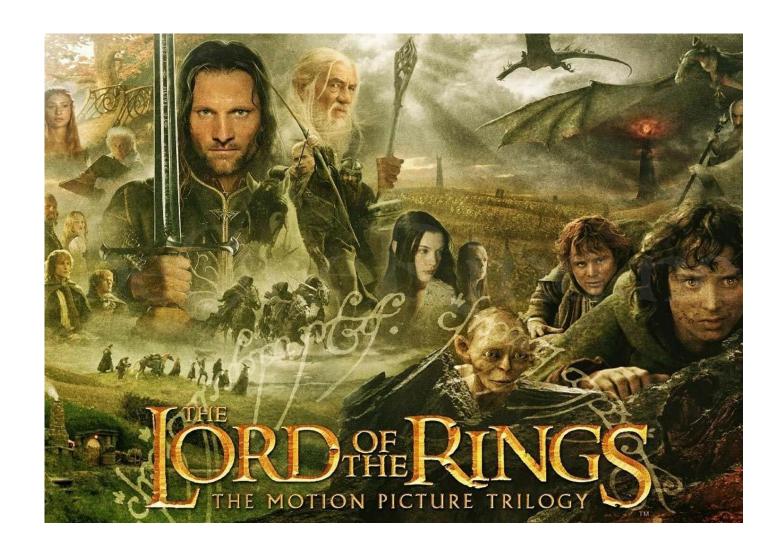
## MOVIE COMMENT SENTIMENT ANALYSIS PROJECT

Uğur Selim ÖZEN 03/16/2022



### **PROBLEM**

Classifying user's comment that made for IMDB movies.

### **METHODOLOGY**

- Writing/Reading Dataset with MongoDB
- Data Cleaning & Data Processing
- Data Visualization for EDA via Plotly
- Different NLP Techniques
- Model Building & Evaluation
- Streamlit App & Deployment

## UTILIZED TECHNOLOGIES

**Dataset Overview** 

• Reviews : **573913** 

■ Target1 : **Sentiment** 

■ Target2 : **is\_spoiler** 

■ Data Size : ~ 900 MB





















# WRITING & READING DATASET FROM MONGODB

```
import json
from pymongo import MongoClient
import pandas as pd
client = MongoClient("mongodb://localhost:27017/")
imdbDB = client["IMDB"]
movie details collection = imdbDB["Movie Details"]
movie_reviews_collection = imdbDB["Movie Reviews"]
movieDetails = [json.loads(line) for line in open('IMDB_movie_details.json', 'r')]
movieReviews = [json.loads(line) for line in open('IMDB reviews.json', 'r')]
movie details collection.insert many(movieDetails)
movie reviews collection.insert many(movieReviews)
query1 = movie details collection.find()
query2 = movie reviews collection.find()
movieDetailsDF = pd.json normalize(list(query1))
movieReviewsDF = pd.json_normalize(list(query2))
movieDetailsDF.to_csv("movieDetails.csv", index=False)
movieReviewsDF.to_csv("movieReviews.csv", index=False)
```

## DATA CLEANING & DATA PROCESSING

```
movieReviews_DF['sentiment'] = np.where(movieReviews_DF['rating'] >= 8, 'positive', 'negative')
movieReviews_DF.sentiment.value_counts(normalize=True)
```

positive 0.543826 negative 0.456174 Name: sentiment, dtype: float64

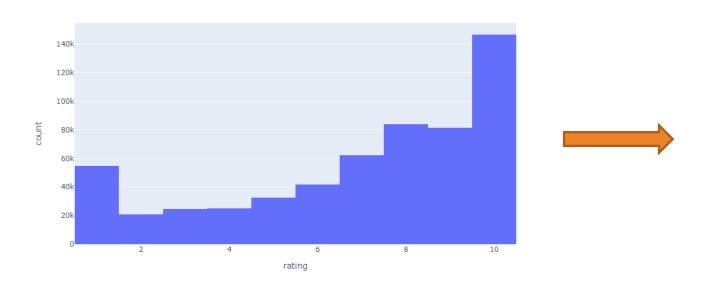
Name: Sentiment, dtype: 110at64

```
# Text preprocessing steps - remove numbers, capital letters and punctuation
alphanumeric = lambda x: re.sub('\w*\d\w*', ' ', x)
punc_lower = lambda x: re.sub('[%s]' % re.escape(string.punctuation), ' ', x.lower())
movieReviews_DF['review_text'] = movieReviews_DF.review_text.map(alphanumeric).map(punc_lower)
movieReviews_DF.head()
```

