GrandPad Reviews Analysis

October 28, 2024

1 Retrieving and Preprocessing GrandPad Google Play Reviews

Import required libraries.

```
[1]: import pandas as pd
     from langdetect import detect_langs
     import re
     import seaborn as sns
     import matplotlib.pyplot as plt
     import os
     from pathlib import Path
     from utils.review_utils import get_google_play_data
     from sklearn.feature_extraction.text import CountVectorizer
     from sklearn.decomposition import LatentDirichletAllocation
     import re
     import nltk
     from nltk.corpus import stopwords
     from nltk.stem import WordNetLemmatizer
     from pathlib import Path
     import os
     import pyLDAvis
     import pyLDAvis.lda_model
     from utils.review_utils import find_reviews_by_keyword_list
```

1.1 Scraping Data

First, scrape the reviews from the Google Play Store.

```
[2]: # Get reviews
APP_ID = 'net.grandpad.puma'
reviews = get_google_play_data(APP_ID)
df = pd.DataFrame(reviews)

# Print a few reviews (just the score and content)
print('Score Content')
print('-----')
for i in range(5):
    print(f"{df.iloc[i]['score']:<7}{df.iloc[i]['content']}")</pre>
```

Score Content

- 5 Love the app for keeping my family in touch.
- 5 Grandpad is a great way to stay connected to those near and far.
- 5 My 94 year old aunt always tells me how much she loves her GrandPad! She's become very proficient at making phone calls, & sending voice & text messages. And she loves the photo display!
- 5 Enjoyed and used by the whole family
- 4 Great medium to remain in contact with our loved ones.

```
[3]: # Print number of rows in the data
rows0 = df.shape[0]
print(f'Initially there are {rows0:,} reviews')
```

Initially there are 1,823 reviews

The length of the reviews might give important insight into the quality of the review.

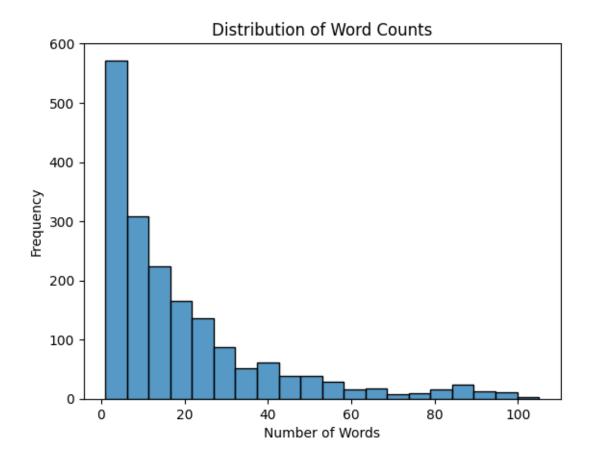
```
[4]: # Make column for word count (of the review content)

df['word_count'] = df['content'].str.split().apply(lambda s: 0 if s == None

else len(s))
```

Plotting the distribution of the word counts of the reviews, we see that the vast majority of the reviews contain fewer than 10 words.

```
[5]: # Histogram of the word count of the reviews
sns.histplot(df['word_count'], bins=20)
plt.xlabel('Number of Words')
plt.ylabel('Frequency')
plt.title('Distribution of Word Counts')
plt.show()
```



1.2 Removing Short Reviews

We will remove short reviews, since they probably don't contain much useful information.

```
[6]: # Make column for word count (of the review content)
df['word_count'] = df['content'].str.split().apply(lambda s: 0 if s == None
else len(s))

# Remove short reviews
REVIEW_WORD_COUNT_THRESHOLD = 10
df = df[df['word_count'] >= REVIEW_WORD_COUNT_THRESHOLD]
```

```
[7]: # Print number of rows in the data after filtering out short reviews
rows1 = df.shape[0]
print(f'After filtering out short reviews, we have {rows1:,} reviews (removed

Grows0 - rows1:,} reviews)')
```

After filtering out short reviews, we have 1,063 reviews (removed 760 reviews)

1.3 Splitting Reviews By Rating

It is likely that reviews that rate the app highly will have positive things to say about the app, whereas reviews with lower ratings will contain complaints about the app. Positive reviews can be used to detect requirements that are being satisfied and negative reviews can point to ignored requirements (or new, previously unspecified requirements).

