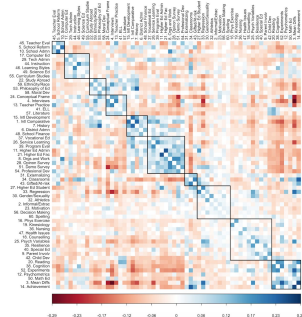


# Topic Models

Sebastian Munoz-Najar

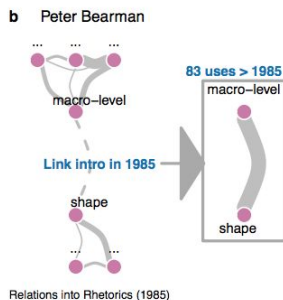
# My research with topic models

## Education Research 1980-2010 AERJ '19



How did education research discourse change over time? **TMs for research trends.**

## Diversity Breeds Innovation Arxiv '19



How are grad students rewarded for innovation? **TMs for phrase extraction.**

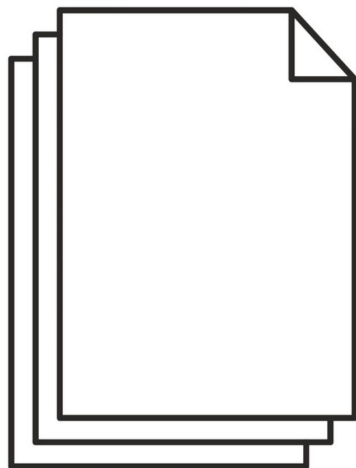
# Today

- What is a topic model?
  - Intuition
  - Inputs and Outputs
  - Notation
- Kinds of topic models
- Validation
  - Exclusivity
  - Coherence
- Examples

What is a topic model?

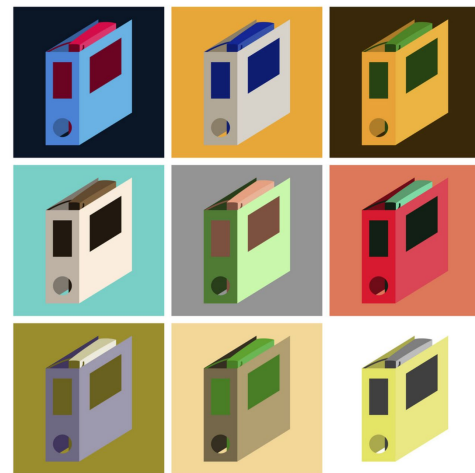
# The short answer

You have...



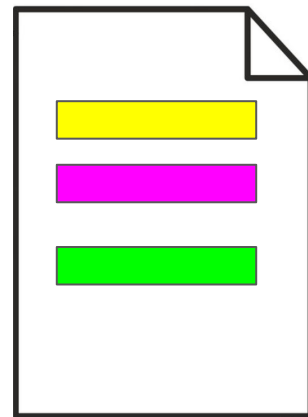
many texts

What the  
collection is  
about



You want  
to know...

What each  
text is about



# Some involved answers

Technical Answer	Translation
Unsupervised Learning Algorithm	Classifies text without a predetermined classification scheme
Generative Probabilistic Model	Answers, “How is a text put together?” with probability distributions
Mixture Model	Represents texts as combinations of topics

# Intuition

What is this text about? ( i.e. Big themes)

This study used structural equation modeling to test a model of ethnic identity development among 513 Mexican-origin adolescents living in the United States.  
(Umana-Taylor & Fine, 2004)



What did you need to know about  
language and the world?

