# Who wrote it?

Speech Technologies Project, Spring 2016



Lucas Rodés

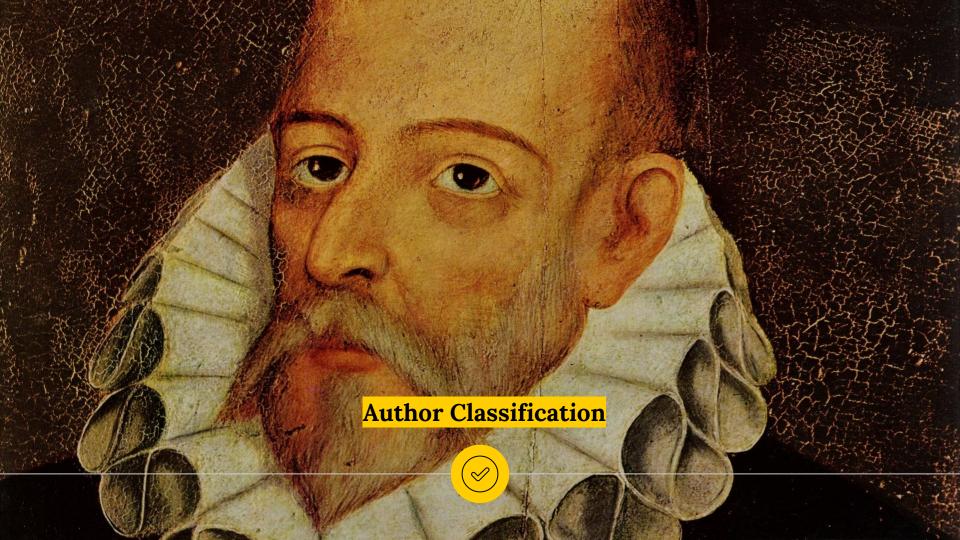
Sergi Liesegang

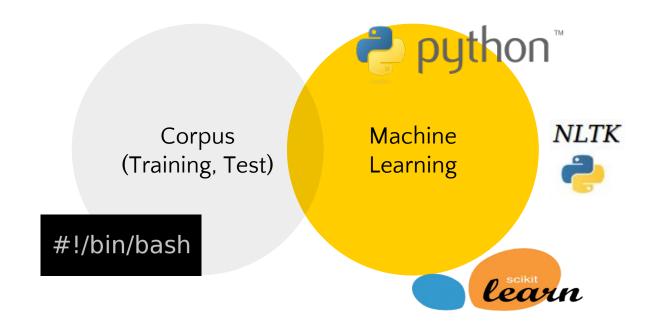
# Outline

- Introduction
- Corpus
- Classifiers
- Results
- Conclusions. Future Work.

### 1 Introduction

primero que gozó el fruto de enteramente, como deseaba, y sido otro mi deseo que po aborrecimiento de los mos fingidas y disparatada les libros Spanish Literature Caballería i verdade o don (2)





# 2 — Corpus

### **Database**

Author	Epoch	Genre
Miguel de Cervantes	Renaissance (C16-17th)	Novel - Poetry - Drama
P. Calderón de la Barca	Golden Age - Baroque (C17th)	Drama - Poetry
Felix <b>Lope de Vega</b>	Golden Age - Baroque (C17th)	Novel - Poetry - Drama
Francisco de <b>Quevedo</b>	Baroque (C17th)	Poetry
G. Adolfo <b>Becquer</b>	Romanticism (C19th)	Poetry - Novel
Emilia <b>Pardo</b> Bazan	Realism - Naturalism (C19th)	Novel
Benito Pérez <b>Galdós</b>	Realism (C19th)	Novel - Drama
Federico Garcia Lorca	Generation of 27 (C20th)	Poetry - Drama

