Text classification in the wild

March 24, 2017

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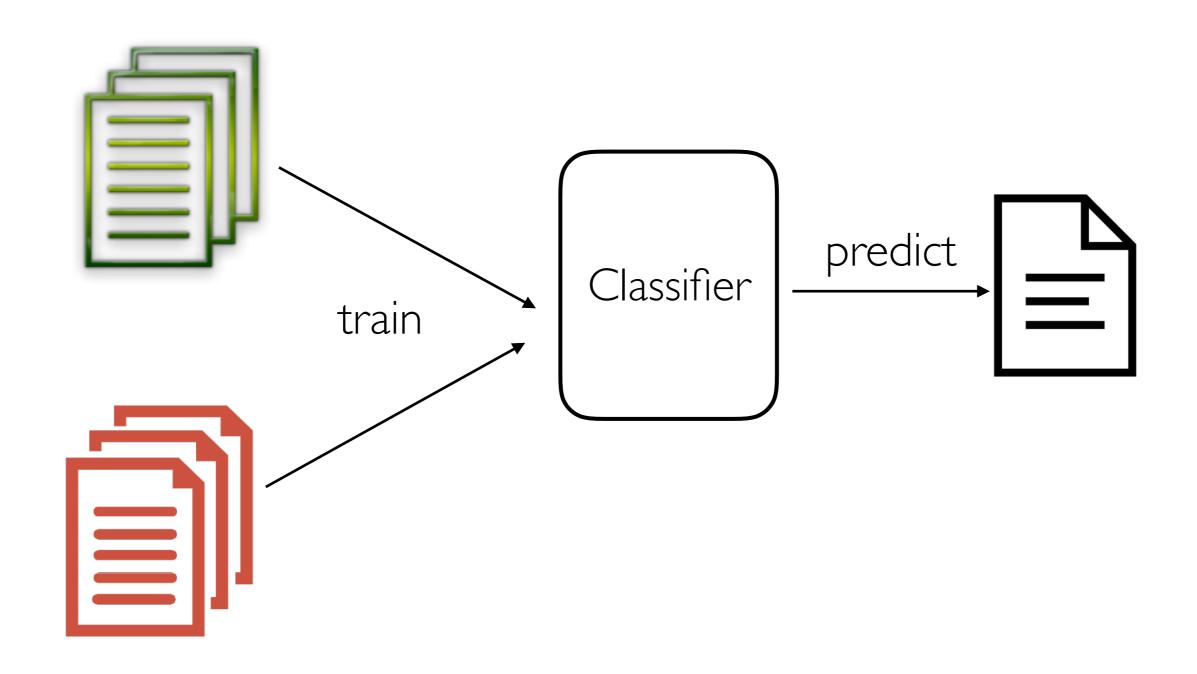
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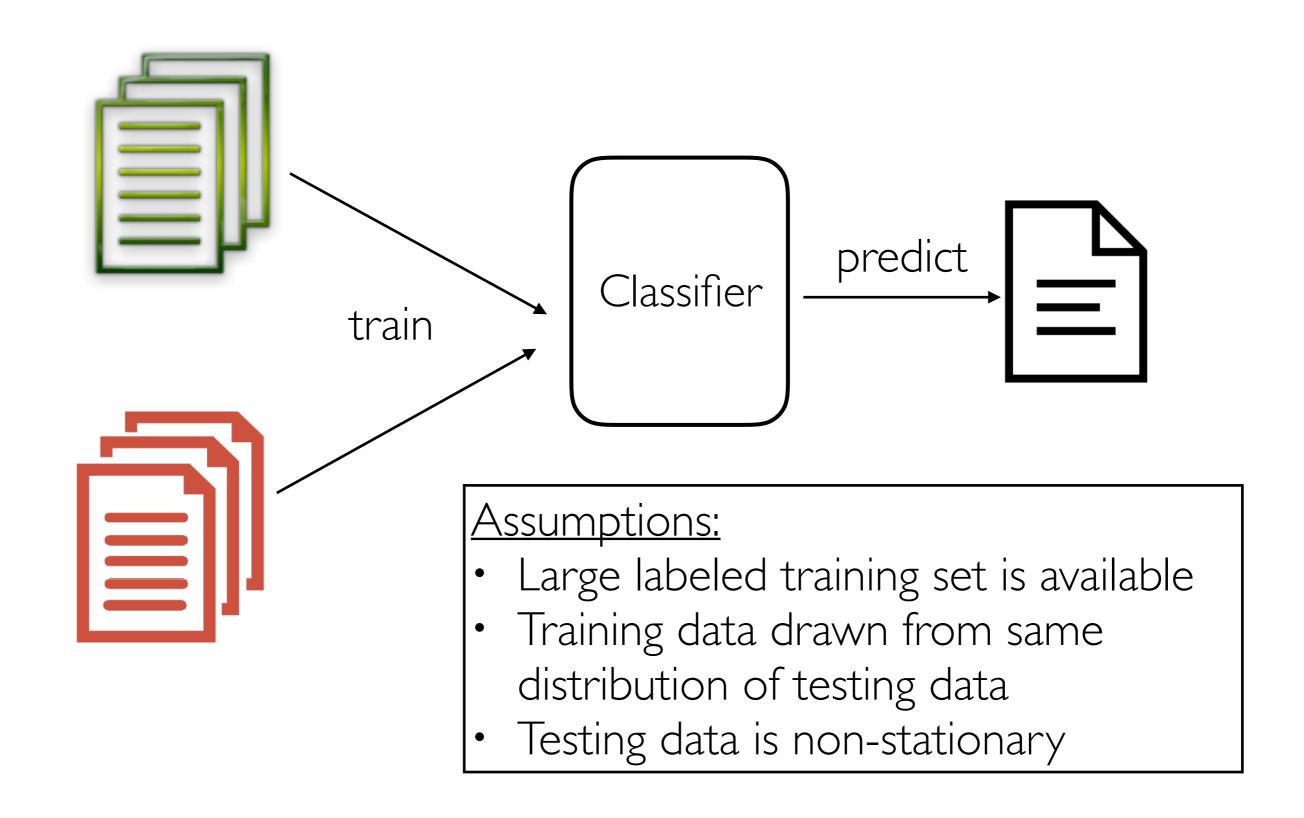
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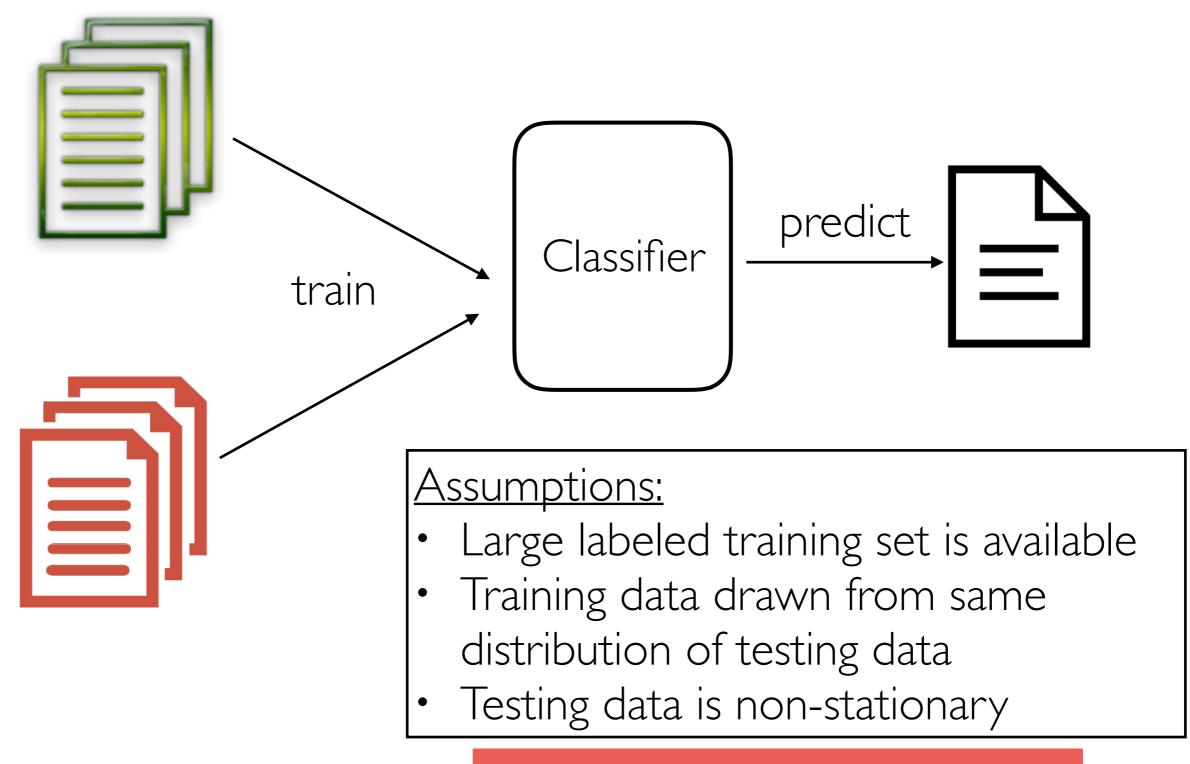