# What you know

- Language
  - Grammar
  - Vocabulary
  - Pragmatics
- Domain Knowledge
  - Intertextuality
  - Style
- Context Knowledge
  - Year
  - Author
  - Institution

# What a TM knows (approximately)

- Language
  - ~~Grammar~~
  - **Vocabulary**
  - **~Pragmatics**
- Domain Knowledge
  - **~Intertextuality**
  - ~~Style~~
- Context Knowledge
  - **Year\***
  - **Author\***
  - **Institution\***

~ : approximates

\* : depends on model

# What is this text about? ( i.e. Big themes)

**model,**  
structural\_equation\_model,  
factor\_analysis,  
....

**adolescent,**  
identity\_development,  
socialization  
....

**ethnic,**  
mexican-origin,  
chicano  
....

This study used **structural equation modeling** to test a **model** of **ethnic identity development** among 513 **Mexican-origin adolescents** living in the United States.  
(Umana-Taylor, 2001)

# What is this text about? ( i.e. Big themes)

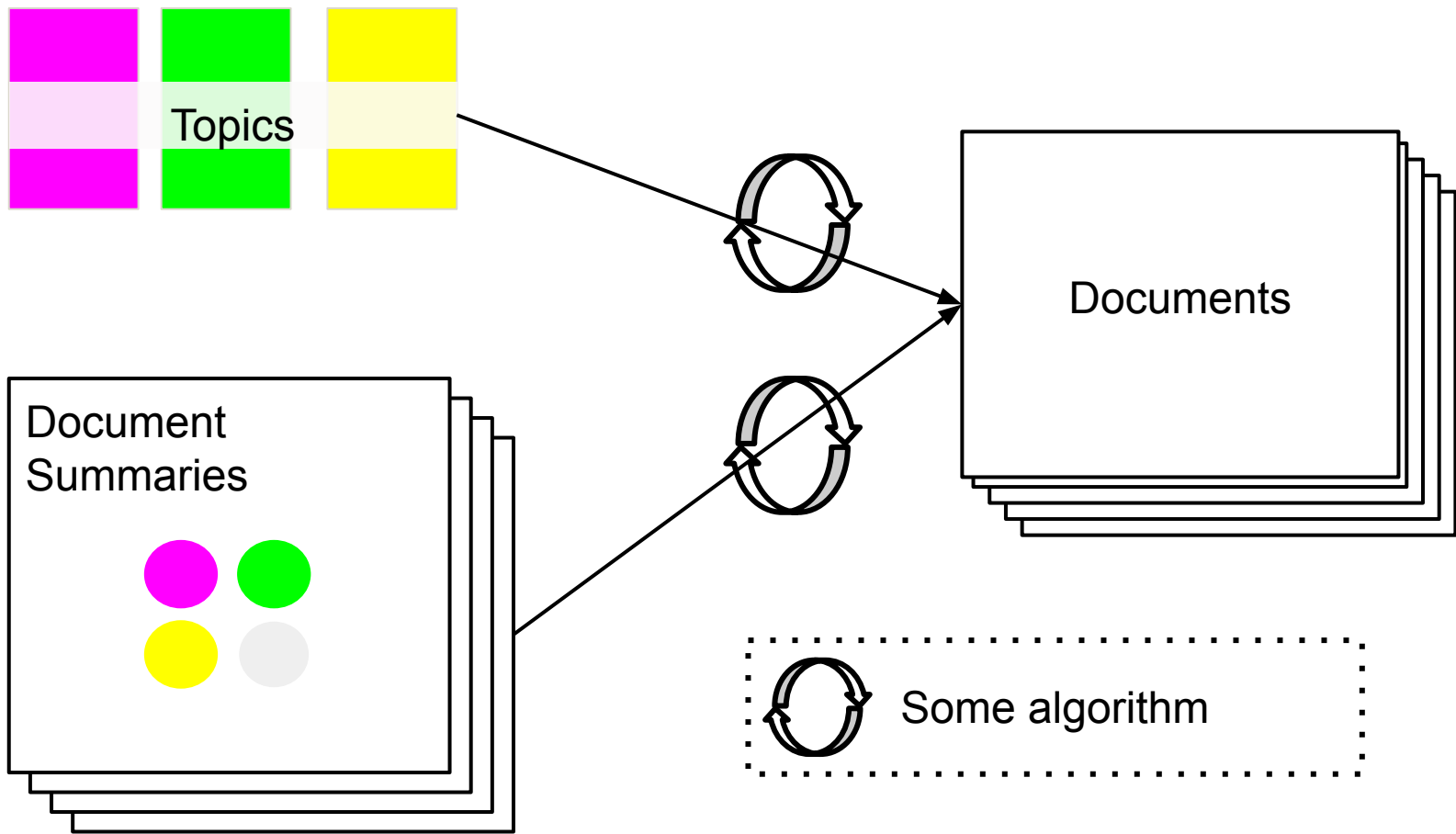
Umana-Taylor & Fine:

