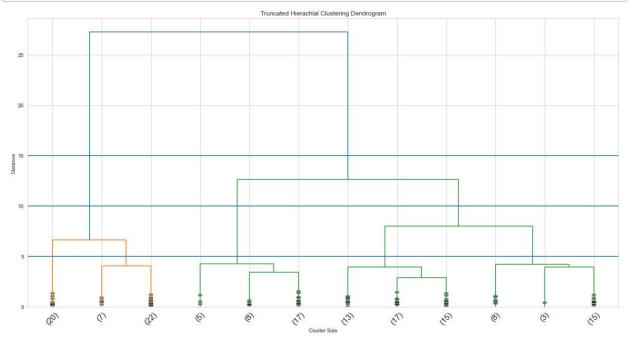
```
In [1]:
        import numpy as np
        import pandas as pd
        import scipy
        from scipy.cluster.hierarchy import dendrogram,linkage
        from scipy.cluster.hierarchy import fcluster
        from scipy.cluster.hierarchy import cophenet
        from scipy.spatial.distance import pdist
        import matplotlib.pyplot as plt
        from pylab import rcParams
        import seaborn as sb
        import sklearn
        from sklearn import datasets
        from sklearn.cluster import AgglomerativeClustering
        import sklearn.metrics as sm
        from sklearn.preprocessing import scale
In [2]: #Configure the output
        np.set_printoptions(precision=4, suppress=True)
        %matplotlib inline
        rcParams["figure.figsize"] =20,10
        sb.set style("whitegrid")
In [3]: iris = datasets.load iris()
        #scale the data
        data = scale(iris.data)
        target = pd.DataFrame(iris.target)
        variable names = iris.feature names
        data[0:10]
Out[3]: array([[-0.9007, 1.019, -1.3402, -1.3154],
               [-1.143, -0.132, -1.3402, -1.3154],
               [-1.3854, 0.3284, -1.3971, -1.3154],
               [-1.5065, 0.0982, -1.2834, -1.3154],
               [-1.0218, 1.2492, -1.3402, -1.3154],
               [-0.5372, 1.9398, -1.1697, -1.0522],
               [-1.5065, 0.7888, -1.3402, -1.1838],
               [-1.0218, 0.7888, -1.2834, -1.3154],
               [-1.7489, -0.3622, -1.3402, -1.3154],
               [-1.143, 0.0982, -1.2834, -1.4471]])
```

```
In [4]: z = linkage(data,"ward")
#generate dendrogram
dendrogram(z,truncate_mode= "lastp", p =12, leaf_rotation=45,leaf_font_size=15, s
plt.title("Truncated Hierachial Clustering Dendrogram")
plt.xlabel("Cluster Size")
plt.ylabel("Distance")
#divide the cluster
plt.axhline(y=15)
plt.axhline(5)
plt.axhline(10)
plt.show()
```



```
In [5]: #based on the dendrogram we have two clusetes
k =3
#build the model
HClustering = AgglomerativeClustering(n_clusters=k , affinity="euclidean",linkage
#fit the model on the dataset
HClustering.fit(data)
#accuracy of the model
sm.accuracy_score(target,HClustering.labels_)
```

Out[5]: 0.0133333333333333334

```
In [6]: #based on the dendrogram we have two clusetes
k =3
    #build the model
HClustering = AgglomerativeClustering(n_clusters=k , affinity="euclidean",linkage
#fit the model on the dataset
HClustering.fit(data)
#accuracy of the model
sm.accuracy_score(target,HClustering.labels_)
```

Out[6]: 0.0133333333333333333

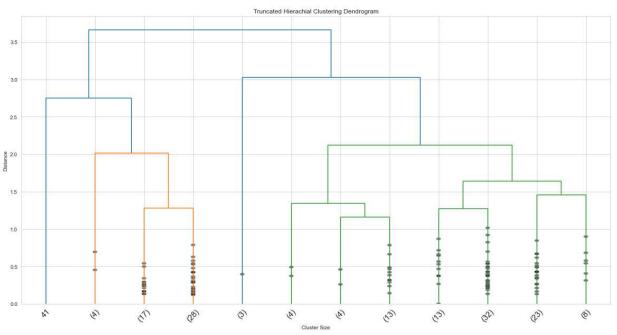
```
In [7]: #based on the dendrogram we have two clusetes
k =3
    #build the model
HClustering = AgglomerativeClustering(n_clusters=k , affinity="euclidean",linkage
#fit the model on the dataset
HClustering.fit(data)
#accuracy of the model
sm.accuracy_score(target,HClustering.labels_)
```

Out[7]: 0.686666666666666

```
In [8]: #based on the dendrogram we have two clusetes
k =3
    #build the model
HClustering = AgglomerativeClustering(n_clusters=k , affinity="euclidean",linkage
#fit the model on the dataset
HClustering.fit(data)
#accuracy of the model
sm.accuracy_score(target,HClustering.labels_)
```