### 2 Database

	Author	Epoch	Genre
correlated	Miguel de Cervantes	Renaissance (C16-17th)	Novel - Poetry - Drama
	P. Calderón de la Barca	Golden Age - Baroque (C17th)	Drama - Poetry
	Felix Lope de Vega	Golden Age - Baroque (C17th)	Novel - Poetry - Drama
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### 3 — Classifiers

3 — Naive Bayes

**Mainstay** in text Processing

Simple Categorization

Based on Bag of Words model

Compares elements in texts

Simple Categorization

Based on Bag of Words model
Compares elements in texts

Sparse Matrix as input feature
Word occurrences

P(A|B)

### **Naive Bayes**

**Mainstay** in text Processing

**Simple** Categorization

P(B|A)P(A)

**P(B)** 

Based on **Bag of Words** model

Compares elements in texts

**Sparse Matrix** as input feature

Word occurrences

**Mainstay** in text Processing

**Simple** Categorization

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

Based on **Bag of Words** model

Compares elements in texts

Sparse Matrix as input feature

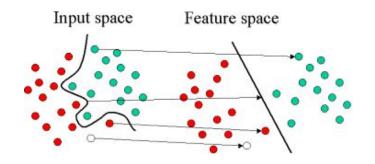
Word occurrences

Implemented in Python

Scikit-learn Library

### 3

### **Support Vector Machine**



Well-known classifier

Finds boundary with maximum class separability.

**Kernel** 

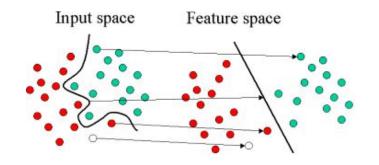
We used **Linear** 

**Sparse Matrix** as input feature

Word occurrences

### 3

### **Support Vector Machine**



Well-known classifier

Finds boundary with maximum class separability.

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**Sparse Matrix** as input feature

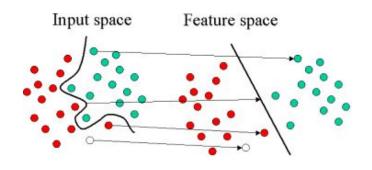
Word occurrences

**Grid Search** 

Optimization method to find best classifier parameters

### 3

### **Support Vector Machine**



#### Well-known classifier

Finds boundary with maximum class separability.

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Optimization method to find best classifier parameters

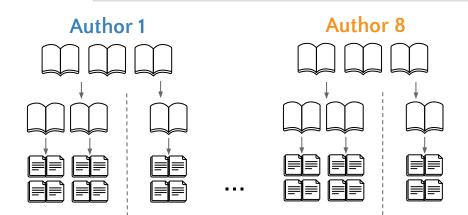
- 3 Approaches
- Download books



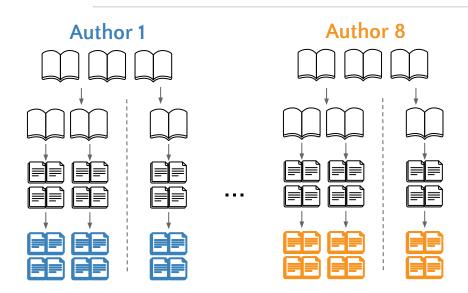
- Download books
- 66% Training, 33% Test



- Download books
- 66% Training, 33% Test
- Split in chunks of L lines

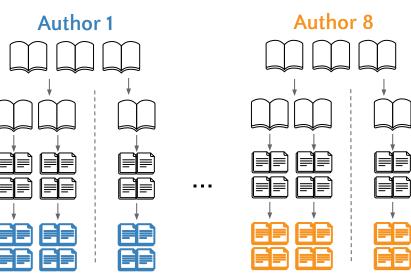


- Download books
- 66% Training, 33% Test
- Split in chunks of L lines
- Labelling