14  
%

14  
%

9%

This study used structural equation modeling to test a model of ethnic identity development among 513 Mexican-origin adolescents living in the United States.  
(Umana-Taylor, 2001)



Inputs and outputs

# Input

- Document-Term Matrix ( $M \times V$ )
  - $M$  documents
  - $V$  words
- Number of Topics ( $K$ )
- Covariates ( $M \times X$ )

ID	educ	...	stud	V
1	3	...	1	10
...	...	...	...	...
M	5	...	5	0



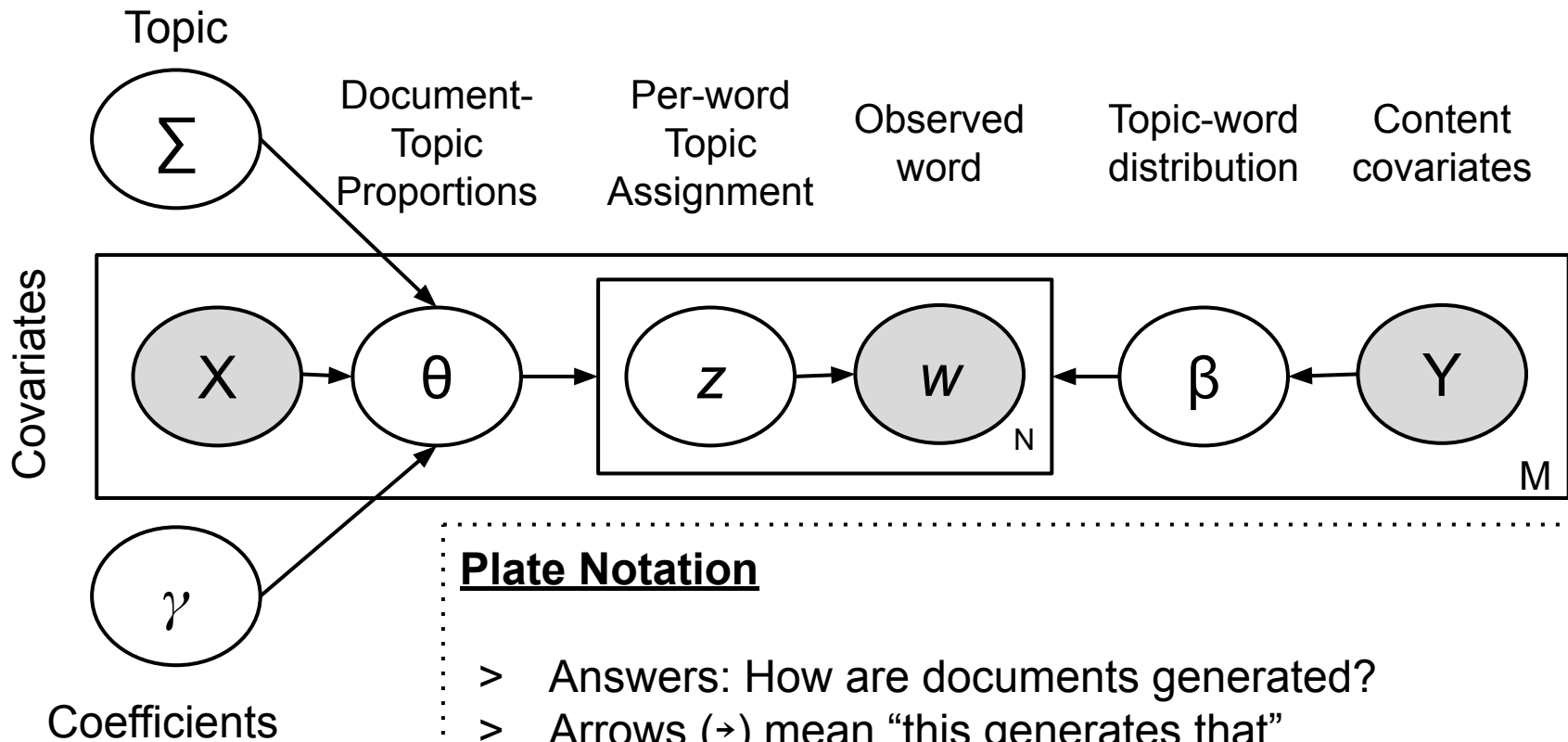
# Output

- Topics ( $K \times V$ )
  - Word Proportions
- Document Summaries ( $M \times K$ )
  - Topic Proportions

ID	Topic 1	...	Topic K
1	.001	...	.2
...	...	...	...
M	.1	...	.00009

Topics	educ	...	stud	V
1	.2	...	.2	.001
...	...	...	...	...
K	.2	...	.001	.6

# Notation



### Plate Notation

- > Answers: How are documents generated?
- > Arrows ( $\rightarrow$ ) mean “this generates that”
- > Plates say how many times you repeat a process:
  - ◆  $M$  documents,  $N$  terms per document

# Kinds of TM

Name	Gist	Learns
<a href="#"><u>LDA</u></a>	Vanilla	Topics, Document Summaries
<a href="#"><u>Correlational TM</u></a>	Topics are correlated!	Topic Correlations
<a href="#"><u>Dynamic TM</u></a>	Words in topics change!	Dynamic Topics
<a href="#"><u>Hierarchical TM</u></a>	Topics have hierarchy!	Topic Hierarchy
<a href="#"><u>Bayesian Hierarchical TM</u></a>	Documents are nested in authors!	Author topical priorities
<a href="#"><u>Labelled LDA</u></a>	Documents have labels!	Labelled Topics
<a href="#"><u>Relational TM</u></a> , <a href="#"><u>Topic-Link LDA</u></a>	Documents have relations!	Links
<a href="#"><u>Structural TM</u></a>	Documents have all kinds of metadata!	Coefficients of covariates

Validation

Goal: Select Number of Topics ( $K$ )

# What is wrong with these topics?

**education,**  
**model,**  
structural\_equation\_model,  
factor\_analysis,

**education,**  
**adolescent,**  
identity\_development,  
socialization

**education**  
**ethnic,**  
mexican-origin,  
chicano



# What is wrong with these topics?

**education,  
model,  
structural\_equation\_model,  
factor\_analysis,**

**education,  
adolescent,  
identity\_development,  
socialization**

**education,  
ethnic,  
mexican-origin,  
chicano**

They lack exclusivity!

