Ideology detection



Left Wing

Right Wing

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

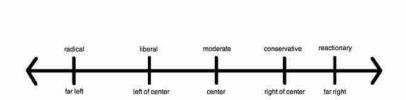
Using quotes of people with a strong political bias try to identify and capture key-words and pattern to detect ideology hidden behind

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

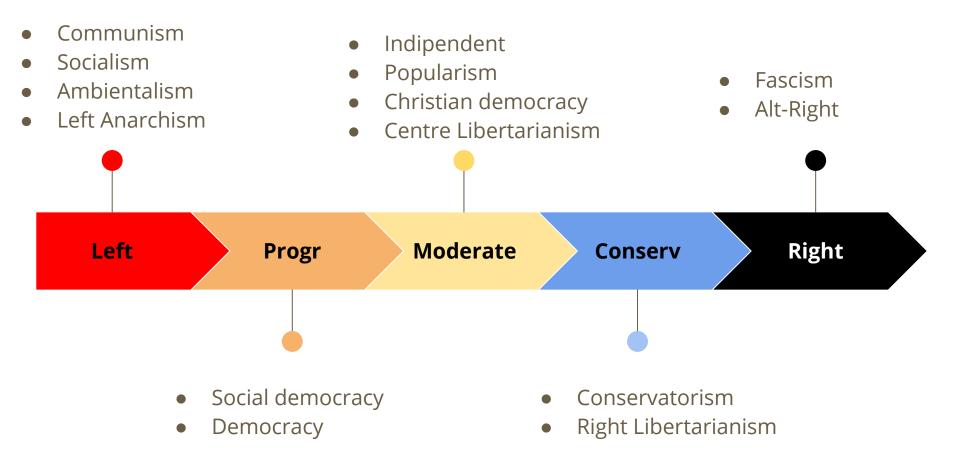
Criteria

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

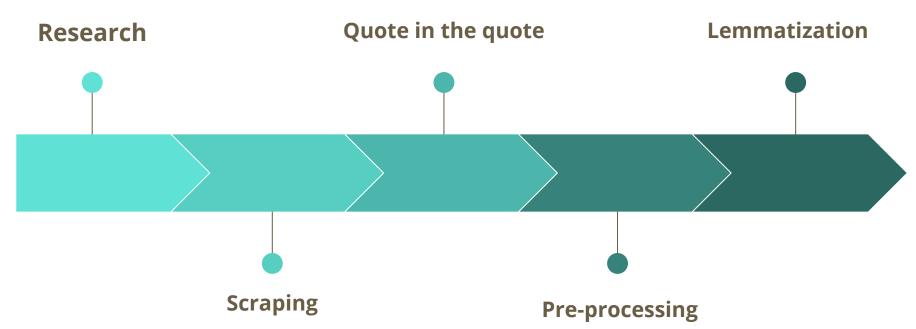
Selection

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

The target



First steps



Balance

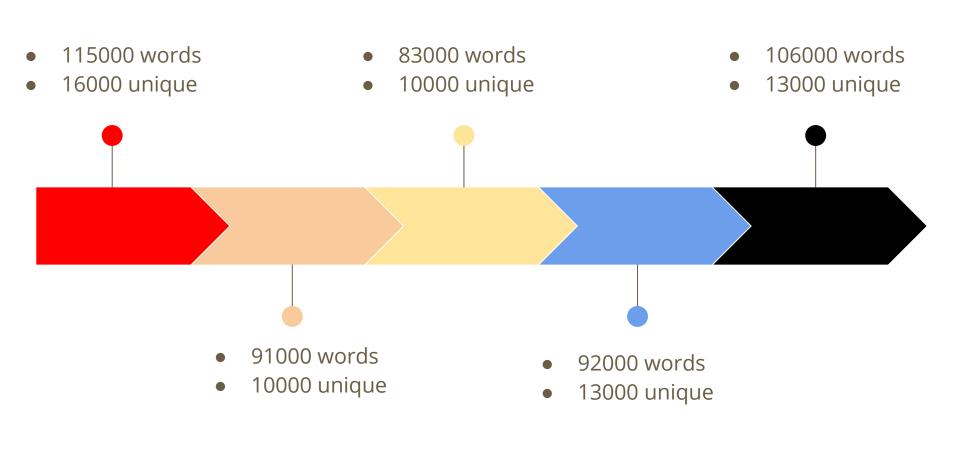


- The weakest category choose the number of the sample
- External Dataset with four new authors for each category

