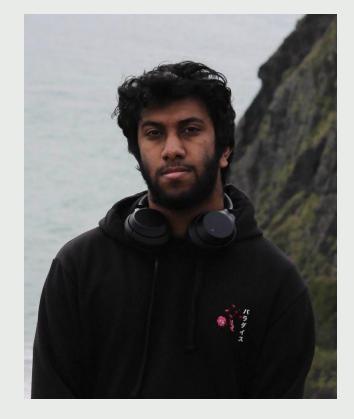
Game Sense: Understanding Player Sentiment Through NLP

Exploring models to predict player sentiment



Bio

- Bachelor of Computing and Information Sciences
- Experience in Customer Service,
 Customer Relations and Logistics
- Aspiring Data Analyst



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Overview

Business Statement

- Reviews are relied on when deciding where to spend
- Misleading reviews can damage trust and game reputation
- Benefits players and developers

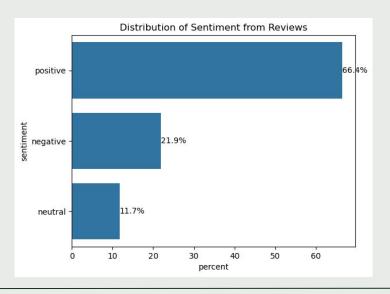
Data Science Statement

- Preprocessed textual review
- Testing various models
- Achieve an acceptable score between 70/80%

Dataset Overview

- Source: Kaggle
- Scrapped from backloggd.com
- 1512 Samples, 13 Features
- Features:
 - Title
 - Release Date
 - o Team
 - o <u>Rating</u>
 - Times Listed
 - o Number of Reviews
 - o Genres
 - o <u>Summary</u>
 - o <u>Reviews</u>
 - o Plays
 - o Playing
 - o Backlogs
 - Wishlist

- Feature Engineering:
 - Vader Sentiment Analysis
 - Count Vectors
 - TF-IDF Word Level
 - TF-IDF N-gram level
 - TF-IDF Character Level



Methods

Train / Test Split	70% Training Set / 30% Test Set
Logistic Regression	A simple, interpretable baseline model
Naive Bayes	A simple, easy to implement and scalable model
Support Vector Machine	Effective in high dimensional spaces while robust to noise and outliers.

Prashil Patel

Models - Baseline (without tuning)

F1-Score	Count Vectors	WordLevel TF-IDF	N-Gram Vectors	<u>CharLevel</u> <u>Vectors</u>
Logistic Regression	74.02	69.93	58.55	70.43
Naive Bayes	60.63	53.68	53.47	55.87
<u>SVM</u>	72.7	73.72	63.57	72.6

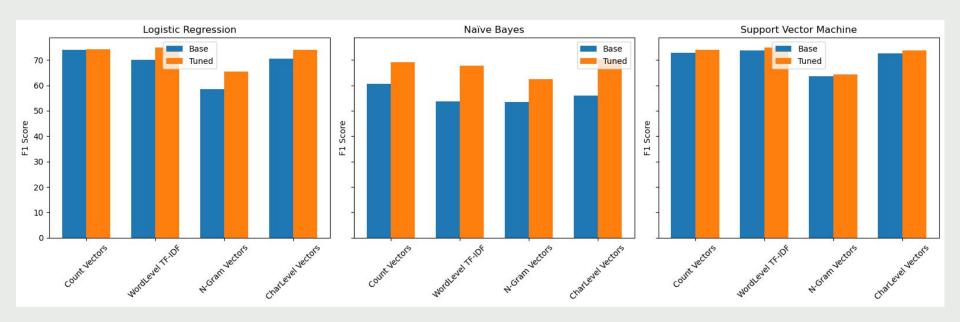
Prashil Patel

Models - Baseline (with tuning)

F1-Score	Count Vectors	WordLevel TF-IDF	N-Gram Vectors	<u>CharLevel</u> <u>Vectors</u>
Logistic Regression	74.2	74.98	65.5	73.9
Naive Bayes	69.22	67.75	62.36	70.29
SVM	74.07	74.96	64.2	73.73

