



Unlocking Game Success: *Can We Predict Game Reviews from Key Attributes?*

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Research Question

Can we predict the review of a game based on its features such as price, genres, publishers, and other relevant features?

Interest: Guidance for developers to create popular and well received games with respect to metadata (writing descriptions), OS support, discount timeline and strategy, etc based on “successful” games, while exploring consumer habits in game preference.

95 - 100		500+ reviews		positive		overwhelming
85 - 100		50+ reviews		positive		very
80 - 100		1+ reviews		positive		
70 - 79		1+ reviews		positive		mostly
40 - 69		1+ reviews		mixed		
20 - 39		1+ reviews		negative		mostly
0 - 19		1+ reviews		negative		
0 - 19		50+ reviews		negative		very
0 - 19		500+ reviews		negative		overwhelming

Data

Date: 19 May 2024

Source: [Kaggle. Steam Store: a site dedicated to “playing, discussing, and creating games”](#)

Rows: 42,496

Columns: 24



Summary of Data

	genres	release_date	awards	discounted_price
count	42497.000000	42495.000000	42497.000000	42497.000000
mean	2.877685	737419.252006	0.309528	371.064969
std	1.351021	1227.812971	1.264100	1050.559189
min	1.000000	729205.000000	0.000000	0.000000
25%	2.000000	736594.000000	0.000000	80.000000
50%	3.000000	737643.000000	0.000000	250.000000
75%	4.000000	738445.000000	0.000000	480.000000
max	11.000000	739138.000000	41.000000	150000.000000
unique	NaN	NaN	NaN	NaN
top	NaN	NaN	NaN	NaN
freq	NaN	NaN	NaN	NaN

	overall_review
count	42497
mean	NaN
std	NaN
min	NaN
25%	NaN
50%	NaN
75%	NaN
max	NaN
unique	10
top	Very Positive
freq	11146

Total number of null values in column

app_id	0
title	0
release_date	57
genres	87
categories	45
developer	190
publisher	211
original_price	37638
discount_percentage	37638
discounted_price	240
dlc_available	0
age_rating	0
content_descriptor	40122
about_description	138
win_support	0
mac_support	0
linux_support	0
awards	0
overall_review	2477
overall_review_%	2477
overall_review_count	2477
recent_review	36994
recent_review_%	36994
recent_review_count	36994

Number of values in column

app_id	42497
title	42497
release_date	42440
genres	42410
categories	42452
developer	42307
publisher	42286
original_price	4859
discount_percentage	4859
discounted_price	42257
dlc_available	42497
age_rating	42497
content_descriptor	2375
about_description	42359
win_support	42497
mac_support	42497
linux_support	42497
awards	42497
overall_review	40020
overall_review_%	40020
overall_review_count	40020
recent_review	5503
recent_review_%	5503
recent_review_count	5503

Percentage of null values in column

app_id	0.000000
title	0.000000
release_date	0.134127
genres	0.204720
categories	0.105890
developer	0.447090
publisher	0.496506
original_price	88.566252
discount_percentage	88.566252
discounted_price	0.564746
dlc_available	0.000000
age_rating	0.000000
content_descriptor	94.411370
about_description	0.324729
win_support	0.000000
mac_support	0.000000
linux_support	0.000000
awards	0.000000
overall_review	5.828647
overall_review_%	5.828647
overall_review_count	5.828647
recent_review	87.050851
recent_review_%	87.050851
recent_review_count	87.050851

Features of Interest

overall_review: The overall review category classification based on the rating score of the game (target)

genres: List of genres the game belongs to

release_date: Date the game was published

awards: Number of awards the game has received

discounted_price: Price after the discount (as of the time the data was scraped)



Additional Features

categories: Types of game content / features

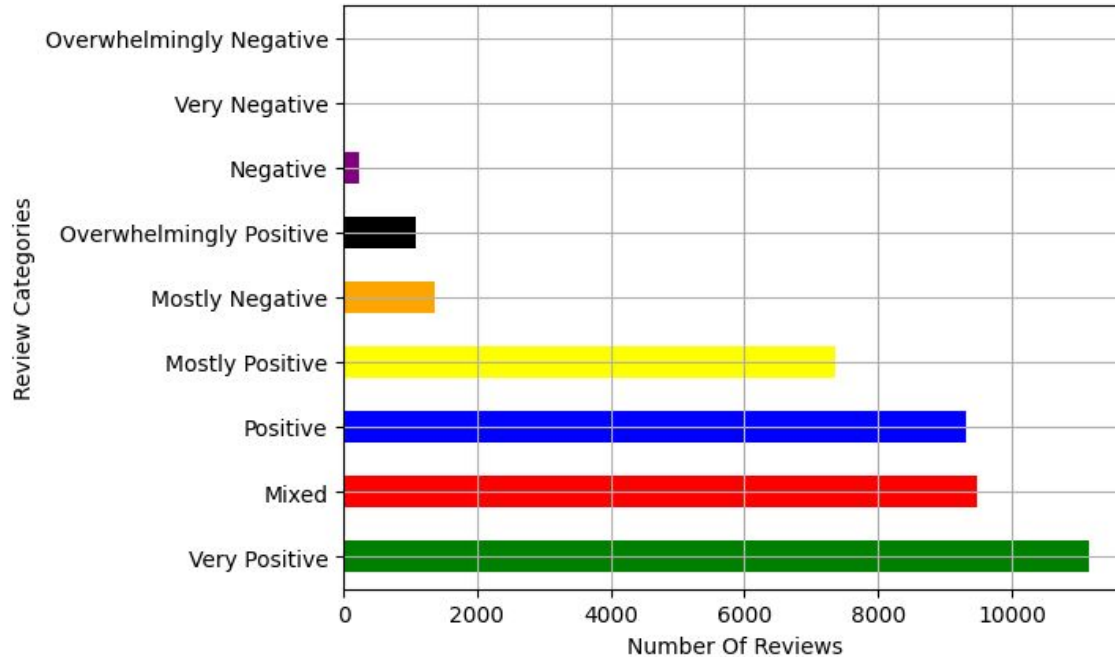
about_description: A textual description about the game

