DLNLP Project - Amazon Food Reviews Dataset

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```
In [1]:
        import pandas as pd
        import numpy as np
        import re
        import nltk
        from nltk.tokenize import word tokenize, sent tokenize
        from nltk.corpus import stopwords
        from nltk.stem.porter import PorterStemmer
        from nltk.stem import WordNetLemmatizer
        import matplotlib.pyplot as plt
        import seaborn as sns
        from string import punctuation
        from sklearn import svm
        from nltk import ngrams
        from itertools import chain
        from wordcloud import WordCloud
```

In [2]: #read the data
df = pd.read_csv("C:/Users/chira/Desktop/food_subset.csv")

In [3]: df.head()

Out[3

3]:		ld	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score	Time	Summary	
	0	1	B001EO5QW8	AOVROBZ8BNTP7	S. Potter	19	19	4	1163376000	Best of the Instant Oatmeals	McCa In Oat is gr you
	1	2	B001EO5QW8	A3PMM0NFVEJGK9	Megan "Bad at Nicknames"	13	13	4	1166313600	Good Instant	in oat fror bes
	2	3	B003ZFRKGO	A2VOZX7YBT0D6D	Johnnycakes "Johnnycakes"	15	15	5	1325635200	Forget Molecular Gastronomy - this stuff rocke	pro title Mole Gas
	3	4	B000ITVLE2	A3NID9D9WMIV01	Louie Arrighi "Lou da Joo"	17	19	5	1260057600	tastes very fresh	expir date mo fror d
	4	5	B001UJEN6C	A1XM65S80UQ2MD	Joseph Kagan	13	13	5	1276214400	Great Natural Energy	This fant pro
											•

In [4]: df['Text'].values[0]

"McCann's Instant Oatmeal is great if you must have your oatmeal but can only scrape together two or three minu tes to prepare it. There is no escaping the fact, however, that even the best instant oatmeal is nowhere near a s good as even a store brand of oatmeal requiring stovetop preparation. Still, the McCann's is as good as it g ets for instant oatmeal. It's even better than the organic, all-natural brands I have tried. All the varieties in the McCann's variety pack taste good. It can be prepared in the microwave or by adding boiling water so it is convenient in the extreme when time is an issue.

| The product is a convenient of the convenience product. Real sugar tastes better and is not as harmful as the other stuff. One thing I do not like, though, is McCann's use of thickeners. Oats plus water plus heat should make a creamy, tasty oatmeal without the need for guar gum. But this is a convenience product. Maybe the guar gum is why, after sitting in the bowl a while, the instant McCann's becomes too thick and gluey."

In [5]: df.shape

```
Out[5]: (17244, 10)
In [6]: df= df.head(17244)
        print(df.shape)
        (17244, 10)
In [7]:
        df['Score'].value_counts()
             10562
Out[7]:
              2929
        4
              1664
        3
              1178
        2
               911
        Name: Score, dtype: int64
In [8]: ax= df['Score'].value_counts().sort_index().plot(kind='bar',title ='Count of reviews by Stars',figsize =(10,5))
        ax.set xlabel('Review Stars')
Out[8]: Text(0.5, 0, 'Review Stars')
                                                    Count of reviews by Stars
         10000
          8000
          6000
          4000
          2000
```

In []:

Most of the review are 5 stars but less 1 stars so we can say that most customers are tend to positive review

