

GOOGLE PLAY STORE APPS

Group 7

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WHY ANALYZE THE GOOGLE PLAY STORE?



Mobile App Market is set to grow 20% by 2023



Android Apps comprise 90% of the Mobile App Market



What makes an App popular? Can we predict how popular it's going to be?



What are some interesting patterns in user behavior related to app usage & feedback



OVERVIEW OF ANALYSIS

Data Cleaning

Data Exploration

Predictive Modeling



Understand the structure of the dataset and clean data before analysis



Uncover initial patterns, characteristics, and points of interest using visual exploration



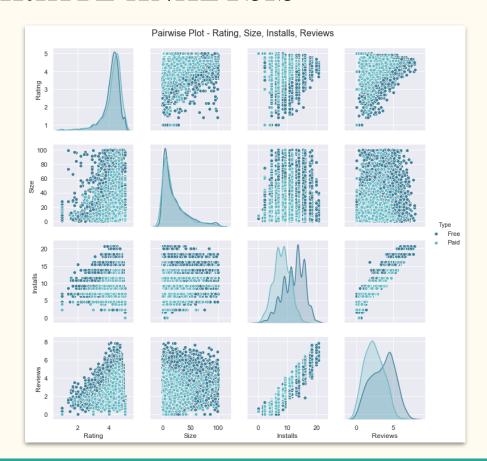
Formulate a statistical model to forecast an outcome using relevant predictors



EXPLORATORY ANALYSIS

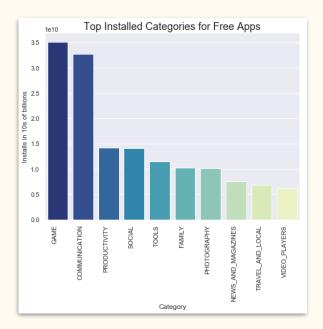


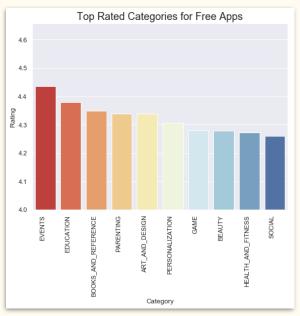
BIVARIATE ANALYSIS





CATEGORIES IN DEMAND

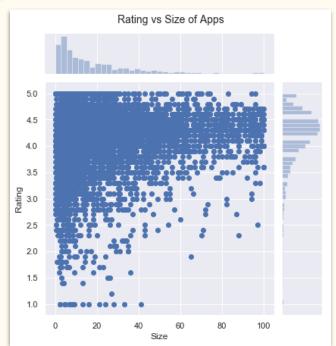


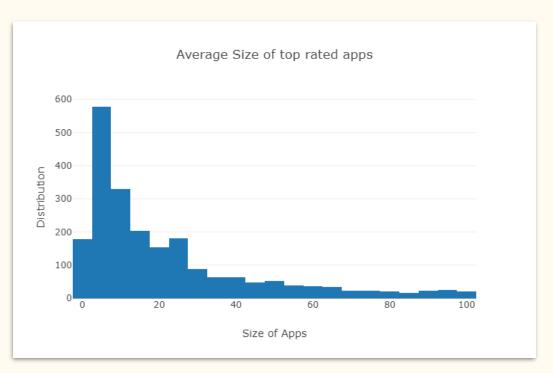


- Communication, Productivity, Tools, Family, Photography, News & Magazines, Travel & Local, Video Players are untapped free app categories
- High Installs because of a high underlying demand
- Low user satisfaction