Out[8]: 0.0

```
In [9]: z = linkage(data, "average")
#generate dendrogram
dendrogram(z,truncate_mode= "lastp", p =12, leaf_rotation=45,leaf_font_size=15, s
plt.title("Truncated Hierachial Clustering Dendrogram")
plt.xlabel("Cluster Size")
plt.ylabel("Distance")
#divide the cluster
plt.axhline(y=15)
plt.axhline(5)
plt.axhline(10)
plt.show()
```

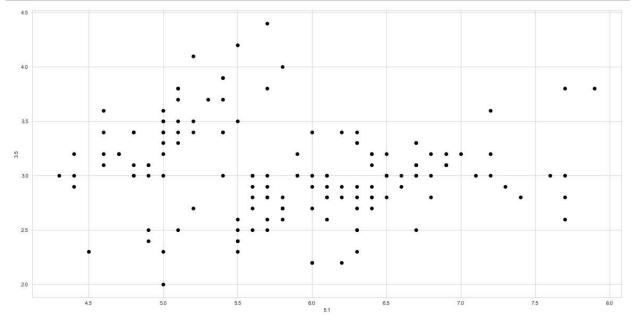


```
In [10]: import random as rd
```

```
In [11]: data = pd.read_csv('E:\iris.data')
    data.head()
```

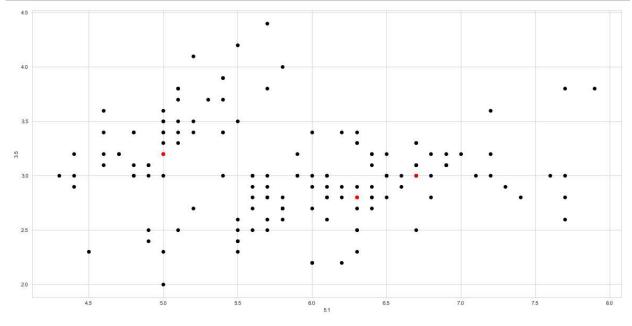
Out[11]:

	5.1	3.5	1.4	0.2	Iris-setosa	
0	4.9	3.0	1.4	0.2	Iris-setosa Iris-setosa	
1	4.7	3.2	1.3	0.2		
2	4.6	3.1	1.5	0.2	Iris-setosa	
3	5.0	3.6	1.4	0.2	Iris-setosa	
4	5.4	3.9	1.7	0.4	Iris-setosa	



```
In [13]: #number of clusters
K=3

# Select random observation as centroids
Centroids = (X.sample(n=K))
plt.scatter(X["5.1"],X["3.5"],c='black')
plt.scatter(Centroids["5.1"],Centroids["3.5"],c='red')
plt.xlabel('5.1')
plt.ylabel('3.5')
plt.show()
```



```
In [14]: # Step 3 - Assign all the points to the closest cluster centroid
         # Step 4 - Recompute centroids of newly formed clusters
         # Step 5 - Repeat step 3 and 4
         diff = 1
         j=0
         while(diff!=0):
             XD=X
             i=1
             for index1,row c in Centroids.iterrows():
                  for index2,row_d in XD.iterrows():
                      d1=(row_c["5.1"]=row_d["3.5"])**2
                      d2=(row_c["3.5"]=row_d["3.5"])**2
                      d=np.sqrt(d1+d2)
                      ED.append(d)
                 X[i]=ED
                  i=i+1
             C=[]
             for index,row in X.iterrows():
                 min dist=row[1]
                 pos=1
                  for i in range(K):
                      if row[i+1] < min dist:</pre>
                          min dist = row[i+1]
                          pos=i+1
                  C.append(pos)
             X["Cluster"]=C
             Centroids_new = X.groupby(["Cluster"]).mean()[["3.5","5.1"]]
             if j == 0:
                  diff=1
                  j=j+1
             else:
                  diff = (Centroids new['3.5'] - Centroids['3.5']).sum() + (Centroids new[
                  print(diff.sum())
             Centroids = X.groupby(["Cluster"]).mean()[["3.5","5.1"]]
         0.0
         <ipython-input-14-44df8a06da73>:18: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
```

