

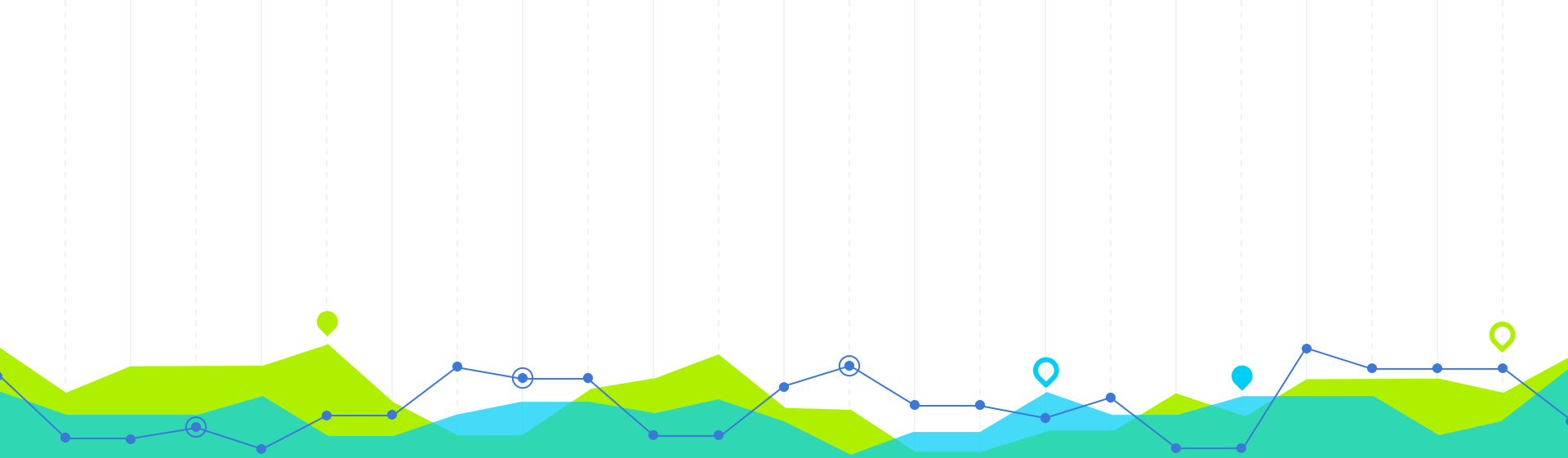


GREENPASS

Sentiment Analysis

Cancello Tortora Giuseppe
Macrì Armando

a.a. 2021-2022

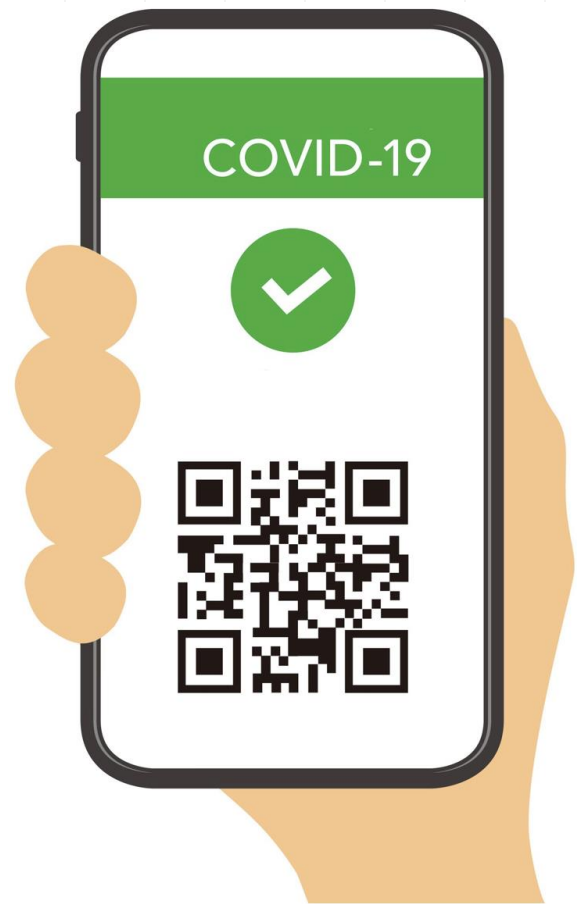


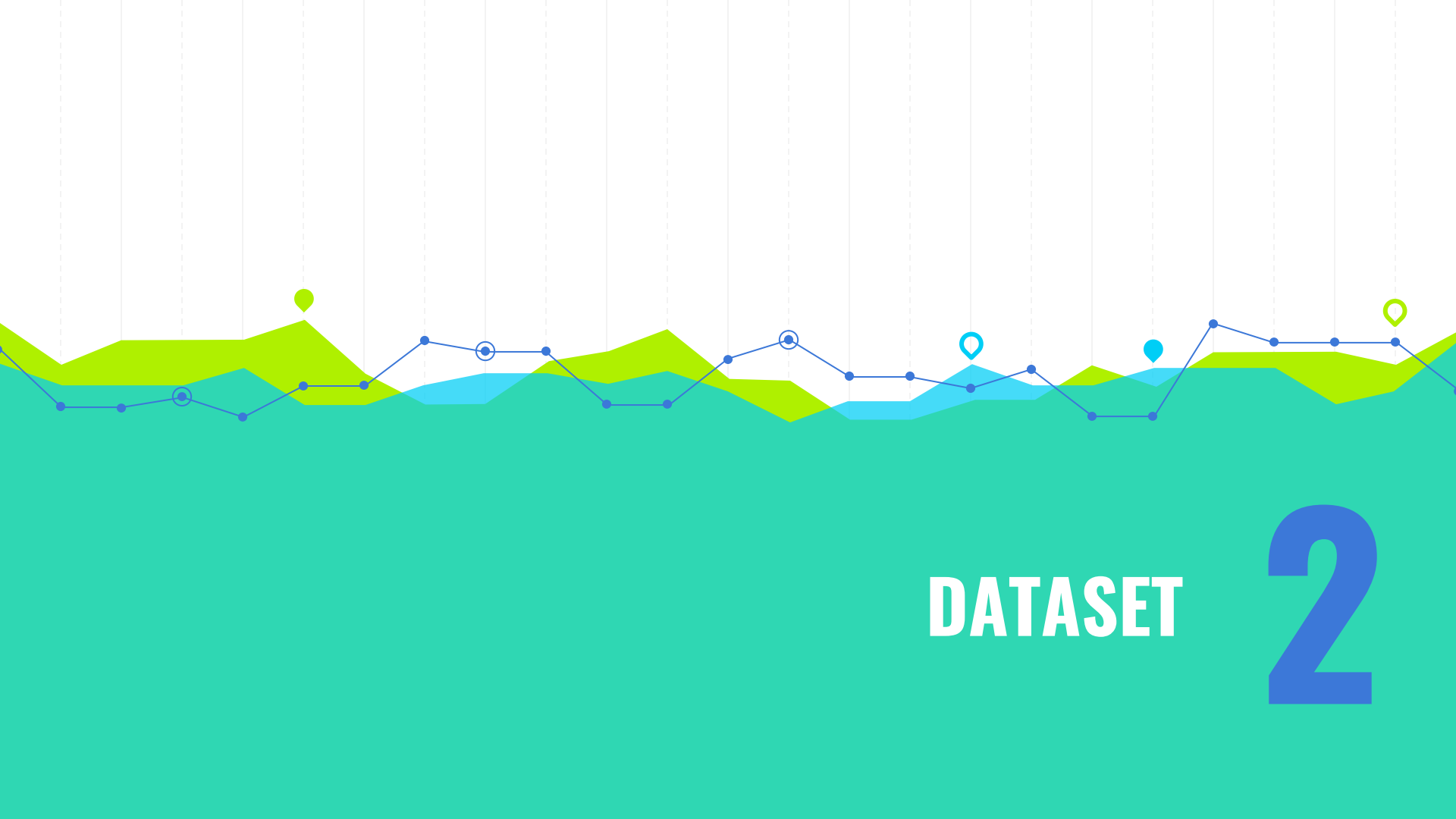
INTRODUCTION

1

Goal of the project

The goal of this project is to develop a monitoring analysis in order to extract useful information and retrieve what is the popular opinion about **Green Pass**.



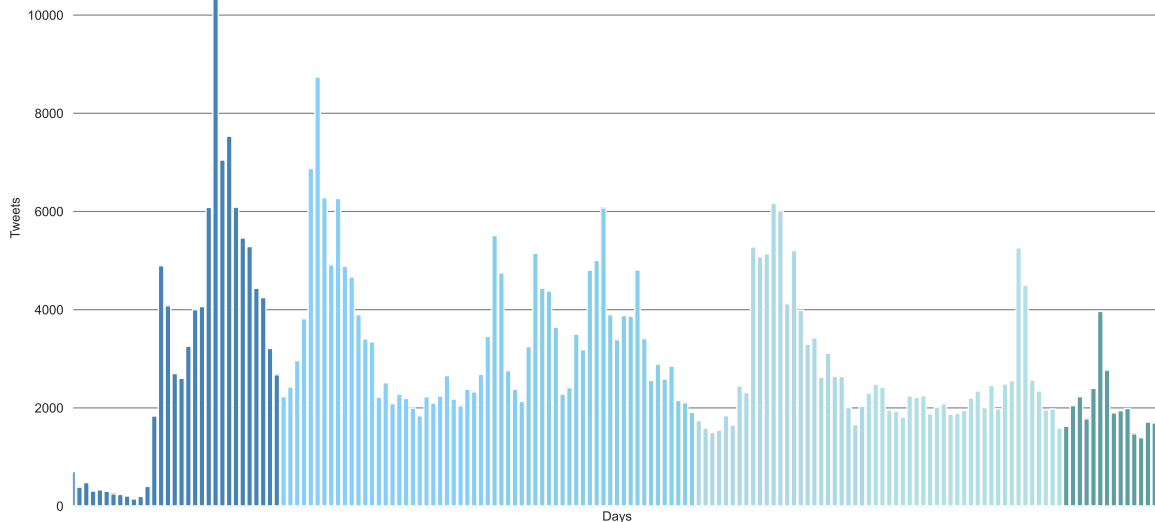


DATASET 2

DATASET - BUILDING

Tweets distribution (OVERALL)

July September November
August October December



● Tweets are collected by using a Python script

● From July 1st to December 15th

● ~ 500.000+ tweets



DATASET - CLEANING

- Keep only italian tweets
- Removing URL, mentions, emoticons
- Replacing multiple spaces with single space
- Removing punctuation marks
- Lower case

Clean tweets content from useless stuff and try to **standardize** them as much as possible.



