```
In [33]: !pip install textblob
```

Requirement already satisfied: textblob in c:\users\arnav\appdata\local\programs\python \python37\lib\site-packages (0.17.1)

Requirement already satisfied: nltk>=3.1 in c:\users\arnav\appdata\local\programs\python \python37\lib\site-packages (from textblob) (3.5)

Requirement already satisfied: regex in c:\users\arnav\appdata\local\programs\python\python37\lib\site-packages (from nltk>=3.1->textblob) (2020.6.8)

Requirement already satisfied: joblib in c:\users\arnav\appdata\local\programs\python\py thon37\lib\site-packages (from nltk>=3.1->textblob) (0.15.1)

Requirement already satisfied: click in c:\users\arnav\appdata\local\programs\python\python\python37\lib\site-packages (from nltk>=3.1->textblob) (8.1.3)

Requirement already satisfied: tqdm in c:\users\arnav\appdata\local\programs\python\pyth on37\lib\site-packages (from nltk>=3.1->textblob) (4.55.1)

Requirement already satisfied: colorama in c:\users\arnav\appdata\local\programs\python \python37\lib\site-packages (from click->nltk>=3.1->textblob) (0.4.6)

Requirement already satisfied: importlib-metadata in c:\users\arnav\appdata\local\progra ms\python\python37\lib\site-packages (from click->nltk>=3.1->textblob) (6.3.0)

Requirement already satisfied: typing-extensions>=3.6.4 in c:\users\arnav\appdata\local \programs\python\python37\lib\site-packages (from importlib-metadata->click->nltk>=3.1-> textblob) (4.4.0)

Requirement already satisfied: zipp>=0.5 in c:\users\arnav\appdata\local\programs\python \python37\lib\site-packages (from importlib-metadata->click->nltk>=3.1->textblob) (3.1. 0)

```
[notice] A new release of pip is available: 23.0.1 -> 23.1
[notice] To update, run: python.exe -m pip install --upgrade pip
```

```
In [1]: import pandas as pd
    from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
    import numpy as np
    from scipy.stats import pearsonr
```

Reading data from dataset

```
In [2]: df=pd.read_csv('15uni.csv',encoding='utf-8')
    df.head()
```

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1	Visio	Mission	University name	World Rank	Serial Number	
	The University's core values are a follows: f	The mission of the University of Cambridge is	Cambridge University	2	1	0
	To be a global center of excellent for lifelo	We inspire people locally, nationally, and glo	University of Oxford	4	2	1
	Enabling social good: societ advancement is	To transform people's lives through education,	UCL	8	3	2
	To recruit and develop the world most promis	The mission of our University is the creation,	University of Edinburg	15	4	3
	We envision a future in whic society thrives	We, the undersigned alliance partners, Budapes	Universit PSL	26	5	4

Initializing the sentiment analyzer

```
In [4]: analyzer = SentimentIntensityAnalyzer()
```

Getting sentiment values of Mission Statements

```
In [5]: mission=df['Mission']
        neg=[]
        pos=[]
        neu=[]
        comp=[]
        for sentence in mission:
          vs = analyzer.polarity scores(sentence)
          # print("{:-<65} {}".format(sentence, str(vs)))</pre>
          neg.append(vs['neg'])
         pos.append(vs['pos'])
          neu.append(vs['neu'])
          comp.append(vs['compound'])
        df['neg mission']=neg
        df['pos mission']=pos
        df['neu mission']=neu
        df['comp mission']=comp
        df.head()
```

	ar, nead()											
Out[5]:		Serial Number	World Rank	University name	Mission	Vision	neg_mission	pos_mission	neu_mission	comp_r		
	0	1	2	Cambridge University	The mission of the University of Cambridge is	The University's core values are as follows: f	0.000	0.136	0.864			
	1	2	4	University of Oxford	We inspire people locally, nationally, and glo	To be a global center of excellence for lifelo	0.000	0.272	0.728			
	2	3	8	UCL	To transform people's lives through education,	Enabling social good: societal advancement is	0.023	0.164	0.813			
	3	4	15	University of Edinburg	The mission of our University is the creation,	To recruit and develop the world's most promis	0.000	0.191	0.809			
	4	5	26	Universit PSL	We, the undersigned alliance partners, Budapes	We envision a future in which society thrives	0.000	0.143	0.857			

