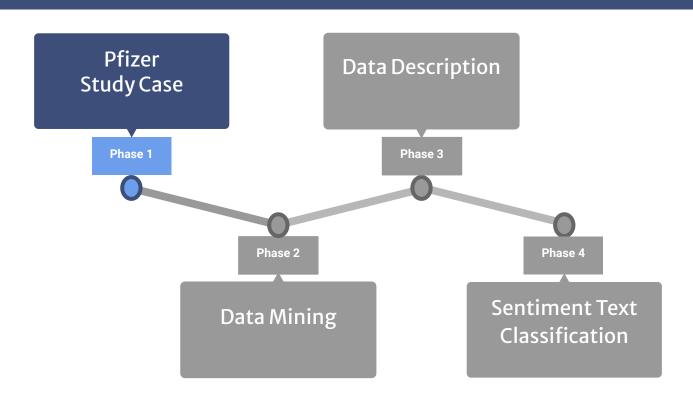
Natural Language Processing



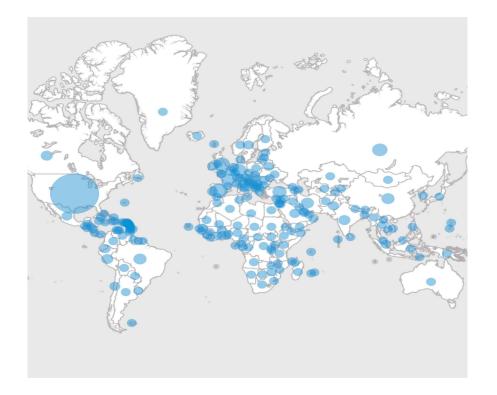
Text Classification on Pfizer' tweets for Sentiment Analysis

