Predicting political orientation from Twitter contents

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8 January 2020



Summary



- Data Collection and Database Creation
- Data Preprocessing
- Model Creation through fastText
- Results
- M5S Prediction

Data collection



- 5 italian political parties
- 1000 italian users randomly selected from each of the 5 parties Twitter's followers
- 3000 Twitter's users randomly selected











Database Creation



- Each user was classified by each member of the groups as a supporter of one of the 5 parties or as Non-Classifiable (NC)
- \blacksquare Data collected by each group were merged together, leading to a ≈ 8000 users dataset

Twitter ID	Username	Sex	SID1	SID2	SID3	SID4
Ruggier98623365	Ruggiero	M	1	1	1	1
EricSalicetti	Eric Salicetti	M	0	0	0	0
MassimoDolore	Massimo Dolore	M	4	4	4	4
Rivalevante	Sergio Stagnaro	M	2	0	2	2
Iolanda07658172	Islanda	U	1	3	1	3
manuelamimosa	Manuela Mimosa	F	2	2	4	2



Aim: to construct a model through fastText that is able of predicting whether a user belongs to the right parties (*Lega, Fratelli d'Italia, Forza Italia*) or to the left ones (*Partito Democratico*)

- Label at 100% vs Label at 75%
- What about *Movimento Cinque Stelle*?

Data Preprocessing



The most important step was cleaning up text data.

Among others, we removed:

- Punctuation
- Stopwords
- Emoji and links
- Insignificant words

We also performed Lemmatization.

Model Creation



To create our NLP models we chose fastText.

We trained models in two different ways:

- Labelling each user and considering all their tweets together
- Labelling each tweet of a user with their same orientation

In order to overcome the variability and the unbalance of the dataset, we tried performing the k-fold cross validation, the stratification and the upsampling.

Result - Basic



	Parameters		I hreshold $= 75\%$	I hreshold = 100%
NC	Lemmatization	Person	Accuracy	Accuracy

NC	Lemmatization	Person	Accuracy	Accuracy
Х	Х	Х	0,787	0,790
X	X	✓	0,699	0,701
X	✓	Х	0,781	0,786
X	✓	✓	0,697	0,699
1	X	Х	0,653	0,651
1	X	✓	0,627	0,599
1	✓	Х	0,648	0,646
✓	✓	✓	0,617	0,598

- \blacksquare NC: if X, then the possible lables are just DX (Right) and SX (Left), while if \checkmark , then there are three possible lables, DX, SX, and NC
- Person: if X the model classifies each tweet, while if ✓ it classifies the single persons considering all their tweets

Result - Upsampled

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	NC	Lemmatization	Person	Accuracy	Accuracy	
	X	Х	Х	0,706	0,694	
	v	V	/	0.622	0.576	

Throshold = 75% Throshold = 100%

IVC	Lemmanzanon	reison	Accuracy	Accuracy
Х	Х	Х	0,706	0,694
X	X	✓	0,623	0,576
X	✓	X	0,685	0,679
X	✓	✓	0,650	0,611
1	X	Х	0,605	0,595
1	X	✓	0,535	0,543
1	✓	Х	0,596	0,591
/	✓	✓	0,530	0,533

- \blacksquare NC: if X, then the possible lables are just DX (Right) and SX (Left), while if \checkmark , then there are three possible lables, DX, SX, and NC
- Person: if X the model classifies each tweet, while if ✓ it classifies the single persons considering all their tweets

K-Fold Cross Validation



	Parameters		I hreshold = 75%	I hreshold $= 100\%$
NC	Lemmatization	Person	Accuracy	Accuracy
v	V	V	0.042	0.001

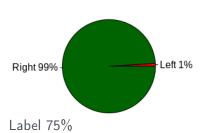
NC	Lemmatization	Person	Accuracy	Accuracy
Х	Х	Х	0,943	0,801
X	X	✓	0,971	0,971
X	✓	Х	0,899	0,949
X	✓	✓	0,970	0,970
1	X	Х	0,609	0,611
1	X	✓	0,648	0,621
1	✓	X	0,621	0,623
1	✓	✓	0,645	0,679

- \blacksquare NC: if X, then the possible lables are just DX (Right) and SX (Left), while if \checkmark , then there are three possible lables, DX, SX, and NC
- Person: if X the model classifies each tweet, while if ✓ it classifies the single persons considering all their tweets

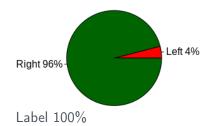
Results - M5S users



How does the best models classify Movimento Cinque Stelle users?



- NC X
- Lemmatization X
- Person ✓



- NC X
- Lemmatization X
- Person ✓

Thanks for your attention.