È passata la mezzanotte, ho
ufficialmente il 'GREEN PASS'



Cleaning

è passata la mezzanotte ho
ufficialmente il green pass

DATASET – TRAINING SET

■ Positive ■ Negative ■ Neutral



● ~ 2400 tweets labelled by hand

● Balanced dataset, about 800 instances for each class

● Timeline for training set:
01/07/2021 to 22/07/2021



DATA PREPROCESSING

TOKENIZATION



STOP-WORDS FILTERING

Words from file (stopwords_it)

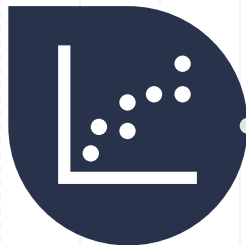


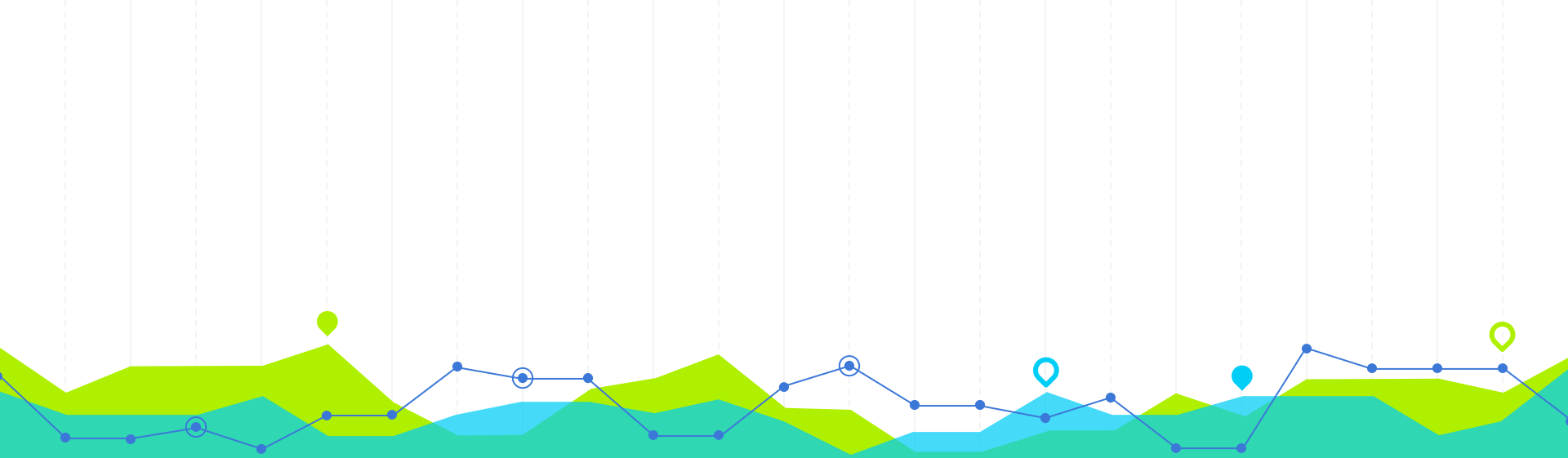
STEMMING



FEATURES EXTRACTION

TFIDF



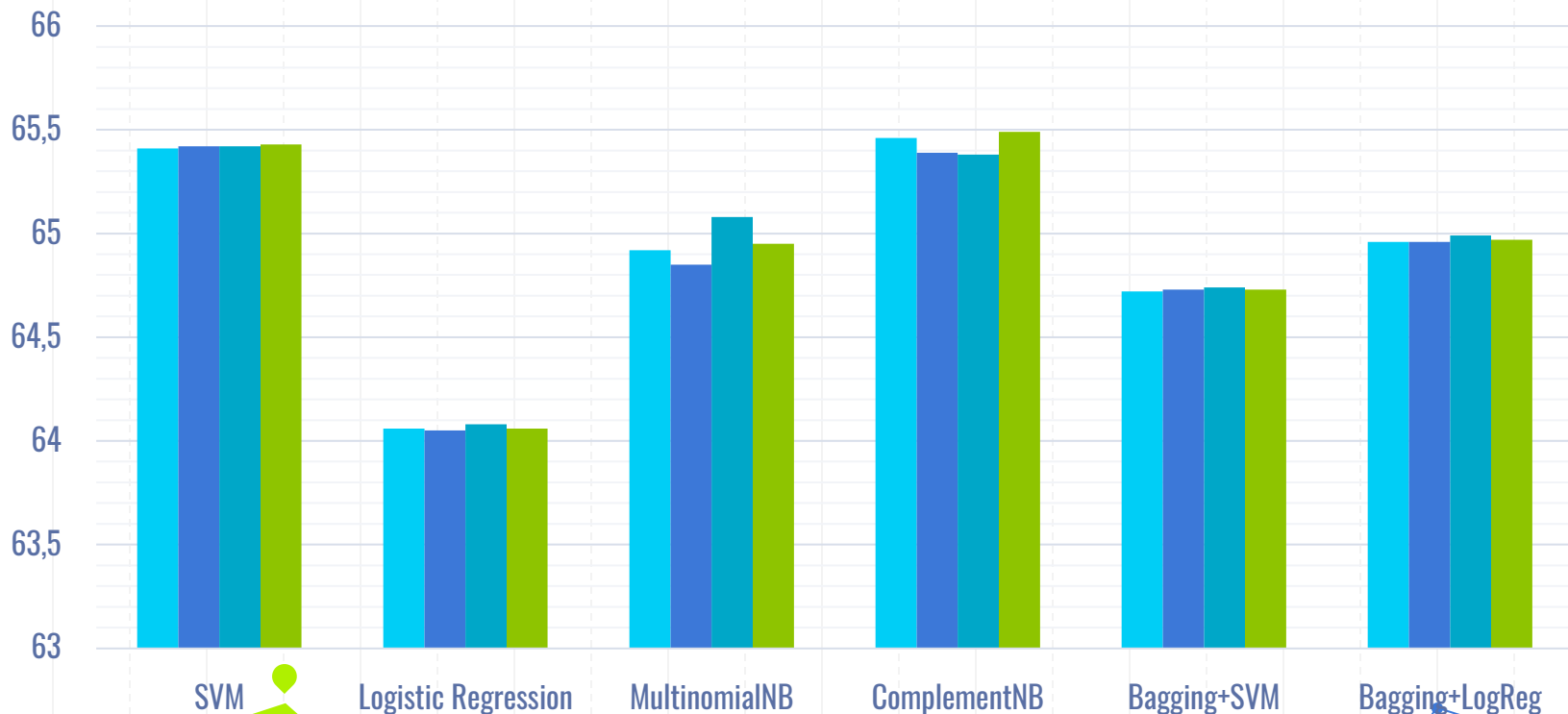


CLASSIFICATION

3

MODEL PERFORMANCE ON TRAINING SET

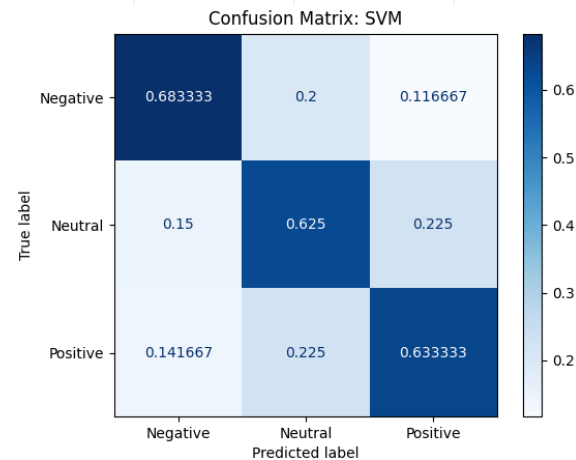
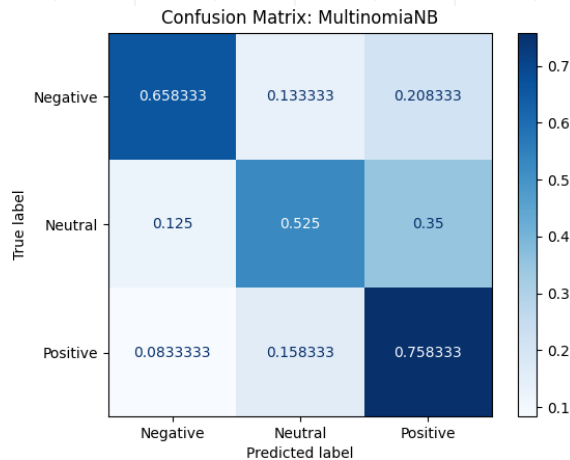
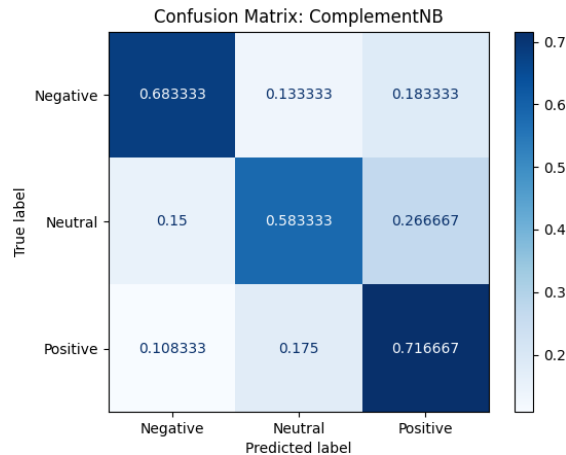
Accuracy Precision Recall F1-score



MODEL SELCTION USING PAIRED T-TEST

CLASSIFIER	LOG REG	MULTI_NB	COMP_NB	BAG_SVM	BAG_LOG
SVM	4.3883	1.5077	-1.2245	7.6961	7.7386
	0.6467 / 0.6522	0.6494 / 0.6522	0.6545 / 0.6522	0.6456 / 0.6522	0.6444 / 0.6522
LOG REG		-1.5650	-6.0565	0.8210	2.6656
		0.6494 / 0.6467	0.6545 / 0.6467	0.6456 / 0.6467	0.6444 / 0.6467
MULTU_NB			-3.459	1.7161	2.2671
			0.6545 / 0.6494	0.6456 / 0.6494	0.6444 / 0.6494
COMP_NB				4.3025	5.5733
				0.6456 / 0.6545	0.6444 / 0.6545
BAG_SVM					1.0879
					0.6444 / 0.6456

CLASSIFICATION – SELECTION



66,11%

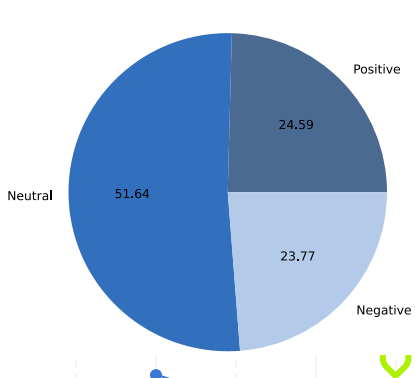
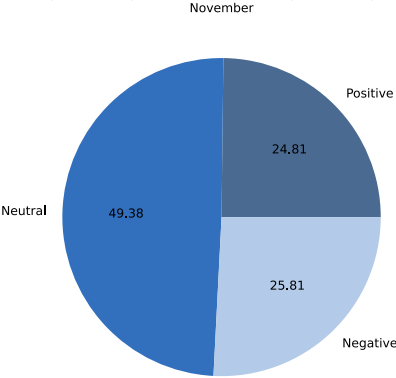
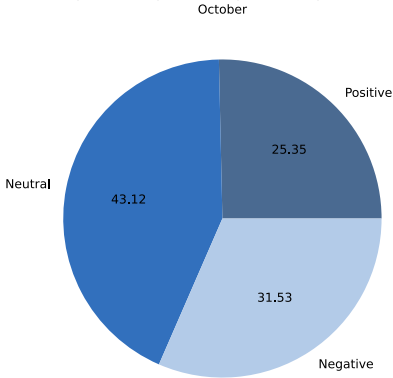
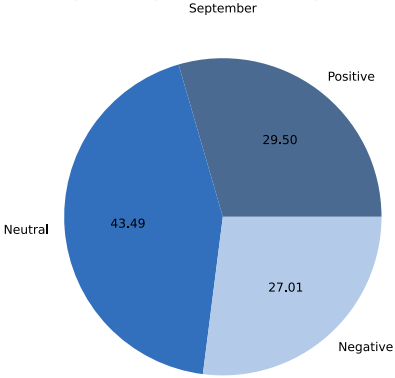
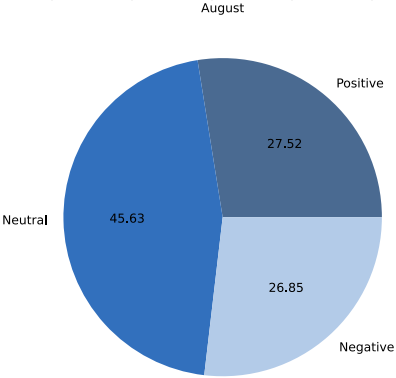
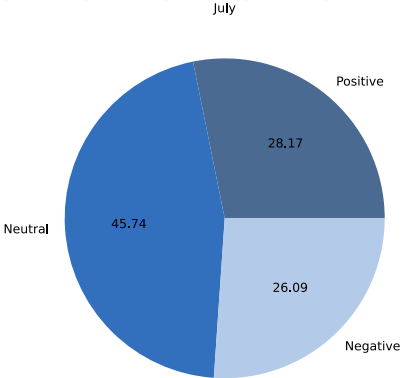


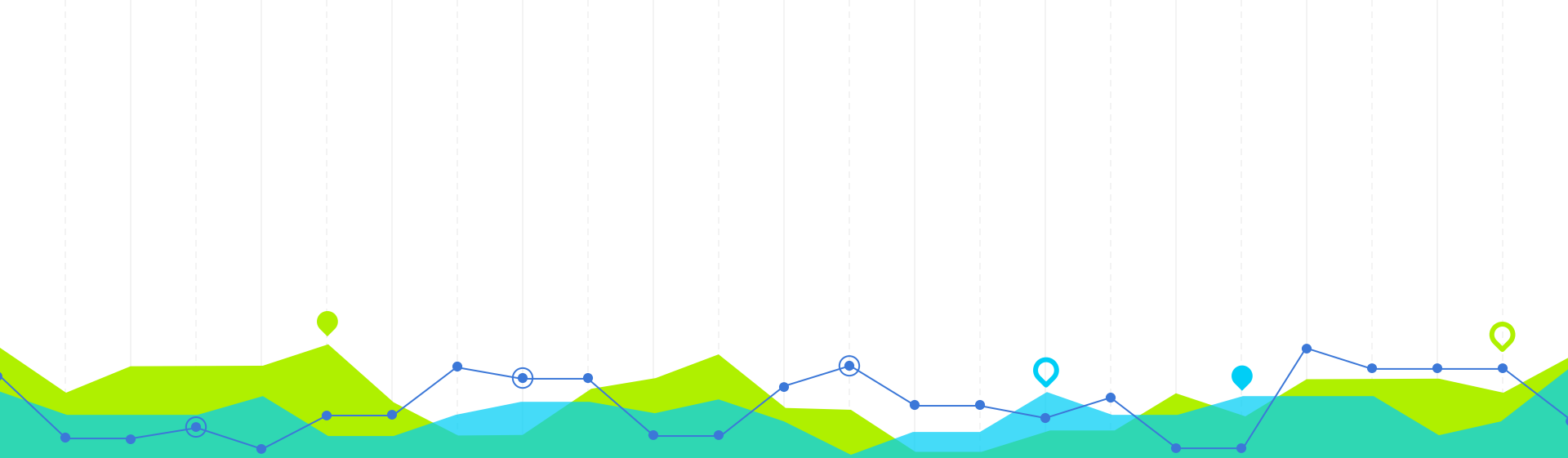
64,72%



64,71%

DISTRIBUTION OF SENTIMENT

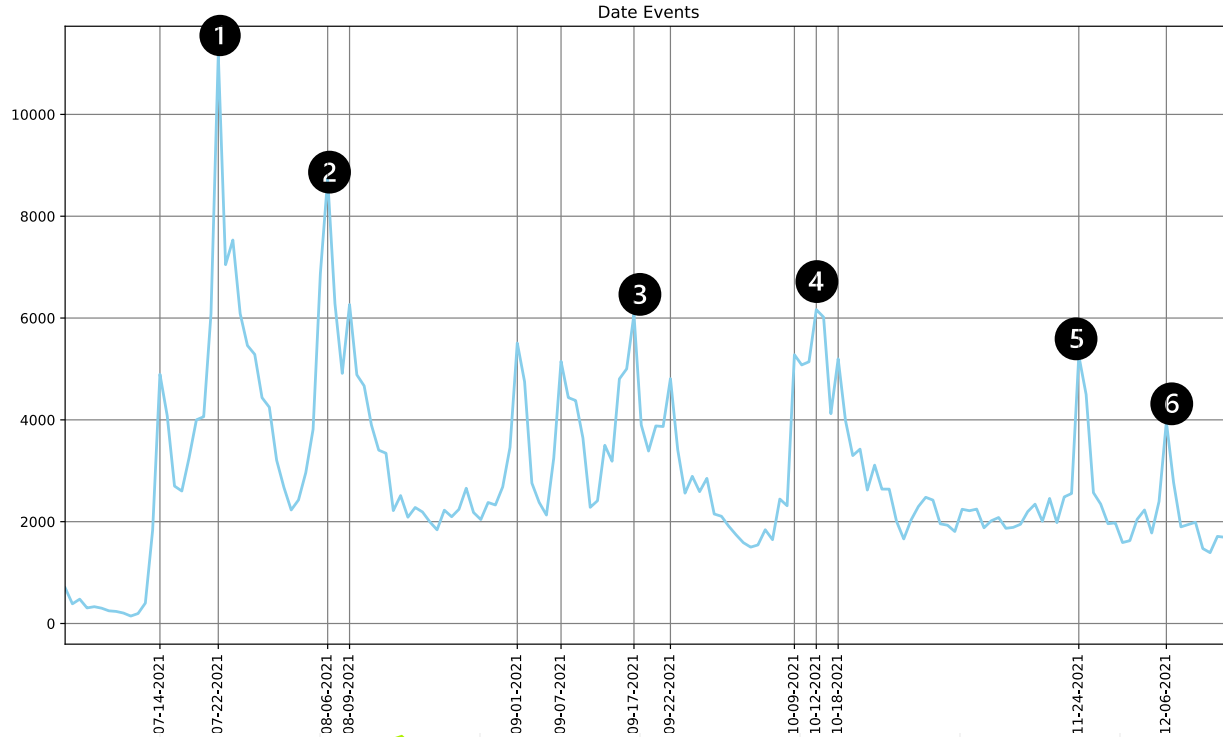




MONITORING ANALYSIS

4

MONITORING ANALYSIS – TIMELINE



Found 13 key events on the timeline considered

Peak of tweets of those events

We focused on 6 events

EVENTS





MONITORING ANALYSIS – SCHEMAS

3 different learning settings!

INCREMENTAL MODEL

Trained with the initial training set and all the hand labelled data of all the previous events before testing on a new event.

SLIDING MODEL

Retrained each time with the most recent 2400 tweets, removing the oldest 360 and adding the newest 360.

STATIC MODEL

The initial training set composed by 2400 tweets

MONITORING ANALYSIS – RESULTS

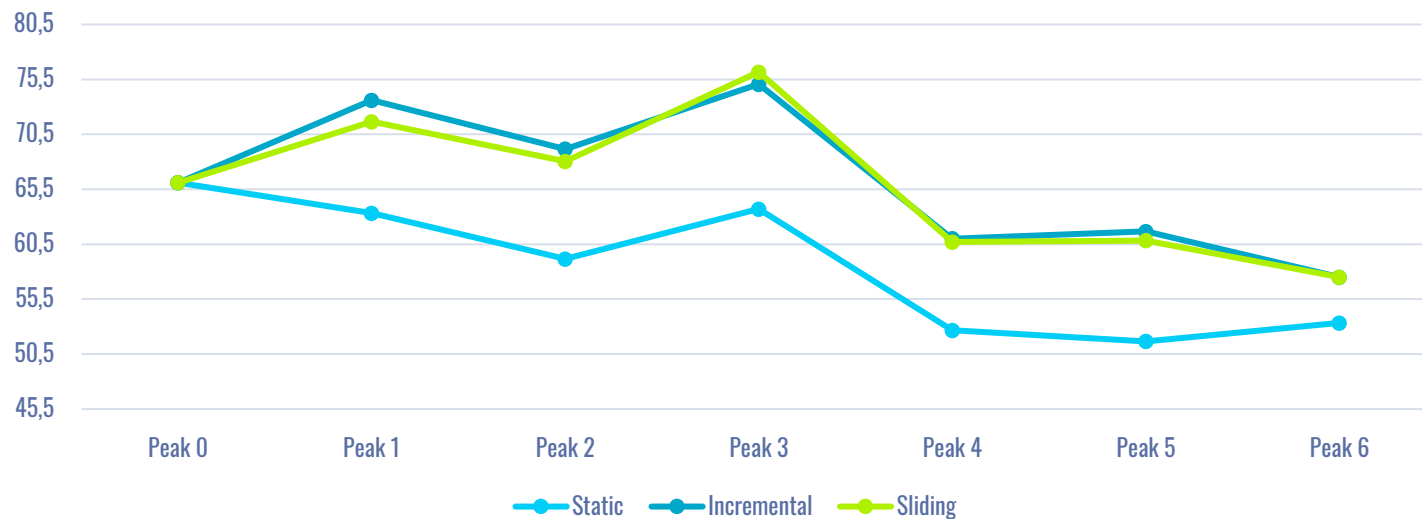
SHEMAS	Accuracy (%)		
	Incremental	Sliding	Static
Peak 1	73,61	71,66	63,33
Peak 2	69,16	68,05	59,16
Peak 3	75,06	76,17	63,71
Peak 4	61,01	60,71	52,67
Peak 5	61,66	60,83	51,66
Peak 6	57,50	57,50	53,33
Average	66,33	65,82	57,31

These are the results in terms of accuracy of the three considered approaches.

Incremental and sliding model perform quite well, while the static model is affected by the concept drift.

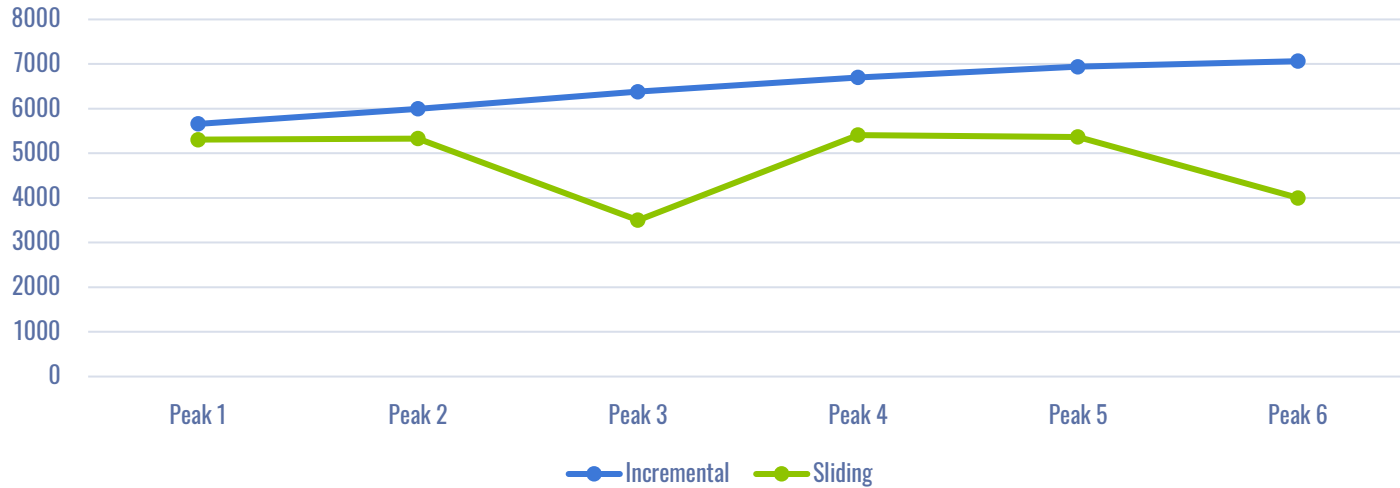
MONITORING ANALYSIS - RESULTS

Accuracy trend with respect to the different approaches



MONITORING ANALYSIS - RESULTS

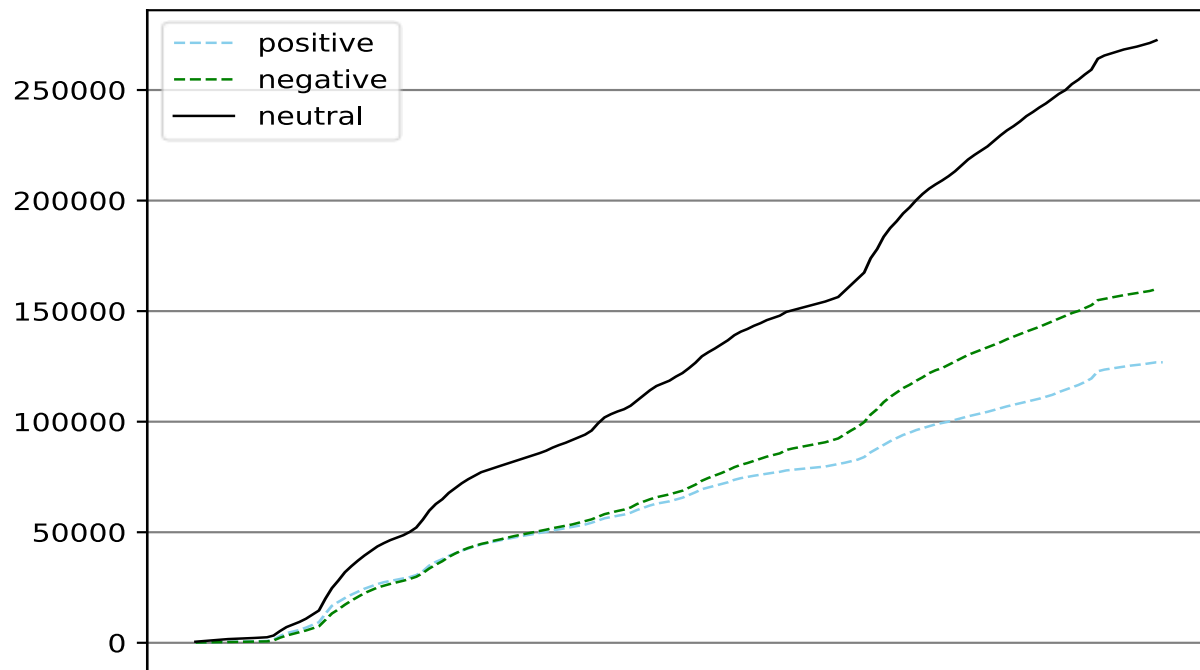
Number of features



We select the *sliding* approach to conclude our analysis because, on average, it gave us results comparable to those obtained with the incremental approach, and it requires, also, a smaller number of features.

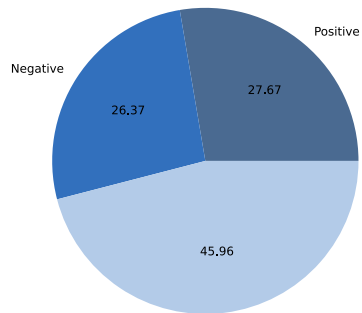
MONITORING ANALYSIS - CONCLUSION

Cumulated tweets

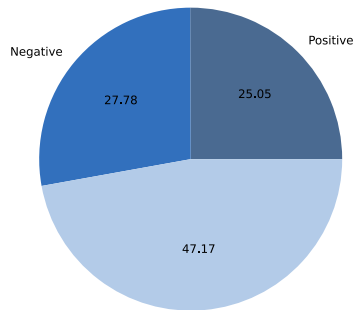


DISTRIBUTION OF SENTIMENT

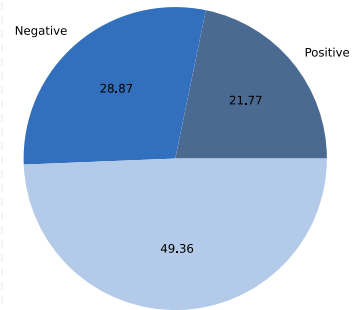
July



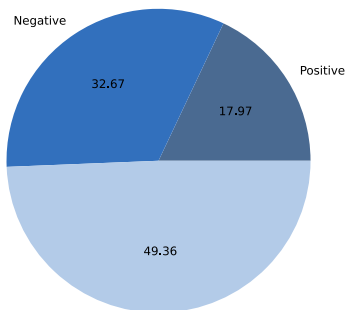
August



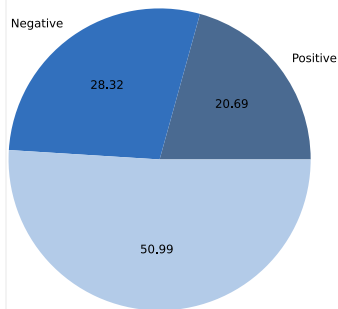
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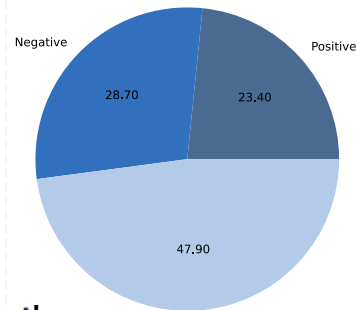
October



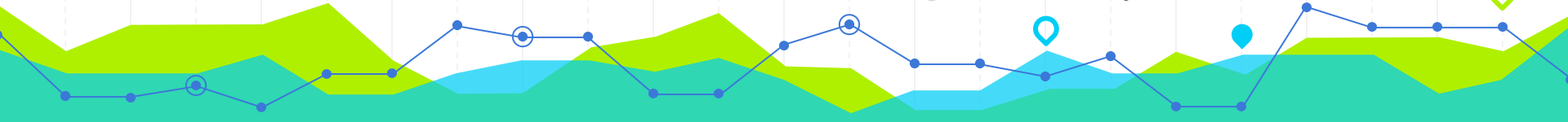
November



December



Negative tweets are predominant!



THANKS!

