Data Pipeline Anteraja Review

TextBlob-based Sentiment Analysis

Our Team



Hakim 440307



Adhitya 444033



Hafidz 444052



Samatha 444071

Background



Platform *e-commerce* menggunakan strategi promo di tanggal cantik.

(Aria, 2020)



Terjadi peningkatan aktivitas logistik yang signifikan.

(Asosiasi Logistik Indonesia, 2019)



Diperlukan *insight* apakah peningkatan aktivitas logistik mempengaruhi sentimen publik terhadap perusahaan *low-cost* logistic.

Kenapa Anteraja?



Anteraja melayani pengiriman gratis ongkir (Subsidi pengiriman Rp20.000 tetapi dengan durasi pengiriman yang lebih lama).

ETL Process



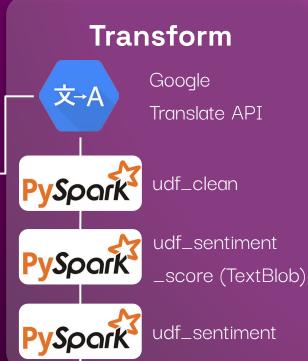




Table of contents

01 Extract

Get user reviews of Anteraja app from Twitter & Google Play Store

03 Load

Loading the transformed data into data warehouse

02 Transform

Data cleaning and translation to get well-defined data

04 Visualization

Create sentiment analysis charts



O'1 Extract

Get user reviews of Anteraja app from Twitter & Google Play Store

Twitter + Play Store Review

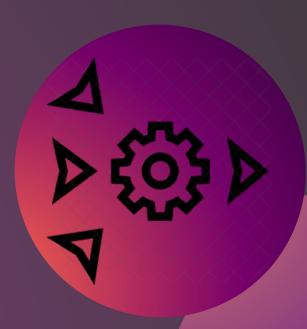
```
id_reviews = reviews_all(
  'id.anteraja.aca',
    sleep_milliseconds=0,
    lang='en',
    country='id',
    sort=Sort.NEWEST,
)
```

Tweepy

google-play-scraper

02 Transform

Data cleaning and translation to get well-defined data. Also getting sentiment score from each data.



Transform

Translate

Translate reviews from Twitter and Google Play

Get Sentiment Score

Get sentiment scores from each review



Clean

Remove unnecessary signs so that only user reviews are left

Get Sentiment

Get sentiment category for every review based on the sentiment scores

translator()

Translating from lang:id to lang:en with Google Translator API

udf_clean

```
def cleaner(text):
    text = re.sub("@[A-Za-z0-9]+", "", text) #Remove @ sign
    text = ''.join(c for c in text if c not in emoji.UNICODE_EMOJI) #Remove Emojis
    text = text.replace("#", "").replace("_", "") #Remove hashtag sign but keep the text
    text = " ".join(w for w in nltk.wordpunct_tokenize(text) \
        if w.lower() in words or not w.isalpha()) #Remove non-english tweets (not 100% success)
    return text

udf_clean = udf(lambda x:cleaner(x), StringType())
```

Clean data from @username, emoji, #Hashtag

udf_sentiment_score

```
def sentiment_score(text):
   blob = TextBlob(str(text))
   score = blob.polarity
   return score

udf_sentiment_score = udf(lambda x:sentiment_score(x), FloatType())
```

Get Sentiment Score from TextBlob Polarity ranged from -1 to 1

udf_sentiment

```
def sentiment(sentiment_score):
    if sentiment_score > 0:
        return "positive"
    elif sentiment_score < 0:
        return "negative"
    else:
        return "neutral"

udf_sentiment = udf(lambda x:sentiment(x), StringType())</pre>
```

Categorization of Sentiment Score (between "positive", "negative", and "neutral")

Transformed Data

```
tweet|sentiment_score|sentiment|
     username
                             date
       GrabID 2021-10-11 23:09:02 If it on the poli...
                                                                         neutral
       GrabID 2021-10-11 22:59:46 Hi Sis :) for int...
                                                                  0.25 positive
 marsupilapila 2021-10-11 21:36:13 id want to ask fo...
                                                                         neutral
    aLpyPutraa 2021-10-11 21:19:56 : Hello id I orde...
                                                                         neutral
   Naanganang 2021-10-11 21:13:43 id this min why s...
                                                                   0.0
                                                                         neutral
 nanang indra 2021-10-11 21:09:07 id hello sis rece...
                                                                   0.0
                                                                         neutral
cabangpurnamaa 2021-10-11 19:16:04 , I have to be fu...
                                                                         neutral
```

Transformed Data

Google Play Store

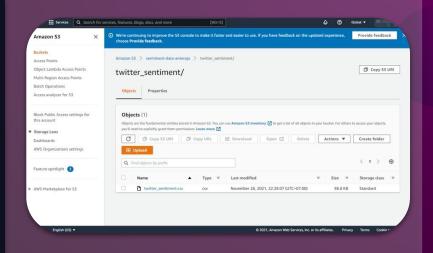


O3 Load

Loading the transformed data into data warehouse (AWS S3)

