Project 3: Marvel/DC Reddit Post Classifier & Sentiment Analysis

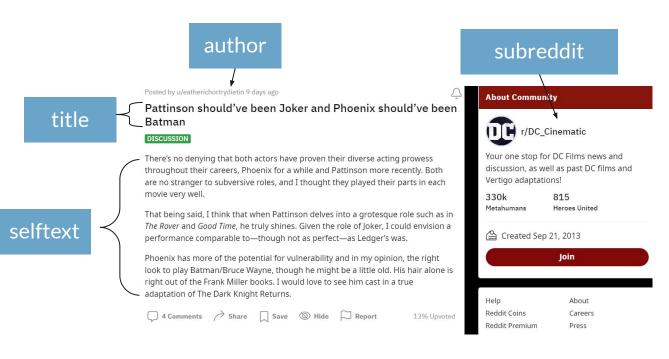
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Problem Statement

We are employees of a marketing agency hired by a toy company to perform market research to **classify posts related to either Marvel or DC** movies in order to:

- a. Build a classifier model that can be applied to other platforms (e.g. Twitter, Facebook) with text data to determine public interest in either movie franchise
- b. Identify which top heroes to create toys that gives most returns

Data Collection & Cleaning - Features



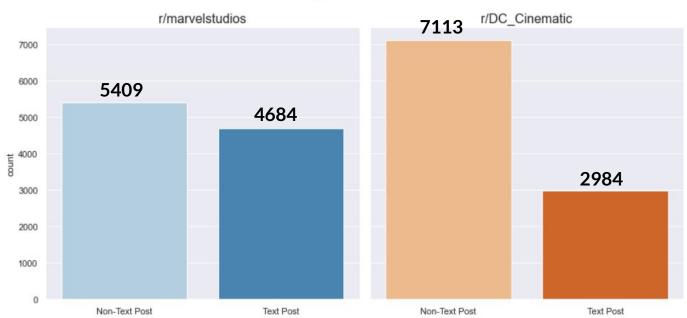
- 1. Total data = 20,000 rows
- 2. Duplicates = 6% of 20,000 post (subsequently removed)

Summary:

EDA: Type of Posts

 To retain the data classification of the post from the two subreddits, the selftext and the title of the post was combined.

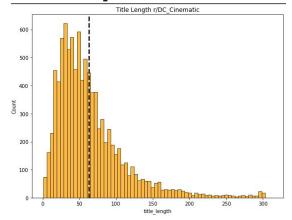


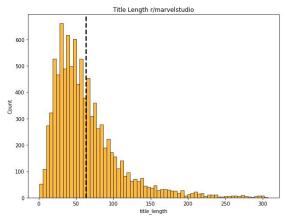


EDA: Text title and selftext of posts



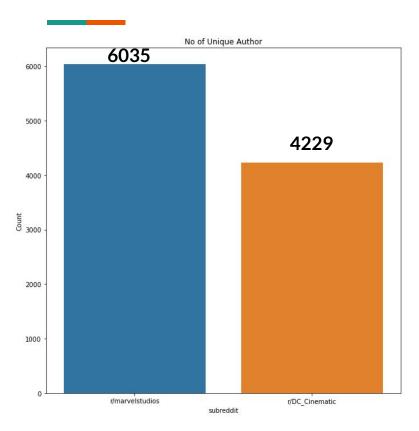






Posts in both subreddits have a similar average characters and title and selftext.

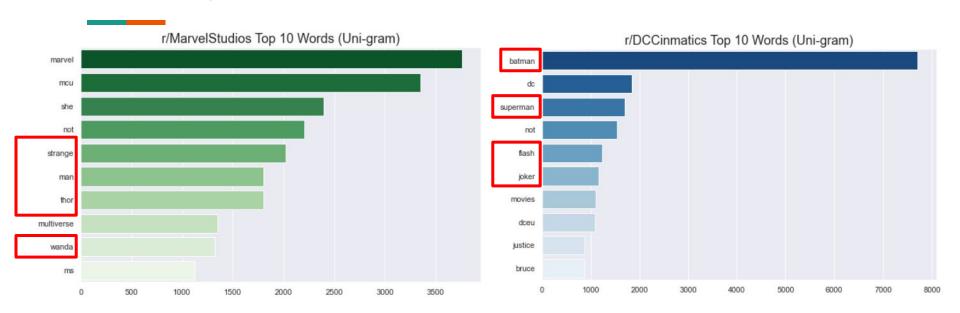
EDA: Unique redditors



Summary:

1. r/marvelstudios has a larger active fanbase on reddit.

