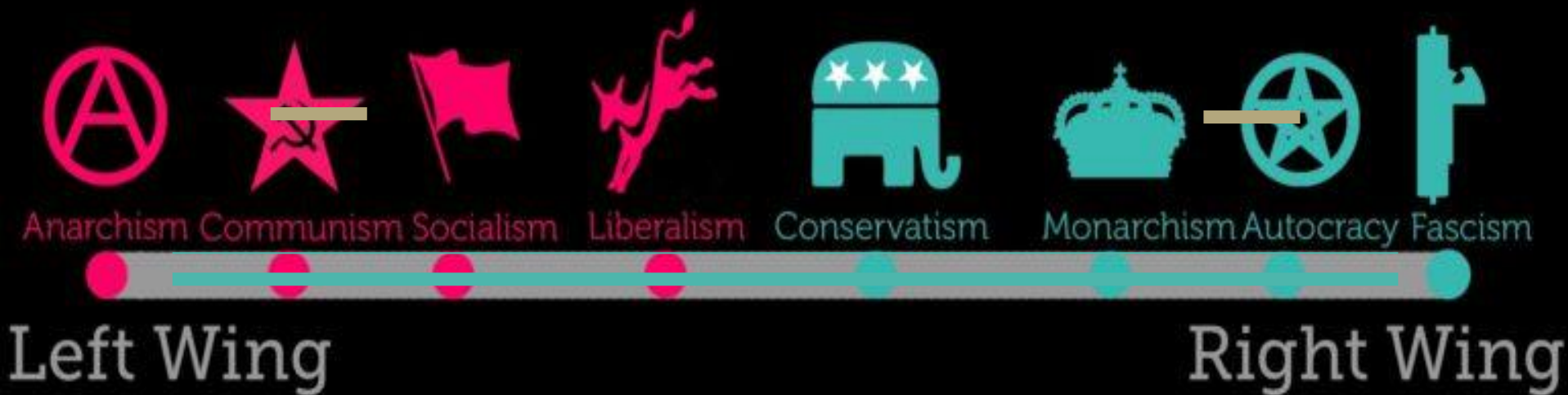


# Ideology detection



# The words are important!

## Political Lexicon

From the sophists vs Socrates to Cambridge Analytica, From the speech in the street to the tweet: The art of manipulation of language.

## Text classification

Opinion mining and text classification are still very restricted. Text, words and thoughts are dynamic and fluid.

# The goal

Detect the ideology  
of a sentence

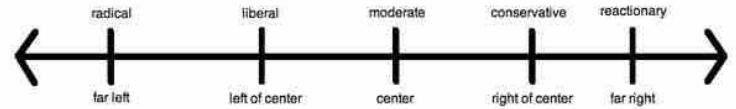
Using quotes of people with a  
strong political bias try to  
identify and capture key-words  
and pattern to detect ideology  
hidden behind

---

# The Dataset

# The political spectrum

Authoritarianism			
Stalin Communist Dictators	Generic Tyrants		Nazis Fascists
State Socialism	Social Democracy	Democrat Party	Republican Party
Democratic Socialism	Social Liberalism	Conservative Liberalism	Tea Party
Libertarian Socialism	Classical Liberalism		Right-wing Libertarianism and Minarchism
Liberalism			



# Data: Selection and Criteria

## Criteria

- Research of not politician people with strong political bias
- English mother tongue
- Wikipedia confirmation of the bias
- Allsides.com
- Not get involved