Load Twitter Data to AWS S3

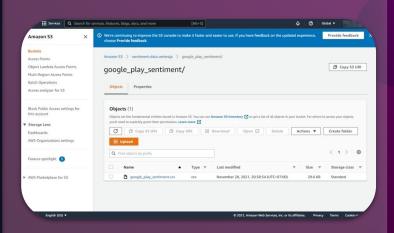
```
# Load df Twitter to S3
filename= "twitter sentiment.csv"
bucketName = 'sentiment-data-anteraja'
folder name = "twitter sentiment/"
csv_buffer = StringIO()
pd_Twitter.to_csv(csv_buffer, index=False)
client = boto3.client("s3")
response = client.put object(
   Body = csv_buffer.getvalue(),
   Bucket = bucketName,
   Key = folder_name+filename
```



twitter_sentiment.csv

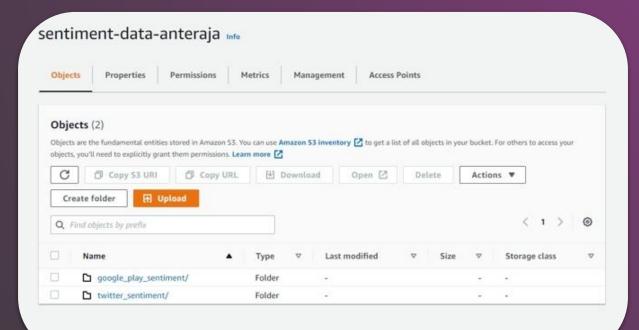
Load Google Play Data to AWS S3

```
# Load df google play to S3
filename= "google play sentiment.csv"
bucketName = 'sentiment-data-anteraja'
folder_name = "google_play_sentiment/"
csv_buffer = StringIO()
pd_google_play.to_csv(csv_buffer, index=False)
client = boto3.client("s3")
response = client.put object(
   Body = csv_buffer.getvalue(),
   Bucket = bucketName,
   Key = folder name+filename
```



google_play_sentiment.csv

AWS S3

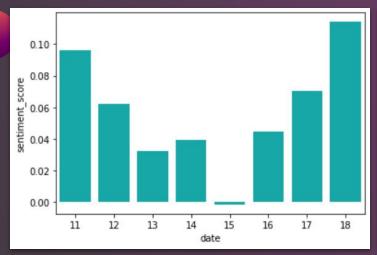


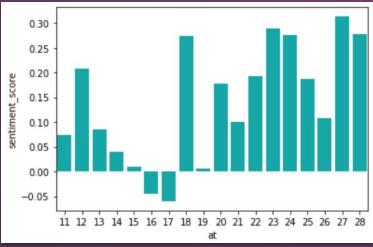


04 Visualization

Gathering insight from the twitter and playstore review data

Visualization (Barplot)





Twitter

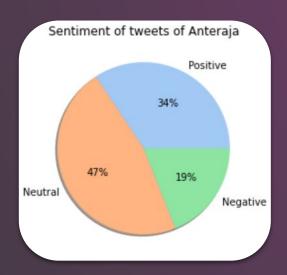
Google Play Store

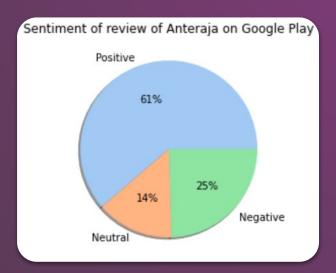
Visualization (Barplot)

Terdapat kesamaan penurunan sentimen pada data Twitter dan data Playstore, yaitu penurunan signifikan pada tanggal 15 November 2021. (Twitter = -0.001666; Playstore = 0.010236)

Hal ini diduga dikarenakan ekspektasi pelanggan mengenai waktu pengiriman selama 3-5 hari. Jika dihitung dari tanggal 11, maka 3-5 hari jatuh pada tanggal 14-16 November 2021. Dapat diduga penurunan tersebut diakibatkan oleh belum sampainya paket dikarenakan tingginya arus logistik setelah *event* 11.11.

Visualization (Pie Chart)





Twitter

Google Play Store

Visualization (Pie Chart)

Terlihat dari *pie chart* sebelumnya, sentimen negatif pada kedua sumber bukan merupakan sentimen yang dominan. Pada Twitter, sentimen yang dominan adalah *neutral* (sebanyak 47%), sedangkan pada Google Play Store adalah *positive* (sebanyak 61%). Hal ini dikarenakan sentimen negatif hanya terjadi pada beberapa hari setelah 11.11, yaitu pada saat pengiriman barang ke pelanggan.

Summary

The promotional strategy from e-commerce on November 11, 2021 led to an increase in logistics activities, one of which was Anteraja. Not a few customers use Anteraja's services for freight delivery. However, the delivery time is relatively longer than usual (up to 3 to 5 days) causing negative public sentiment as shown in user reviews.

Referensi

Aria, 2020. 11.11 dan Promo Rutin Tanggal Cantik, Strategi Shopee Dominasi Pasar.

https://katadata.co.id/pingitaria/digital/5faa3b9049e9a/1111-dan-promo-rutin-tanggal-cantik-strategi-shopee-dominasi-pasar

Asosiasi Logistik Indonesia, 2019. Transaksi Harbolnas 2019 Tercatat Rp9,1 Triliun. https://www.ali.web.id/web2/news_detail.php?id=1736

Thanks!

Do you have any questions?

CREDITS: This presentation template was created by **Slidesgo**, including icons by **Flaticon** and infographics & images by **Freepik**