```
In [9]: text = df['Text'][45]
  print(text)
```

0

I received this mix along with a waffle maker as a gift. It's so good that I keep buying this same brand mix w hen I run out.
br />sbr />I cut the Farmhouse Waffles recipe in half (so that's 1 cup Farmhouse Pancake and Waffle Mix; 1 egg; 5/8 cup water; 2 tablespoon melted butter) and it's more than enough batter for 2 waffles in a full-size Belgian waffle maker.
br />Easy to make and very tasty.

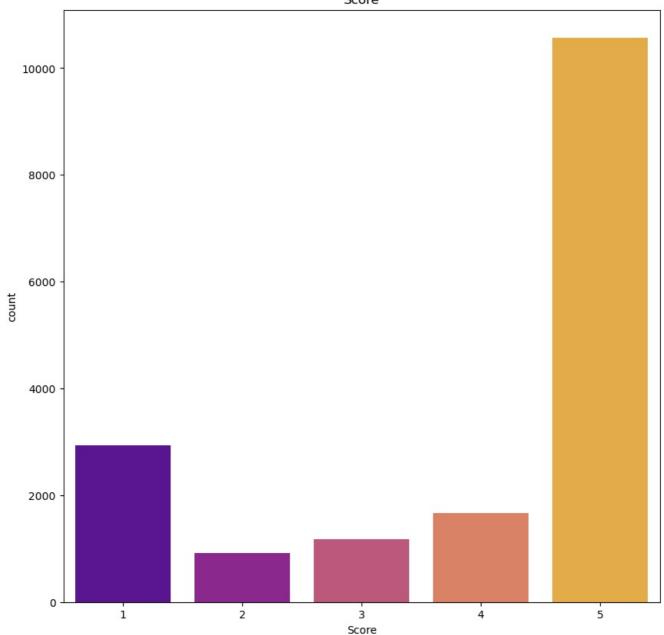
Review Stars

4

2

```
In [10]: Amazon Review =df
In [11]: Amazon Review.dtypes
                                      int64
Out[11]:
         ProductId
                                     object
         UserId
                                     object
         ProfileName
                                     object
         {\tt HelpfulnessNumerator}
                                      int64
         HelpfulnessDenominator
                                      int64
                                      int64
         Score
         Time
                                      int64
         Summary
                                     object
         Text
                                     object
         dtype: object
In [12]: Amazon Review.info()
```

```
<class 'pandas.core.frame.DataFrame'>
          RangeIndex: 17244 entries, 0 to 17243
          Data columns (total 10 columns):
                                         Non-Null Count Dtype
              Column
                                         -----
                                         17244 non-null int64
           0
               Ιd
               ProductId
                                         17244 non-null object
                                         17244 non-null object
17243 non-null object
           2
               UserId
           3
               ProfileName
               HelpfulnessNumerator 17244 non-null int64
HelpfulnessDenominator 17244 non-null int64
Score 17244 non-null int64
           4
           5
           6
           7
               Time
                                         17244 non-null int64
           8
               Summary
                                         17244 non-null object
                                         17244 non-null object
           9
               Text
          dtypes: int64(5), object(5)
          memory usage: 1.3+ MB
In [13]: ##Removing the Duplicates if any
          Amazon_Review.duplicated().sum()
          Amazon Review.drop duplicates(inplace=True)
In [14]: #Check for the null values in each column
          Amazon Review.isnull().sum()
Out[14]: Id
          ProductId
          UserId
                                      0
          ProfileName
                                      1
          HelpfulnessNumerator
                                      0
          HelpfulnessDenominator
                                      0
          Score
                                      0
          Time
                                      0
          Summary
                                      0
          Text
                                      0
          dtype: int64
In [15]: ##Remove the NaN values from the dataset
          Amazon_Review.isnull().sum()
          Amazon_Review.dropna(how='any',inplace=True)
In [16]:
          import seaborn as sns
          sns.countplot(Amazon_Review['Score'], palette="plasma")
          fig = plt.gcf()
          fig.set size inches(10,10)
          plt.title('Score')
          C:\Users\chira\anaconda3\lib\site-packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variabl
          e as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other
          arguments without an explicit keyword will result in an error or misinterpretation.
warnings.warn(
Out[16]: Text(0.5, 1.0, 'Score')
```



```
In [17]: Amazon_Review = pd.DataFrame(Amazon_Review , columns=['UserId','Score', 'Text'])
    print(Amazon_Review.shape)
    Amazon_Review.head()
    (17243, 3)
```

 Userid
 Score
 Text

 0
 AOVROBZ8BNTP7
 4
 McCann's Instant Oatmeal is great if you must ...

 1
 A3PMM0NFVEJGK9
 4
 This is a good instant oatmeal from the best o...

 2
 A2VOZX7YBT0D6D
 5
 I know the product title says Molecular Gastro...

 3
 A3NID9D9WMIV01
 5
 The expiration date is 21 months from the day ...

 4
 A1XM65S80UQ2MD
 5
 This is a fantastic product, and I wish it was...