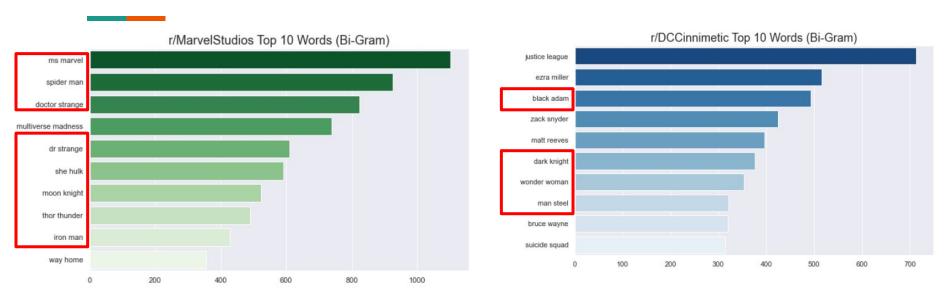
EDA: Uni-grams



EDA Summary:

- 1. Marvel identified characters such as strange, thor, man, and wanda
- 2. DC identified characters such as batman, superman, flash, and joker

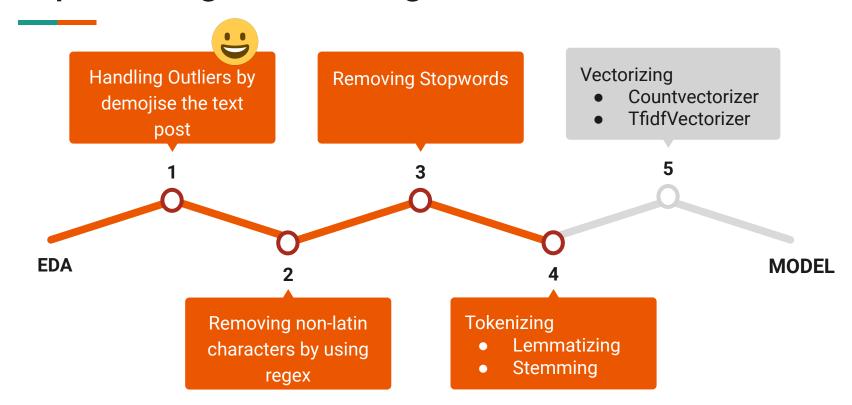
EDA: Bi-grams



EDA Summary:

- Marvel identified characters such as ms marvel, spiderman, dr. strange, she hulk, moon knight, thor and iron man.
- 2. DC identified characters such as black adam, dark knight, wonder woman, man steel.