## Selection

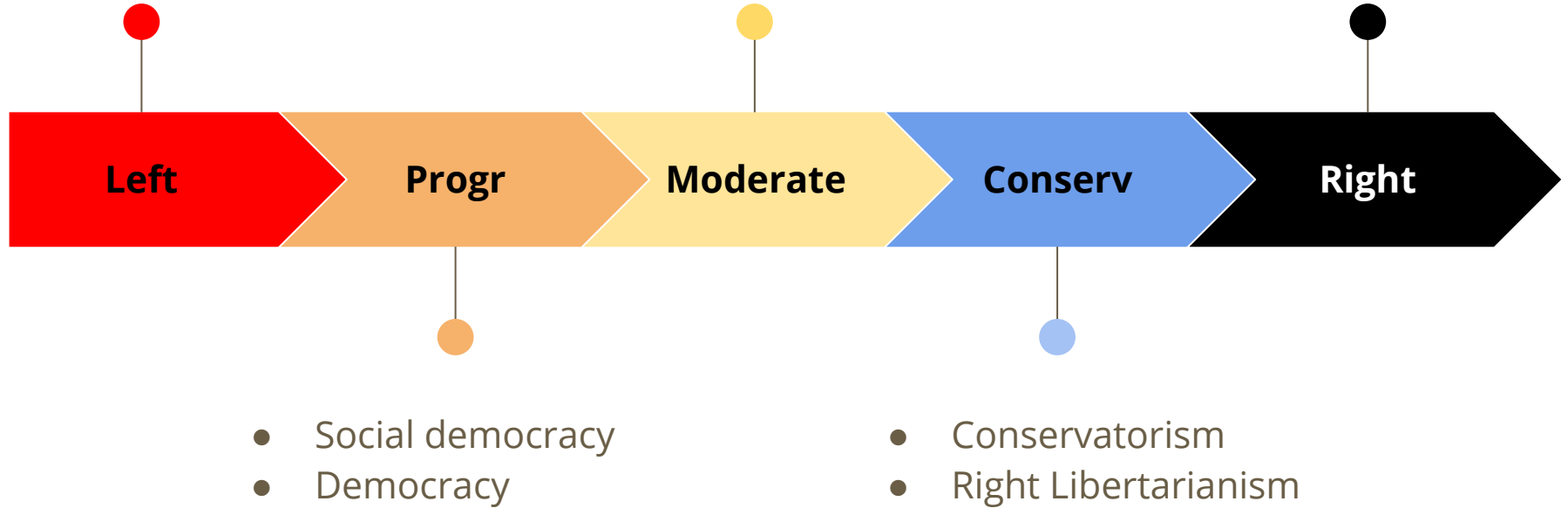
- Wikiquote
- Goodreads
- BrainQuote
- Random sample to balance classes