```
import re
import nltk
nltk.download('punkt')

from nltk.tokenize import word_tokenize

nltk.download('stopwords')
from nltk.corpus import stopwords
stop_words = set(stopwords.words("english"))

#stop_words.extend(['crypto','even','early'])
import nltk
nltk.download('wordnet')
nltk.download('wordnet')
from nltk.stem import WordNetLemmatizer
```

```
lemmatizer = WordNetLemmatizer()
          [nltk_data] Downloading package punkt to
          [nltk data]
                           C:\Users\chira\AppData\Roaming\nltk data...
                          Package punkt is already up-to-date!
          [nltk_data]
          [nltk_data] Downloading package stopwords to
          [nltk data]
                            C:\Users\chira\AppData\Roaming\nltk data...
          [nltk_data]
                         Package stopwords is already up-to-date!
          [nltk_data] Downloading package wordnet to
          [nltk_data]
                            C:\Users\chira\AppData\Roaming\nltk_data...
          [nltk data]
                          Package wordnet is already up-to-date!
          [nltk_data] Downloading package omw-1.4 to
          [nltk_data]
                            C:\Users\chira\AppData\Roaming\nltk_data...
          [nltk data]
                        Package omw-1.4 is already up-to-date!
In [24]: !pip install demoji
          Requirement already satisfied: demoji in c:\users\chira\anaconda3\lib\site-packages (1.1.0)
In [25]:
          import demoji
          def handle emoji(string):
               #x = string.to_string(header=False, index=False)
               emojis = demoji.findall(string)
               #print(emojis)
               for emoji in emojis:
                   string = string.replace(emoji, " " + emojis[emoji].split(":")[0])
               return string
In [21]: def text_cleaner(review):
             # removing the not required texts
             cleaned_review = re.sub(re.compile('<.*?>'), '', review) #removing HTML tags
cleaned_review = re.sub('[^A-Za-z]+', ' ', cleaned_review) #taking only words
             cleaned_review = handle_emoji(cleaned_review)
             cleaned_review = re.sub(r"http\S+", "", cleaned_review)
             cleaned_review = cleaned_review.lower()
             tokens = nltk.word_tokenize(cleaned_review)
             filtered review = [word for word in tokens if word not in stop words] # removing stop words
             lemm_review = [lemmatizer.lemmatize(word) for word in filtered_review]
review = " ".join(lemm_review)
             return(review)
In [26]: cleanText=[]
          for t in Amazon_Review['Text']:
           cleanText.append(text_cleaner(t))
          Amazon Review["cleanText"] = cleanText
          Amazon Review.head()
Out[26]:
                       Userld Score
                                                                      Text
                                                                                                          cleanText
          0 AOVROBZ8BNTP7
                                   4 McCann's Instant Oatmeal is great if you must ... mccann instant oatmeal great must oatmeal scra...
          1 A3PMM0NFVEJGK9
                                  4 This is a good instant oatmeal from the best o... good instant oatmeal best oatmeal brand us can...
          2 A2VOZX7YBT0D6D
                                  5 I know the product title says Molecular Gastro...
                                                                             know product title say molecular gastronomy le...
             A3NID9D9WMIV01
                                  5 The expiration date is 21 months from the day ... expiration date month day bought product tuna ...
          4 A1XM65S80UQ2MD
                                       This is a fantastic product, and I wish it was...
                                                                               fantastic product wish readily available store...
In [27]: #WORD CLOUD
In [28]: # importing all necessary modules
          from wordcloud import WordCloud, STOPWORDS
          import matplotlib.pyplot as plt
In [29]: wordcloud = WordCloud(
                   background color = 'white',
                   max words = 200,
                   max_font_size = 40,
                   scale = 3,
                   random state = 42,
                   stopwords= stop_words
               ).generate(str(cleanText))
          # plot the WordCloud image
          plt.figure(figsize = (8, 8), facecolor = None)
          plt.imshow(wordcloud)
          plt.axis("off")
          plt.tight_layout(pad = 0)
          plt.show()
```