SIZING STRATEGY

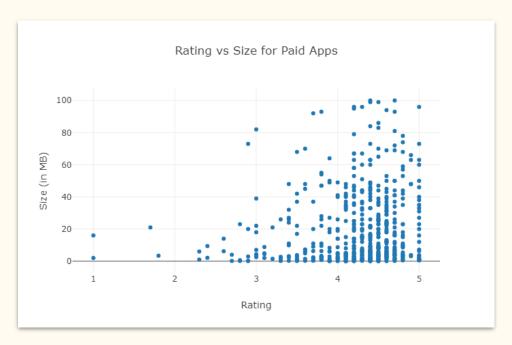


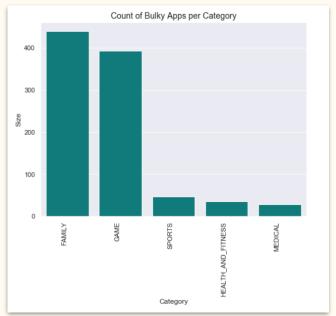


Highly rated apps are optimally sized between ~2 MB and ~40 MB



SIZING STRATEGY

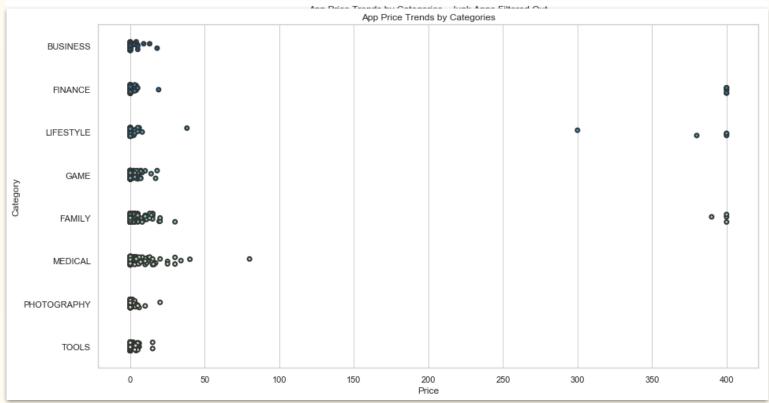




Users prefer to pay for apps that are light-weighted

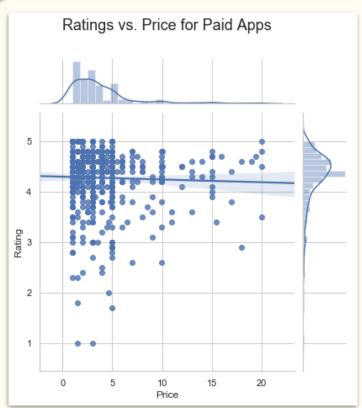
PRICING STRATEGY

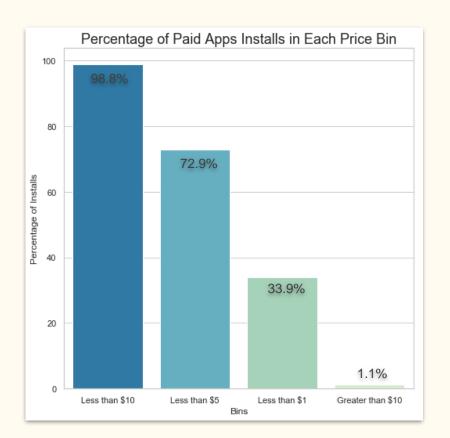






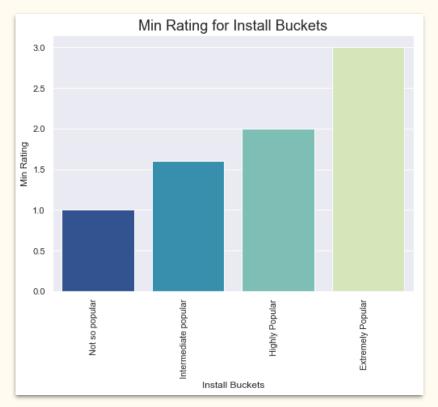
PRICING STRATEGY







HOW DOES MINIMUM RATING VARY WITH INSTALLS?