# The target

- Communism
- Socialism
- Ambientalism
- Left Anarchism

- Indipendent
- Popularism
- Christian democracy
- Centre Libertarianism

- Fascism
- Alt-Right



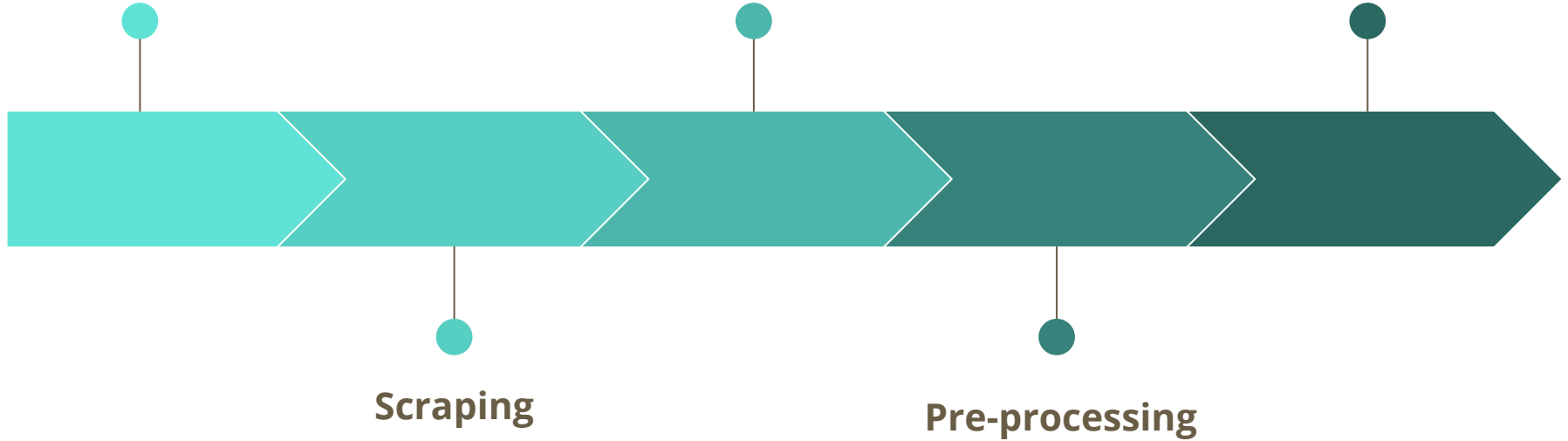


# First steps

Research

Quote in the quote

Lemmatization



## Balance

Selection

Baseline

- The weakest category choose the number of the sample
- External Dataset with four new authors for each category

0.2

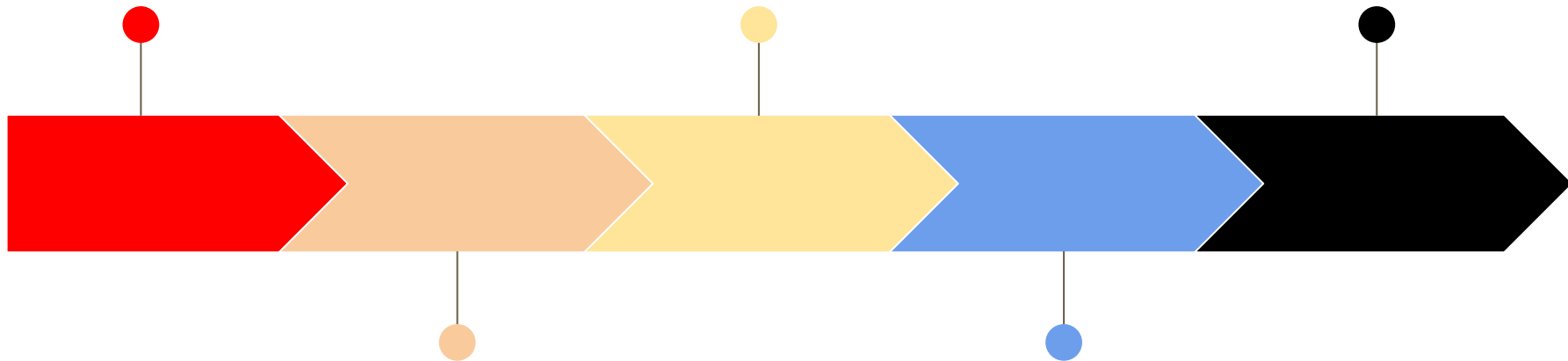
# The Vocabulary

- 500 000 total words (after stopwords)
  - 27000 unique
  - Left is the most chatty: 115000 words
  - Moderate the quiet one: 83000
  - Longest quote 761 words from Right
  - 140 Authors in total
  - Right most represented
-

- 115000 words
- 16000 unique

- 83000 words
- 10000 unique

- 106000 words
- 13000 unique



- 91000 words
- 10000 unique

- 92000 words
- 13000 unique

# Confusion will be my epitaph

	Left	Progressive	Moderate	Conservative	Right
0	(people, 368)	(people, 302)	(people, 404)	(people, 355)	(people, 473)
1	(world, 271)	(state, 191)	(world, 230)	(government, 212)	(state, 226)
2	(human, 248)	(world, 170)	(make, 184)	(american, 151)	(world, 211)
3	(state, 234)	(time, 169)	(thing, 163)	(right, 151)	(american, 193)
4	(life, 216)	(government, 160)	(life, 157)	(life, 148)	(right, 186)
5	(time, 213)	(right, 141)	(way, 153)	(state, 146)	(america, 185)
6	(way, 200)	(life, 140)	(moral, 148)	(make, 141)	(muslim, 176)
7	(war, 193)	(way, 128)	(time, 146)	(time, 139)	(white, 164)
8	(right, 182)	(thing, 127)	(good, 145)	(know, 138)	(government, 159)
9	(new, 177)	(know, 127)	(human, 128)	(thing, 135)	(black, 156)
10	(power, 156)	(country, 127)	(think, 128)	(year, 129)	(know, 155)