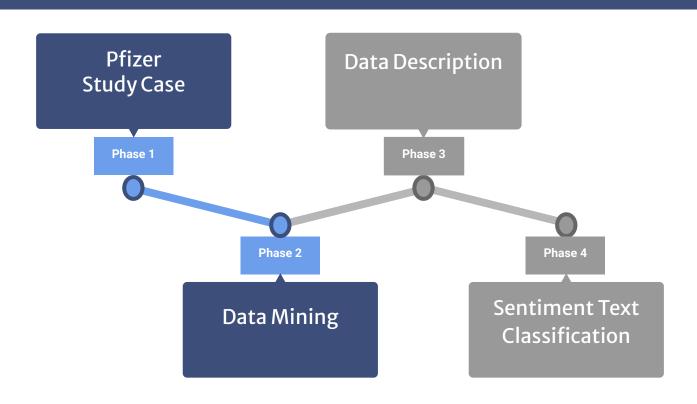
Student: Pietro Morichetti



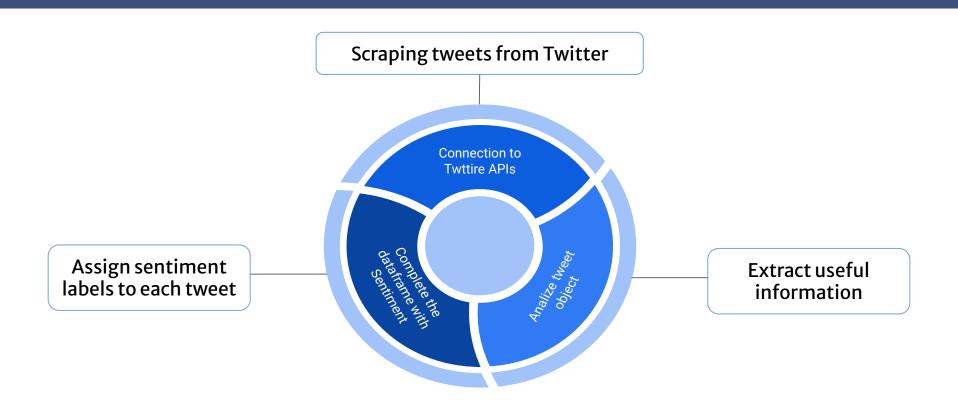
Pfizer & Covid-19 case study

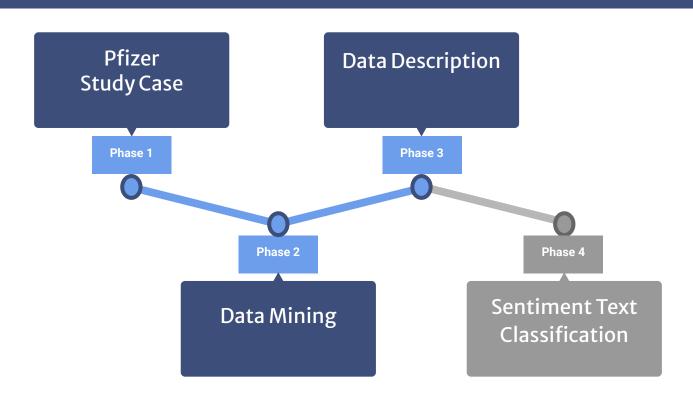






Data Mining: get access to Twitter Uninverse





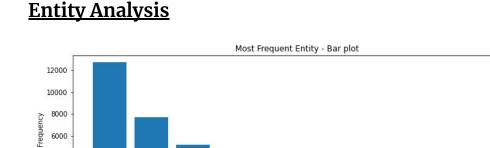
Data Description: NLP Analysis

Topic Modelling

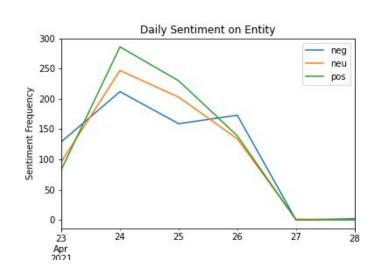
Topics		Related Words									
T1	jab	wait	fulli	think	go	work	look	better	tomorrow	week	
T2	dose	got	shot	second	effect	today	get	feel	arm	day	
Т3	covid	peopl	receiv	israel	heart	case	india	inflamm	world	examin	
T4	moderna	deal	time	know	said	booster	jampj	canada	come	problem	
T5	avail	appoint	covid	new	need	moderna	pfizerbiontech	amp	astrazeneca	read	

- Why Topic Modelling? Be sure most of the tweets are related to Pfizer (and Covid-19, vaccines, ...)
- A preliminary phase is preferred to reduce the variance on tweet text
- Topic Modelling is a stochastic procedure: new iteration... new terms!
- Improvement by tuning the hyper-parameters

Data Description: NLP Analysis



4000 2000



- Useful to better understand what people are more focused on
- Pfizer, Moderna, Vaccin, India, Washington, Jeff Bezos,...
- Trend Sentiments plot for pfizer entity: positive sentiment is mostly dominant!

Data Description: NLP Analysis

Table shows the collections set of the most likely terms to be seen together

In this PMI there is metrics on the score

Seems people are quite worried about the negative effects of the Covid-19 and vaccines.

PMI Analysis

Terms	Score
heart_inflamm	331.863
raw_materi	238.647
israel_examin	198.439
examin_heart	182.204
inflamm_case	165.408
pfizer_vaccin	142.228
neurolog_degen	137.768
covid_vaccin	116.6277
caus_neurodegen	112.576
al_neurolog	109.096

Terms	Score
alzheim_al	106.588
sore_arm	104.236
$second_dose$	100.465
$trigger_alzheim$	97.504
bee_gee	85.0
pfizer_shot	84.327
neurodegen_diseas	82.586
gee_singer	76.246
barri_gibb	73.751
singer_barri	71.579

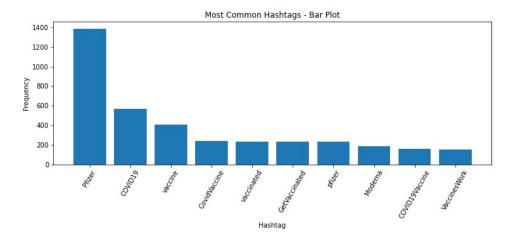
Data Description: Descriptive Analysis

Pre-processing applied, most frequent terms are related with the pandemic situation

Also from the point of view of the most used hashtags...