# What is wrong with this topic?

**All Documents:**

**model,**  
structural\_equation\_model,  
factor\_analysis, ...

model,  
factor\_analysis,

model,  
structural\_equation\_model

model,  
factor\_analysis

# What is wrong with this topic?

**All Documents:**

**model,**  
structural\_equation\_model,  
factor\_analysis,

model  
factor\_analysis,

model,  
structural\_equation\_model

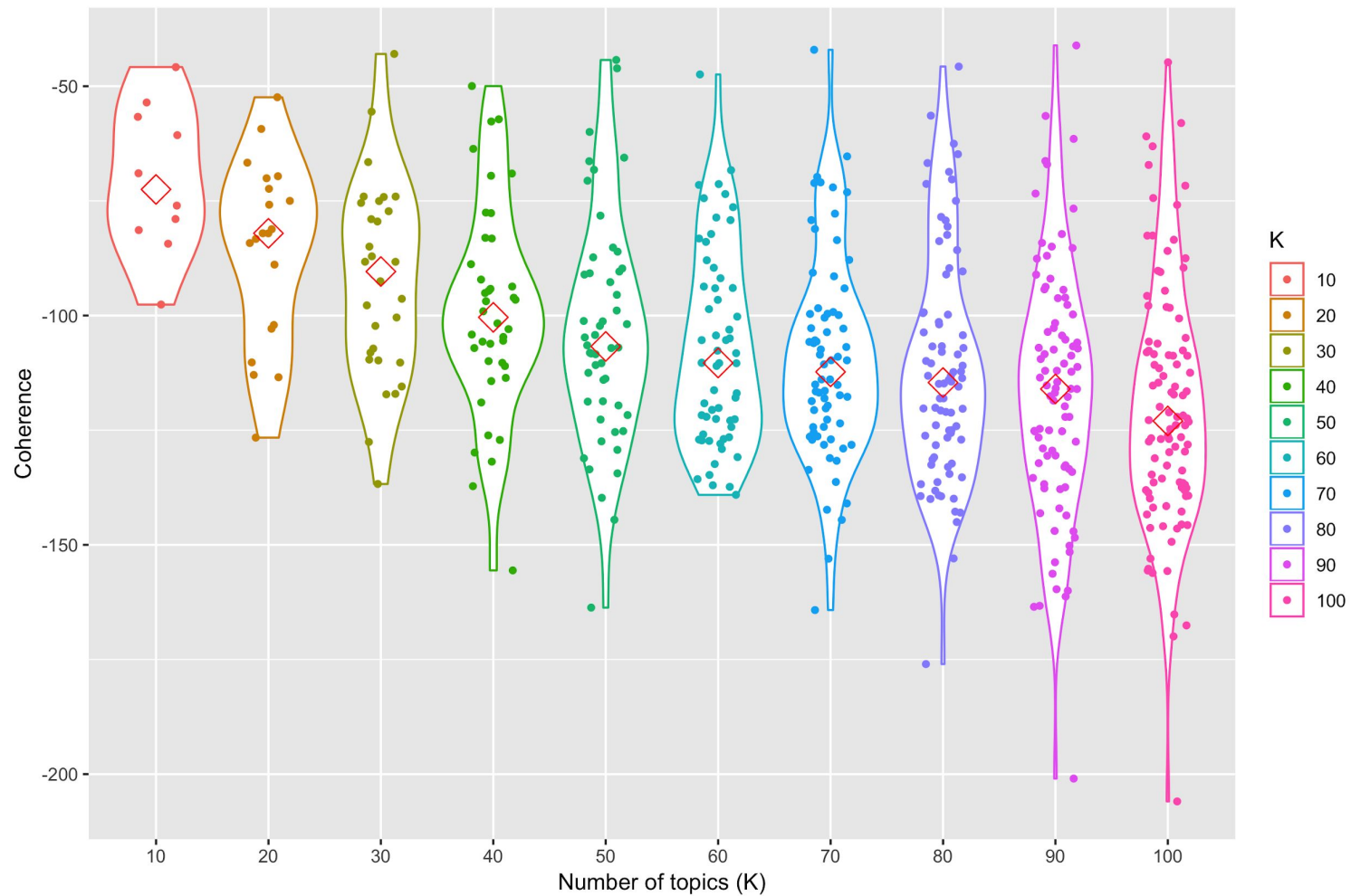
model,  
factor\_analysis

They lack coherence!

Can we trivially maximize these metrics?