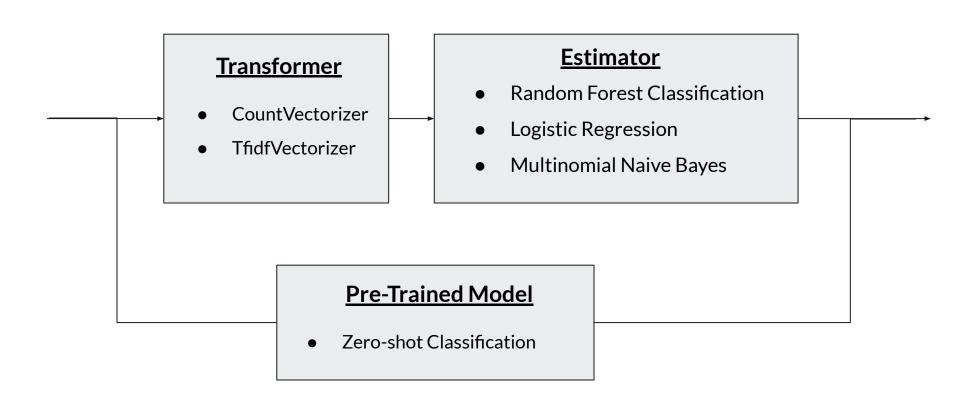
Preprocessing & Vectorizing



Modelling - Baseline Model

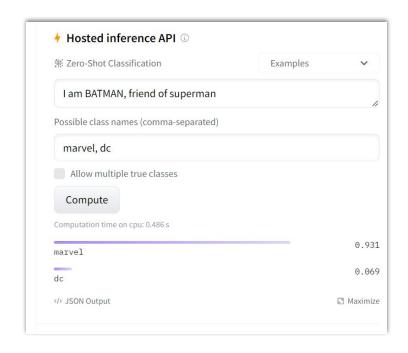
- 1. Normalized value count of target data set as baseline model
- 2. Shows 50/50% distribution between two classes
- 3. Dataset is balance

Modelling - Model Testing



Modelling - Zero shot Classification Model

- 1. Bad test score of around 0.49
- 2. Equivalent to randomly assigning a post to a class
- 3. The word "dc" is too short and generic

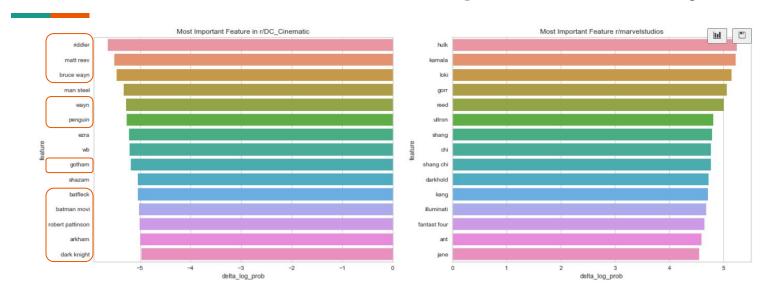


Production Model: Evaluation and Selection

	train_score	cv_score	test_score
logr_cvec	0.910798	0.910798	0.914783
logr_tvec	0.914306	0.914111	0.914783
nb_cvec	0.912422	0.911967	0.916342
nb_tvec	0.913786	0.913526	0.916342
rf_cvec	0.903261	0.902612	0.906989
rf_tvec	0.904496	0.903131	0.905170

- Both transformer perform very similarly
- The model from the GridSearchCV are well fitted
- Narrowed down to Naive Bayes and Logistic Regression model based on the scores

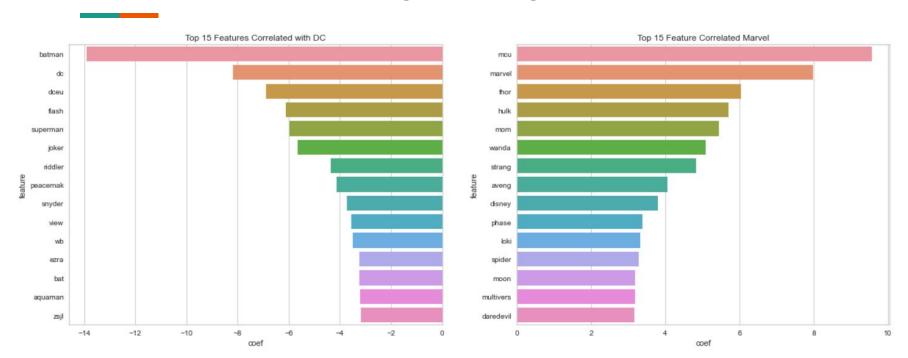
Production Model: Disadvantages of Naive Bayes



There are 2 **significant disadvantages** of Naive Bayes:

- Assumption of independence between words
- Determining feature importance requires the use of predicted probabilities which are known to be unreliable hence the top feature list from NB model may not be a representative list