... having the *potential* to move the conscious of a lot of people



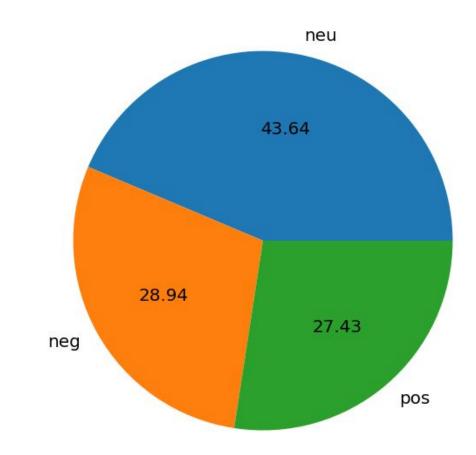


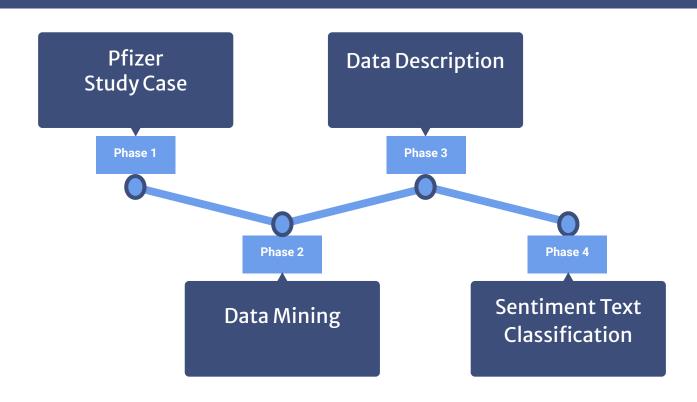
Data Description: Tweet Sentiment

Pie chart shows the percentage of sentiment among: negative, neutral and positive

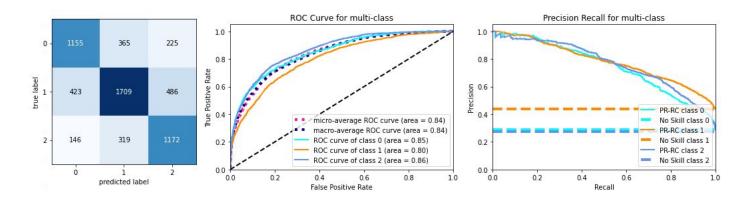
Most of the tweets have a neutral sentiment

The dataset is balanced enough to avoid oversampling or undersampling





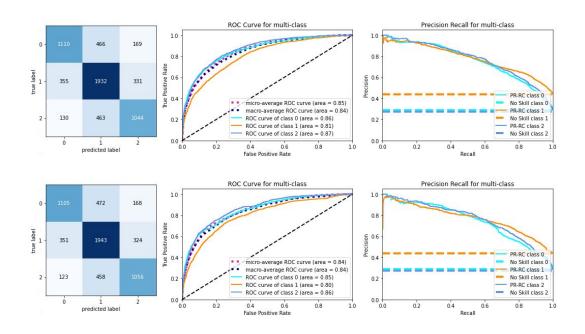
Logistic Regression



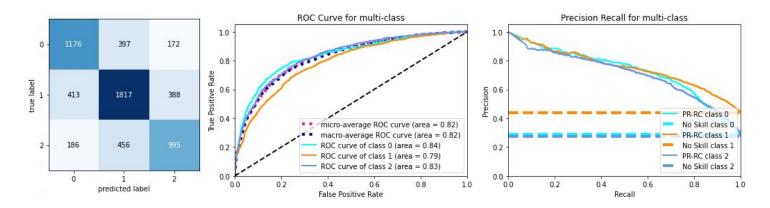
- All classes have mostly the same scores
- Baseline has score 0.43
- Class 2 is the class better predicted, with AUC of 0.86

OnevsAll & OnevsOne

- All classes have mostly the same scores
- Baseline has score 0.43
- Class 2 is the class better predicted, with AUC of 0.87 and 0.88 for OnevsOne and OnevsAll



Convolutional NN



- All classes have mostly the same scores
- Baseline has score 0.43
- Class 2 is the class better predicted, with AUC of 0.86

		gularize ure Sele		CNN			
	Precision	Recall	F1 Score	Precision	Recall	F1 Score	
Class 0	0.69	0.67	0.68	0.66	0.70	0.68	
Class 1	0.72	0.68	0.70	0.71	0.65	0.68	
Class 2	0.66	0.74	0.70	0.64	0.69	0.66	

Accuracy	177	-	0.69	S=3		0.67
Macro Avg	0.69	0.70	0.69	0.67	0.68	0.67
Micro Avg	0.69	0.69	0.69	0.68	0.67	0.67

	O	ne vs A	.11	One vs One			
	Preicsion	Recall	F1 Score	Precision	Recall	F1 Score	
Class 0	0.71	0.64	0.67	0.71	0.64	0.67	
Class 1	0.69	0.75	0.72	0.69	0.75	0.72	
Class 2	0.70	0.67	0.69	0.70	0.67	0.68	

Accuracy	7.7		0.70	S-2	1 2	0.70
Macro Avg	0.70	0.69	0.69	0.70	0.69	0.69
Micro Avg	0.70	0.70	0.70	0.70	0.70	0.69

Natural Language Processing



Text Classification on Pfizer' tweets for Sentiment Analysis

Student: Pietro Morichetti

Thanks for your Attention