Shown below, most of the reviews are 1 star, with 5 star reviews following close behind. Reviews with 2, 3, and 4 stars are significantly smaller in number, but when combined together have almost the same amount as the 1 and 5 star reviews.

```
[8]: # Print the counts of each rating
print(df['score'].value_counts().sort_index())
```

```
1 46
2 17
3 36
4 80
5 884
```

score

Name: count, dtype: int64

```
[9]: # Split the reviews into negative (1 star), mixed (2, 3, and 4 stars), and # positive (5 stars)

df_negative = df[df['score'] == 1].copy()

df_mixed = df[(df['score'] == 2) | (df['score'] == 3) | (df['score'] == 4)].

→copy()

df_positive = df[df['score'] == 5].copy()
```

Shown below, the 1 star reviews are mostly negative,

```
[10]: for i in range(5):
    print(f"{df_negative.iloc[i]['content']}\n")
```

Speakerphone no longer working I would give zero stars if possible

95% of the time, it seems, this is worthless. My aunt is in a nursing home -- that's why I bought a grandpad, but I usually cannot hear her. Yes, I have tried numerous times w the help center.

This app is not functioning as it should. I was on a regular phone call and the Grandpad rang through while I was on the phone call, interrupting my call. Additionally, I had my phone on no not disturb and the Grandpad app still somehow rang through. There are bugs that need to be worked out because this app is terrible.

I am so upset right now because I put in my email address and they say it's wrong what does this app want lies a am starting to believe that this app is a scam I am Uninstalling they won't let me do my email account because they say they don't have it developers please fix this issue I am feeling sad right now

so disappointed this app looked like fun only problem is that they don't like my email account

Unable to Personalize GrandPad as Administrator. Calling Support hasn't helped.

the 2, 3, and 4 star reviews are mixed,

```
[11]: for i in range(5):
    print(f"{df_mixed.iloc[i]['content']}\n")
```

Great medium to remain in contact with our loved ones.

Once you have figured out how to set it up it's an awesome device. It almost got thrown across the room while I was trying to get it up and running for my 86 yo mother who has dementia. Once it's set up though it's easy. I highly recommend to get this before the senior citizen is as advanced in dementia as my mom is. She is scared of it and even though I tell her to practice and reassure her she can't mess it up she still won't use it alone.

So far, this app is great. My grandma is having an easier time with this than other social apps

I have a few questions on posting things & submitted it. Waiting on a response.

The app is fantastic. It's a great way to keep in touch with loved ones. I love that I can post videos and pictures or do video calls with anyone in my family who has access to the app. Also I love some of the recent changes. Keep up the good work, Grandpad!

and the 5 star reviews are mostly positive.

```
[12]: for i in range(5):
    print(f"{df_positive.iloc[i]['content']}\n")
```

Grandpad is a great way to stay connected to those near and far.

My 94 year old aunt always tells me how much she loves her GrandPad! She's become very proficient at making phone calls, & sending voice & text messages. And she loves the photo display!

Easy to use! Large variety of music, easy to load photos!

This makes sharing photos and memories with my senior citizen father much easier. We are able to connect very easily and he can call me without memorizing a phone number. That adds to helping him feel more independent as he ages.

The screen and user interface are simple, and easy to use for mom. Just the right features to stay in touch and share photos/videos.

1.4 Saving Preprocessed Data to Files

As the last step, we will save the dataframes to CSV files so that they can be used later.

```
[13]: # Path of this notebook
NOTEBOOK_PATH = Path(os.path.abspath(''))

# Path of data files
DATA_PATH = NOTEBOOK_PATH.parent / 'data'

# Ensure data directory exists
DATA_PATH.mkdir(exist_ok=True)

# Save dataframes
df_positive.to_csv(DATA_PATH / 'positive_reviews.csv', index=False)
df_mixed.to_csv(DATA_PATH / 'mixed_reviews.csv', index=False)
df_negative.to_csv(DATA_PATH / 'negative_reviews.csv', index=False)
```

2 Performing LDA Analysis

Now that the reviews have been preprocessed, we will analyze each category of reviews by applying an LDA (Latent Dirichlet Allocation) model.

First, we define a function for preparing text for LDA analysis.

```
[14]: nltk.download('wordnet')
      numbers_re = re.compile(r'\d+')
      punctuation_re = re.compile(r'[^\w\s]')
      extra_spaces_re = re.compile(r'\s+')
      lemmatizer = WordNetLemmatizer()
      def preprocess_text(text):
          11 11 11
          Cleans `text` by:
            - setting it to lower case,
            - removing numbers,
            - removing punctuation,
            - removing extra spaces,
            - lemmatizing words, and
            - removing stopwords.
          11 11 11
          text = text.lower()
          text = numbers_re.sub(r'', text)
          text = punctuation_re.sub('', text)
```

```
text = extra_spaces_re.sub(' ', text)
text = ' '.join([lemmatizer.lemmatize(word) for word in text.split() if
word not in stopwords.words('english')])
return text
```

[nltk_data] Downloading package wordnet to /home/daniel/nltk_data...
[nltk_data] Package wordnet is already up-to-date!