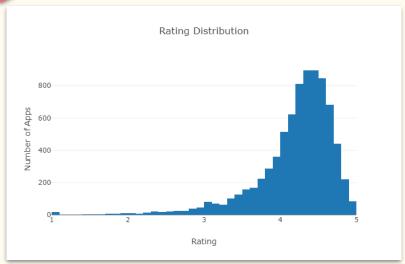


Installs	Install Buckets
0 - 10000	Not so popular
10000 - 100000	Intermediate popular
100000 - 500000	Highly Popular
500000+	Extremely Popular

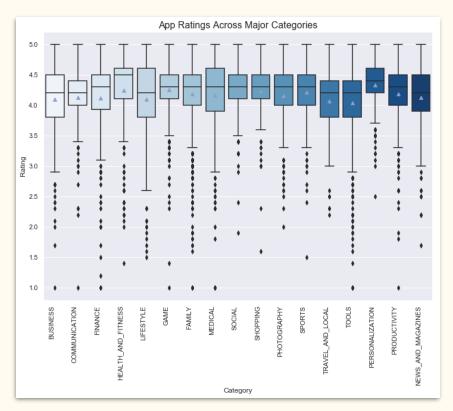
Finding: An app that finds a popular audience is less likely to be critically rated.



RATING DISTRIBUTION

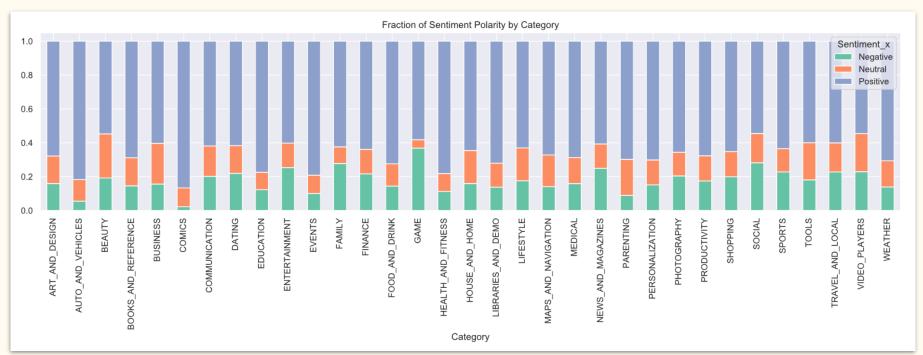


- Average rating of apps is 4.17
- One-way Anova test revealed that the average ratings of categories are statistically different
- Best performing apps Health and Fitness
- Worst performing apps Dating



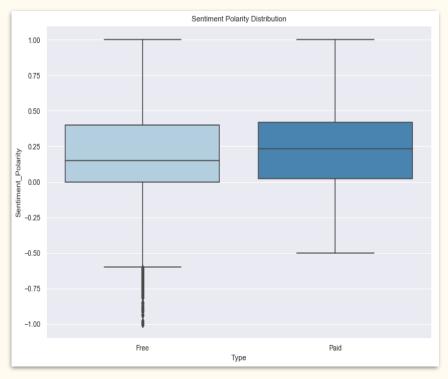


SENTIMENT POLARITY BY CATEGORY





SENTIMENT POLARITY BY TYPE (FREE/PAID)



People are harsher towards free apps whereas users are more tolerant when they are paying for it.



WORD CLOUD - POSITIVE REVIEWS



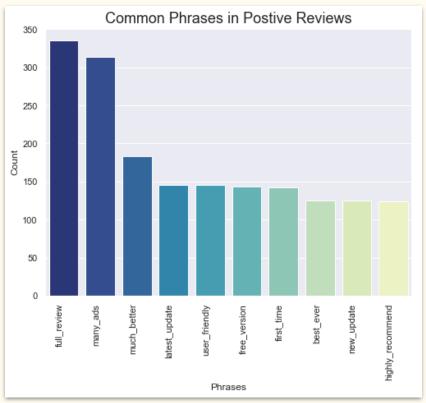


WORD CLOUD - NEGATIVE REVIEWS





COMMON PHRASES - POSITIVE REVIEWS



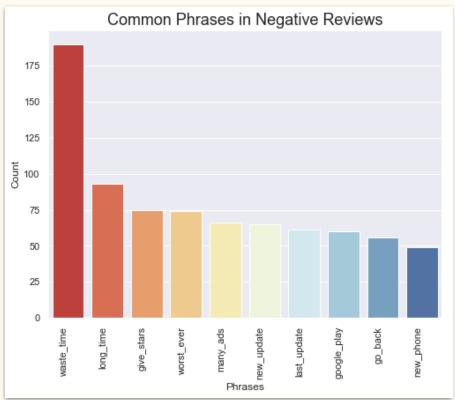
Common phrases in positive reviews:

- User-friendly
- Free version
- Works great
- Highly recommend

Usability is one of the main reasons for positive reviews



COMMON PHRASES - NEGATIVE REVIEWS



Common phrases in negative reviews:

- Waste time
- Many ads
- Spend money
- Takes forever

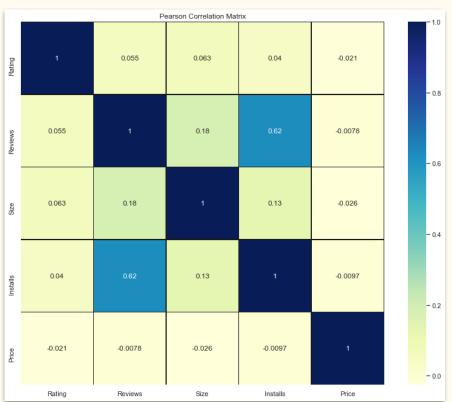
Loading time is one of the main reasons for negative reviews



PREDICTIVE MODELING



HOW TO PREDICT THE POPULARITY OF AN APP



Based on the correlation plot, we identified the following variables that we could use in our classification problem:

- 1. Category
- 2. Genre
- 3. Reviews
- Content Rating
- 5. Size
- 6. Type



LOGISTIC REGRESSION AND DECISION TREES

Logistic Confusion Matrix					
Not so popular	204	4	31	29	-400
Intermediate Popular	0	21	189	67	-320
Popular	0	18	472	43	- 240 - 160
Extremely Popular	24	24	152	128	-80
	Not so popular	Intermediate Popular	Popular	Extremely Popular	-0

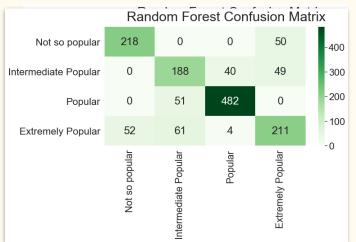
Decision Tree Confusion Matrix							
Not so popular	210	2	0	56	-400		
Intermediate Popular	0	179	34	64	- 320		
Popular	0	52	480	1	- 240 - 160		
Extremely Popular	59	66	2	201	-80		
	Not so popular	Intermediate Popular	Popular	Extremely Popular	-0		

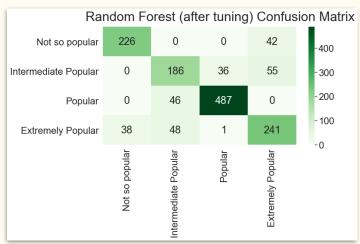
We bucketed installs into 4 bins based on quartile values & used test & hold out sets to perform the following classification algorithms:

- 1. Logistic Regression ~ 58.67%
- 2. Decision Trees ~ 76.10%



HYPERTUNING RANDOM FOREST





We used hypertuning & 3 fold cross validation to obtain a optimal classification accuracy of 81.08% on Random Forest Classifier

Parameters used:

- No of trees
- 2. Level of depth of the trees
- 3. Minimum number of samples required to split a node

- 4. No of variables to consider at each split
- 5. Min/Max no of samples required at each node





SIZE

Highly rated apps were optimally sized between 2MB to 40 MB

Paid apps with specific functionality were lighter



PRICE

Free apps outperform paid apps

If an app is paid, the likelihood of it being popular is higher if it's priced under \$10



RATING

Positive relation between installs and rating.

However, highly installed apps are not always highly rated



Issues like loading time and positive features like usability were revealed by sentiment analysis.

People tend to review harsher for paid apps



QUESTIONS?