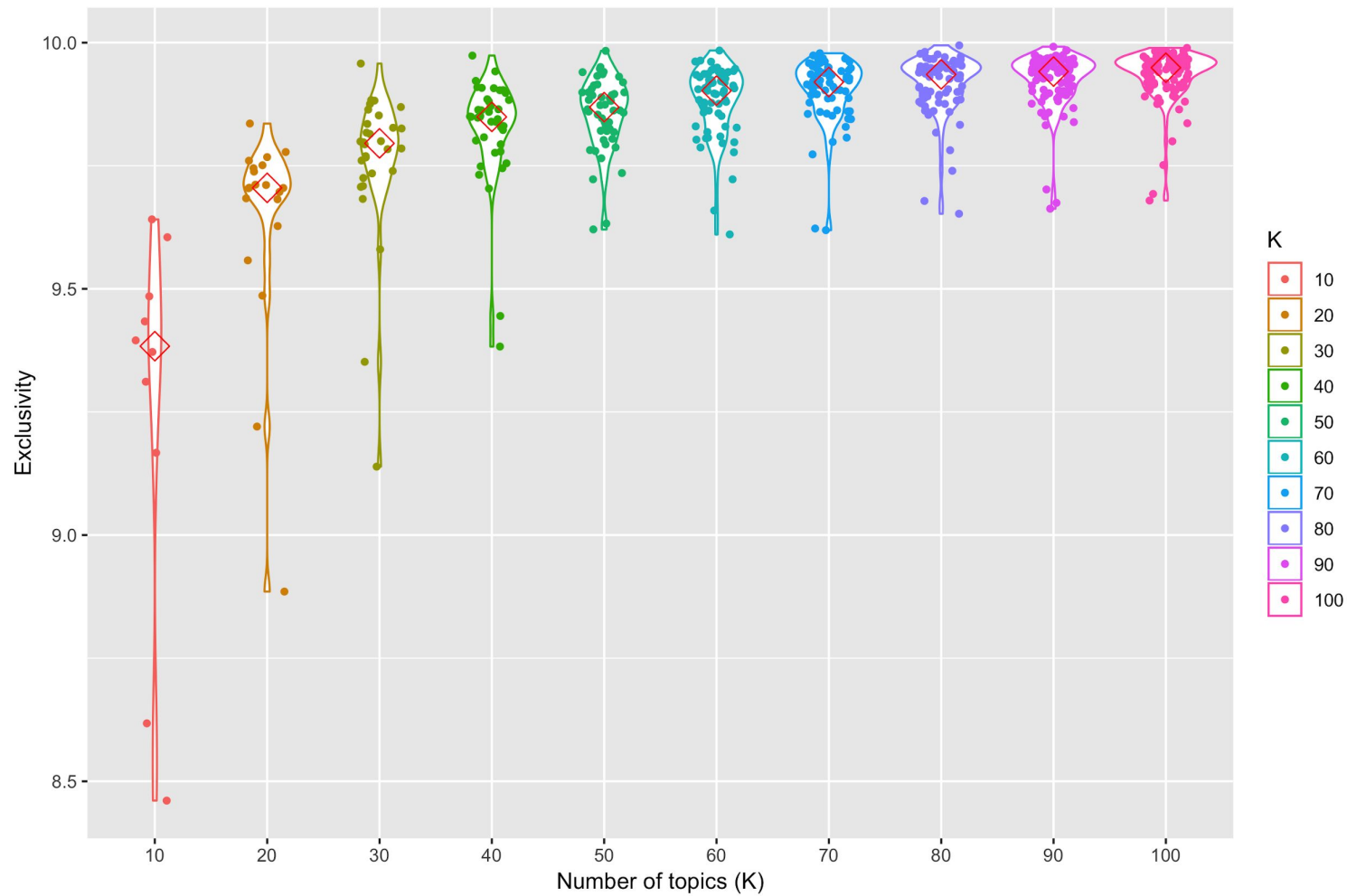
## Model Coherence

Do the pairs of most probable words in each topic have high co-document frequencies?



## Model Exclusivity

Do most probable words in topics tend also to be exclusive to them?



