```
# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES
# TO THE CORRECT LOCATION (<a href="//kaggle/input">/kaggle/input</a>) IN YOUR NOTEBOOK,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.
import os
import sys
from tempfile import NamedTemporaryFile
from urllib.request import urlopen
from urllib.parse import unquote, urlparse
from urllib.error import HTTPError
from zipfile import ZipFile
import tarfile
import shutil
CHUNK SIZE = 40960
DATA_SOURCE_MAPPING = 'amazon-fine-food-reviews:https%3A%2F%2Fstorage.googleapis.com%2Fkaggle-data-sets%2F18%2F2157%2Fbundle%
KAGGLE_INPUT_PATH='/kaggle/input'
KAGGLE_WORKING_PATH='/kaggle/working'
KAGGLE_SYMLINK='kaggle
!umount /kaggle/input/ 2> /dev/null
shutil.rmtree('/kaggle/input', ignore_errors=True)
os.makedirs(KAGGLE INPUT PATH, 0o777, exist ok=True)
os.makedirs(KAGGLE_WORKING_PATH, 0o777, exist_ok=True)
  os.symlink(KAGGLE_INPUT_PATH, os.path.join("..", 'input'), target_is_directory=True)
except FileExistsError:
 pass
trv:
  os.symlink(KAGGLE_WORKING_PATH, os.path.join("..", 'working'), target_is_directory=True)
except FileExistsError:
  pass
for data source mapping in DATA SOURCE MAPPING.split(','):
    directory, download_url_encoded = data_source_mapping.split(':')
    download url = unquote(download url encoded)
    filename = urlparse(download_url).path
    destination_path = os.path.join(KAGGLE_INPUT_PATH, directory)
        with urlopen(download_url) as fileres, NamedTemporaryFile() as tfile:
            total_length = fileres.headers['content-length']
            print(f'Downloading {directory}, {total_length} bytes compressed')
            dl = 0
            data = fileres.read(CHUNK SIZE)
            while len(data) > 0:
                dl += len(data)
                tfile.write(data)
                done = int(50 * dl / int(total_length))
                sys.stdout.write(f'' r[{'=' * done}{' ' * (50-done)}] {dl} bytes downloaded")
                sys.stdout.flush()
                data = fileres.read(CHUNK_SIZE)
            if filename.endswith('.zip'):
              with ZipFile(tfile) as zfile:
                zfile.extractall(destination_path)
              with tarfile.open(tfile.name) as tarfile:
                tarfile.extractall(destination path)
            print(f'\nDownloaded\ and\ uncompressed:\ \{directory\}')
    except HTTPError as e:
        print(f'Failed to load (likely expired) {download_url} to path {destination_path}')
        continue
    except OSError as e:
        print(f'Failed to load {download_url} to path {destination_path}')
print('Data source import complete.')
    Downloading amazon-fine-food-reviews, 253873708 bytes compressed
                                             ========] 253873708 bytes downloaded
    Downloaded and uncompressed: amazon-fine-food-reviews
    Data source import complete.
```

Sentiment Analysis in Python

This notebook is part of a tutorial that can be found on my youtube channel here, please check it out!

In this notebook we will be doing some sentiment analysis in python using two different techniques:

- 1. VADER (Valence Aware Dictionary and sEntiment Reasoner) Bag of words approach
- 2. Roberta Pretrained Model from 🤗
- 3. Huggingface Pipeline

Step 0. Read in Data and NLTK Basics

		Id	ProductId	UserId	ProfileName	HelpfulnessNumerator	Helpfu
	0	1	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	
	1	2	B00813GRG4	A1D87F6ZCVE5NK	dll pa	0	
4	2	3	B000LQOCH0	ABXLMWJIXXAIN	Natalia Corres "Natalia	1	>

Quick EDA



Basic NLTK

like/VB

```
example = df['Text'][50]
print(example)
     This oatmeal is not good. Its mushy, soft, I don't like it. Quaker Oats is the way to go.
nltk.download('punkt')
tokens = nltk.word_tokenize(example)
tokens[:10]
      [nltk_data] Downloading package punkt to /root/nltk_data...
      [nltk_data] Unzipping tokenizers/punkt.zip.
['This', 'oatmeal', 'is', 'not', 'good', '.', 'Its', 'mushy', ',', 'soft']
 nltk.download('averaged_perceptron_tagger')
tagged = nltk.pos_tag(tokens)
tagged[:10]
      [nltk_data] Downloading package averaged_perceptron_tagger to
                        /root/nltk data...
      [nltk_data]
                       Unzipping taggers/averaged_perceptron_tagger.zip.
      [nltk data]
      [('This', 'DT'),
      [('This', 'DT'),
  ('oatmeal', 'NN'),
  ('is', 'VBZ'),
  ('not', 'RB'),
  ('good', 'JJ'),
  ('.', '.'),
  ('Its', 'PRP$'),
  ('mushy', 'NN'),
  (',', ','),
  ('soft', 'JJ')]
nltk.download('maxent_ne_chunker')
nltk.download('words')
entities = nltk.chunk.ne_chunk(tagged)
entities.pprint()
      (S
        This/DT
        oatmeal/NN
        is/VBZ
        not/RB
        good/JJ
        ./.
        .
Its/PRP$
        mushy/NN
        ,/,
        soft/JJ
        I/PRP
        do/VBP
        n't/RB
```

```
it/PRP
./.
(ORGANIZATION Quaker/NNP Oats/NNPS)
is/VBZ
the/DT
way/NN
to/TO
go/VB
./.)
[nltk_data] Downloading package maxent_ne_chunker to
[nltk_data] /root/nltk_data...
[nltk_data] Package maxent_ne_chunker is already up-to-date!
[nltk_data] Downloading package words to /root/nltk_data...
[nltk_data] Unzipping corpora/words.zip.
```