Labeled training data

Unlabeled testing data







Rarely (if ever) true in practice

Outline

- Domain adaptation
 - Identifying hazardous products from Amazon reviews

- Confounding variables
 - Predicting health attributes of social media users

- Learning from bags of instances
 - Predicting demographics of social media users

Identifying Amazon reviews that precede product recalls

[Bhat & Culotta 2017]

Contours Options Tandem Stroller, Ruby



106 customer reviews





















cpsc.gov

Recall date: July 24, 2012 Recall number: 12-233



Recall Summary

Name of product:

Contours Options LT Tandem Strollers

Hazard:

The front wheel assembly can break, posing a fall hazard to the child in the stroller. In addition, for strollers manufactured in January and February 2012, the nuts that hold the stroller's basket support screws in place can detach. Detached nuts can pose a choking hazard to young children.



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Customer Review

4 of 6 people found the following review helpful

Flimsy, and the company is unresponsive - waste of \$, December 10, 2010

By **Madfoot**

This review is from: Contours Options Tandem Stroller, Ruby (Older Version) (Discontinued by Manufacturer) (Baby Product)

I was so excited when I found this stroller. I loved that I could turn the seats forward or back, so the kids could face each other or me. It was a bit of a bear to get in and out of the back of my car, but I didn't mind because it was really handy. But after less than 4 months of use, it fell apart: the front end collapsed because the two pins holding it in place popped out. This happened just as I was leaving the house with my parents, and it was very embarrassing; they clearly thought I was trying to take their granddaughters out in a death-trap, and I am inclined to agree.

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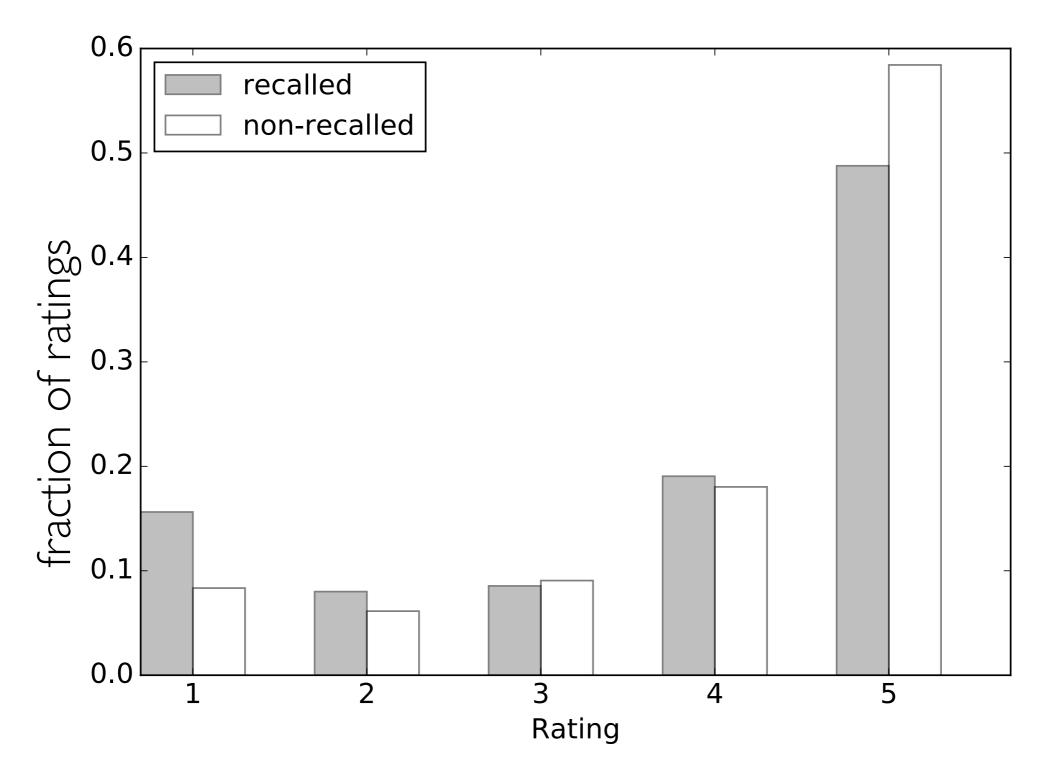
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Recalled products also have good reviews.



Can we build a text classifier to identify Amazon reviews indicating a potential health or safety hazard?

Desired labeled data

```
"...My son almost suffocated to death in this..."
```

"...My baby rolled forward and hit his head..."

"...My daughter got a bump on her head..."

positive review mentioning a hazard

"...awesome stroller!!..."