```
In [30]:
         #After Word Cloud
         new_stopwords = ['also','get','one','use','since','got','even','u','say','k','still']
In [31]:
         stpwrd = nltk.corpus.stopwords.words('english')
         stpwrd.extend(new_stopwords)
 In [ ]:
         new clean review
         wordcloud = WordCloud(
In [32]:
                  background_color = 'white',
                 \max \text{ words} = 200,
                  max_font_size = 40,
                  scale = 3,
                  random state = 42,
                  stopwords= stpwrd
              ).generate(str(cleanText))
         # plot the WordCloud image
         plt.figure(figsize = (8, 8), facecolor = None)
         plt.imshow(wordcloud)
         plt.axis("off")
         plt.tight_layout(pad = 0)
         plt.show()
```



In [38]: cleanText

Out[38]:

['mccann instant oatmeal great must oatmeal scrape together two three minute prepare escaping fact however even best instant oatmeal nowhere near good even store brand oatmeal requiring stovetop preparation still mccann good get instant oatmeal even better organic natural brand tried variety mccann variety pack taste good prepared m icrowave adding boiling water convenient extreme time issue mccann use actual cane sugar instead high fructose corn syrup helped decide buy product real sugar taste better harmful stuff one thing like though mccann use this ckener oat plus water plus heat make creamy tasty oatmeal without need guar gum convenience product maybe guar

'bought case coconut milk based lack bpa packaging happy product smell strange quite watery gave wife upset st omach two occasion tried can came sri lanka dented',

'dairy free family use coconut milk substitute pretty familiar coconut milk purchased several can native fores t retail store liked use coffee creamer however stuff delivered amazon completely different can bought store milk thin mind incredibly oily shook can thoroughly opened see oily floating top milk separate fridge coconut oil forming thick layer top odd flavor milk bummed go can order thai kitchen',

'reading every review checked see country origin listed description thailand tried one replacement case well h orribly greasy product sri lanka amazon immediately reimbursed used product amazon can must thailand product dr eamy really wanted decided call company description amazon false advertising woman spoke absolute brat borderli ne rude excuse excuse maybe tired fielding complaint isnt job surprised great conversation learned much small c ompany like like nutiva complained leaking coconut oil devleoped great relationship one representative sent rep lacement even though told amazon already reembursed wanted let know problem biggest excuse among several side s tepping maneuver sample sent match product speaks volume lack quality control company make wonder can amazon po or consumer suffer edward son faulty business product sri lankan used another month seriously small company poor customer service doomed fail matter great product lousy product like unwilling try thing stand behind product

'ordering native forest coconut milk amazon quite frequently delectable coconut milk could find anywhere five star review spot loved sudden started receiving completely different product label seemed like watered coconut milk coconut oil added result liquid oil solid translucent like coconut oil throughout matter much shook heated unopened hot water shaking disgusting unusable oily slop refrigerated open oil would rise top form solid top la yer never seen brand coconut milk can say product sri lanka spite amazon listing stating thailand suspect reaso n poor quality would bet can label stating thailand real native forest coconut milk sri lanka can sort evil dop pelg auml nger update due new review decided give product try look like amazon native forest addressed issue in ferior product sri lanka recieved several pack thailand super creamy delicious thank goodness love product also wanted thank amazon always refunded money complained sri lankan slop',

'use canned clam longer live northeast use linguine clam sauce found crown prince chopped clam like minced cho pped also packed water rather clam juice add bottled clam juice make better overall adequate returned snow chop ped clam packed clam juice',

'looking supplement breastmilk horrible time finding something worked tried least different formula including new alimentum formula son could tolerate physically still tasted good buy formula cheaper walmart',

'reluctant buy product bad review thought maybe good review put family team concocted product day co worker ga ve chocolate brownie made erythritol enjoyed waited weird aftertaste waited laxative effect neither grocery sto re today picked sugar free oreo saw bottle sugar free hershey chocolate syrup instantly thought back review low carber use stevia soy milk hemp milk tea coffee think stevia terrible chocolate chocolate product usually made maltitol sucralose mix enjoy product like would buy explosive diarrhea upset stomach related episode people mal titol maybe consume lot maybe special tried recently chocolate brownie thought heck mixed three teaspoon syrup soy milk absolutely thought great delicious naysayer exaggerating flavor sweet expected okay since commercial p roduct way sweet like sugar free chocolate market thought genius sweet enough go low carb everything start tast e sweet obnoxious awhile notice slight bitter aftertaste attributable natural chocolate bitterness fleeting unp leasant mind lingering taste mouth slightly sweet chocolate chemical taste aftertaste taste natural though orga nic type buy ready side five star raters wanted sure took teaspoon chocolate syrup ate straight sudden one star raters right sour people right taste terrible bottle say lowcarber diabetic looking chocolate treat form drink like chocolate milk example though milk carbs would want drink milk add unsweetened soy milk hemp milk mean get great whatever tempted eat pour throat otherwise think gone bad throw treasure close world sugar free chocolate syrup forever sure hershey trying make money increasingly hard niche market cater least know need trying make h appy thank hershey give product chance come organic carb free non diarrhea inducing syrup pleasant aftertaste m outh feel compromise deal hope try see value product like',

'tried bottle new improved strong chemical sour taste little chocolate flavor bad bottle throw away',

