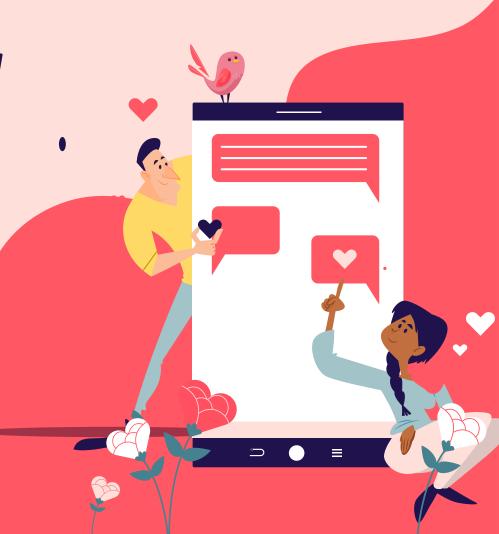
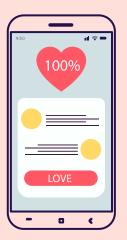
DATING APP REVIEW ANALYSIS & RECOMMENDATION

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Group 5
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Background & Research Question

Over **323 million** people worldwide are using dating apps. Tinder, Bumble and Hinge are the top three players in the market, occupied a **69%** worldwide market share in total.



"Is terrible- there are 50 different tabs to show you the same people-but in different ways and it never syncs- go back to chronological"

-USER###

In the research, we aims to analyze comments and ratings of dating apps and investigate what features of the dating apps users care the most about.

The research means to **help the dating app companies** to understand
the users' demand and favor to
improve the satisfaction by developing
the apps.



Data Source & Analysis Techniques





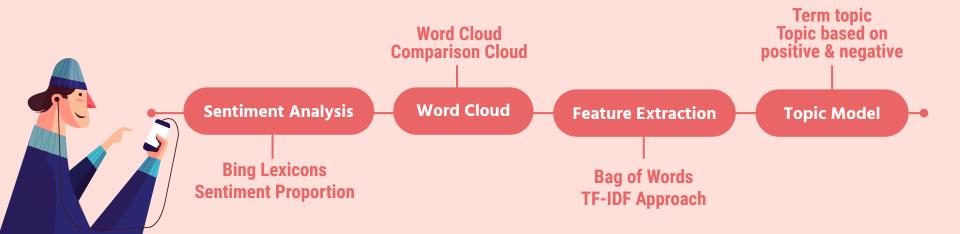


DATA EXTRACTION

- 'Apprler' package to extract review and rating data from Apple App Store.
- Collected reviews were posted from January 20th 2022 to April 14th 2022.
- Nearly 1000 reviews for each app.

DATA PREPARATION

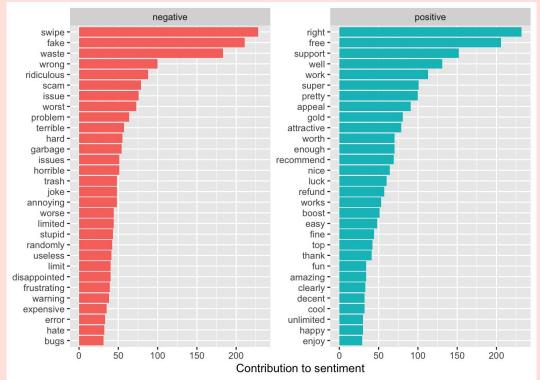
- Create Corpus
- Clean Text (particularly, remove meaningfulness stopwords)
- Tokenize Document Term Matrix





SENTIMENT ANALYSIS

We use bing lexicons to yield sentiment score for each word, determining the positive and negative words of user reviews.



Proportion of positive (and negative words) for each rating

Words such as "swipe", "fake", "scam", "expensive", and "error" show negative sentiments.