"I'm sending this product back today!!! ... This seat was so heavy I could hardly get it out of the box."

"It's cheaply made. I washed it on the gentle cycle and it began to fall apart :("

negative
no mention of hazard
(though can be other
problems with product)

Problem Positive examples are rare.



- Report Details
- Product Details
- Incident Details

Incident Description

Britax B-ready stroller seat hinge came apart completely as my child was climbing into the stroller. The footrest dropped and the child fell from the stroller, scratching her leg and arm as she fell. There is no way to repair the hinge. The footrest no longer works properly.

Incident Date 5/21/2015

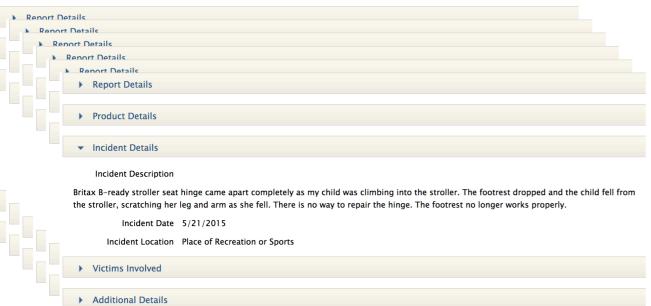
Incident Location Place of Recreation or Sports

- Victims Involved
- Additional Details

Positive/Unlabeled Learning

[Li & Liu 2005]





Positive examples from one domain



Customer Daview
Customer Daview
Customer Daview
Customer Daview
Customer Review

4 of 4 people found the following review helpful

February 15, 2013

By Erica Helf

Verified Purchase (What's this?)

This review is from: Phil&teds TS3 Car Seat Adapter for Graco Classic Connect To Sport, Dot, Classic and Navigator Strollers - Main Seat (Baby Product)

After much searching I found wording that all accessories made for phil&teds explorer will fit the newer model P&T's

Unlabeled examples from another domain assumed to be negative

Baseline Classifier

Logistic regression

- Positive examples from <u>SaferProducts.gov</u>
 - "Babies & Kids" category

- · Negative examples: random sample from Amazon reviews
 - "Baby" category
 - Optionally restrict to those with star rating > T

Problem with the baseline classifier

```
Top terms for positive class:
            mold
          pampers
             fell
            crib
            rock
         dangerous
         night light
           hazard
            gate
```