Getting sentiment values of Vision Statements

```
In [6]: vision=df['Vision']
    neg=[]
    pos=[]
    comp=[]
    for sentence in vision:
       vs = analyzer.polarity_scores(sentence)
       # print("{:-<65} {}".format(sentence, str(vs)))
       neg.append(vs['neg'])
       pos.append(vs['pos'])
       neu.append(vs['neu'])</pre>
```

```
comp.append(vs['compound'])

df['neg_vision']=neg
df['pos_vision']=pos
df['neu_vision']=neu
df['comp_vision']=comp
df.head()
```

Out[6]:

:		Serial Number	World Rank	University name	Mission	Vision	neg_mission	pos_mission	neu_mission	comp_r
	0	1	2	Cambridge University	The mission of the University of Cambridge is	The University's core values are as follows: f	0.000	0.136	0.864	
	1	2	4	University of Oxford	We inspire people locally, nationally, and glo	To be a global center of excellence for lifelo	0.000	0.272	0.728	
	2	3	8	UCL	To transform people's lives through education,	Enabling social good: societal advancement is	0.023	0.164	0.813	
	3	4	15	University of Edinburg	The mission of our University is the creation,	To recruit and develop the world's most promis	0.000	0.191	0.809	
	4	5	26	Universit PSL	We, the undersigned alliance partners, Budapes	We envision a future in which society thrives	0.000	0.143	0.857	

Formulating the compound values of mission and vision statements and the world rank

```
In [7]: world_rank=df['World Rank']
    mission_compound=df['comp_mission']
    vision_compound=df['comp_vision']
    print(type(world_rank[0]))
    df1=pd.DataFrame()
    df1['World Rank']=[float(x) for x in world_rank]
    df1['Mission Compound']=[float(x) for x in mission_compound]
    df1['Vision Compound']=[float(x) for x in vision_compound]
    df1[head()
```

<class 'numpy.int64'>

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	World Rank	Mission Compound	Vision Compound
0	2.0	0.6249	0.9022
1	4.0	0.8020	0.9828
2	8.0	0.9287	0.9287
3	15.0	0.9633	0.8803
4	26.0	0.9985	0.9571

Finding the pearsons correlation coefficient

```
In [8]: corr, pval = pearsonr(df1['World Rank'], df1['Mission Compound'])
        print(f"Pearson correlation coefficient between World Rank and Mission Compound: {corr:.
        corr, pval = pearsonr(df['World Rank'], df['pos mission'])
        print(f"Pearson correlation coefficient between World Rank and pos mision: {corr:.2f}, p
        corr, pval = pearsonr(df['World Rank'], df['neu mission'])
        print(f"Pearson correlation coefficient between World Rank and neu mision: {corr:.2f}, p
        corr, pval = pearsonr(df['World Rank'], df['neg mission'])
        print(f"Pearson correlation coefficient between World Rank and neg mision: {corr:.2f}, p
        corr, pval = pearsonr(df1['World Rank'], df1['Vision Compound'])
        print(f"Pearson correlation coefficient between World Rank and Vision Compound: {corr:.2
        corr, pval = pearsonr(df['World Rank'], df['pos vision'])
        print(f"Pearson correlation coefficient between World Rank and pos vision: {corr:.2f}, p
        corr, pval = pearsonr(df['World Rank'], df['neu vision'])
        print(f"Pearson correlation coefficient between World Rank and neu vision: {corr:.2f}, p
        corr, pval = pearsonr(df['World Rank'], df['neg vision'])
        print(f"Pearson correlation coefficient between World Rank and neg vision: {corr:.2f}, p
        Pearson correlation coefficient between World Rank and Mission Compound: -0.08, p-value:
        0.66
        Pearson correlation coefficient between World Rank and pos mision: 0.04, p-value: 0.81
        Pearson correlation coefficient between World Rank and neu mision: -0.01, p-value: 0.94
        Pearson correlation coefficient between World Rank and neg mision: -0.25, p-value: 0.14
        Pearson correlation coefficient between World Rank and Vision Compound: -0.33, p-value:
        0.05
        Pearson correlation coefficient between World Rank and pos vision: -0.11, p-value: 0.54
        Pearson correlation coefficient between World Rank and neu vision: 0.04, p-value: 0.80
        Pearson correlation coefficient between World Rank and neg vision: 0.34, p-value: 0.05
```