# Examples

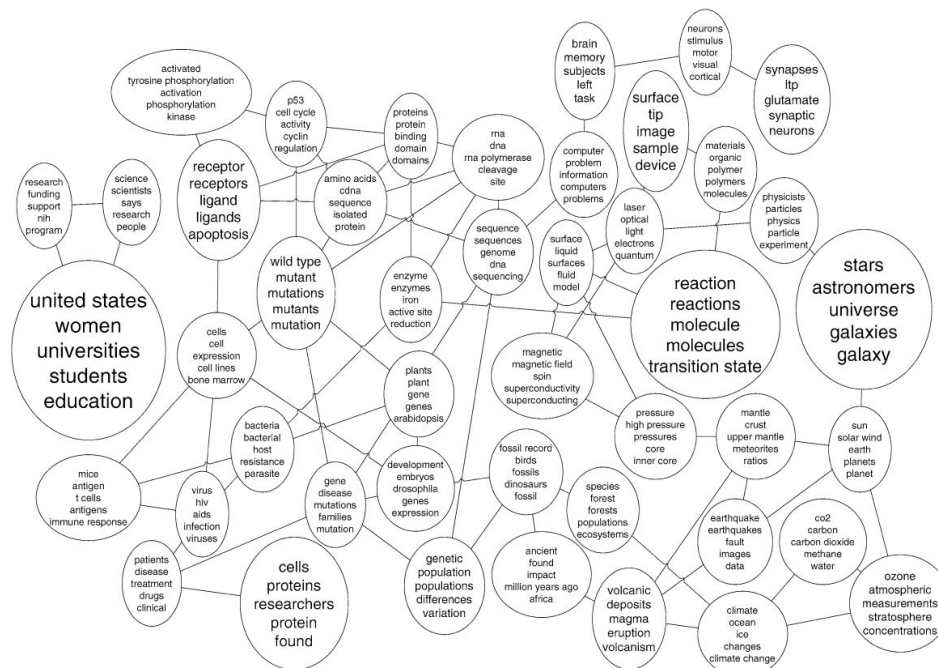
## David Blei and John Lafferty (2007), A Correlated Topic Model of Science

**Question:** Can we model topic correlations?

### Model: Correlational Topic Model

**Data:** articles from Science published from 1990–1999

**Result:** Yes





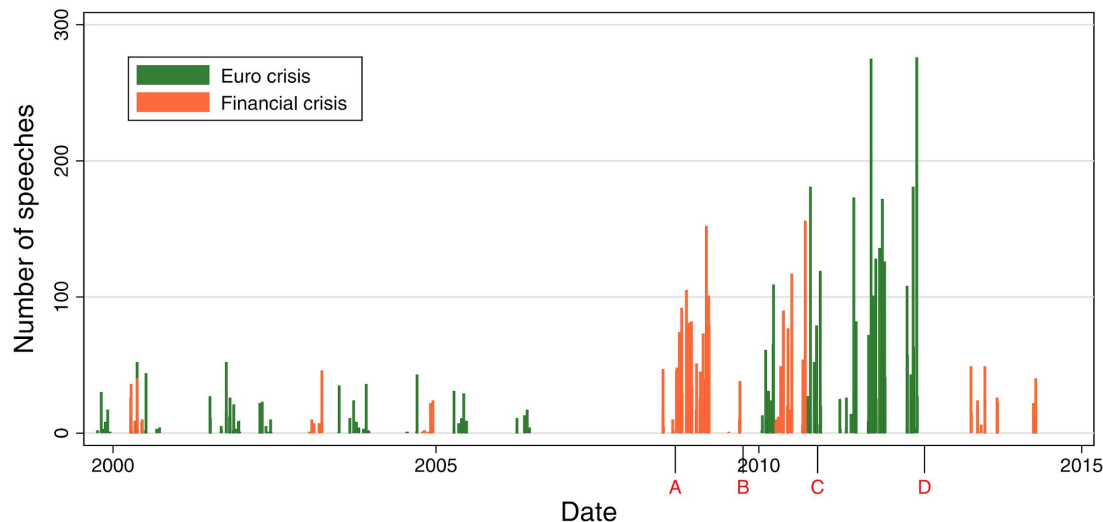
## Derek Greene and James P., (2017) Cross Exploring the Political Agenda of the European Parliament

**Question:** How has the political agenda of the European Parliament (EP) plenary evolved over time?

**Model:** Dynamic Topic Model

**Data:** all English language legislative speeches in the EP plenary from the period 1999 to 2014

**Result:** Agenda responds to external shocks!



**Figure 7.** “Financial & Euro crises” dynamic topics.

## Justin Farrell (2016), Corporate funding and ideological polarization about climate change

**Question:** How does corporate funding influence the production and content of polarization efforts in climate change debates?

**Model:** Structural Topic Model

**Data:** every text about climate change produced by 164 organizations between 1993 and 2013

**Result:** organizations that received corporate funding were more likely to have written and disseminated contrarian texts.