rash

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            rash
```

Selection bias

- Certain products are more common in complaints to CPSC than on Amazon
 - cribs, night lights, gates

 Traditional algorithms assume training data are drawn from same distribution as testing data

• Here, $p_{train}(y|x) := p_{test}(y|x)$

Solution: Feature Weighting

Idea:

 increase the feature values for terms that correlate with positive instances in the Amazon data

"This crib is very dangerous"

"crib"	"is"	"very dangerous"	"zebra"	"very"	
1	1	1	0	1	
1.2	1.01	17.4	0	1.5	

Solution: Feature Weighting

Of course, we don't know which reviews are positive. So, we estimate with the baseline classifier.

$$p(y = 1 \mid x_i = 1) \approx \frac{C(\hat{y} = 1 \land x_i = 1)}{C(x_i = 1)}$$

- \hat{y} class label predicted by baseline
- x_i feature i
- $C(\cdot)$ document count

E.g.,
$$p(y = 1 \mid pampers = 1) = .02$$

Compared with .91 in original training data

Solution: Feature Weighting

With additional transformations, we scale probabilities for each term to be non-negative and have mean=1.

We use the transformed values and train on the same data as the baseline model.

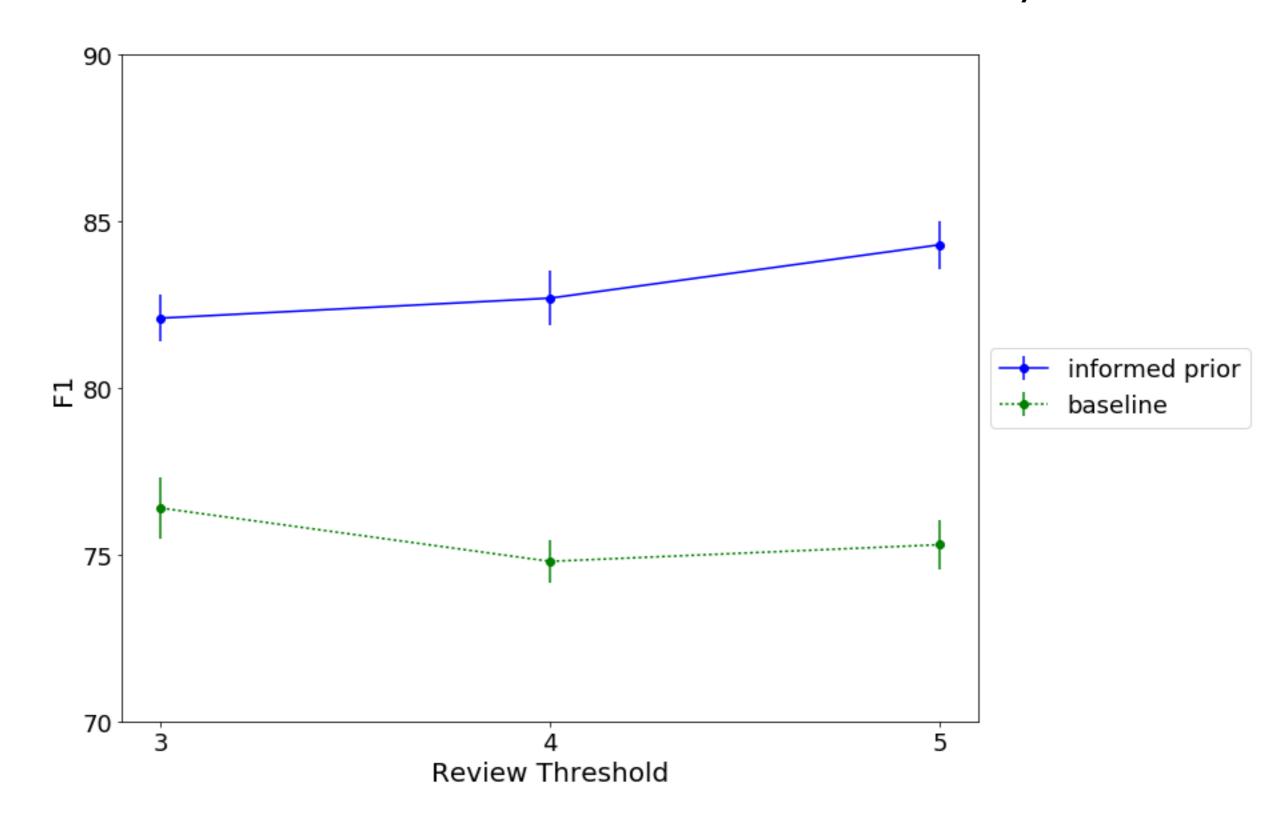
We also transform the values of the testing data.

We call this the "informed prior" method.

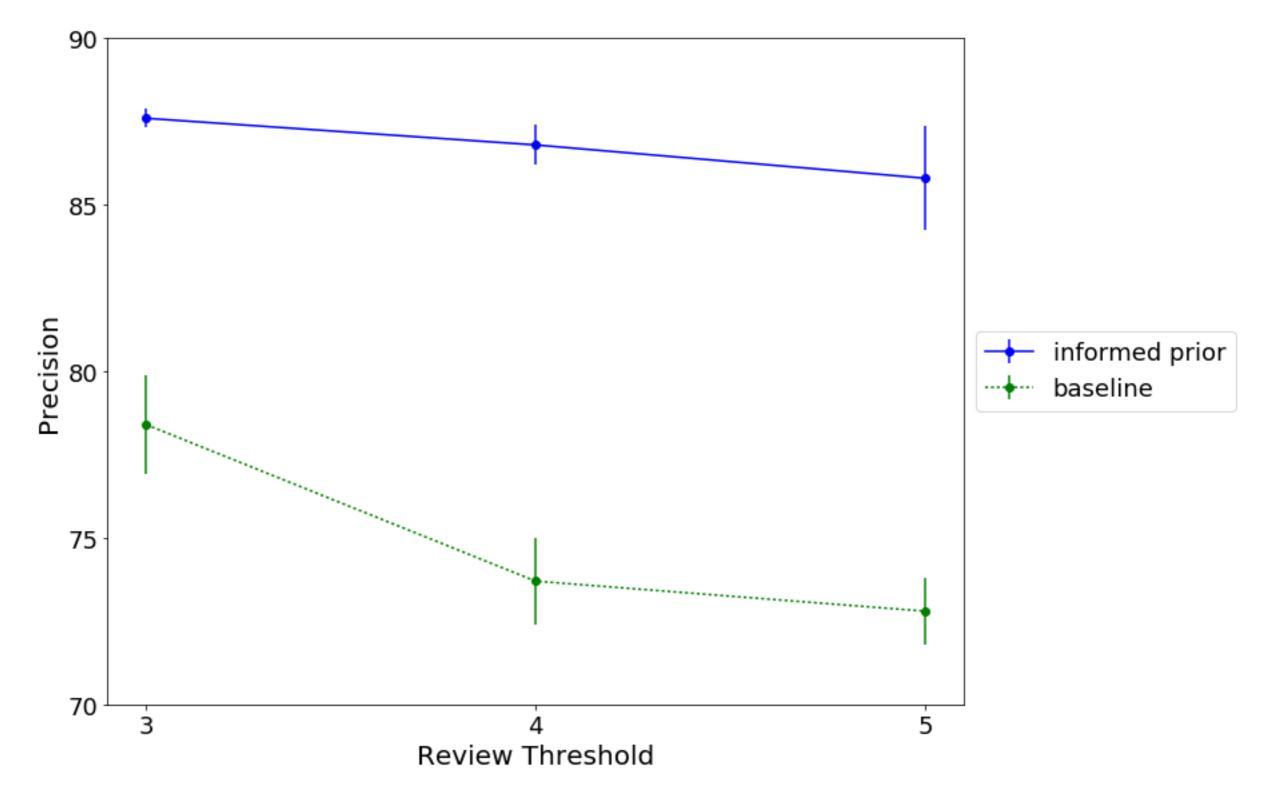
"crib"	"is"	"very dangerous"	"zebra"	"very"	
1.2	1.01	17.4	0	1.5	

Model	Top terms	
Informed	very dangerous, cpsc, mold, smacked, swal-	
prior	low it, emergency room, recalled, recall, was playing, hazard, is unsafe, snapped, leaned forward, the consumer, got stuck, was hanging, burnt, injured, exploded, was chewing	
Baseline	mold, <i>pampers</i> , fell, <i>crib</i> , rock, dangerous, <i>night light</i> , hazard, broke, happened, <i>gate</i> , rash, <i>light</i> , recall, <i>model</i> , stuck, unsafe, caused, noticed, choking	