Step 1. VADER Seniment Scoring

We will use NLTK's SentimentIntensityAnalyzer to get the neg/neu/pos scores of the text.

- This uses a "bag of words" approach:
 - 1. Stop words are removed
 - 2. each word is scored and combined to a total score.

```
nltk.download('vader_lexicon')
from nltk.sentiment import SentimentIntensityAnalyzer
from tqdm.notebook import tqdm
sia = SentimentIntensityAnalyzer()
    [nltk_data] Downloading package vader_lexicon to /root/nltk_data...
sia.polarity_scores('I am so happy!')
    {'neg': 0.0, 'neu': 0.318, 'pos': 0.682, 'compound': 0.6468}
sia.polarity_scores('This is the worst thing ever.')
    {'neg': 0.451, 'neu': 0.549, 'pos': 0.0, 'compound': -0.6249}
sia.polarity_scores(example)
    {'neg': 0.22, 'neu': 0.78, 'pos': 0.0, 'compound': -0.5448}
# 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)
    100%
                                              500/500 [00:00<00:00, 981.74it/s]
vaders = pd.DataFrame(res).T
vaders = vaders.reset index().rename(columns={'index': 'Id'})
vaders = vaders.merge(df, how='left')
# Now we have sentiment score and metadata
vaders.head()
```

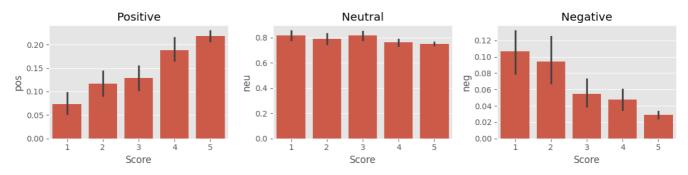
	Id	neg	neu	pos	compound	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominato
0	1	0.000	0.695	0.305	0.9441	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	
1	2	0.138	0.862	0.000	-0.5664	B00813GRG4	A1D87F6ZCVE5NK	dll pa	0	
2	3	0.091	0.754	0.155	0.8265	B000LQOCH0	ABXLMWJIXXAIN	Natalia Corres "Natalia Corres"	1	
3	4	0.000	1.000	0.000	0.0000	B000UA0QIQ	A395BORC6FGVXV	Karl	3	
4	5	0.000	0.552	0.448	0.9468	B006K2ZZ7K	A1UQRSCLF8GW1T	Michael D. Bigham "M. Wassir"	0	