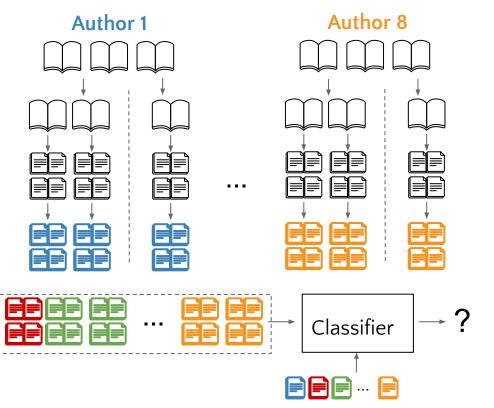


- Download books
- 66% Training, 33% Test
- **Split** in chunks of *L* lines
- Labelling
- Train Classifier





- Download books
- 66% Training, 33% Test
- **Split** in chunks of *L* lines
- Labelling
- Train Classifier
- Test Classifier



Popular in speech processing Based on Statistical Learning

Word statistics

#### Popular in speech processing

Based on Statistical Learning

Word statistics

#### N-gram Model

Unigram, Bigram, Trigram.

Popular in speech processing Based on Statistical Learning

Word statistics

N-gram Model

Unigram, Bigram,

Trigram.

**Smoothing Methods** 

We have used Add one Smoothing

Popular in speech processing

Based on **Statistical Learning** 

Word statistics

N-gram Model

Unigram, Bigram,

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**Smoothing Methods** 

We have used Add one Smoothing

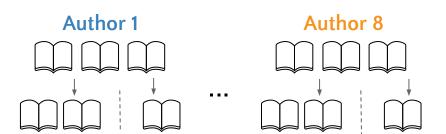
Implemented in Python

**NLTK Library** 

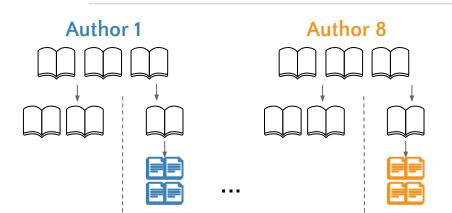
- 3 Approaches
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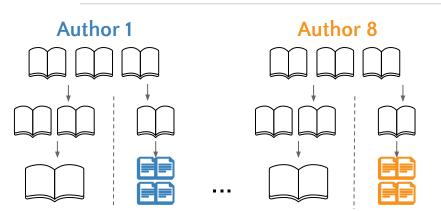
- 3 Approaches
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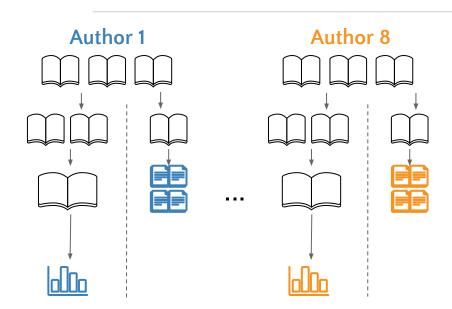
- Download books
- 66% Training, 33% Test
- Split Test in chunks & label



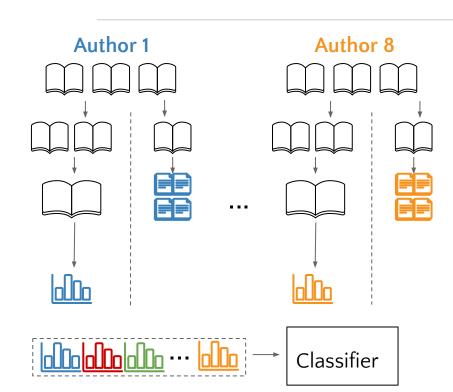
- Download books
- 66% Training, 33% Test
- Split Test in chunks & label
- Merge Training books



