

# **Uptake Data Fellows Natural Language Processing Workshop**



# ***Hello! I'm Michael Miller Yoder***

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1

# **What is natural language processing (NLP)?**

*Computational processing of human language.*

**Examples:** machine translation, dialogue systems, question answering, speech recognition, search engines



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*Computational processing of human language.*

Often involves applying machine learning  
techniques to text or speech data



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Use case: **free-text survey data**



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Many questions are categorical (yes/no/maybe) or on a numerical scale.



## Use case: **free-text survey data**

Many questions are categorical (yes/no/maybe) or on a numerical scale. But others may be open text response ("please explain...")





# Topic modeling



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Statistical models for finding "topics" that occur in a collection of documents.



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Statistical models for finding "topics" that occur in a collection of documents. Common approach is Latent Dirichlet Allocation (LDA) [Blei et al, 2003]



## LDA (Latent Dirichlet Allocation)

- Unsupervised: no "true" topics



## LDA (Latent Dirichlet Allocation)

topic 0  
topic 1  
topic 2  
topic 3  
topic 4



- Unsupervised: no "true" topics
- Each document mixture of topics



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- Each document mixture of topics
- Each topic mixture of words



## LDA (Latent Dirichlet Allocation)

topic 0  
topic 1  
topic 2  
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topic 4



- Unsupervised: no "true" topics
- Each document mixture of topics
- Each topic mixture of words
- Based on word co-occurrence



## Exercise: **topic modeling**

### **tokenization**

What is a word?

What words count?





## Exercise: **topic modeling**

### **tokenization**

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What words count?

### **feature extraction**

Words to numbers  
(bag-of-words).



## Exercise: topic modeling

### tokenization

What is a word?  
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### feature extraction

Words to numbers  
(bag-of-words).

### LDA interpretation

Do some  
unsupervised ML!  
Play around, interpret  
results.

**BIKE  
PGH!**



**Data:**  
**Autonomous vehicle survey**





## Autonomous vehicle survey from cyclists and pedestrians

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### Context

Pittsburgh is a testing ground for AVs from Uber, ArgoAI and other companies.

### Bike Pittsburgh

Bike Pittsburgh, bike and pedestrian advocacy organization, made an online survey in [2017](#) and [2019](#).





Download:

[bit.ly/2EJ3kLv](https://bit.ly/2EJ3kLv)





## Choose your environment

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### Python

Jupyter Notebook:

[https://github.com/michaelmilleryoder/av-survey-topic-modeling/av-survey-topic-modeling\\_python.ipynb](https://github.com/michaelmilleryoder/av-survey-topic-modeling/av-survey-topic-modeling_python.ipynb)

### R

Jupyter Notebook:

[https://github.com/michaelmilleryoder/av-survey-topic-modeling/av-survey-topic-modeling\\_r.ipynb](https://github.com/michaelmilleryoder/av-survey-topic-modeling/av-survey-topic-modeling_r.ipynb)



## Choose a text field

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- interaction\_details
- positive\_av\_interaction
- negative\_av\_interaction
- other\_av\_regulations
- elaborate\_bikepgh\_position
- other\_comments



## Workflow

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- Tokenize: split into words
- Extract features: words to word IDs (bag-of-words model)
- Run LDA with varying numbers of topics
- Interpret topics
  - Look at high-ranking words for each topic
  - Look at high-ranking documents for each topic





## ...If you get to it

- Correlate topics with categorical and numerical fields
- Predict non-text fields with a machine learning algorithm such as logistic regression from topic distributions or text features
- Look into the [Structural Topic Model](#) (R package)



# Thanks!

*Any* **questions** ?

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