



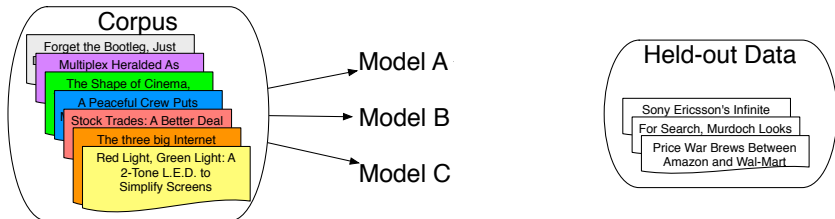
## Topic Models

Advanced Machine Learning for NLP

Jordan Boyd-Graber

EVALUATION

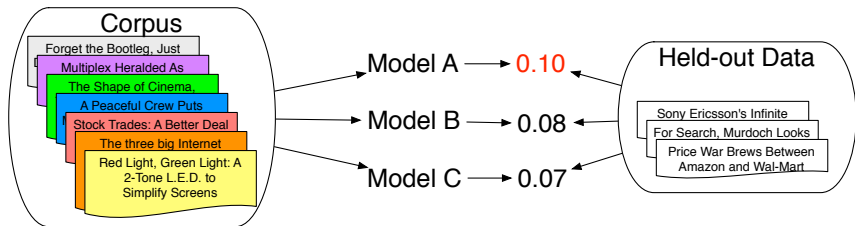
## Evaluation



$$P(\mathbf{w} | \mathbf{w}', \mathbf{z}', \alpha \mathbf{m}, \beta \mathbf{u}) = \sum_{\mathbf{z}} P(\mathbf{w}, \mathbf{z} | \mathbf{w}', \mathbf{z}', \alpha \mathbf{m}, \beta \mathbf{u})$$

How you compute it is important too

## Evaluation



Measures predictive power, not what the topics are

$$P(\mathbf{w} | \mathbf{w}', \mathbf{z}', \alpha \mathbf{m}, \beta \mathbf{u}) = \sum_{\mathbf{z}} P(\mathbf{w}, \mathbf{z} | \mathbf{w}', \mathbf{z}', \alpha \mathbf{m}, \beta \mathbf{u})$$

How you compute it is important too

### TOPIC 1

computer,  
technology,  
system,  
service, site,  
phone,  
internet,  
machine

### TOPIC 2

sell, sale,  
store, product,  
business,  
advertising,  
market,  
consumer

### TOPIC 3

play, film,  
movie, theater,  
production,  
star, director,  
stage

## Word Intrusion

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- 1 Take the highest probability words from a topic

### Original Topic

dog, cat, horse, pig, cow

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- 2 Take a high-probability word from another topic and add it

### Topic with Intruder

dog, cat, **apple**, horse, pig, cow

## Word Intrusion

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- 1 Take the highest probability words from a topic

### Original Topic

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- 2 Take a high-probability word from another topic and add it

### Topic with Intruder

dog, cat, **apple**, horse, pig, cow

- 3 We ask users to find the word that doesn't belong

### Hypothesis

If the topics are interpretable, users will consistently choose true intruder

## Word Intrusion

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1 / 10

crash

accident

board

agency

tibetan

safety

2 / 10

commercial

network

television

advertising

viewer

layoff

3 / 10

arrest

crime

inmate

pitcher

prison

death

4 / 10

hospital

doctor

health

care

medical

tradition



## Word Intrusion

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**1 / 10** Reveal additional response

crash	accident	board	agency	tibetan	safety
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**2 / 10**

commercial	network	television	advertising	viewer	layoff
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**3 / 10**

arrest	crime	inmate	pitcher	prison	death
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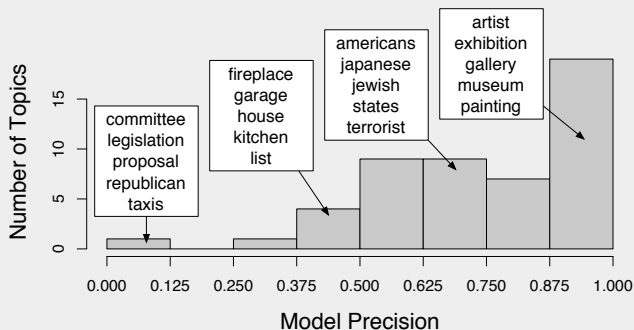
**4 / 10**

hospital	doctor	health	care	medical	tradition
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- Order of words was shuffled
- Which intruder was selected varied
- Model precision: percentage of users who clicked on intruder

## Word Intrusion: Which Topics are Interpretable?

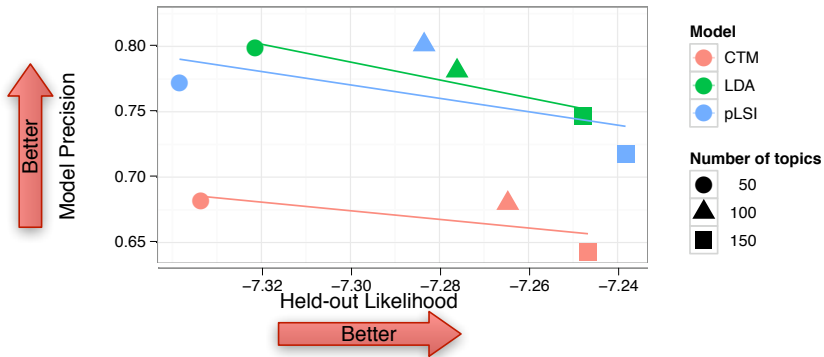
### New York Times, 50 LDA Topics



Model Precision: percentage of correct intruders found

## Interpretability and Likelihood

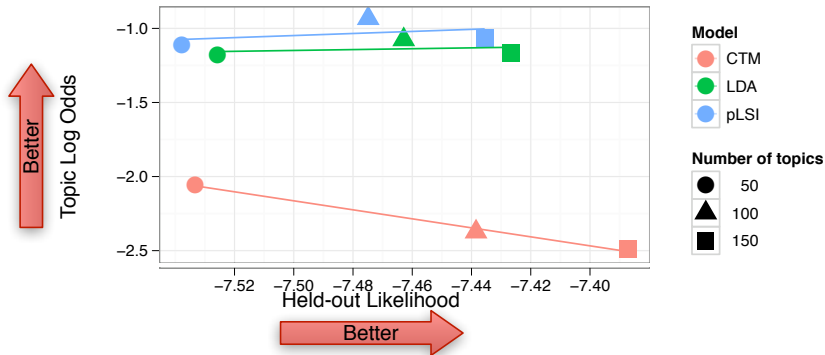
### Model Precision on New York Times



within a model, higher likelihood  $\neq$  higher interpretability

## Interpretability and Likelihood

Topic Log Odds on Wikipedia



across models, higher likelihood  $\neq$  higher interpretability

## Downstream Tasks

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- Classification
- Machine Translation
- Political Polarization/Framing

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