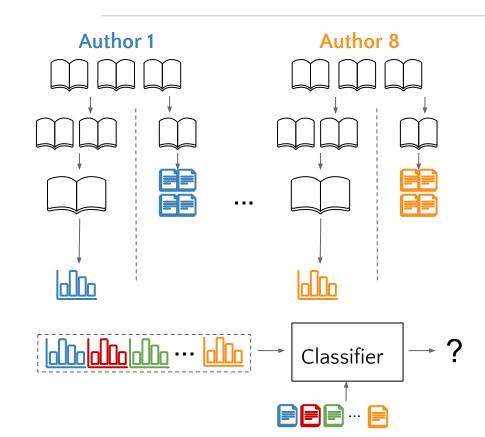
- Download books
- 66% Training, 33% Test
- Split Test in chunks & label
- Merge Training books
- Obtain Language Model



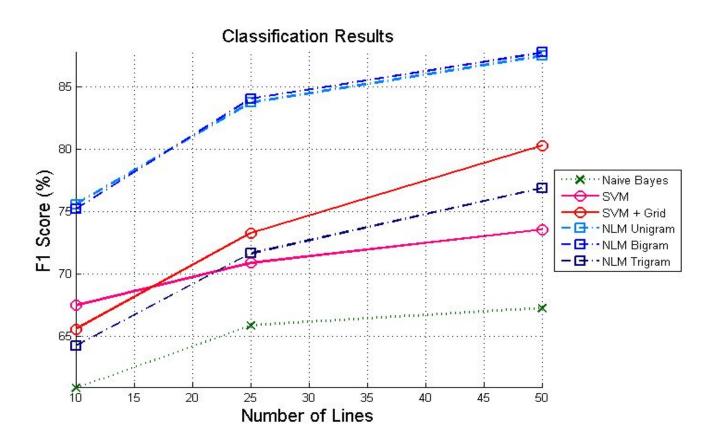
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- Perplexity based classifier

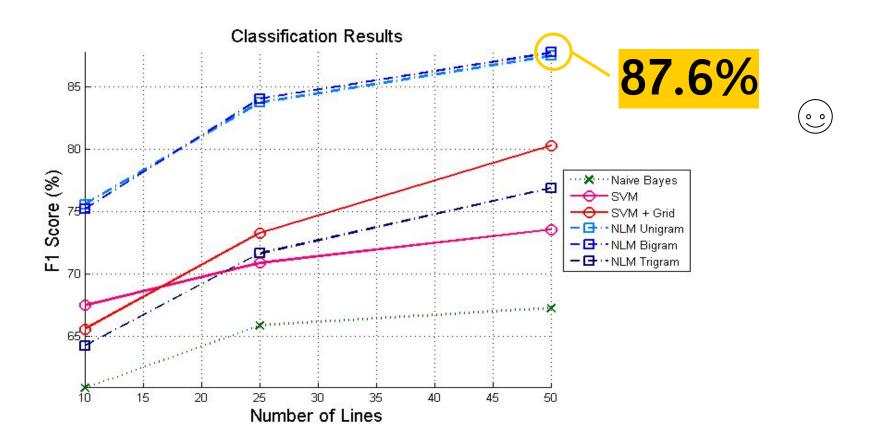


- Download books
- 66% Training, 33% Test
- Split Test in chunks & label
- Merge Training books
- Obtain Language Model
- Perplexity based classifier
- Test Classifier



### 4 — Results





### Conclusions.

**Future Work** 

- NLM is, by far, the best option
- Larger chunks (more lines) lead to better results
- Correlation represents a problem in simple approaches

- **Increase** database (training and test)
- Further analyze relation between chunk size and accuracy
- Equilibrate number of samples per author
- **Explore** other **smoothing** methods of NLM
- Use of **Neural Networks**



# Thanks!



Special thanks to all the people who made and released these awesome resources for free:

- Presentation template by <u>SlidesCarnival</u>
- Photographs by <u>Unsplash</u>

To evaluate the system

 $Precision = \frac{}{True\ Positives + False\ Positives}$ 

True Positives

 $Recall = \frac{True \ Positives}{True \ Positives + False \ Negatives}$ 

More Compact way:

F<sub>1</sub> Score = 
$$2 \cdot \frac{\text{Precision} \cdot \text{Recall}}{\text{Precision} + \text{Recall}}$$