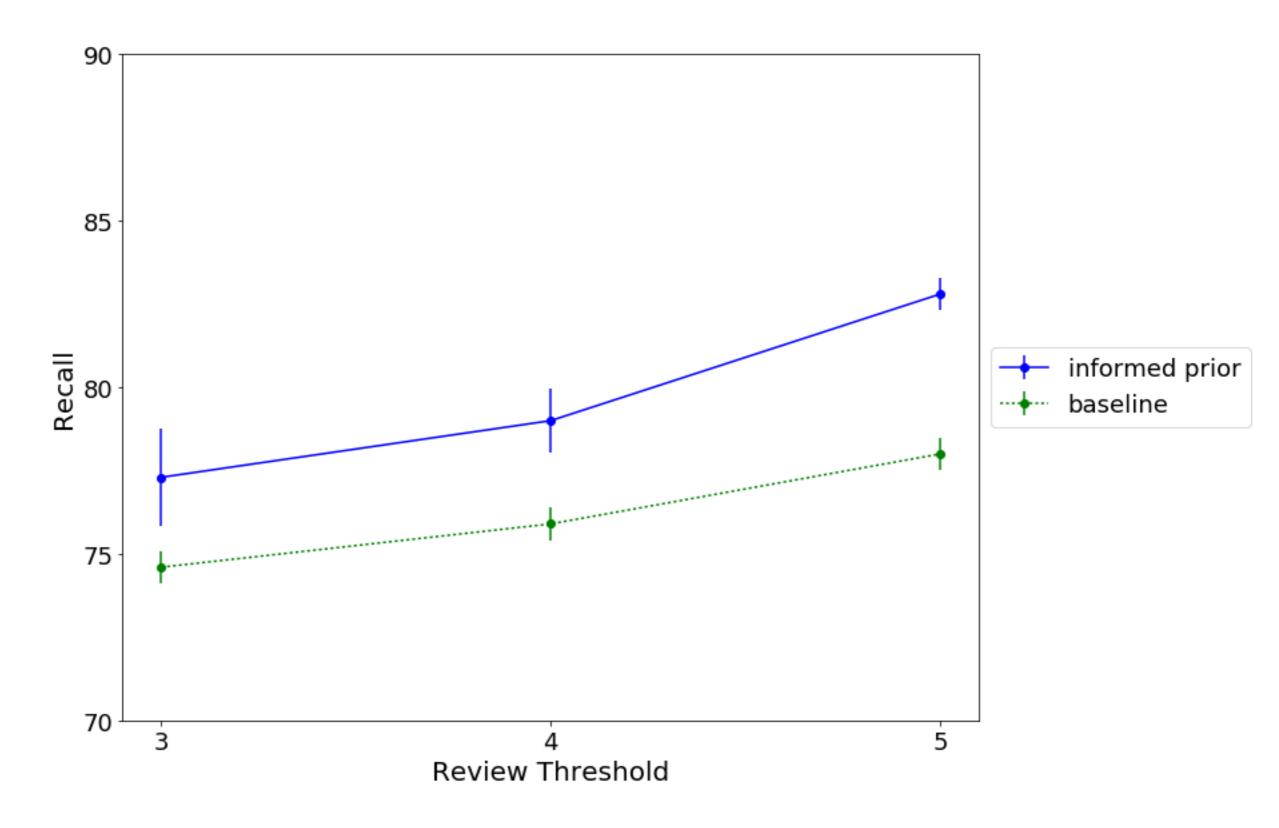
Review classification accuracy

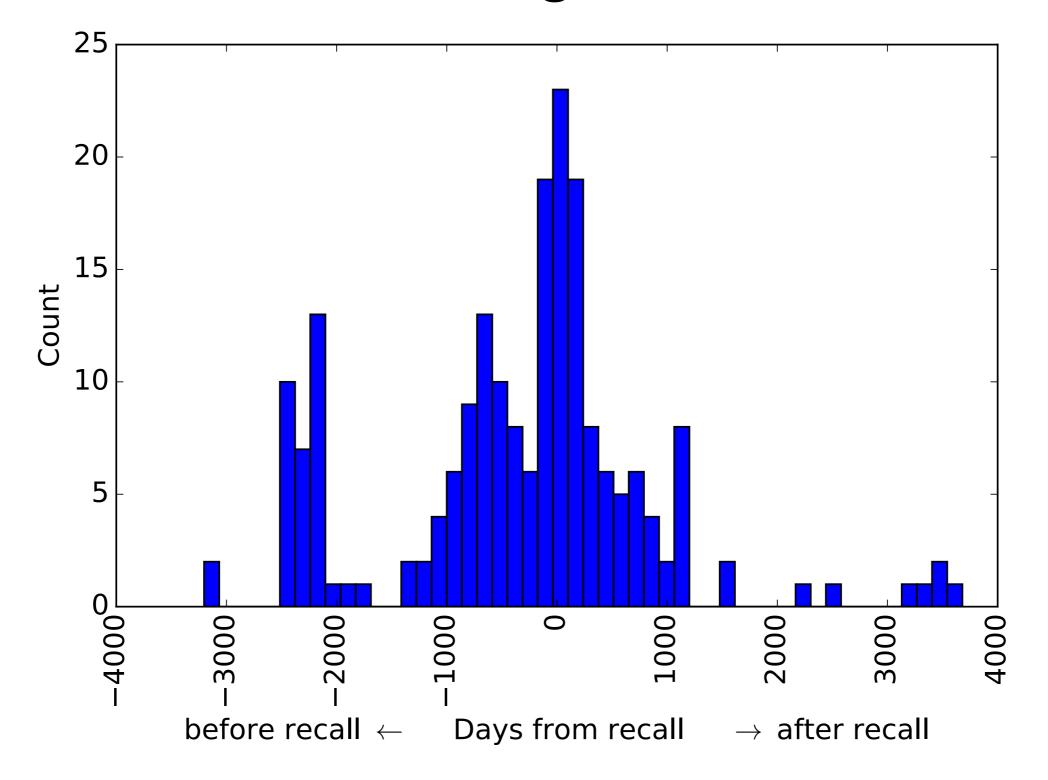


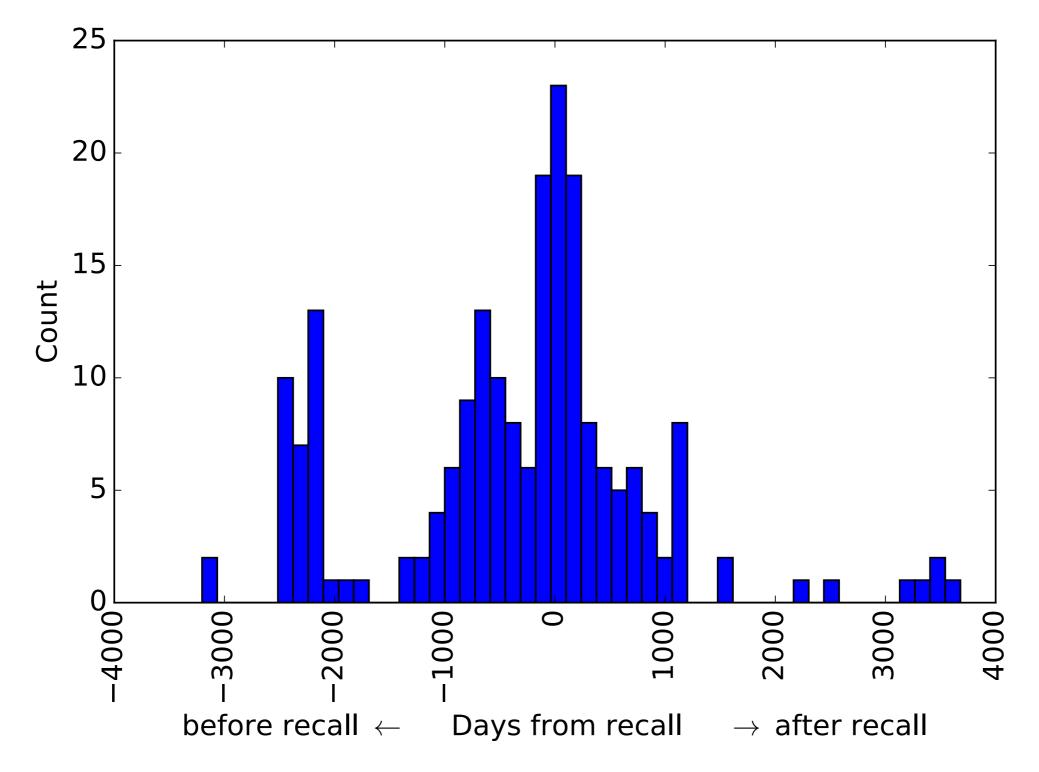
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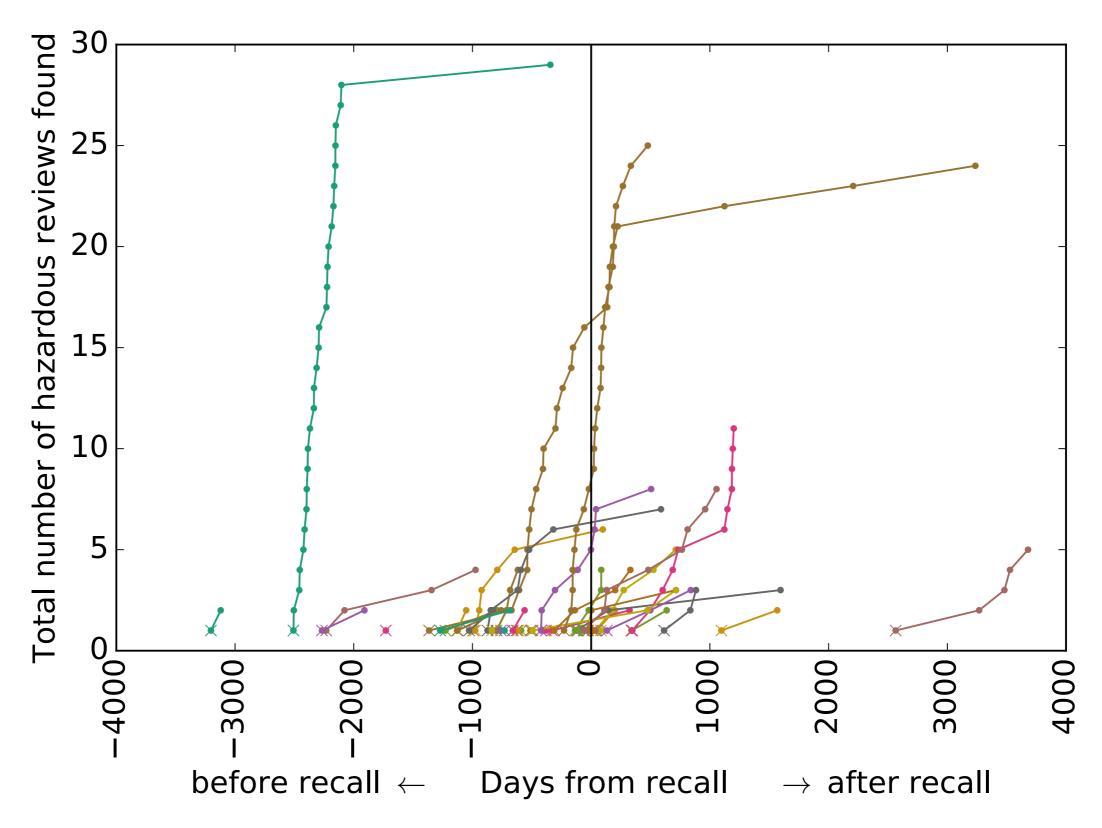
Review classification accuracy







Hazardous review found prior to recall for 45% of products



Detected reviews often relate to recall

Review, 4/30/2013

"I'm not sure if this attachment has a defect or if it is only supposed to have one button on the adapter, but it makes the carseat very wobbly and unstable...Is mine defective? Everyone else seems to have great reviews, but mine is so unstable it seems dangerous."

Recall, 6/4/2014

"The plastic adaptors used to connect an infant car seat to a stroller can crack, become unstable and break during use, posing a fall hazard to infants."

Detected reviews often relate to recall

Review, 1/14/2007

"It's a very poor design and needs a LOT of work. And my daughter ends up in a crumbled up ball on one side of the swing more times than not."

Recall, 5/30/2007

"Infants can shift to one side of the swing and become caught between the frame and seat, posing an entrapment hazard."

Conclusions: Review Classification

- Feature reweighting can adapt a model from one domain to another
- Allows us to train a classifier using "found" data
- Amazon reviews appear to provide leading indicators of subsequent product recalls

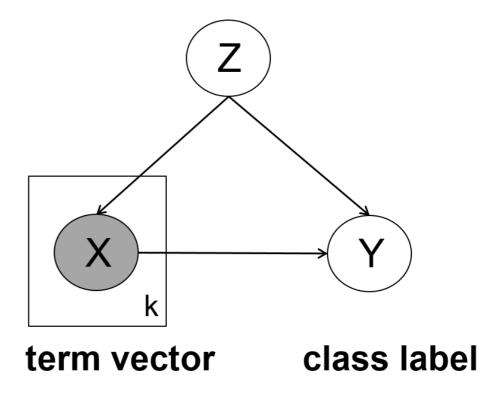
Code/data: https://github.com/tapilab/icwsm-2017-recalls

Controlling for confounds in text classification

[Landeiro & Culotta '16, '17]

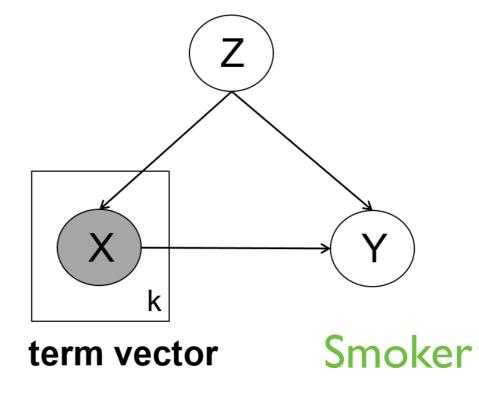
• In the review classification domain, we had $p_{train}(y|x) := p_{test}(y|x)$