Making a column for mission and vision statements combined

```
In [9]: | df["Total"] = df[["Mission", "Vision"]].apply("-".join, axis=1)
```

Getting the sentiment values of mission and vision statements combined

```
In [10]: total=df['Total']
         neg=[]
         pos=[]
         neu=[]
          comp=[]
         for sentence in total:
           vs = analyzer.polarity scores(sentence)
            # print("{:-<65} {}".format(sentence, str(vs)))</pre>
           neg.append(vs['neg'])
           pos.append(vs['pos'])
           neu.append(vs['neu'])
           comp.append(vs['compound'])
         df['neg total']=neg
         df['pos total']=pos
          df['neu total']=neu
          df['comp total']=comp
          df.head()
```

	Number	Rank	name						
0	1	2	Cambridge University	The mission of the University of Cambridge is	The University's core values are as follows: f	0.000	0.136	0.864	
1	2	4	University of Oxford	We inspire people locally, nationally, and glo	To be a global center of excellence for lifelo	0.000	0.272	0.728	
2	3	8	UCL	To transform people's lives through education,	Enabling social good: societal advancement is	0.023	0.164	0.813	
3	4	15	University of Edinburg	The mission of our University is the creation,	To recruit and develop the world's most promis	0.000	0.191	0.809	
4	5	26	Universit PSL	We, the undersigned alliance partners, Budapes	We envision a future in which society thrives	0.000	0.143	0.857	

Finding the pearsons correlation coefficients for mission and vision statements combined

```
In [11]: corr, pval = pearsonr(df1['World Rank'], df['neg_total'])
    print(f"Pearson correlation coefficient between World Rank and neg_total: {corr:.2f}, p-
    corr, pval = pearsonr(df['World Rank'], df['pos_total'])
    print(f"Pearson correlation coefficient between World Rank and pos_total: {corr:.2f}, p-
    corr, pval = pearsonr(df['World Rank'], df['neu_total'])
    print(f"Pearson correlation coefficient between World Rank and neu_total: {corr:.2f}, p-
    corr, pval = pearsonr(df['World Rank'], df['comp_total'])
    print(f"Pearson correlation coefficient between World Rank and comp_total: {corr:.2f}, p-
    Pearson correlation coefficient between World Rank and neg_total: 0.09, p-value: 0.60
    Pearson correlation coefficient between World Rank and pos_total: -0.01, p-value: 0.96
    Pearson correlation coefficient between World Rank and neu total: -0.00, p-value: 0.99
```

Finding sentiment values using NLTK

```
In [29]: def sentiment_analyzer(sentence):
    sia=SentimentIntensityAnalyzer()
    sent=sia.polarity_scores(sentence)
    # print(sent)
    return sent

# sentiment_analyzer("Our mission is to improve peoples health and wellbeing through out
In [22]: pltk pog = []
```

Pearson correlation coefficient between World Rank and comp total: -0.10, p-value: 0.58

```
In [32]: nltk_neg = []
nltk_pos = []
```

```
nltk_neu = []
nltk_comp = []
for sentence in total:
    sent = sentiment_analyzer(sentence)
    nltk_neg.append(sent['neg'])
    nltk_pos.append(sent['pos'])
    nltk_neu.append(sent['neu'])
    nltk_comp.append(sent['compound'])

df['nltk_neg']=nltk_neg
df['nltk_pos']=nltk_pos
df['nltk_neu']=nltk_neu
df['nltk_comp']=nltk_comp
df.head()
```