```
In [44]: example = cleanText[15]
print(example)
```

[nltk_data]
[nltk_data]

u made version cadbury chocolate distinct taste uk version chocolate eating since childhood like new taste frie nd india buy chocolate either made uk india otherwise disappointed

```
In [46]:
         tokens = nltk.word tokenize(example)
          tokens[:12]
         ['u',
Out[46]:
           'made'
           'version',
           'cadbury'
           'chocolate',
           'distinct'.
           'taste',
           'uk',
           'version'
           'chocolate',
           'eating',
           'since']
         nltk.download('averaged_perceptron_tagger')
In [60]:
          nltk.pos tag(tokens)
          [nltk data] Downloading package averaged perceptron tagger to
```

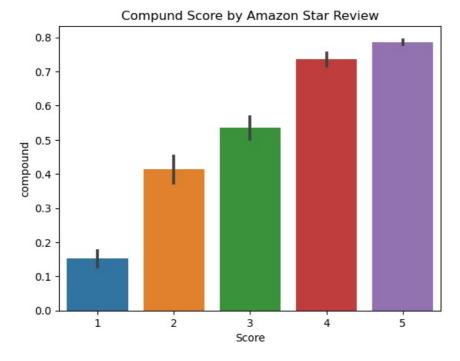
C:\Users\chira\AppData\Roaming\nltk data...

Unzipping taggers\averaged perceptron tagger.zip.

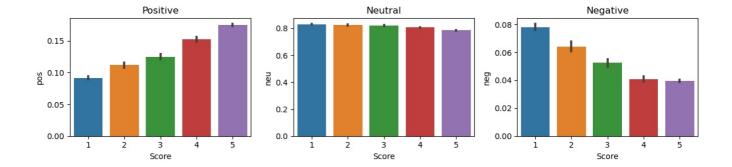
```
Out[60]: [('u', 'JJ'),
             ('made', 'VBN'),
             ('version', 'NN'), ('cadbury', 'NN'),
             ('chocolate', 'NN'), ('distinct', 'JJ'),
             ('taste', 'NN'),
             ('uk', 'JJ'),
             ('version', 'NN'),
             ('chocolate', 'NN'), ('eating', 'VBG'),
             ('since', 'IN'),
             ('childhood', 'NN'),
             ('like', 'IN'),
('new', 'JJ'),
            ('new', JJ',,
('taste', 'NN'),
('friend', 'NN'),
('india', 'NN'),
('buy', 'VB'),
('chocolate', 'NN'),
             ('chocolate', 'NN
('either', 'CC'),
('made', 'VBD'),
('uk', 'JJ'),
             ('india', 'JJ'),
             ('otherwise', 'RB'),
             ('disappointed', 'JJ')]
In [57]:
           from nltk.sentiment import SentimentIntensityAnalyzer
            from tqdm.notebook import tqdm
            import nltk
           nltk.download('vader_lexicon')
            [nltk_data] Downloading package vader_lexicon to
            [nltk_data] C:\Users\chira\AppData\Roaming\nltk_data...
           True
In [58]: sia = SentimentIntensityAnalyzer()
In [61]: sia.polarity_scores('I am so happy!')
           {'neg': 0.0, 'neu': 0.318, 'pos': 0.682, 'compound': 0.6468}
Out[61]:
In [62]: sia.polarity_scores('This is the worst thing ever.')
           {'neg': 0.451, 'neu': 0.549, 'pos': 0.0, 'compound': -0.6249}
Out[62]:
In [64]:
           analyzer = SentimentIntensityAnalyzer()
            Amazon\_Review['compound'] = [analyzer.polarity\_scores(x)['compound'] \  \, \textbf{for} \  \, x \  \, \textbf{in}
            Amazon Review['cleanText']]
            Amazon_Review['neg'] = [analyzer.polarity_scores(x)['neg'] for x in
           Amazon Review['cleanText']]
            Amazon_Review['neu'] = [analyzer.polarity_scores(x)['neu'] for x in
            Amazon_Review['cleanText']]
           \label{lem:amazon_Review['pos'] = [analyzer.polarity\_scores(x)['pos'] for x in $$Amazon\_Review['cleanText']$$]
In [70]: Amazon_Review.head()
                          Userld Score
                                                                       Text
                                                                                                        cleanText compound
Out[70]:
                                                                                                                                neg
                                                                                                                                       neu
                                                                                                                                             pos
                                         McCann's Instant Oatmeal is great if you
                                                                                    mccann instant oatmeal great must
               AOVROBZ8BNTP7
                                                                                                                       0.9323 0.057 0.767 0.176
                                           This is a good instant oatmeal from the
                                                                               good instant oatmeal best oatmeal brand
            1 A3PMM0NFVEJGK9
                                                                                                                       0.9781 0.032 0.557 0.411
                                           I know the product title says Molecular
                                                                                      know product title say molecular
              A2VOZX7YBT0D6D
                                      5
                                                                                                                       0.9859 0.097 0.678 0.225
                                                                    Gastro...
                                                                                                    gastronomy le...
                                            The expiration date is 21 months from expiration date month day bought product
                A3NID9D9WMIV01
                                                                                                                       0.9062 0.000 0.725 0.275
                                                                   the day .
                                                                                                           tuna.
                                           This is a fantastic product, and I wish it
                                                                                 fantastic product wish readily available
           4 A1XM65S80UQ2MD
                                                                                                                       0.9565 0.066 0.661 0.273
                                                                       was...
                                                                                                            store..
In [71]: # Run the polarity score on the entire dataset
            res = {}
            for i, row in tqdm(df.iterrows(), total=len(df)):
                text = row['Text']
                myid = row['Id']
                 res[myid] = sia.polarity_scores(text)
              0%|
                             | 0/17243 [00:00<?, ?it/s]
```