• In other domains, there may be a confounding variable **z** such that $p_{train}(y|z) := p_{test}(y|z)$ confounder



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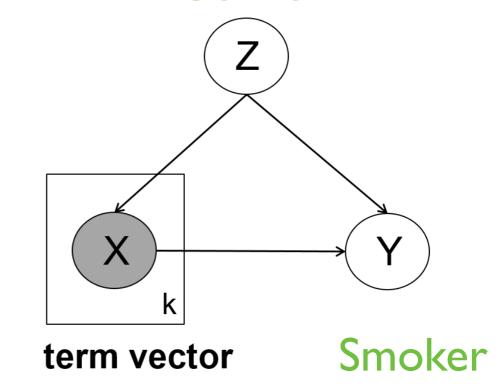


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"Observational studies from social media"

[Yom-Tov '16; De Choudhury & Kıcıman '17; Altoff et. al '17; ...]



- Why does $p_{train}(y|z) != p_{test}(y|z)$ happen?
 - Limited training sample
 - Data drift over time

Adjusting for confounds in classification

• Covariate adjustment ("backdoor adjustment")

- Logistic regression estimates $p(y \mid \mathbf{x})$
 - Omitting z leads to omitted variable bias

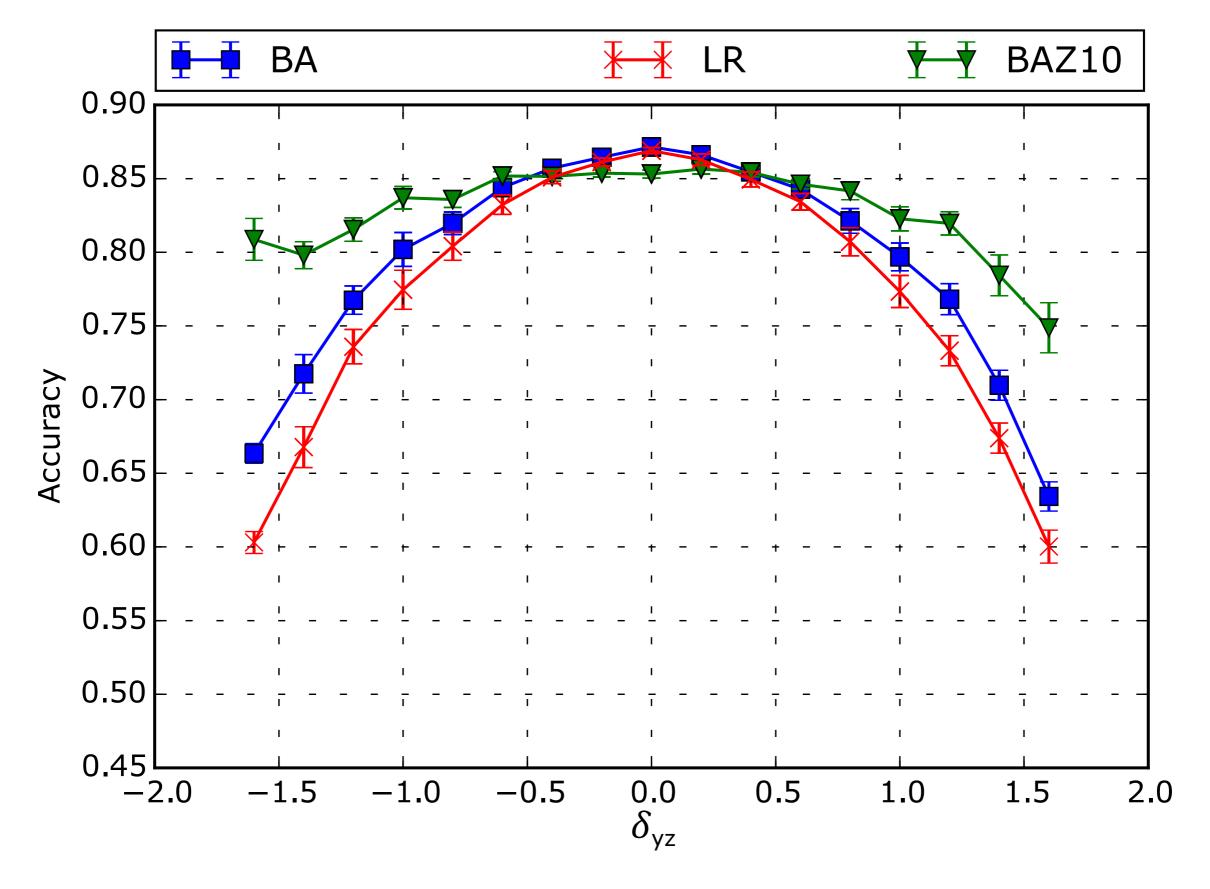
- Instead, estimate $p(y \mid \mathbf{x}, z)$
- At test time, when z is unknown, compute:

$$\sum_{z \in Z} p(y \mid \mathbf{x}, z) p(z)$$

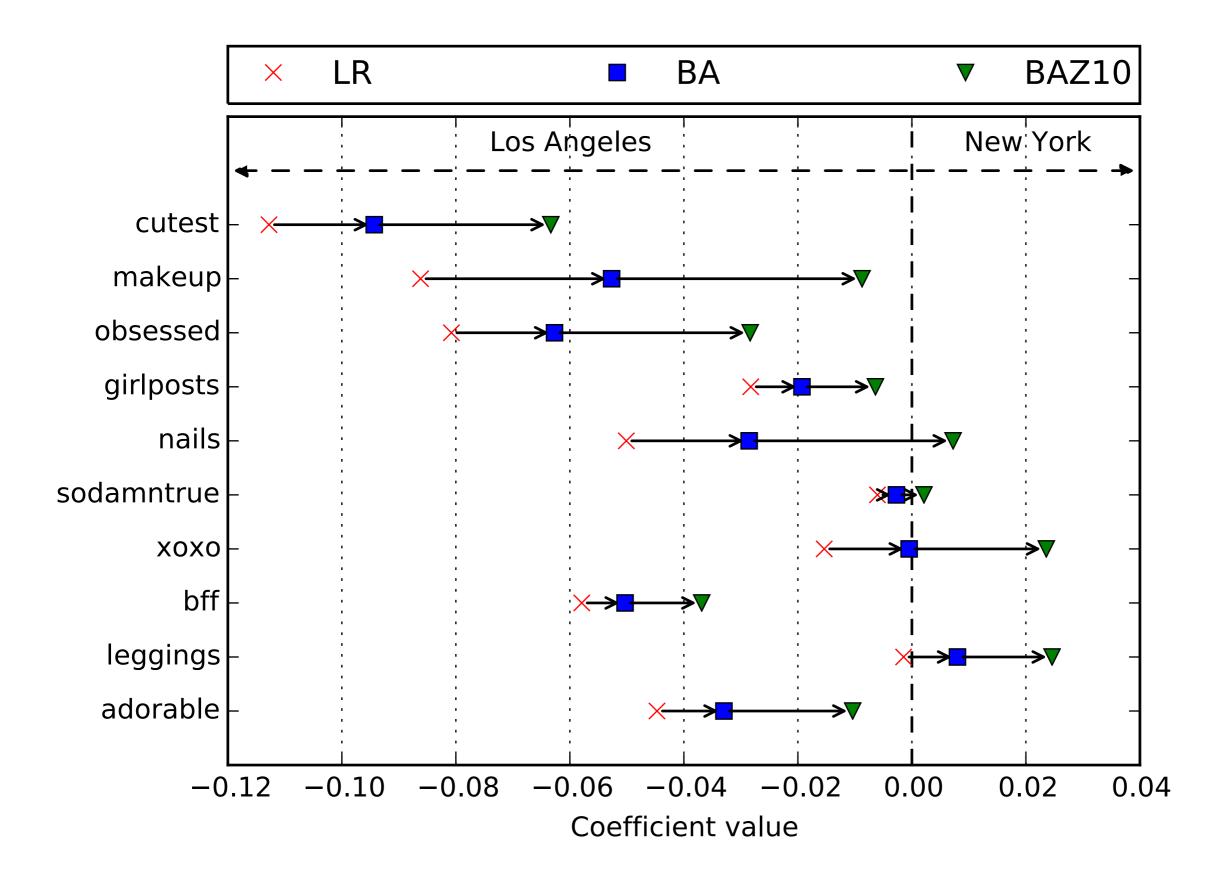
Experiments with confounders

- <u>Task</u>: predict the location of a Twitter user based on their tweets
 - Gender as a confounder: observed at training time, but not testing time

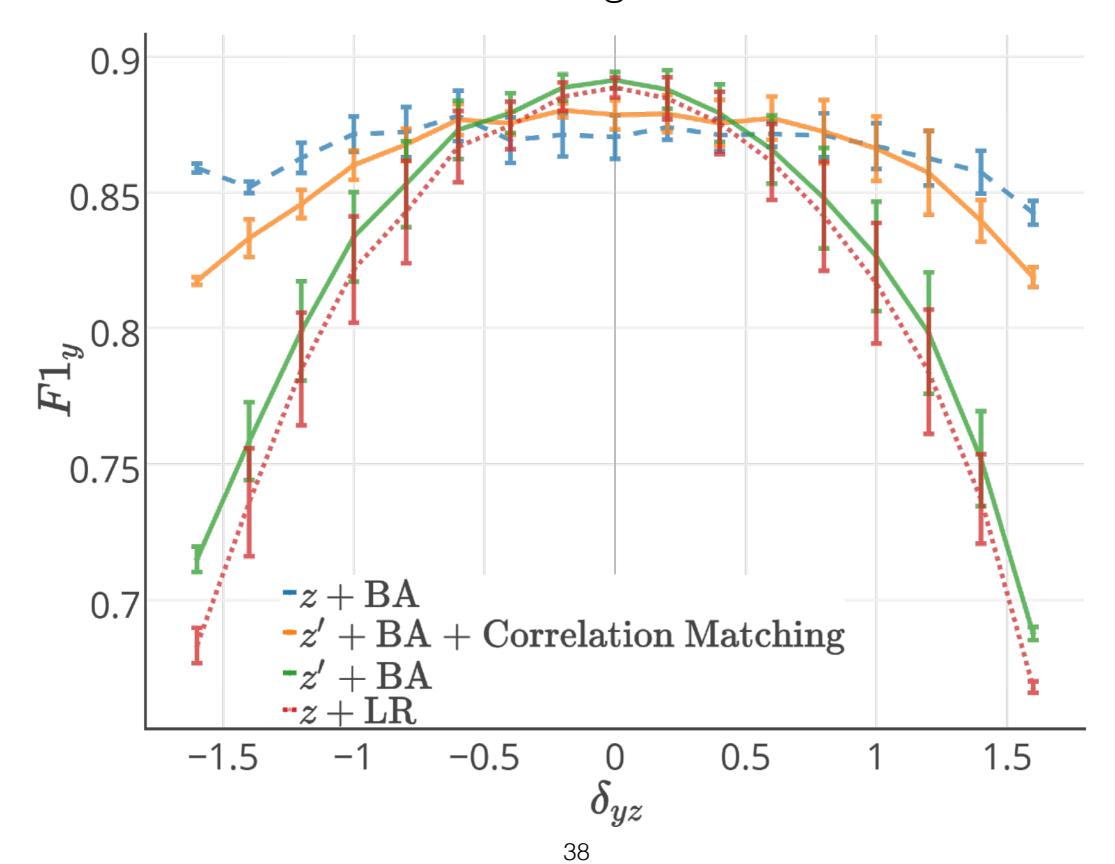
• To measure robustness to shift in p(y|z), sample different train/test splits to vary $[p_{train}(y|z) - p_{test}(y|z)]$



direction of shift in p(y|z)



Recently extended to cases where **z** is observed with measurement error at training time. [Landeiro & Culotta '17]



Conclusions: Confounds in classification

- Omitted variable bias can significantly degrade accuracy of a text classifier.
- Covariate adjustment makes classifier robust to changes in confounding relationships
- Allows us to train a classifier using small, possibly biased datasets

Code/data: https://github.com/tapilab/aaai-2016-robust

Predicting Twitter user demographics with distant supervision

[Culotta et. al '16; Ardehaly & Culotta '16]

Learning from label proportions