# Modelling

- CountVectorizer
    - LogisticRegression
    - MultinomialNB
  - TfidfVectorizer
    - LogisticRegression
    - MultinomialNB
-

# Metrics

- Score not too important
- Precision / Recall
- Probability and threshold
- Difference in the 2 Dataset
- Checking manually

---

## Score

Train / Test

External Dataset

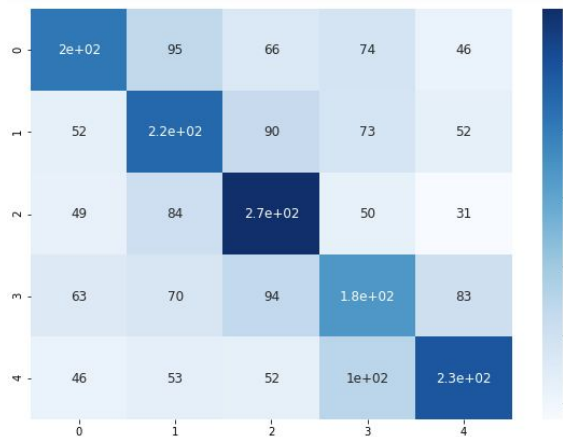
0.45 - 0.52

0.28 - 0.32

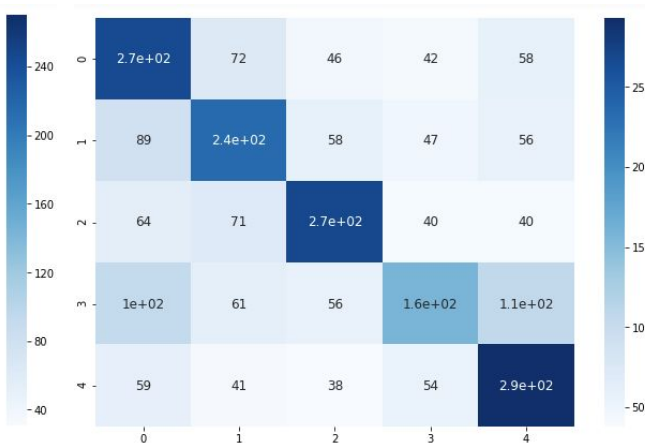


# Confusion will be my epitaph pt. 2

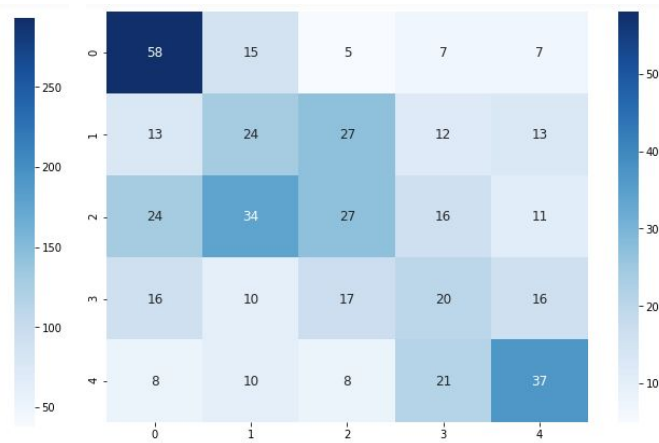
Logistic Regression  
Train-Test



MultinomialNB  
Train-Test



External DF



# Confusion will be my epitaph pt. 2

## MultinomialNB

### Train-Test

	precision	recall	f1-score	support
0	0.45	0.55	0.50	485
1	0.50	0.48	0.49	485
2	0.58	0.54	0.56	484
3	0.45	0.33	0.38	485
4	0.52	0.61	0.56	485
micro avg	0.50	0.50	0.50	2424
macro avg	0.50	0.50	0.50	2424
weighted avg	0.50	0.50	0.50	2424