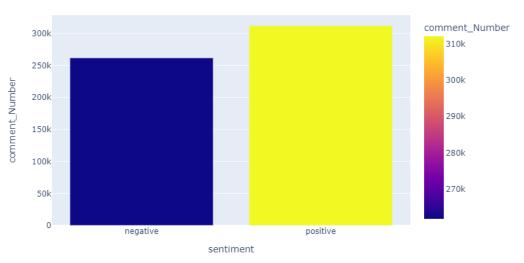
sentiment	review_summary	rating	review_text	is_spoiler	user_id	movie_id	review_date	_id	
positive	A faithful adaption	8	this film starring johnny depp and directed by	True	ur5281145	tt0120669	9 May 2005	622e4fc53acad0a55303b857	547988
negative	If you want to be scared don't watch this	1	this movie was supposed to be scary will	False	ur3688874	tt0185937	9 May 2005	622e4fc53acad0a553041c76	573619
positive	I wanted to CHEER!	10	i enjoy most types of films i seek out epics	False	ur5281697	tt0320661	9 May 2005	622e4fc43acad0a55301206f	378028
positive	Hong Kong Cinema at its Best!	10	i will not review this film as such but i will	False	ur5238145	tt0338564	9 May 2005	622e4fc33acad0a553fceee5	103202
negative	An interesting concept hastily put together 🛘	6	so i m watching the butterfly effect with eyes	False	ur4609782	tt0289879	9 May 2005	622e4fc43acad0a55300eb19	364374

## DATA VISUALIZATION FOR EDA VIA PLOTLY



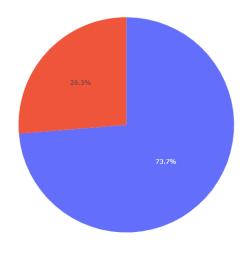


#### Review's Sentiment Distribution

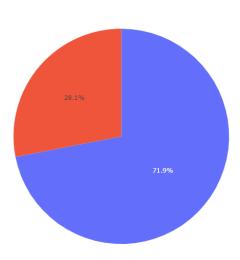


## DATA VISUALIZATION FOR EDA VIA PLOTLY

All Reviews Spoiler Distribution







Spoiler
Not-Spoiler

## DIFFERENT NLP TECHNIQUES – COUNT VECTORIZER

0 0 ...

0 0 ...

0

0

5 rows × 179550 columns

0

0

0

0

## **DIFFERENT NLP TECHNIQUES – TF-IDF**

```
# Create TF-IDF versions of the Count Vectorizers created earlier in the exercise

tfidf1 = TfidfVectorizer(stop_words='english')
X_train_tfidf1 = tfidf1.fit_transform(X_train)
X_test_tfidf1 = tfidf1.transform(X_test)

tfidf2 = TfidfVectorizer(ngram_range=(1,2), binary=True, stop_words='english')
X_train_tfidf2 = tfidf2.fit_transform(X_train)
X_test_tfidf2 = tfidf2.transform(X_test)
```

 $\verb|pd.DataFrame(X_train_tfidf2.toarray(), columns=tfidf2.get_feature_names()).head()|\\$ 

D:\Program Files\Python\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWarning:

Function get\_feature\_names is deprecated; get\_feature\_names is deprecated in 1.0 and will be removed in 1.2. Please use get\_feature\_names\_out instead.

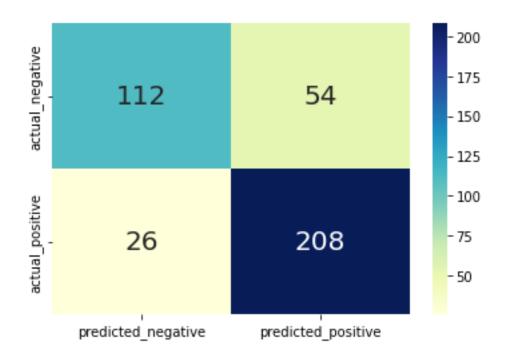
	aan	aan translated	aaron	aaron character	aaron johnson	aaron stamford	aaron taylor	aasif	aasif mandvi	ab	Z	zuckovsky	zuckovsky real	zurer	zurer scientist	zurer tried	zuzu	zuzu petals	zwick	zwick admirable	zwick does
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

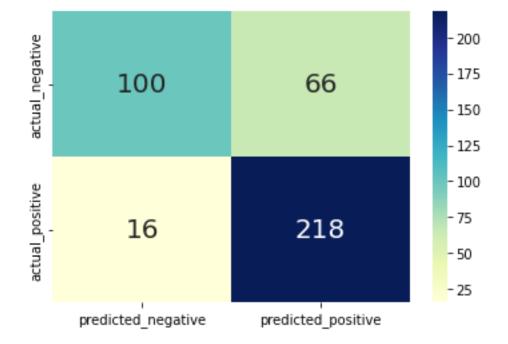
5 rows × 179550 columns

## **MODEL BUILDING & EVALUATION**

	LogReg1	LogReg2	NB1	NB2	LR1-TFIDF	LR2-TFIDF	NB1-TFIDF	NB2-TFIDF
Accuracy	0.765	0.800	0.758	0.595	0.795	0.742	0.668	0.595
Precision	0.780	0.794	0.764	0.591	0.768	0.698	0.639	0.591
Recall	0.833	0.889	0.846	1.000	0.932	0.987	0.991	1.000
F1 Score	0.806	0.839	0.803	0.743	0.842	0.818	0.777	0.743

## TWO BEST MODEL'S COMPARISON





**Logistic Regression – Count Vectorizer with ngram** 

**Logistic Regression – TF-IDF without ngram** 

THANK YOU!