```
In [72]:
          vaders = pd.DataFrame(res).T
          vaders = vaders.reset_index().rename(columns={'index': 'Id'})
          vaders = vaders.merge(df, how='left')
In [73]: # Now we have sentiment score and metadata
          vaders.head()
            ld
                             pos compound
                                               ProductId
                                                                   Userld
                                                                           ProfileName HelpfulnessNumerator HelpfulnessDenominator Score
Out[73]:
          0 1 0.069 0.839 0.092
                                     0.7103 B001EO5QW8 AOVROBZ8BNTP7
                                                                              S. Potter
                                                                                                       19
                                                                                                                            19
                                                                           Megan "Bad
             2 0.024 0.720 0.256
                                     0.9779 B001EO5QW8 A3PMM0NFVEJGK9
                                                                                                       13
                                                                                                                            13
                                                                          at Nicknames'
                                                                           Johnnycakes
          2 3 0.040 0.794 0.165
                                     0.9957 B003ZFRKGO A2VOZX7YBT0D6D
                                                                                                       15
                                                                                                                            15
                                                                          "Johnnycakes"
                                                                           Louie Arrighi
                                             B000ITVLE2 A3NID9D9WMIV01
            4 0.070 0.885 0.045
                                    -0.4721
                                                                                                       17
                                                                                                                            19
                                                                           "Lou da Joo"
          4 5 0.035 0.801 0.163
                                                                                                       13
                                     0.9676 B001UJEN6C A1XM65S80UQ2MD Joseph Kagan
                                                                                                                            13
```

In [74]: ax = sns.barplot(data=vaders, x='Score', y='compound')
 ax.set_title('Compund Score by Amazon Star Review')
 plt.show()



```
fig, axs = plt.subplots(1, 3, figsize=(12, 3))
sns.barplot(data=vaders, x='Score', y='pos', ax=axs[0])
sns.barplot(data=vaders, x='Score', y='neu', ax=axs[1])
sns.barplot(data=vaders, x='Score', y='neg', ax=axs[2])
axs[0].set_title('Positive')
axs[1].set_title('Neutral')
axs[2].set_title('Negative')
plt.tight_layout()
plt.show()
```



Textblob