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:		Serial Number	World Rank	University name	Mission	Vision	neg_mission	pos_mission	neu_mission	comp_r
	0	1	2	Cambridge University	The mission of the University of Cambridge is	The University's core values are as follows: f	0.000	0.136	0.864	
	1	2	4	University of Oxford	We inspire people locally, nationally, and glo	To be a global center of excellence for lifelo	0.000	0.272	0.728	
	2	3	8	UCL	To transform people's lives through education,	Enabling social good: societal advancement is	0.023	0.164	0.813	
	3	4	15	University of Edinburg	The mission of our University is the creation,	To recruit and develop the world's most promis	0.000	0.191	0.809	
	4	5	26	Universit PSL	We, the undersigned alliance partners, Budapes	We envision a future in which society thrives	0.000	0.143	0.857	

5 rows × 22 columns

Finding sentiment values using textblob

```
In [34]: from textblob import TextBlob
In [38]: text_polarity = []
    text_subjectivity = []
    for tot in total:
        sent=TextBlob(tot)
        text_polarity.append(sent.sentiment.polarity)
        text_subjectivity.append(sent.sentiment.subjectivity)

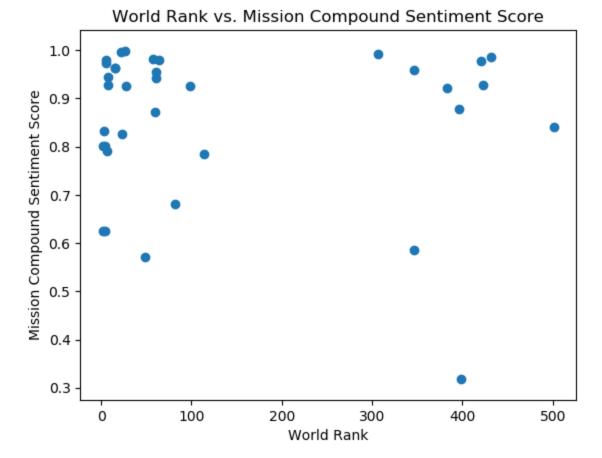
df['text_subjectivity']=text_polarity
    df['text_subjectivity']=text_subjectivity
    df.head()
```

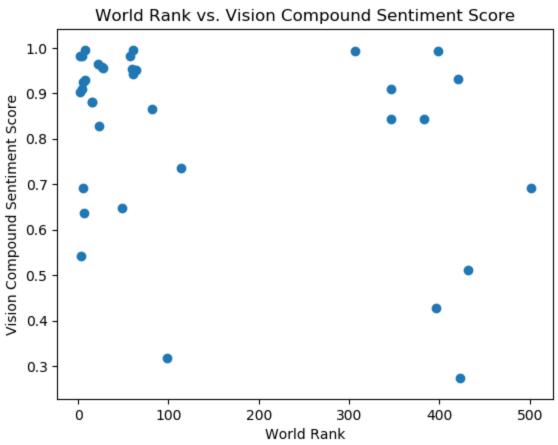
		Serial Number	World Rank	University name	Mission	Vision	neg_mission	pos_mission	neu_mission	comp_r
	0	1	2	Cambridge University	The mission of the University of Cambridge is	The University's core values are as follows: f	0.000	0.136	0.864	
	1	2	4	University of Oxford	We inspire people locally, nationally, and glo	To be a global center of excellence for lifelo	0.000	0.272	0.728	
	2	3	8	UCL	To transform people's lives through education,	Enabling social good: societal advancement is	0.023	0.164	0.813	
3	3	4	15	University of Edinburg	The mission of our University is the creation,	To recruit and develop the world's most promis	0.000	0.191	0.809	
	4	5	26	Universit PSL	We, the undersigned alliance partners, Budapes	We envision a future in which society thrives	0.000	0.143	0.857	

