## SMA LAB 8 from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer import pandas as pd import matplotlib.pyplot as plt sia = SentimentIntensityAnalyzer() df =pd.read csv("https://raw.githubusercontent.com/ArtaXerxess/Social-Media-Analytics-Lab/main/SMA%20Lab%208/20191226-reviews.csv") df.head() rating date verified \ asin name B0000SX2UC October 11, 2005 Janet 3 False 1 B0000SX2UC Luke Wyatt 1 January 7, 2004 False 5 December 30, 2003 B0000SX2UC Brooke False March 18, 2004 3 B0000SX2UC amy m. teague False 4 August 28, 2005 False B0000SX2UC tristazbimmer title \ 0 Def not best, but not worst 1 Text Messaging Doesn't Work 2 Love This Phone 3 Love the Phone, BUT...! Great phone service and options, lousy case! helpfulVotes body I had the Samsung A600 for awhile which is abs... 1.0 Due to a software issue between Nokia and Spri... 17.0 This is a great, reliable phone. I also purcha... 5.0 I love the phone and all, because I really did... 1.0 The phone has been great for every purpose it ... 1.0 cleaning, removing unnecessary columns df.drop(['date', 'asin', 'name', 'rating', 'verified', 'title', 'helpfulVote s'],axis = 1, inplace = True) df.head() bodv I had the Samsung A600 for awhile which is abs...

Due to a software issue between Nokia and Spri...
This is a great, reliable phone. I also purcha...
I love the phone and all, because I really did...
The phone has been great for every purpose it ...

Assigining scores to every review

```
df['scores']=df['body'].apply(lambda body:
sia.polarity scores(str(body)))
df.head()
                                                  body \
  I had the Samsung A600 for awhile which is abs...
  Due to a software issue between Nokia and Spri...
  This is a great, reliable phone. I also purcha...
  I love the phone and all, because I really did...
  The phone has been great for every purpose it ...
                                                scores
   {'neg': 0.076, 'neu': 0.825, 'pos': 0.1, 'comp...
  {'neg': 0.03, 'neu': 0.87, 'pos': 0.1, 'compou...
{'neg': 0.048, 'neu': 0.831, 'pos': 0.121, 'co...
   {'neg': 0.0, 'neu': 0.857, 'pos': 0.143, 'comp...
   {'neg': 0.067, 'neu': 0.816, 'pos': 0.117, 'co...
df['compound']=df['scores'].apply(lambda
score dict:score dict['compound'])
df['pos']=df['scores'].apply(lambda pos dict:pos dict['pos'])
df['neg']=df['scores'].apply(lambda neg dict:neg dict['neg'])
df.head()
                                                  body \
  I had the Samsung A600 for awhile which is abs...
  Due to a software issue between Nokia and Spri...
  This is a great, reliable phone. I also purcha...
  I love the phone and all, because I really did...
  The phone has been great for every purpose it ...
                                                        compound
                                                scores
                                                                    pos
  {'neg': 0.076, 'neu': 0.825, 'pos': 0.1, 'comp...
                                                          0.8629
                                                                 0.100
0.076
1 {'neg': 0.03, 'neu': 0.87, 'pos': 0.1, 'compou...
                                                          0.8720
                                                                  0.100
0.030
  {'neg': 0.048, 'neu': 0.831, 'pos': 0.121, 'co...
                                                          0.8966
                                                                  0.121
0.048
  {'neg': 0.0, 'neu': 0.857, 'pos': 0.143, 'comp...
3
                                                          0.9592
                                                                  0.143
0.000
  {'neg': 0.067, 'neu': 0.816, 'pos': 0.117, 'co...
                                                          0.7432
                                                                  0.117
0.067
```

The compound score is computed by summing the valence scores of each word in the lexicon, adjusted according to the rules, and then normalized to be between -1 (most extreme negative) and +1 (most extreme positive). This is the most useful metric if you want a single unidimensional measure of sentiment for a given sentence. Calling it a 'normalized, weighted composite score' is accurate.