mac_support: Binary indicators of platform support
(25% of the games support mac)

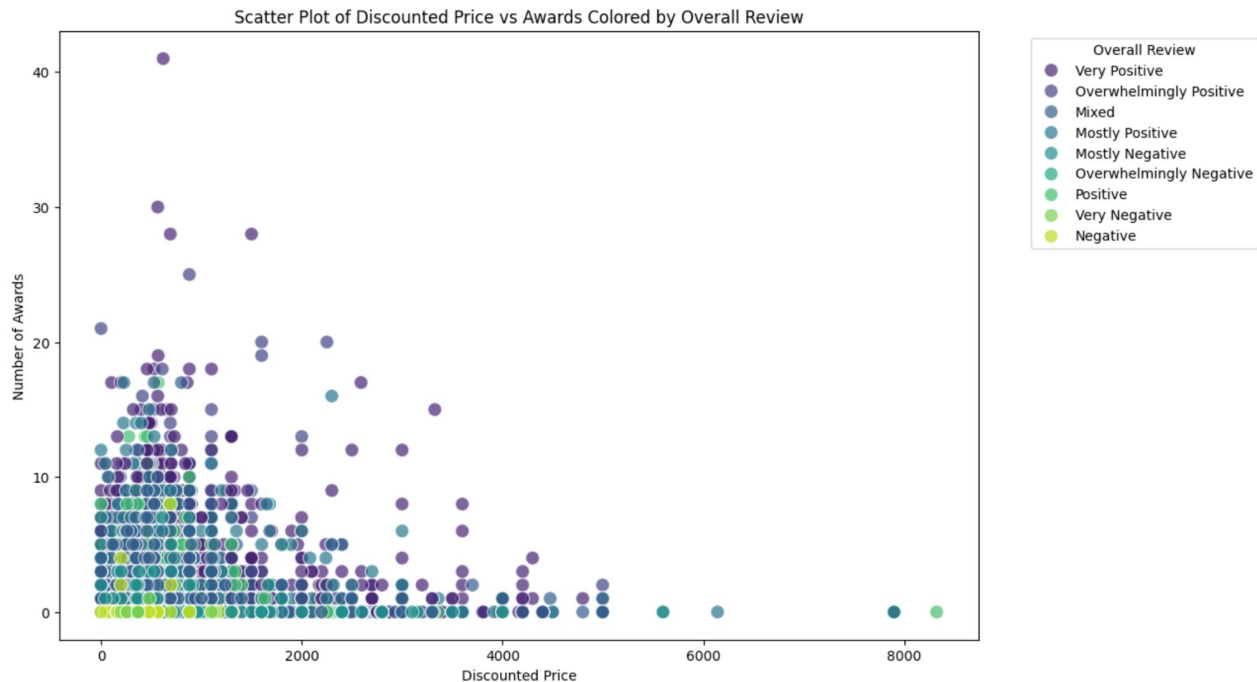
developer: The company or individual who developed the game



Outcome Variable: Review



Relationship between Price, Awards and Review



Baseline Prediction Model

Multi-class Logistic Regression

- Simple, interpretable
- Probabilistic outputs
- Feature importance



Prediction Algorithms

- **Decision Trees**


Are intuitive and easy to interpret, allowing us to visualize how different features impact the final decision, which facilitates explaining the results to non-technical users.

- **XGBoost**

An optimized version of decision trees, is highly effective in classification due to its ability to handle large datasets and enhance performance through boosting techniques. We will assign a higher penalty to misclassifications of the minority class using scale weights parameter.

- **Neural networks + hidden layers**

Powerful for capturing complex non-linear relationships and patterns in the data. Their ability to model intrinsic complexities is essential in a domain as variable as video game preferences, where feature interactions can be highly sophisticated.



Performance Metrics

- Accuracy: Measures the ratio of correctly predicted instances to the total instances, can be misleading with class imbalance.
- Precision, Recall, and F1-Score
- Confusion Matrix
- Balanced Accuracy
- ROC-AUC and Precision-Recall AUC



Questions?