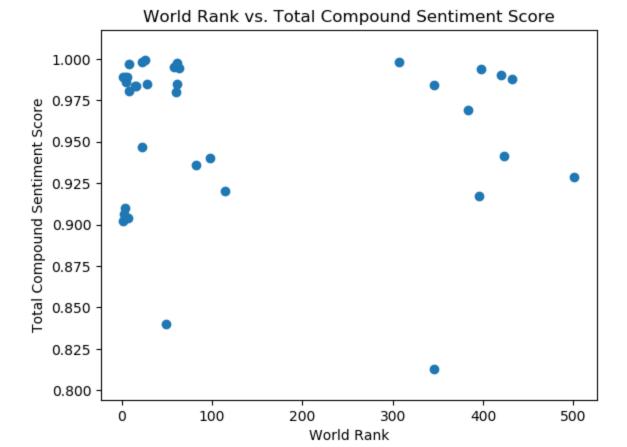
5 rows × 24 columns

Out[38]:

```
In [13]: import matplotlib.pyplot as plt
         # Plotting mission compound sentiment score against world rank
         plt.scatter(df1['World Rank'], df1['Mission Compound'])
         plt.title('World Rank vs. Mission Compound Sentiment Score')
         plt.xlabel('World Rank')
         plt.ylabel('Mission Compound Sentiment Score')
         plt.show()
         # Plotting vision compound sentiment score against world rank
         plt.scatter(df1['World Rank'], df1['Vision Compound'])
         plt.title('World Rank vs. Vision Compound Sentiment Score')
         plt.xlabel('World Rank')
         plt.ylabel('Vision Compound Sentiment Score')
         plt.show()
         # Plotting total compound sentiment score against world rank
         plt.scatter(df['World Rank'], df['comp total'])
         plt.title('World Rank vs. Total Compound Sentiment Score')
         plt.xlabel('World Rank')
         plt.ylabel('Total Compound Sentiment Score')
         plt.show()
```







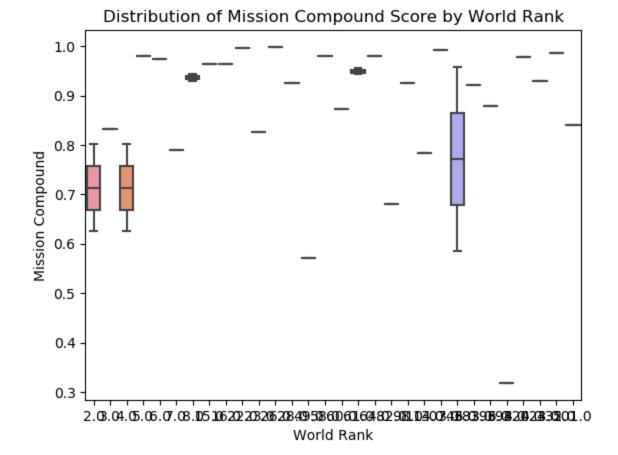
```
In [14]: import seaborn as sns

# Creating a correlation matrix
corr_matrix = df[['World Rank', 'comp_mission', 'pos_mission', 'neu_mission', 'neg_missi

# Plotting a heatmap
sns.heatmap(corr_matrix, cmap='coolwarm', annot=True, fmt='.2f')
plt.title('Correlation between World Rank and Sentiment Scores')
plt.show()
```

```
Correlation between World Rank and Sentiment Scores
                                                                                    1.00
  World Rank -1.00-0.080.04-0.010.250.330.110.040.34-0.100.01-0.000.09
comp_mission -0.081.000.36-0.360.000.020.03-0.040.060.590.24-0.220.10
                                                                                    - 0.75
  pos_mission -0.040.361.00-0.920.120.040.32-0.360.170.360.79-0.780.03
                                                                                    - 0.50
  neu mission -0.010.360.991.000.010.04-0.320.35-0.160.360.780.79-0.09
  neg mission -0.250.000.120.01 1.00-0.040.050.07-0.090.060.060.010.58
                                                                                    - 0.25
  comp vision -0.330.020.040.04-0.041.000.45-0.400.330.450.17-0.140.25
    pos vision -0.110.030.32-0.320.050.451.00-0.980.130.270.75-0.730.12
                                                                                    - 0.00
   neu vision -0.04-0.040.360.35 0.07-0.400.981.00-0.050.300.770.77-0.01
                                                                                    - -0.25
   neg_vision -0.340.060.17-0.160.090.330.130.051.000.120.10-0.190.70
   comp_total -0.100.590.360.360.060.450.27-0.300.121.000.46-0.460.03
                                                                                     -0.50
     pos_total -0.010.240.79-0.780.060.170.75-0.770.100.461.00-0.990.03
    neu_total -0.000.220.780.790.01-0.140.730.77-0.190.460.991.000.15
                                                                                    - -0.75
    neg_total -0.09-0.100.03-0.09<mark>0.58-0.250.12-0.01</mark>0.70 0.030.03-0.15
                      comp_mission
                                        comp_vision
                           oos_mission
                               neu_mission
                                              pos_vision
                                                  neu_vision
                                                       neg_vision
                                                           comp_total
                                    neg_mission
                                                                oos_total
                                                                     neu_total
                                                                          neg_total
```

```
In [15]: # Boxplot
         sns.boxplot(x='World Rank', y='Mission Compound', data=df1)
         plt.title('Distribution of Mission Compound Score by World Rank')
         plt.show()
         # Line plot
         sns.lineplot(y='Vision Compound', x='World Rank', data=df1,label='Vision Compound')
         plt.title('Trend of Vision Compound Score and World Rank')
         plt.ylabel('Vision Compound')
         plt.show()
         sns.lineplot(y='Mission Compound', x='World Rank', data=df1,label='Mission Compound')
         plt.title('Trend of Mission Compound Score and World Rank')
         plt.ylabel('Mission Compound')
         plt.show()
         sns.lineplot(y='comp total', x='World Rank', data=df,label='Total Compound')
         plt.title('Trend of Total Compound Score and World Rank')
         plt.ylabel('Total Compound')
         plt.show()
         # Bar plot
         sns.barplot(x='World Rank', y='neg total', data=df)
         plt.title('Average Negative Sentiment Score by World Rank Group')
         plt.show()
         # Scatter plot with regression line
         sns.regplot(x='World Rank', y='pos total', data=df)
         plt.title('Relationship between World Rank and Positive Sentiment Score')
         plt.show()
```