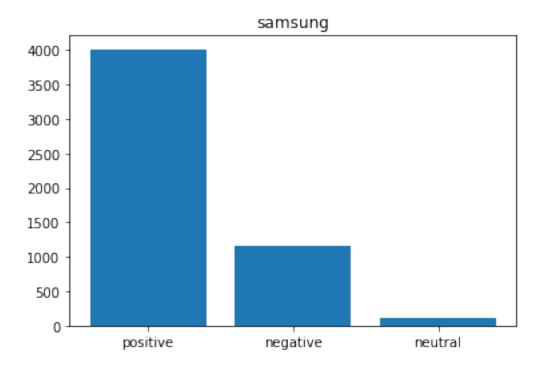
It is also useful for researchers who would like to set standardized thresholds for classifying sentences as either positive, neutral, or negative.

```
# Adding new column that classifies the review as positive , negative
or neutral
df['type']=''
df.loc[df.compound>0,'type']='POS'
df.loc[df.compound==0,'type']='NEUTRAL'
df.loc[df.compound<0,'type']='NEG'</pre>
df.head(10)
                                                body \
  I had the Samsung A600 for awhile which is abs...
1 Due to a software issue between Nokia and Spri...
  This is a great, reliable phone. I also purcha...
3 I love the phone and all, because I really did...
4 The phone has been great for every purpose it ...
                                              scores compound
                                                                  pos
0 {'neg': 0.076, 'neu': 0.825, 'pos': 0.1, 'comp...
                                                        0.8629
                                                                0.100
0.076
1 {'neg': 0.03, 'neu': 0.87, 'pos': 0.1, 'compou...
                                                                0.100
                                                        0.8720
0.030
  {'neg': 0.048, 'neu': 0.831, 'pos': 0.121, 'co...
                                                       0.8966
                                                                0.121
0.048
3 {'neg': 0.0, 'neu': 0.857, 'pos': 0.143, 'comp...
                                                       0.9592
                                                                0.143
0.000
4 {'neg': 0.067, 'neu': 0.816, 'pos': 0.117, 'co...
                                                       0.7432 0.117
0.067
  type
0 P0S
1 P0S
2 P0S
3 P0S
4 P0S
from pandas.io.formats.style_render import List
mobileBrands = ['samsung', 'nokia', 'apple', 'google
pixel','redmi','xiaomi']
# sentence = 'hello i like nokia '
def brandname(a : List,b : str):
  for brand in a:
   if brand in b:
      return brand
# brandname(mobileBrands, sentence)
```

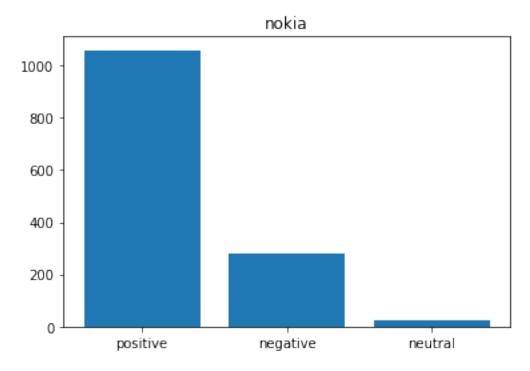
```
rows , cols = df.shape
for row in range(rows):
  df['brands'][row] = brandname(mobileBrands,str(df['body']
[row]).lower())
<ipython-input-53-5ald6a253844>:15: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#
returning-a-view-versus-a-copy
  df['brands'][row] = brandname(mobileBrands,str(df['bodv'])
[row]).lower())
df.head()
  I had the Samsung A600 for awhile which is abs...
1 Due to a software issue between Nokia and Spri...
  This is a great, reliable phone. I also purcha...
  I love the phone and all, because I really did...
4 The phone has been great for every purpose it ...
                                              scores
                                                      compound
                                                                  pos
neg \
0 {'neg': 0.076, 'neu': 0.825, 'pos': 0.1, 'comp...
                                                                0.100
                                                        0.8629
0.076
1 {'neg': 0.03, 'neu': 0.87, 'pos': 0.1, 'compou...
                                                        0.8720
                                                                0.100
0.030
2 {'neg': 0.048, 'neu': 0.831, 'pos': 0.121, 'co...
                                                        0.8966
                                                                0.121
0.048
3 {'neg': 0.0, 'neu': 0.857, 'pos': 0.143, 'comp...
                                                        0.9592
                                                                0.143
0.000
4 {'neg': 0.067, 'neu': 0.816, 'pos': 0.117, 'co...
                                                        0.7432 0.117
0.067
        brands
  type
0 P0S
       samsung
1 P0S
          nokia
2 P0S
       samsung
3 P0S
           None
4 P0S
           None
rows , col = df.shape
def countScores(brandname):
  pos = 0
  neq = 0
  neut = 0
  for i in range(0, rows):
```

```
if df.loc[i]['brands'] == brandname:
      if df.loc[i]['type'] == 'POS':
        pos = pos + 1
      if df.loc[i]['type'] == 'NEG':
        neg = neg + 1
      if df.loc[i]['type'] == 'NEUTRAL':
        neut = neut + 1
  return [pos,neg,neut]
(4011, 1167, 113)
mobileBrands = ['samsung', 'nokia', 'apple', 'google
pixel','redmi','xiaomi']
x = ['positive','negative','neutral']
def plotter(name):
  y = countScores(name)
  plt.title(name)
  plt.bar(x,y)
  plt.show()
  print(name, " positive : ",y[0], " negative : ",y[1], " neutral
",y[2],"\n\n")
for brand in mobileBrands:
```

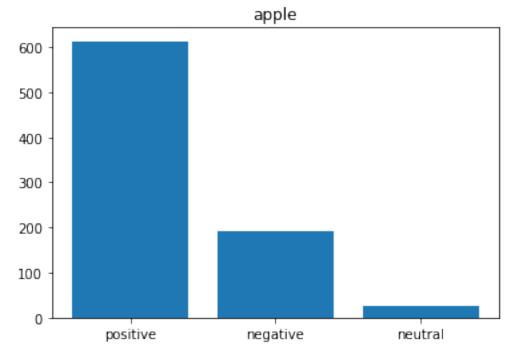
## plotter(brand)



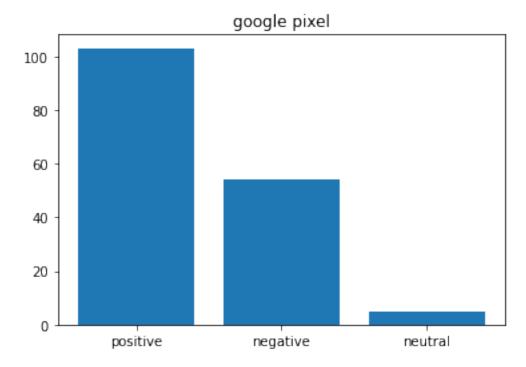
samsung positive: 4011 negative: 1167 neutral 113



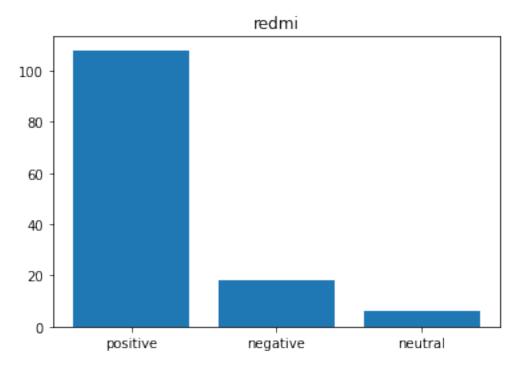
nokia positive: 1058 negative: 282 neutral 26



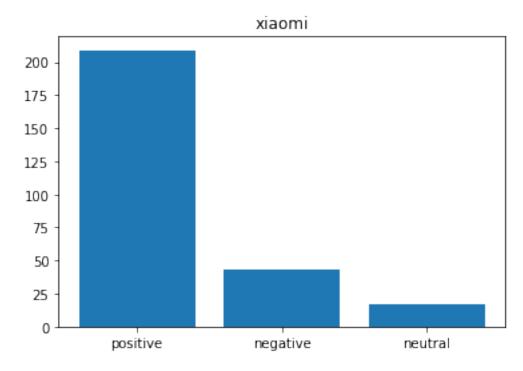
apple positive: 613 negative: 192 neutral 25



google pixel positive : 103 negative : 54 neutral 5



redmi positive: 108 negative: 18 neutral 6



xiaomi positive : 209 negative : 43 neutral 17