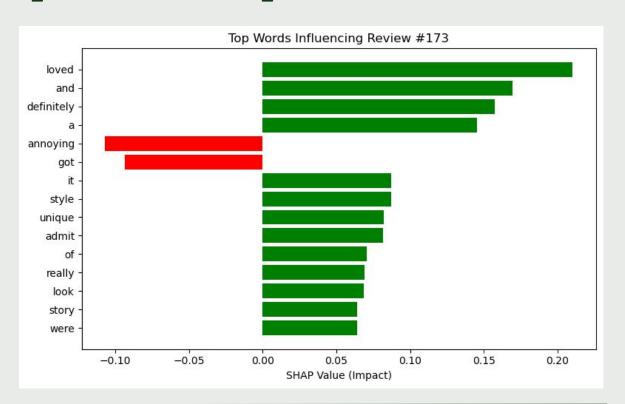
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Findings



Feature importance per Review

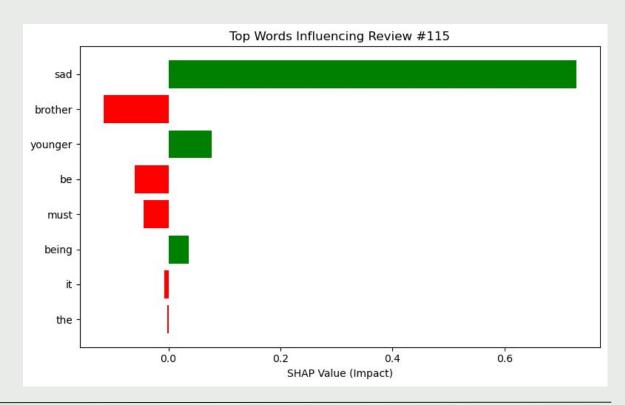
Loved the art style and the story that it told, but a lot of the puzzles were extremely confusing and took a lot of time to figure out. I'll admit, there were a few that I had to look up a guide for, because it got really annoying at a few sections, but it was decent overall and definitely a unique puzzle game. Predicted Sentiment: positive



Feature importance per Review

It must be sad being the overlooked younger brother.

Predicted Sentiment: negative



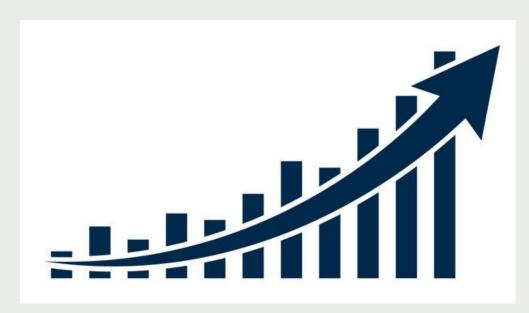
Conclusion

- Best performing model overall is SVM
- Best performing model with vectorizer is Logistic Regression with Word-Level TF-IDF
- Models meets the sentiment analysis requirement at 74.98%
- Models improvement after hyperparameter tuning

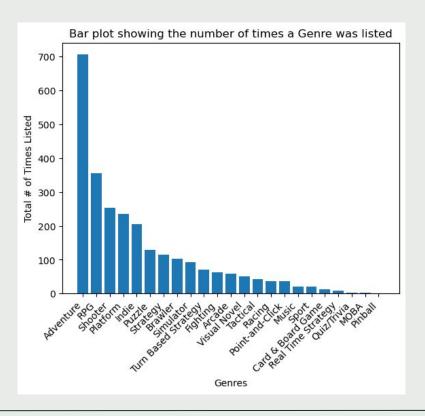


Future Improvements

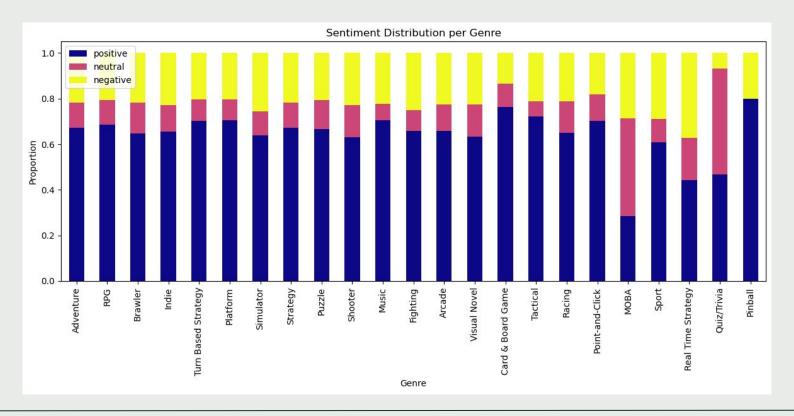
- Deploy an app to predict sentiments
- Improve score
- Bigger dataset
- Deep learning or transformer-based models



Exploratory Data Analysis



Exploratory Data Analysis



Thank You! Q/A?