c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\cbook__init__.py:1377: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, N one]`) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

x[:, None]

c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\axes _base.py:237: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, None] `) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

x = x[:, np.newaxis]

c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\axes
_base.py:239: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, None]
`) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

y = y[:, np.newaxis]

Trend of Vision Compound Score and World Rank 1.0 0.9 0.8 Vision Compound 0.7 0.6 0.5 0.4 0.3 Vision Compound 0 100 200 300 400 500

c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\cbook __init__.py:1377: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, N one]`) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

World Rank

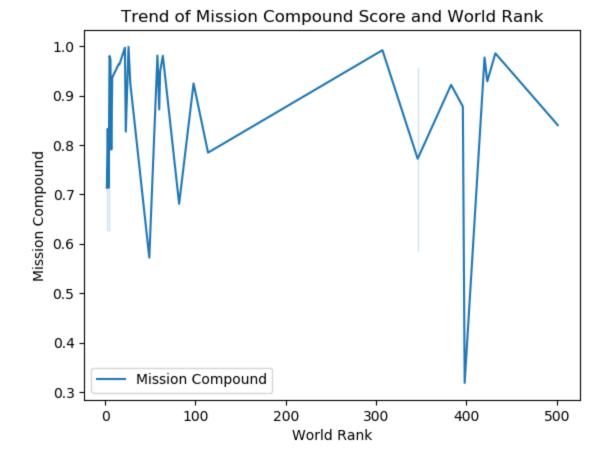
x[:, None]

c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\axes _base.py:237: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, None] `) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

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c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\axes _base.py:239: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, None] `) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

y = y[:, np.newaxis]



c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\cbook__init__.py:1377: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, N one]`) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

x[:, None]

c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\axes _base.py:237: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, None] `) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

x = x[:, np.newaxis]

c:\Users\Arnav\AppData\Local\Programs\Python\Python37\lib\site-packages\matplotlib\axes _base.py:239: FutureWarning: Support for multi-dimensional indexing (e.g. `obj[:, None] `) is deprecated and will be removed in a future version. Convert to a numpy array before indexing instead.

y = y[:, np.newaxis]

