

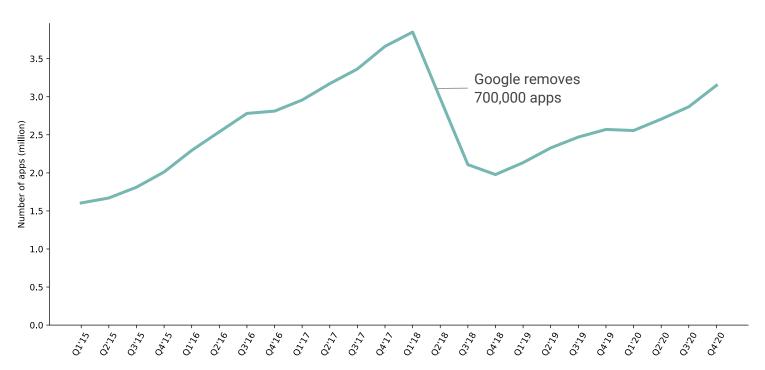
# Exploring the Google Play app store

App analysis and recommender

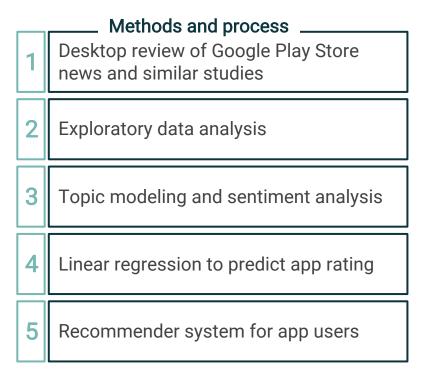
Tawney Lott 22 March 2021

## With over 3 million apps, the Google Play store is a rich resource for app developers and users

Consistent upward trend in number of apps available, despite Google clean up



### The analysis drew on quantitative and qualitative methods and utilized a range of tools



### Analysis tools

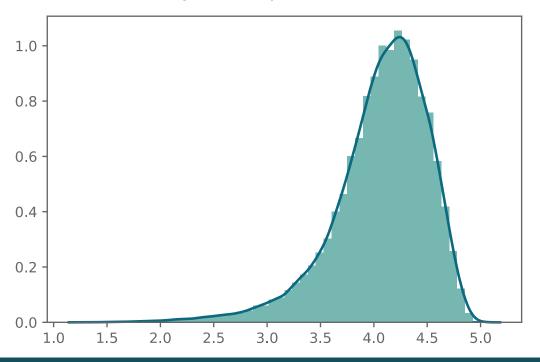
- Python
- Pandas
- Numpy
- Scikit Learn
- Natural Language Toolkit

### **Visualization tools**

- Matplotlib
- Seaborn
- Tableau
- Streamlit

### The sample included data for 22,000 apps which were scraped from the Google Play store

There is a left skew in the distribution of app ratings, which is the target for the predictive model



### **Features**

- App name
- Developer
- Description
- Release date
- Category
- Free (yes / no)
- Price
- Number of ratings
- Number of reviews
- User reviews
- Content rating
- Date of last update

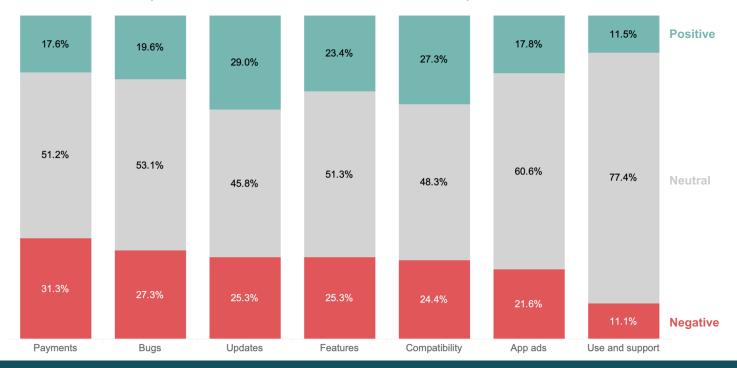
### Topic modeling was used to understand the underlying topics of user reviews

The analysis revealed **seven core topics** and the top words within each topic

Features	Option	Change	Start	Find	Feature
App bugs	Fix	Big	Issue	Crash	Load
Payments	Money	Pay	Real	Card	Purchase
In-app ads	Ad	Watch	Remove	Click	Force
Device compatibility	Work	Stop	Device	Reinstall	Button
App updates	Update	Last	Latest	Year	Before
Useability and support	Help	Learn	Find	Account	Support

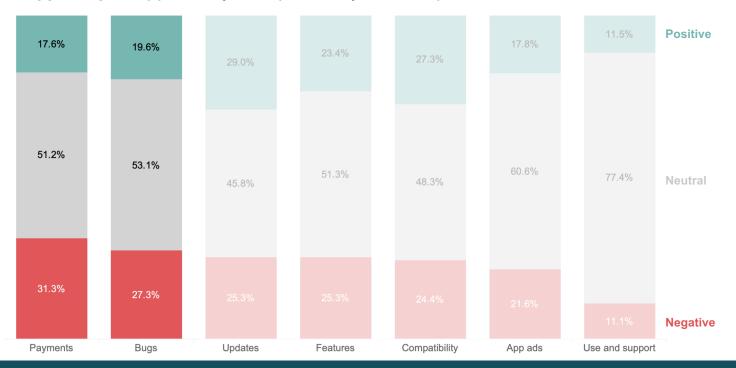
### Sentiment analysis of user reviews was used to provide an additional layer of insight across the seven topics

While the majority of reviews appear to be neutral, the share of negative reviews exceed positive reviews across most of the topics



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The Payments topic has a larger share of negative reviews than Bugs, suggesting an opportunity to improve Payments experience



## These layers of analysis were combined with the data scraped online to predict average user ratings for each app

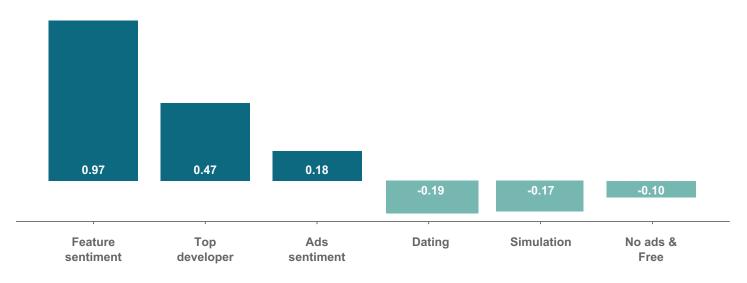
### Results

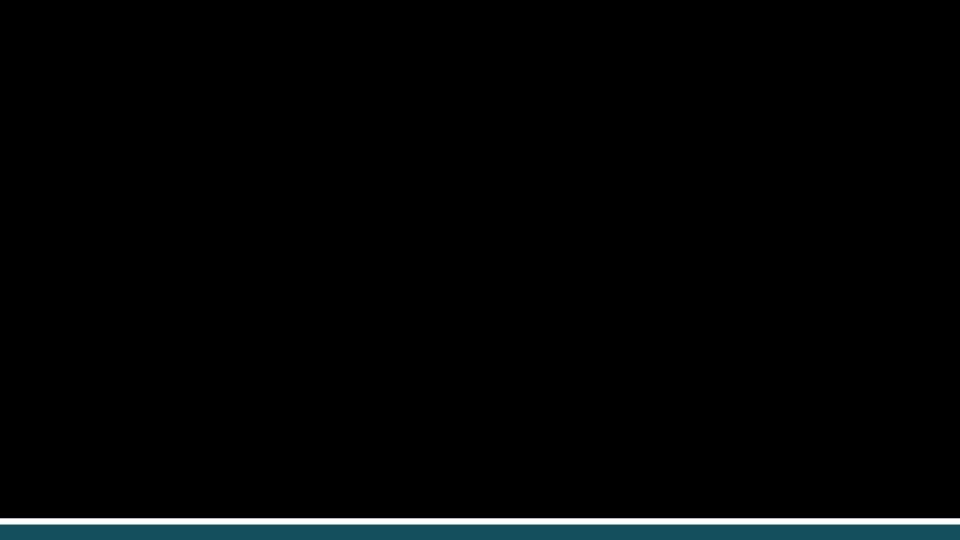
- The model explains 31% of the variance in user ratings
- MAE: 0.28
- Best performance on apps with a score in the range of 3.8 to 4.5 average user rating
- Performed less well on lower-rated apps

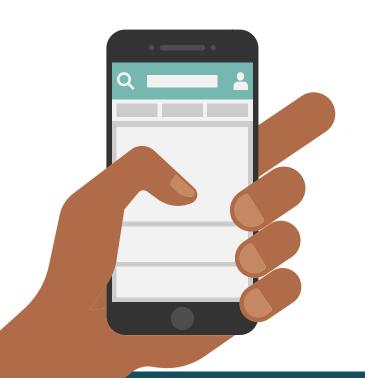


## The model identified the strongest features in predicting appratings

Positive regression coefficients are associated with higher user ratings, while negative coefficients are associated with lower ratings







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 User sentiment regarding Features is strongest predictor of ratings - give the user what they want



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- Payments results suggests there is an opportunity to improve payments experience and process for apps
- Apps by Top Developers tend to have better ratings - important to establish and maintain brand
- Apps that are free and do not contain ads tend to have lower scores - serious developers should investigate ads and / or payment models