Production Model: Logistic Regression



- Logistic regression was chosen as the production model
 - The features are more diverse, and independent from one and another

Sentiment Analysis: Choosing a Model

Hand Labelled Dataset

We manually labelled 150 posts each from Marvel and DC as positive/neutral/negative

Model 1

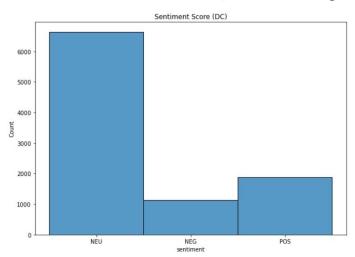
- cardiffnlp/twitter-roberta-ba se-sentiment
- Accuracy: 75.6%

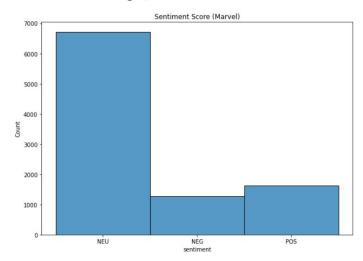
Model 2

- finiteautomata/bertweet-bas e-sentiment-analysis
- Accuracy: 73.3%

- 1. Manually labelled test dataset
 - a. During this process, we noticed that many posts had neutral sentiments so we needed a model that provided neutral labels
- Ran test data on 2 models that were able to produce POS/NEU/NEG labels
- 3. Select the best performing model

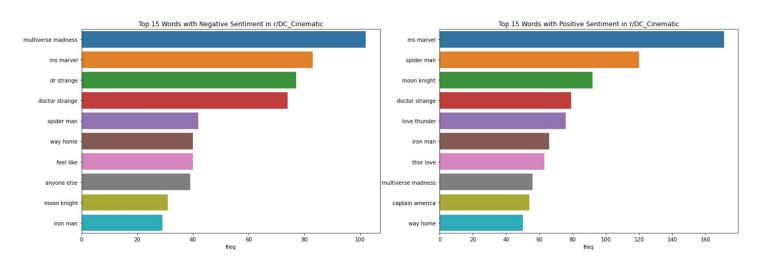
Distribution of posts by sentiment type





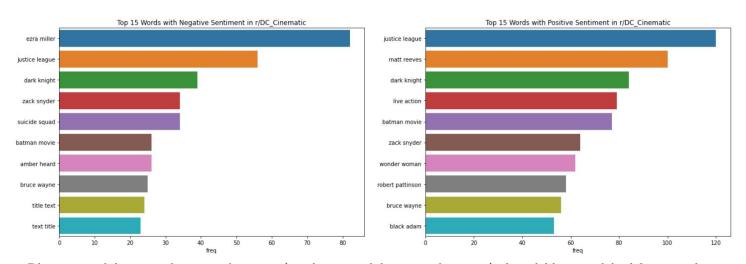
- 1. Majority of posts in both subreddits are neutral
 - a. This is because a lot of the posts are discursive in nature e.g. "What do you think of Dr Strange?"
 - b. However, these posts can still be useful for coming up with new toy/marketing ideas because it's feedback from fans about what they want to see

Top positive and negative bigrams for Marvel



- 1. There's some overlap between top words with negative and positive sentiment
 - a. This is likely because the most popular characters/concepts/movies are likely to have a sizable group of fans and haters

Top positive and negative bigrams for DC



- 1. Phrases with negative sentiments (and no positive sentiments) should be avoided for product releases/marketing
 - a. Certain characters and actors under words with negative sentiments are involved in legal issues/controversy and should be avoided

Conclusion and Future Steps

- The model is able to successfully classify Reddit posts as Marvel/DC content, possible future applications include
 - a. Classify text data from other non-Reddit sources
 - b. Can be used to to determine popularity/public interest in each brand
 - c. Other downstream analysis e.g. sentiment analysis
- 2. Key findings from sentiment analysis:
 - a. Characters to develop toys and marketing initiatives for
 - b. Key characters to avoid