Plot VADER results

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()
```



Step 3. Roberta Pretrained Model

- · Use a model trained of a large corpus of data.
- · Transformer model accounts for the words but also the context related to other words.

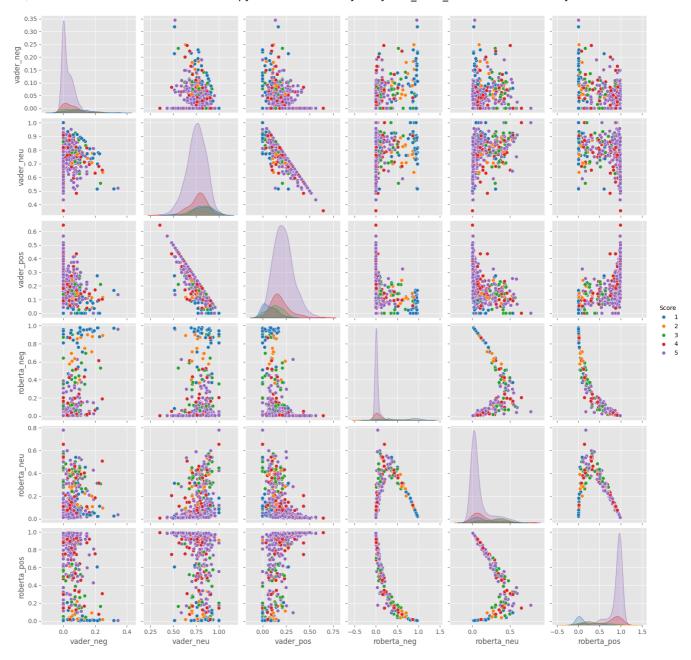
```
from transformers import AutoTokenizer
from transformers import AutoModelForSequenceClassification
from scipy.special import softmax
MODEL = f"cardiffnlp/twitter-roberta-base-sentiment"
tokenizer = AutoTokenizer.from_pretrained(MODEL)
model = AutoModelForSequenceClassification.from_pretrained(MODEL)
     /usr/local/lib/python3.10/dist-packages/huggingface_hub/utils/_token.py:88: UserWarning:
     The secret `HF_TOKEN` does not exist in your Colab secrets.
     To authenticate with the Hugging Face Hub, create a token in your settings tab (https://huggingface.co/settings/tokens),
     You will be able to reuse this secret in all of your notebooks.
     Please note that authentication is recommended but still optional to access public models or datasets.
      warnings.warn(
     config.json: 100%
                                                          747/747 [00:00<00:00, 29.5kB/s]
     vocab.json: 100%
                                                          899k/899k [00:00<00:00, 11.4MB/s]
                                                          456k/456k [00:00<00:00, 18.7MB/s]
     merges.txt: 100%
     special_tokens_map.json: 100%
                                                                    150/150 [00:00<00:00, 4.27kB/s]
     pytorch model.bin: 100%
                                                               499M/499M [00:04<00:00, 102MB/s]
# VADER results on example
print(example)
sia.polarity_scores(example)
    NameFrror
                                                 Traceback (most recent call last)
     <ipython-input-4-959d041be6aa> in <cell line: 2>()
           1 # VADER results on example
     ----> 2 print(example)
           3 sia.polarity_scores(example)
    NameError: name 'example' is not defined
# Run for Roberta Model
encoded_text = tokenizer(example, return_tensors='pt')
output = model(**encoded_text)
scores = output[0][0].detach().numpy()
scores = softmax(scores)
scores_dict = {
    'roberta_neg' : scores[0],
    'roberta_neu' : scores[1],
    'roberta_pos' : scores[2]
print(scores_dict)
```

```
NameError
                                               Traceback (most recent call last)
     <ipython-input-5-032f6db3c850> in <cell line: 2>()
           1 # Run for Roberta Model
     ----> 2 encoded_text = tokenizer(example, return_tensors='pt')
          3 output = model(**encoded_text)
           4 scores = output[0][0].detach().numpy()
          5 scores = softmax(scores)
    NameError: name 'example' is not defined
def polarity_scores_roberta(example):
    encoded_text = tokenizer(example, return_tensors='pt')
    output = model(**encoded_text)
    scores = output[0][0].detach().numpy()
    scores = softmax(scores)
    scores_dict = {
        'roberta_neg' : scores[0],
        'roberta_neu' : scores[1],
        'roberta_pos' : scores[2]
    }
    return scores_dict
res = \{\}
for i, row in tqdm(df.iterrows(), total=len(df)):
    try:
        text = row['Text']
        myid = row['Id']
        vader_result = sia.polarity_scores(text)
        vader_result_rename = {}
        for key, value in vader_result.items():
            vader_result_rename[f"vader_{key}"] = value
        roberta_result = polarity_scores_roberta(text)
        both = {**vader_result_rename, **roberta_result}
       res[myid] = both
    except RuntimeError:
        print(f'Broke for id {myid}')
     100%
                                              500/500 [03:21<00:00, 2.74it/s]
     Broke for id 83
    Broke for id 187
results_df = pd.DataFrame(res).T
results_df = results_df.reset_index().rename(columns={'index': 'Id'})
results df = results df.merge(df, how='left')

    Compare Scores between models

results_df.columns
```

Step 3. Combine and compare



Step 4: Review Examples:

• Positive 1-Star and Negative 5-Star Reviews

Lets look at some examples where the model scoring and review score differ the most.

```
results_df.query('Score == 1') \
    .sort_values('roberta_pos', ascending=False)['Text'].values[0]

'I felt energized within five minutes, but it lasted for about 45 minutes. I paid $3.99 for this drink. I could have ju st drunk a cup of coffee and saved my money.'

results_df.query('Score == 1') \
    .sort_values('vader_pos', ascending=False)['Text'].values[0]

'So we cancelled the order. It was cancelled without any problem. That is a positive note...'
```

```
# nevative sentiment 5-Star view

results_df.query('Score == 5') \
    .sort_values('roberta_neg', ascending=False)['Text'].values[0]
    'this was sooooo deliscious but too bad i ate em too fast and gained 2 pds! my fault'

results_df.query('Score == 5') \
    .sort_values('vader_neg', ascending=False)['Text'].values[0]
    'this was sooooo deliscious but too bad i ate em too fast and gained 2 pds! my fault'
```

Extra: The Transformers Pipeline

• Quick & easy way to run sentiment predictions