Next, we define a function for creating the LDA model given a dataframe containing reviews.

```
[15]: def make_lda_model(df: pd.DataFrame):
          11 11 11
          Makes a LatentDirichletAllocation (LDA) model of the data in `df` DataFrame.
          `df` must have a `'cleaned_content'` column that contains the cleaned
          content of the reviews.
          Returns:
          `lda_model` - the generated LDA model.
          'vectorizer' - the vectorizer that was used on the reviews to convert the
          text data into numbers.
          'X' - the vectorized data of the cleaned reviews.
          # Vectorize the text (turn words into numerical data)
          vectorizer = CountVectorizer(max_df=0.95, min_df=2, stop_words='english')
          X = vectorizer.fit_transform(df['cleaned_content'])
          # Apply LDA for topic modeling
          NUMBER_OF_TOPICS = 5
          lda_model = LatentDirichletAllocation(n_components=NUMBER_OF_TOPICS,_
       ⇒random state=42)
          lda_model.fit(X)
          # Assign topics to reviews
          df['topic'] = lda_model.transform(X).argmax(axis=1)
          # Map topic numbers so that they start at 1 instead of 0 and are sorted in
          # decreasing order
          topic_sizes = df['topic'].value_counts().sort_values(ascending=False)
          size_based_mapping = {old_topic: new_topic + 1 for new_topic, old_topic in_u
       ⇔enumerate(topic_sizes.index)}
          df['topic'] = df['topic'].map(size_based_mapping)
          def print_top_words(model, feature_names, n_top_words=10):
              Prints the top words in each topic.
```

```
topics_map = {}
for i, topic in enumerate(model.components_):
    topics_map[size_based_mapping[i]] = " ".join([feature_names[j] for_
    j in topic.argsort()[:-n_top_words - 1:-1]])
for i in sorted(topics_map.keys()):
    print(f'Topic #{i}: {topics_map[i]}')

# Print top words in each topic
print_top_words(lda_model, vectorizer.get_feature_names_out())
return lda_model, vectorizer, X
```

2.1 LDA Model for Positive Reviews

Creating and analyzing an LDA model for positive reviews.

Topic #2: love family great video picture mom app share able photo
Topic #3: grandpad service family easy use mom customer wonderful phone best
Topic #4: grandpad use able love mother phone easy home mom computer
Topic #5: family life way great able far wonderful away senior time

[18]: <IPython.core.display.HTML object>

2.2 LDA Model for Mixed Reviews

Creating and analyzing an LDA model for mixed reviews.

```
[19]: # Clean the content in the dataframe for mixed reviews

df_mixed['cleaned_content'] = df_mixed['content'].apply(preprocess_text)
```

```
[20]: # Make LDA model for mixed reviews
lda_model_mixed, vectorizer_mixed, X_mixed = make_lda_model(df_mixed)
```

```
Topic #1: video phone app work use time easy love need great
     Topic #2: app able pad contact picture sister family mom like add
     Topic #3: family great love mom easy grandpad app use user picture
     Topic #4: video phone app great family photo elderly internet issue device
     Topic #5: family photo comment touch great app picture delete like posted
[21]: # Visualizing the LDA model for mixed reviews
      panel = pyLDAvis.lda_model.prepare(lda_model_mixed, X_mixed, vectorizer_mixed)
      pyLDAvis.display(panel)
[21]: <IPython.core.display.HTML object>
     2.3 LDA Model for Negative Reviews
     Creating and analyzing an LDA model for negative reviews.
[22]: # Clean the content in the dataframe for negative reviews
      df negative['cleaned content'] = df negative['content'].apply(preprocess text)
[23]: # Make LDA model for negative reviews
      lda_model_negative, vectorizer_negative, X_negative = 
       →make_lda_model(df_negative)
     Topic #1: grandpad account grand pad time installed help try log create
     Topic #2: app phone grandpad answer work suck need way ive fine
     Topic #3: picture say make app link account dont ive notification actually
     Topic #4: app like email account phone dont say problem wont let
     Topic #5: app open registered tried wont let email sign isnt say
[24]: # Visualizing the LDA model for negative reviews
      panel = pyLDAvis.lda_model.prepare(lda_model_negative, X_negative,_
       ⇔vectorizer_negative)
```

[24]: <IPython.core.display.HTML object>

pyLDAvis.display(panel)