```
In [76]: from textblob import TextBlob
In [78]: text_review = pd.DataFrame(Amazon_Review, columns=['Score', 'Text','cleanText'])
          def sentiment_analysis(rating_tb):
In [79]:
           def getSubjectivity(text):
             return TextBlob(text).sentiment.subjectivity
           #Create a function to get the polarity
           def getPolarity(text):
             return TextBlob(text).sentiment.polarity
           #Create two new columns 'Subjectivity' & 'Polarity'
           text_review['TextBlob_Subjectivity'] = text_review['cleanText'].apply(getSubjectivity)
           text_review ['TextBlob_Polarity'] =text_review['cleanText'].apply(getPolarity)
           def getAnalysis(score):
           if score < 0:
    return 'Negative'</pre>
            elif score == 0:
              return 'Neutral'
            else:
              return 'Positive'
           text_review['TextBlob_Analysis'] = rating_tb['TextBlob_Polarity'].apply(getAnalysis)
           return text_review
```

In [80]: sentiment_analysis(text_review)

TextBlob_Analysis	TextBlob_Polarity	TextBlob_Subjectivity	cleanText	Text	Score	80]:
Positive	0.243947	0.538509	mccann instant oatmeal great must oatmeal scra	McCann's Instant Oatmeal is great if you must	0 4	
Positive	0.396667	0.494466	good instant oatmeal best oatmeal brand us can	This is a good instant oatmeal from the best o	1 4	
Positive	0.089922	0.574573	know product title say molecular gastronomy le	I know the product title says Molecular Gastro	2 5	
Positive	0.540000	0.580000	expiration date month day bought product tuna	The expiration date is 21 months from the day	3 5	
Positive	0.071786	0.485952	fantastic product wish readily available store	This is a fantastic product, and I wish it was	4 5	
Neutral	0.000000	0.600000	sister law connaisseur salted caramel given gi	My sister-in-law is a connaisseur of salted ca	9 5	1723
Positive	0.150762	0.529011	review cover stevia raw product packet form tr	My review covers the Stevia In The Raw product	10 5	1724
Positive	0.039013	0.505875	first tie truvia fact decided replace truvia r	First of all, I have no ties with Truvia. In f	i 1 1	1724
Positive	0.017863	0.566010	certain product sold place could find review w	I'm not 100% certain this is the same product	2 5	1724
Positive	0.164683	0.458333	past would buy large quantity baker ammonia wo	In the past, I would have to buy a large quant	3 5	1724

17243 rows × 6 columns

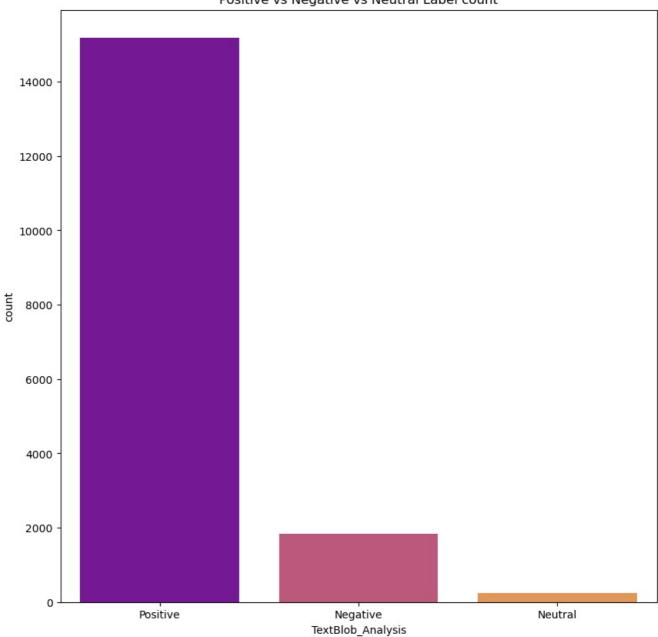
```
import seaborn as sns
sns.countplot(text_review['TextBlob_Analysis'], palette="plasma")
fig = plt.gcf()
fig.set_size_inches(10,10)
plt.title('Positive vs Negative vs Neutral Label count')
```

C:\Users\chira\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings warn(

Out[82]: Text(0.5, 1.0, 'Positive vs Negative vs Neutral Label count')

Positive vs Negative vs Neutral Label count



In []: