Project 3: Sentiment Analysis of Elon Musk

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#### **Background**

We are a **data science-backed PR agency** trying to pitch Elon Musk our services.

A good public image is essential in:

- 1. Building trust,
- 2. Attracting customers, and
- 3. Fostering positive relationships with stakeholders.

#### Elon Musk's Net Worth Has Dropped \$9 Billion Since Buying Twitter

https://www.inc.co

Elon Musk 30 Oct 2018 back on track.



Qz Quan https:

7 Nov 2022 last week,  $\mathsf{T}$  t's been a busy year for Elon Musk, and his net worth is taking a hit.

BY NIK POPLI Y NOVEMBER 1, 2022 2:25 PM EDT

After purchasing Twitter for \$44 billion on Thursday—the largest leveraged buyout of a technology company in history—Musk's net worth has fallen by \$9 billion, according to calculations by the Bloomberg Billionaires Index. His net worth now stands at \$203 billion, down roughly 25% since the beginning of the year.

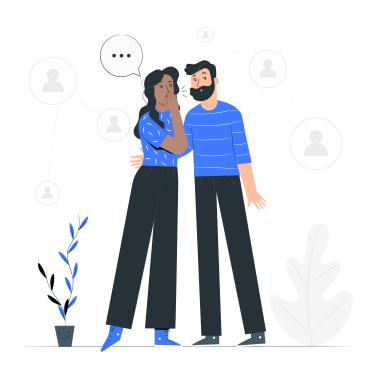
... Elon Musk's

X

mpulse son of tech







#### **Problem Statement**

Can we create an **effective classification model** to identify the **level of negative sentiment** in **text comments?** 

How has public perception changed over time?

## Methodology

#### Methodology



Data Collection

Data Cleaning

**Establishing Ground Truth** 

Pre-modelling, Modelling



1

Pulling comments from Training Video (Youtube API) 2

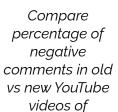
Null values,
Duplicated values,
Data Types,
HTML-encoded
entities, Foreign
Languages



Running VADER model against manual labels, use VADER as ground truth



Lemmatise, Stop word removal → Fitting models to training video



Elon Musk





#### **Training Dataset**

- Streamed on Sep 7, 2018
- **67 million** views
- **140,000** comments





#### **Data Cleaning**

- Null values
- Duplicated values
- Data Types
- HTML-encoded entities
- Foreign Languages

#### Before Cleaning:

- 50000 rows
- 6 columns

#### After Cleaning:

- 44335 rows
- 10 columns



# **Establishing Ground Truth**

- 1. Run VADER sentiment analysis tool
- 2. Label sentiments with VADER tool

- 3. Pick out 100 comments and manually label sentiments
- 4. Compare manually labelled VS VADER labelled comments to see accuracy and recall score

	neg	neu	pos	compound	comment	is_negative
0	0.319	0.460	0.220	-0.6240	I feel like he would be really annoying to tal	1
1	0.096	0.904	0.000	-0.3182	"running an engine with no resistance" i think	1
2	0.460	0.540	0.000	-0.5050	I would not consent to this AI crap	1
3	0.000	0.346	0.654	0.9712	Amen guys. All we need is love, love, love. Gr	0
4	0.000	0.741	0.259	0.7131	6:00 "being in a tunnel is as safe as being in	0

Accuracy score of VADER model: **0.91** Recall score of VADER model: **0.75** 

# Pre-modelling, Modelling

- 1. Tokenization
- 2. Lemmatization for interpretability

3. Removal of Stopwords



#### Test - Old Video



Elon Musk: How I Became The Real 'Iron Man'



- Released on Jun 11, 2014
- **15.6 million** views
- **11,000** comments

#### **Test - New Video**

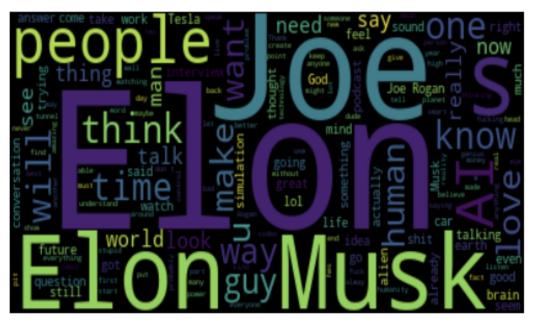


Elon Sues The Lawyers That Forced Him to Buy Twitter



- Released on Jul 19, 2023
- **2.5 million** views
- **9,000** comments

# Analysis (Model) 03

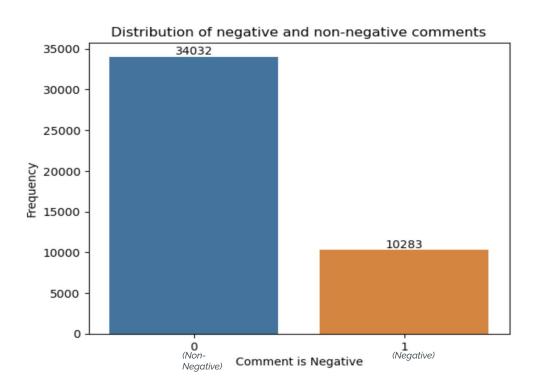














#### **Training Dataset**

Non-negative comment frequency outweighs negative comment frequency





#### **All Models Ran**

	Multinomial Naive Bayes model	Multinomial Naive Bayes model + ROS	Multinomial Naive Bayes model + SMOTE	Random Forest + ROS	Logistic Regression + ROS	Support Vector Machine (Linear) + ROS
Train Set	Recall score: 0.85 Accuracy score: 0.959	Recall score: 0.935 Accuracy score: 0.806	Recall score: 0.860 Accuracy score: 0.864	Recall score: 0.736 Accuracy score: 0.889	Recall score: 0.885 Accuracy score: 0.904	Recall score: 0.936 Accuracy score: 0.935
Test Set	Recall score: 0.346 Accuracy score: 0.811	Recall score: 0.863 Accuracy score: 0.742	Recall score: 0.760 Accuracy score: 0.812	Recall score: 0.666 Accuracy score: 0.839	Recall score: 0.771 Accuracy score: 0.864	Recall score: 0.751 Accuracy score: 0.860
Review	Extremely overfit	Mildly overfit. Chosen.	Mildly overfit. Lower than ROS	Underfit	Strongly overfit.	Very strongly overfit.
Confusion Matrix	1 - 1907 - 1900	0 3885 2600 - 5000 - 6000 - 50	9 - 4866 2337 - 5000 -	9 1023 1023 1023 1023 1023 1023 1023 1023	9 1362 513 400 4000 4000 4000 4000 4000 4000 400	1700 - 17023 000 - 170

#### **Best Model**

We ran 6 models, and found that the best was:

Model 2: Multinomial Naive Bayes model + RandomOverSampler

	Train Set	Test Set
Recall Score	0.935	0.863
Accuracy Score	0.806	0.742

**74.2**% of our model's predictions were correct, and **86.3**% of all negative comments were picked up by our model.

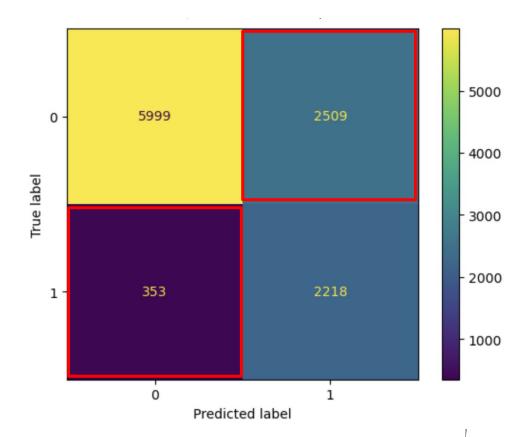
It performed the best, but was still mildly overfitted.



#### Analysis (CM for best model)

We can see that there is quite a lot of false positives compared to false negatives.

This is aligned with what we are trying to do, which is to maximise our recall score.

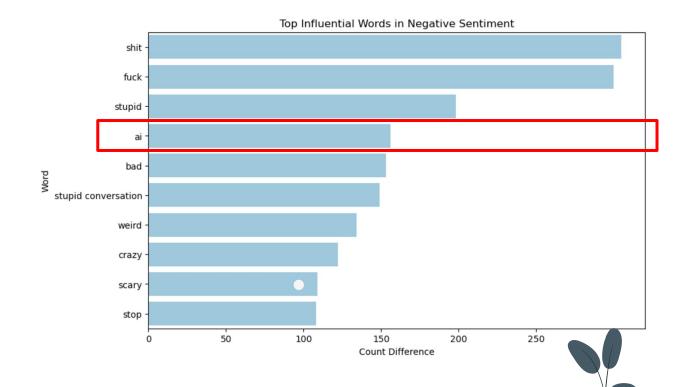




#### **Top Influential Words (Negative)**

These are the most influential words that appear in the negative sentiment.

Count difference refers to the number of times it appears in a negative comment minus the number of times it appears in a non-negative comment.

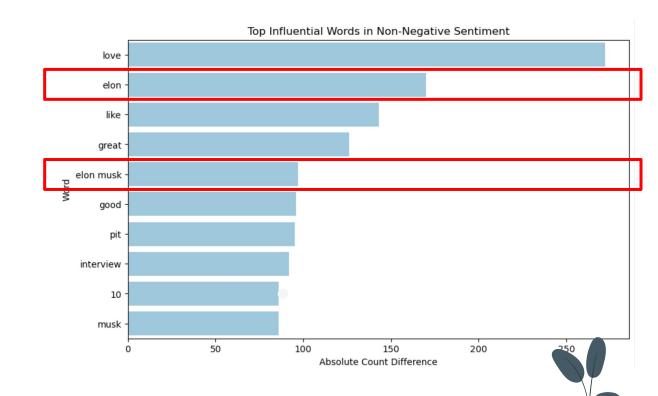




#### Top Influential Words (Non-Negative)

These are the most influential words that appear in the non-negative sentiment.

Count difference refers to the number of times it appears in a non-negative comment minus the number of times it appears in a negative comment.





# O4 Analysis (Test Sets)



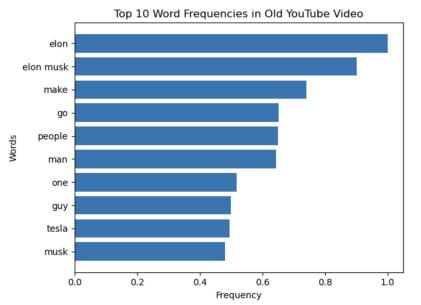
#### **2014 Video**

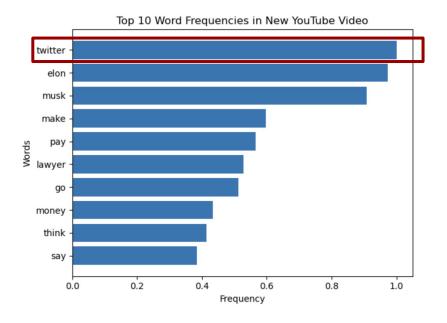
42.2% of comments from a 2014 Elon Musk video are negative



#### 2023 Video

64.8% of comments from a 2023 Elon Musk video are negative









We ran a hypothesis test to see whether the difference is statistically significant.

$$H_0: p_1 = p_2$$

$$H_A$$
:  $p_1 \neq p_2$ 

Using a two-tailed proportion Z-test, we found that the p-value is:

$$p - value = 5.94 \times 10^{-115}$$

So, we can say that there is enough evidence to show a statistically significant difference between the two proportions at the 99% level.







#### **Conclusions**



#### **Best Model**

Multinomial Naive Bayes +

Random Oversampling



### Change in Sentiment

42.2% of comments on older videos about him are negative

Negative comments increased to 64.8% in 2023





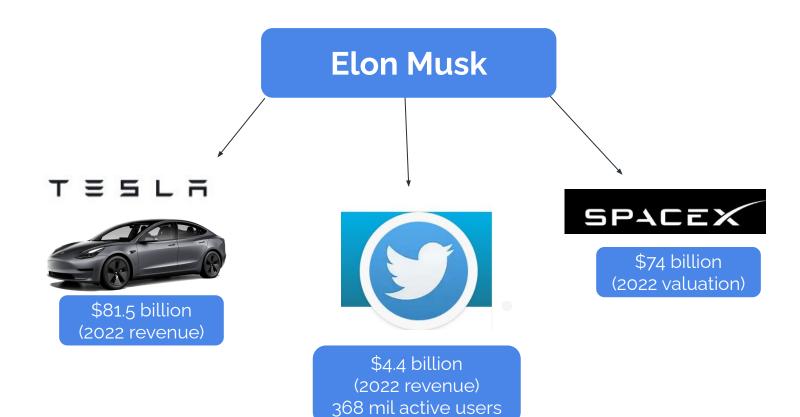
#### **Recommendations & Future Works**

- **1** Using Twitter
  - API limitation on some platforms
  - Best way to represent sentiment analysis is through YouTube videos
  - Twitter's API or online discussion forums
- Balancing Dataset
  - Imbalanced datasets lead to bias in labeling
  - limited exposure to the minority class
  - Random Over-sampling
  - better strategy is to gather more data to train model to improve predictive power
- 3 Deep-diving into VADER
  - VADER lexicon was a decent model to handle emojis and slang
  - Deep dive to check if it attuned to latest slang e.g. 'W rizz', or 'slay'





#### What's at stake



# Thank you!