Features that need to be improved: user profile accuracy audit, pricing, system operation, etc.

positive

wonderfulthoughtful accurate leads appropriate recommended enjoying incredibly strong standout beautiful willing helping significant willing interesting interesting fair liking genuine honest successuseful successful suc favorite amazing top luck recommend safe trust respect luckyfast clearly worth Super nice fun easier meaningful interests cool pretty support refund wow appreciate advanced decent advanced d advanceddecent sensitive happygold welled free enough perfect protect popular enjoy appeal right hot attractive correctly unmatched unmatched violation unable limit lose problem hard error dumb fraud buggy unfortunable limit lose problem hard error dumb fraud buggy refused glitch wasting frustrating unfortunately unable limit lose problem hard error dumb fraud buggy expensive gloss bug line weed confused warning mad bugsdifficult sorry dead weed confused warning inappropriate shady pointless gross bug insane messed shady pointless gross bug lame desperate beware glitches harassment

negative

however attractive esponse conversation show feel things explanation option saying sure send ly corresponds to filter using customer point message appeal t without terms year another ho someone say put already account findsomething never months free - waste, fake

women - months free - waste, fake

never months free - waste, fake

women - months free - waste, fake

never months free - waste, fake

never months free - waste, fake

women - months free - waste, fake

never months free - waste, take talk feature report ole's

ey, dating, time se, someone, free swipe, message, waste



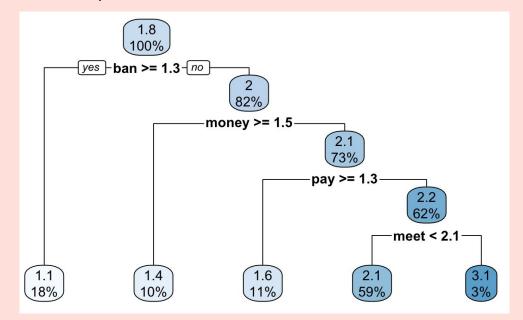


- TF (term frequency):
 Term frequency works by analyzing the frequency of a particular term relative to the document.
- IDF (inverse document frequency):
 Inverse document frequency works by looking at how common (or uncommon) a word is in a corpus.

Next, we built **the decision tree** diagram for prediction. The selected tree contains **4 variables and 4 splits.**



CART is based on the known probability of occurrence of various situations, by forming a decision tree to find the probability that the expected value is greater than or equal to zero.





Topic Modeling(LDA)

Topic modeling is an useful way to discover topics that customers care about from reviews. It is an extension to detect customer attitudes based on sentiment analysis. We can calculate and visualize the probability that each review is associated with each topic.

LDA analysis and categorized all words into four topics.

Topic 1:Payment system, swipe feature

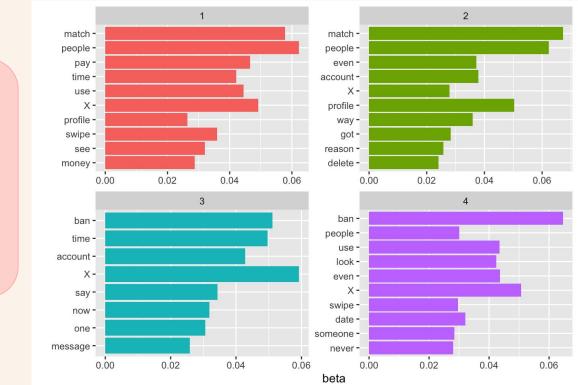
Topic 2: User Accounts, Profiles

Topic 3: User account, message feature,

Prohibited Features

Topic 4: Prohibited Features, swipe

feature





TOPIC MODEL BASED ON POSITIVE AND NEGATIVE

Two-topic models of positive and negative reviews are created to find out which aspects of the customer experience come up in positive and negative reviews.

Positive Reviews:

Topic1: Customer support

Topic2: Free and premium features

(Gold subscription)

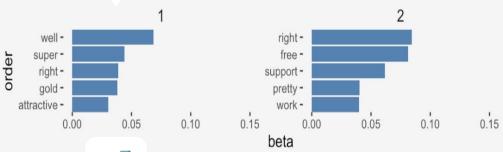
Negative Reviews:

Topic1: User Information authenticity

Topic2: Swipe design of the app



Positive review topics





Negative review topics





CONCLUSION AND RECOMMENDATIONS



Conclusion

Positive emotion: The positive emotions of users come from the commercial use of social software, and the functions of matching users, promoting social interaction with strangers, customer service support, free download and use.

Negative emotion: The negative sentiment comes from other users' untrue information, receiving spam, the design of the app's sliding function, paid-for-use features of apps, payment systems, and account-related activities.

Recommendation

- Enhance users' identity reviews to prevent identity theft and fraud
- Improve the security optimization of the background, especially for specious words
- Develop the questionnaire that is sent to users in the beginning to upgrade the satisfaction of the "swipe" and match function