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



Confusion will be my epitaph

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

Modelling

- CountVectorizer
 - LogisticRegression
- MultinomialNB

- TfiVectorizer
- LogisticRegression
- MultinomialNB

Metrics

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

Score



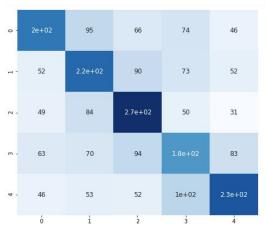
0.45 - 0.52

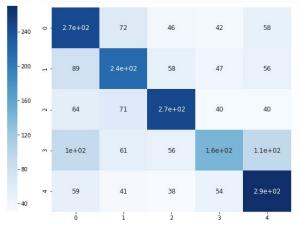
0.28 - 0.32

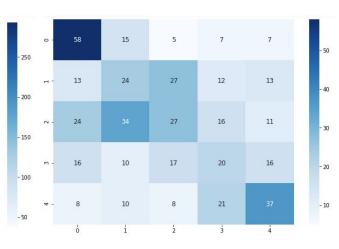
Confusion will be my epitaph pt. 2

Logistic Regression
Train-Test

MultinomialNB Train-Test **External DF**







Confusion will be my epitaph pt. 2

MultinomialNB Train-Test

External DF

		precision	recall	fl-score	support		precision	recall	fl-score	support
	0	0.45	0.55	0.50	485	0	0.49	0.63	0.55	92
	1	0.50	0.48	0.49	485	1	0.26	0.27	0.26	89
	2	0.58	0.54	0.56	484	2	0.32	0.24	0.28	112
		100 mm (100 mm)		35-2-37-37-37-37		3	0.26	0.25	0.26	79
	3	0.45	0.33	0.38	485	4	0.44	0.44	0.44	84
	4	0.52	0.61	0.56	485					
102		8 86	02002020	8 88	10212020	micro avg	0.36	0.36	0.36	456
micro	avg	0.50	0.50	0.50	2424	macro avq	0.35	0.37	0.36	456
macro	avg	0.50	0.50	0.50	2424	weighted avg	0.35	0.36	0.36	456
weighted	avg	0.50	0.50	0.50	2424	norghood dvg	0.05	0.00	0.00	150

Words for Left

LR CV	NB CV	LR Tfi	NB Tfi
POS Climate Capitalism Corporate Class NEG America Muslim	 People 0.21 World 0.18 Human 0.16 State 0.14 Life 0.14 	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	LR Tfi	NB Tfi
•	Open Society	 People 0.19 State 0.12 World 0.11 Time 0.1 Government 0.1 	POS City Economic NSA Copyright NEG	 City 0.003 Book 0.0025 Market 0.0023 Culture 0.0023 Street 0.0021
	Cod		- Black - Human	

Words for Moderate

LR CV	NB CV	LR Tfi	NB Tfi
POS Company Fairness Google Manageme NEG Black Civilization	• Way 0.1	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	LR	Tfi	NB Tfi	
•	Public school NEG Company	 People 0.23 Government 0.12 Life 0.1 American 0.09 Right 0.08 	FreeGodScho	vernment • edom • d • ool • npany	Government 0. Freedom 0.003 War 0.0028 Trump 0.002 Class 0.0018	003
-	Lose		- Wor	rld		

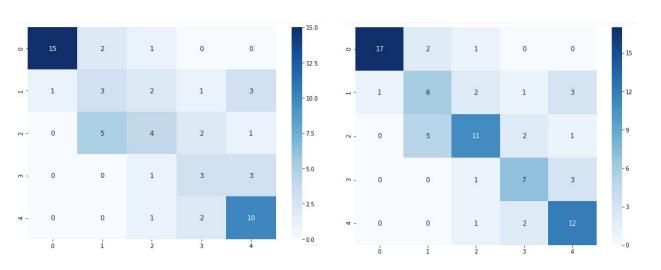
Words for Right

LR CV	NB CV	LR Tfi	NB Tfi
 POS Muslim Hitler Immigration Progressive NEG Place Class 	 People 0.29 State 0.14 World 0.13 America 0.12 Right 0.11 	 POS Muslim Progressive White America NEG Work Kind 	 Muslim 0.004 Black 0.0038 White 0.0035 Democrat 0.0031 Child 0.0024

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

Word2Vec

N.E.R.

Sentiment Analysis = Progressive Embedding and Parsing = Unclear confusion Name Entity
 Recognition =
 Loss of
 important
 features

11

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

11

Left

11

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] *(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

Conclusion

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

- 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!