- Often, it is difficult/unethical to get labeled training data at the instance level
 - E.g., income level of a Twitter user?

- However, it is often possible to associate population statistics with bags of instances.
 - E.g.:
 - 27% of households in Cook County have income >\$100k
 - Create a bag of all Twitter users from Cook County

Can we fit a model to population statistics, then use it to classify individual instances?

Learning from label proportions

- Simple model: ridge regression used for classification
- For bag i, we have
 - The mean feature vector z_i
 - The given label proportions \widetilde{p}_i
- · Set parameters by minimizing penalized squared error

$$E(\theta) = \sum_{i} (\tilde{p}_i - z_i^T \theta)^2 + \lambda ||\theta||^2$$

• Classify a new instance as positive if $x^T \theta > .5$

LLP Experiments

- Predict 7 attributes of Twitter users
 - gender, age, income, political party, education, children, race/ ethnicity
- Use population statistics from QuantCast.com
 - E.g., 19% of web users who visited <u>lifehacker.com</u> have a graduate degree
 - Bag consists of a sample of users that follow @lifehacker on Twitter

• In [Ardehaly & Culotta 'I 6], we also experiment with county statistics and name statistics.

LLP Experiments

FI results on validation data

	Friends		Text		Friends + Text	
	distant	full	distant	full	distant	full
Gender	.75	.66	.86	.84	.87	.84
Ethnicity	.60	.68	.86	.86	.81	.86
Politics	.80	.83	.56	.73	.74	.73
Average	.72	.72	.76	.81	.81	.81

distant = LLP model

full = fully supervised logistic

full = fully supervised logistic regression

friends = use features over who a user follows

text = use features over tweets from the user

Top follow features

Education	No College	YouTube, PlayStation, RockstarGames, Xbox, IGN
	College	StephenAtHome, WIRED, ConanOBrien, mashable