## External DF

	precision	recall	f1-score	support
0	0.49	0.63	0.55	92
1	0.26	0.27	0.26	89
2	0.32	0.24	0.28	112
3	0.26	0.25	0.26	79
4	0.44	0.44	0.44	84
micro avg	0.36	0.36	0.36	456
macro avg	0.35	0.37	0.36	456
weighted avg	0.35	0.36	0.36	456

# Words for Left

LR -- CV

NB -- CV

POS

- Climate
- Capitalism
- Corporate
- Class

NEG

- America
- Muslim

- People 0.21
- World 0.18
- Human 0.16
- State 0.14
- Life 0.14

LR -- Tfi

NB -- Tfi

POS

- Human
- Climate
- Woman
- Capitalism

NEG

- America
- Moral

- Human 0.0027
- Woman 0.0025
- War 0.002
- Climate 0.002
- Class 0.0018

# Words for Progressive

LR -- CV

NB -- CV

POS

- City
- NSA
- Open Society
- Help

NEG

- Evil
- God

- People 0.19
- State 0.12
- World 0.11
- Time 0.1
- Government 0.1

LR -- Tfi

NB -- Tfi

POS

- City
- Economic
- NSA
- Copyright

NEG

- Black
- Human

- City 0.003
- Book 0.0025
- Market 0.0023
- Culture 0.0023
- Street 0.0021

# Words for Moderate

LR -- CV

NB -- CV

POS

- Company
- Fairness
- Google
- Management

NEG

- Black
- Civilization

- People 0.26
- World 0.14
- Make 0.13
- Life 0.12
- Way 0.1

LR -- Tfi

NB -- Tfi

POS

- Company
- Moral
- Work
- Group

NEG

- Black
- White

- Company 0.004
- Work 0.002
- Love 0.002
- Need 0.002
- Fear 0.0019

# Words for Conservative

LR -- CV

NB -- CV

POS

- Darwinism
- God
- Polygamist
- Public school
- People 0.23
- Government 0.12
- Life 0.1
- American 0.09
- Right 0.08

NEG

- Company
- Lose

LR -- Tfi

NB -- Tfi

POS

- Government
- Freedom
- God
- School
- Government 0.003
- Freedom 0.003
- War 0.0028
- Trump 0.002
- Class 0.0018

NEG

- Company
- World

# Words for Right

LR -- CV

NB -- CV

POS

- Muslim
- Hitler
- Immigration
- Progressive
- People 0.29
- State 0.14
- World 0.13
- America 0.12
- Right 0.11

NEG

- Place
- Class

LR -- Tfi

NB -- Tfi

POS

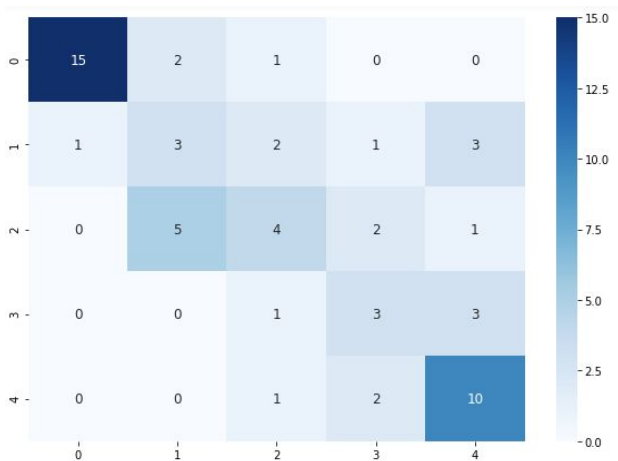
- Muslim
- Progressive
- White
- America
- Muslim 0.004
- Black 0.0038
- White 0.0035
- Democrat 0.0031
- Child 0.0024

