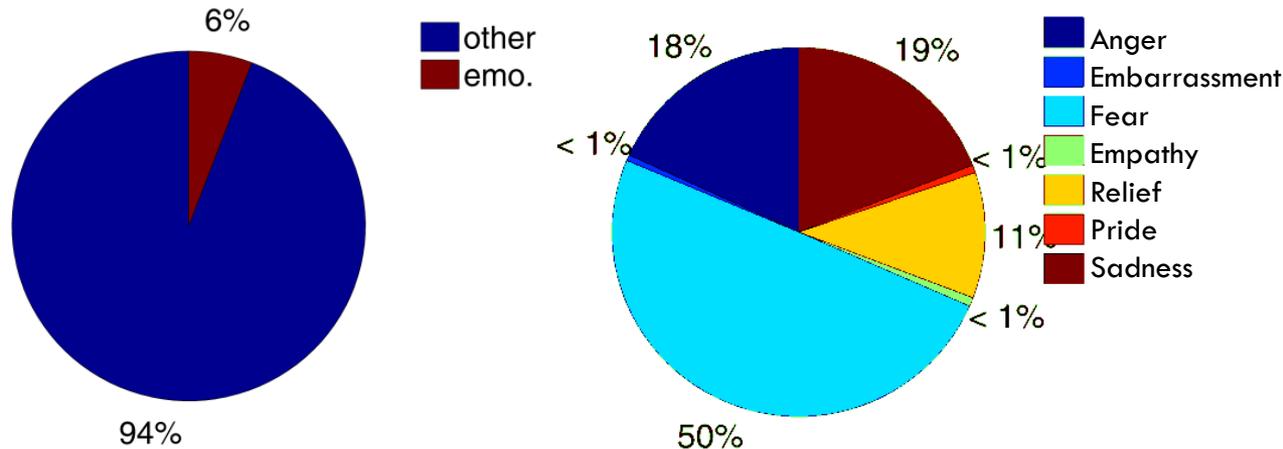
Recognize person mentions: “sister”, “teacher”, “girls”...

Train NER on informal tweets

Task C: Sentiment Analysis

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Emotions: a wide range, some in extremes



Teasing: lacking of severity, orthogonal to other emotions

“I may bully you but I love you lots. Just like jelly tots!”

“@USERNAME lol stop being a cyber bully lol :p.”

Binary text classification as in Task A (accuracy 0.89, baseline 0.86)

Future Works

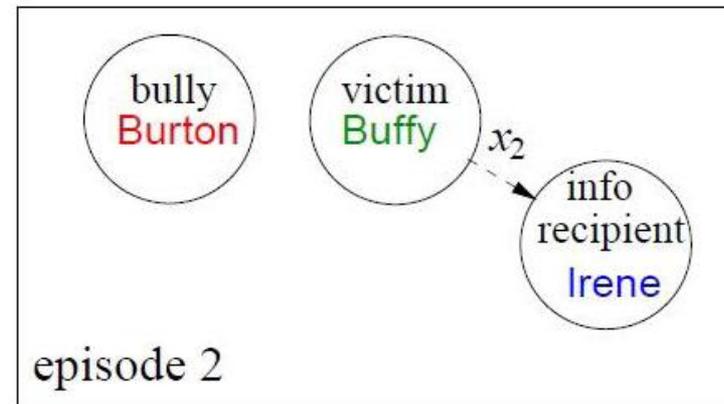
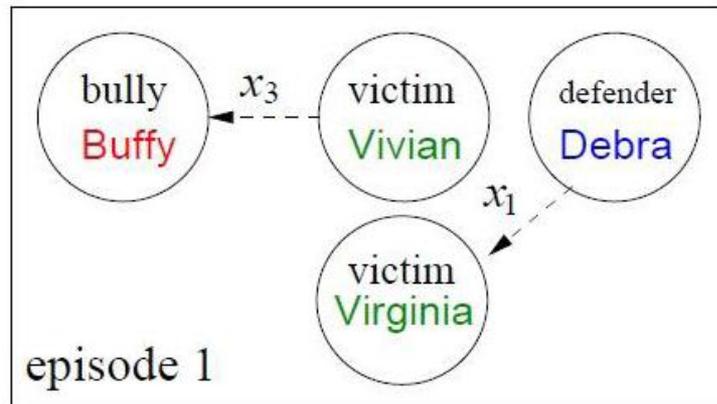
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Recovering the whole structure of an episode

Debra: *“Virginia, I heard Buffy call you and Vivian fat—ignore her!”*

Buffy to Irene: *“Burton picked on me again because I’m only 5 feet”*

Vivian: *“Buffy I’m not fat! Stop calling me that.”*



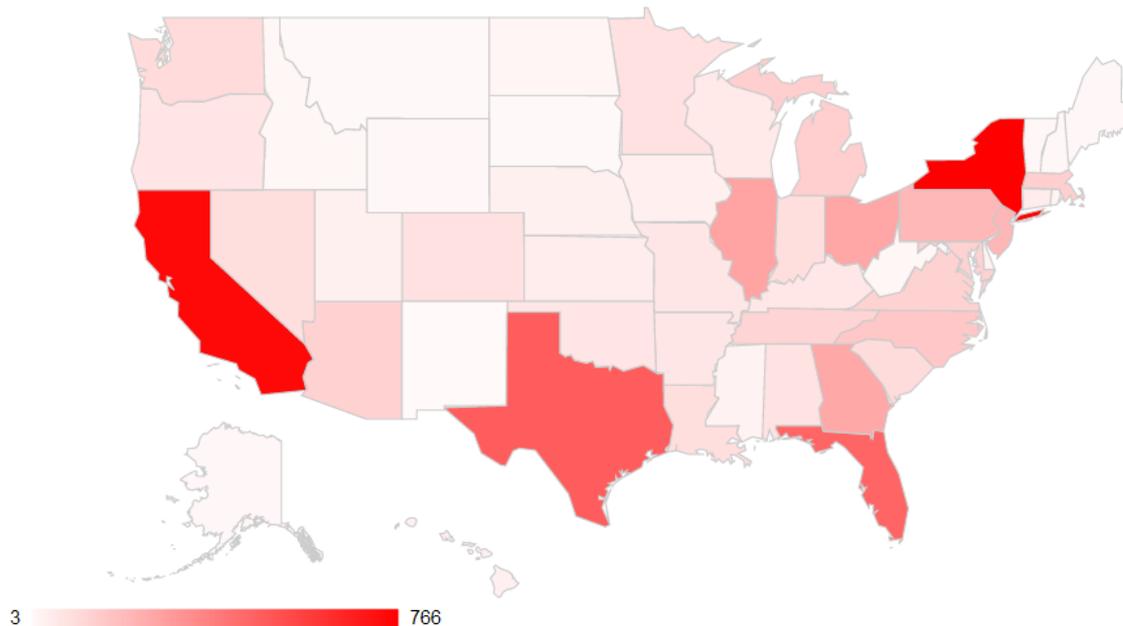
Future Works (Cont.)

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Mild intervention: showing a bullying intensity map

Help victim to cope

Raise public awareness



Conclusions

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Social media is an important data source for the study of bullying

NLP community can contribute more

Data and code available online

<http://research.cs.wisc.edu/bullying>

Thanks!

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