	Grad School	nytimes, WSJ, NewYorker, TheEconomist, washingtonpost
	•	
Children	No Kids	NewYorker, StephenAtHome, nytimes, maddow, pitchfork
	Has Kids	parenting, parentsmagazine, HuffPostParents,
		TheEllenShow, thepioneerwoman
Income	\$0-50k	YouTube, PlayStation, IGN, RockstarGames, Drake
Income		
	\$50-100k	AdamSchefter, cnnbrk, SportsCenter, espn, ErinAndrews
	\$100-150k	WSJ, espn, AdamSchefter, SportsCenter, ErinAndrews
	150k+	WSJ, TheEconomist, Forbes, nytimes, business

Top text features

18-24	d, haha, album, x, xd, _:, actually, stream, wanna, im
25 - 34	super, dc, baby, definitely, nba, pregnancy, wedding, even,
	entire, nyc
35-44	star, fans, kids, tv, bike, mind, store, awesome, screen, son
45-54	wow, vote, american, comes, ca, santa, county, boys, nice,
	high
55-64	vote, golf, red, american, country, north, county, holiday,
	smile, 99,999
65+	vote, golf, @foxnews, holiday, may, american, he, family,
	north, national
	25-34 35-44 45-54 55-64

Politics	Democrat	women, u, ain't, nyc, equality, la, voice, seattle, dc, @nytimes	
	Republican	@foxnews, christmas, #tcot, football, county, morning,	
		family, christians, country, obama's	

Conclusions: Learning from label proportions

- To avoid annotating data, we can fit models to population statistics matched with bags of instances.
- Often meet or exceed accuracy of models trained on instance-level annotations.

Code/data: https://github.com/tapilab/jair-2016-demographics

Overall Conclusions

- · It's difficult to get the data you want for text classification
- Instead, take the imperfect data you have and adapt the models to work with it
 - Domain adaptation: transfer model from one type of data to another
 - Robustness to confounders
 - Learning from label proportions

Thanks!

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http://cs.iit.edu/~culotta

https://github.com/tapilab