NEG

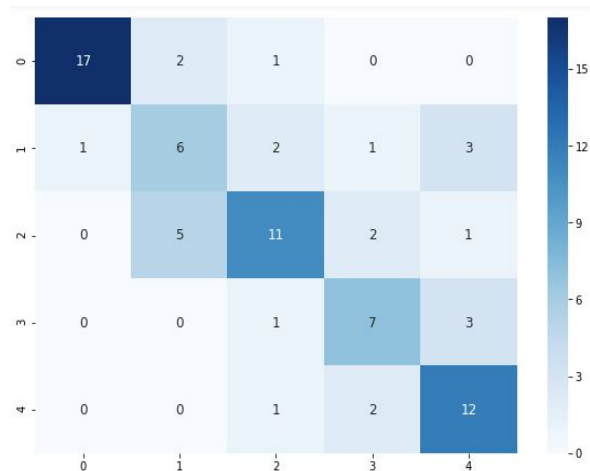
- Work
- Kind

# Confusion will be my epitaph pt. 3

## Threshold Matrix



## Smooth + Threshold



- Confusion Matrix with Threshold on the probability
- Same Matrix considering the neighbor category



## Extra steps

Vader

- **Sentiment Analysis = Progressive**

Word2Vec

- **Embedding and Parsing = Unclear confusion**

N.E.R.

- **Name Entity Recognition = Loss of important features**

“

We don't have to engage in grand, heroic actions to participate in the process of change. Small acts, when multiplied by millions of people, can transform the world.

”

# Man vs. Machine

---

# Left

“

We don't have to engage in grand, heroic actions to participate in the process of change. Small acts, when multiplied by millions of people, can transform the world.

”

Howard Zinn

1. [0.44, 0.18, 0.17, 0.13, 0.06]
2. [0.27, 0.18, 0.22, 0.16, 0.15]
3. [0.9 , 0.003, 0.001, 0.002, 0.09]
4. [0.36, 0.19, 0.14 , 0.14, 0.15]

---

# Example

"

Obama offered Muslims a careful admonition about women's rights, noting how denying women education impoverishes a country — balanced, of course, with this: "Issues of women's equality are by no means simply an issue for Islam." Example? "The struggle for women's equality continues in many aspects of American life....."

"

Charles Krauthammer

ACTUAL = CONSERVATIVE

1.  $[0.006, 0.02, 0.1, 0.65, 0.2] \cdot (L)$
2.  $[0.69, 0.001, 0.06, 0.03, 0.21]^{**}$
3.  $[0.25, 0.1, 0.14, 0.17, 0.3]$
4.  $[0.0, 0.0, 0.0, 0.99]$
5.  $[0.19, 0.13, 0.13, 0.19, 0.33]$

---

# Transformation

“  
Public  
”

PROGRESSIVE

“  
Public school  
”

CONSERVATIVE

---

# Man vs. Machine

“ Capitalism does a number of things very well: it helps create an entrepreneurial spirit; it gets people motivated to come up with new ideas, and that's a good thing “

# Bernie Sanders

PREDICTION

“ Conservative , Moderate, Moderate, Moderate ”

## Future



- Better Dataset
- Different Dataset
- Implement an ordinal categorical classification
- Investigate about Sentiment Analysis
- Topic as feature
- Locate space and time
- Get a better computer

## Conclusion



- Homologation of the words
- There is a light pattern
- A biased person is not always biased
- Limitation of ambiguity
- The 3 categories in the centre are confused
- I have a computational power problem



# Special Thanks to

- My heroic, noisy, slow, dirty, old mac
- And to friends who advised me to change project...They were probably right !