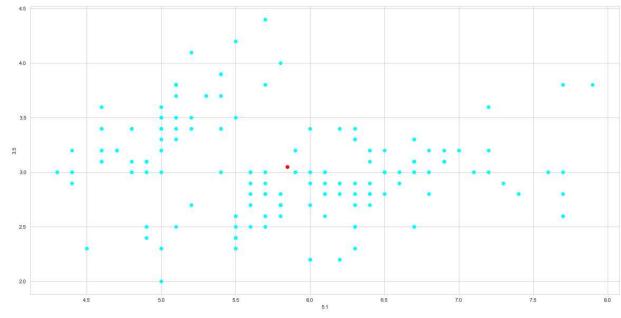
```
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

X[i]=ED
```

```
In [15]: color=['blue','green','cyan']
    for k in range(K):
        data=X[X["Cluster"]==k+1]
        plt.scatter(data["5.1"],data["3.5"],c=color[k])
    plt.scatter(Centroids["5.1"],Centroids["3.5"],c='red')
    plt.xlabel('5.1')
    plt.ylabel('3.5')
    plt.show()
```



```
In [16]: import nltk
from nltk.corpus import stopwords

from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfTransformer

import pandas as pd
import string
import seaborn as sns
```

```
In [17]: df = pd.read_csv("E:\smsspamcollection\SMSSpamCollection", names=["label","messag
df.head(2)
```

Out[17]:

label	message

• ham\tGo until jurong point crazy.. Available only in bugis n great world...

1 ham\tOk lar... Joking wif u oni...

NaN

In [18]: df.info() df.describe()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 5574 entries, 0 to 5573 Data columns (total 2 columns): Column Non-Null Count Dtype ----------0 label 5574 non-null object object 1 message 1318 non-null dtypes: object(2) memory usage: 87.2+ KB

Out[18]:

	label	message
count	5574	1318
unique	4969	1153
top	ham\tSorry	I'll call later
frea	52	30

In [19]: df.groupby('label').describe()

Out[19]:

message

	count	unique	top	freq
label				
ham\t	4	4	im On the snowboarding trip. I was wonderi	1
ham\t &It#> in mca. But not conform.	0	0	NaN	NaN
ham\t &It#> mins but i had to stop somewhere first.	0	0	NaN	NaN
ham\t &ItDECIMAL> m but its not a common car here so its better to buy from china or asia. Or if i find it less expensive. I.II holla	0	0	NaN	NaN
ham\t and picking them up from various points	0	0	NaN	NaN
spam\ttodays vodafone numbers ending with 0089(my last four digits) are selected to received a £350 award. If your number matches please call 09063442151 to claim your £350 award	0	0	NaN	NaN
spam\tu r a winner U ave been specially selected 2 receive £1000 cash or a 4* holiday (flights inc) speak to a live operator 2 claim 0871277810710p/min (18)	0	0	NaN	NaN
spam\tu r subscribed 2 TEXTCOMP 250 wkly comp. 1st wk?s free question follows	1	1	subsequent wks charged@150p/msg.2 unsubscribe	1
spam\twamma get laid?want real doggin locations sent direct to your mobile? join the UKs largest dogging network. txt dogs to 69696 now!nyt. ec2a. 3lp £1.50/msg.	0	0	NaN	NaN
spam\twe tried to contact you re your response to our offer of a new nokia fone and camcorder hit reply or call 08000930705 for delivery	0	0	NaN	NaN

4969 rows × 4 columns

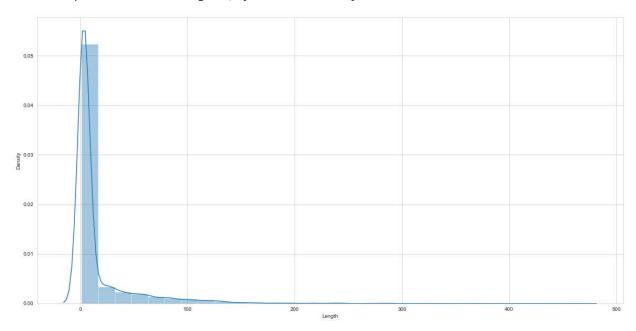
```
In [20]: df['message'] = df['message'].apply(str)
In [21]: df["Length"] = df["message"].apply(len)
```

In [22]: sns.distplot(df["Length"], bins=30)

C:\Users\Priya\anaconda3\lib\site-packages\seaborn\distributions.py:2557: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

warnings.warn(msg, FutureWarning)

Out[22]: <AxesSubplot:xlabel='Length', ylabel='Density'>



```
In [23]: df["Length"].max()
```

Out[23]: 465

```
In [24]: df[df["Length"]==465]["message"].iloc[0]
```

Out[24]: "hope you are having a nice day. I wanted to bring it to your notice that I have been late in paying rent for the past few months and have had to pay a \$ & 1 t; #> charge. I felt it would be inconsiderate of me to nag about something you give at great cost to yourself and that's why i didnt speak up. I however a min a recession and wont be able to pay the charge this month hence my askin well ahead of month's end. Can you please help. Thank you for everything."

```
In [25]: df[df["Length"] == df["Length"].min()]["message"].iloc[0]
```

Out[25]:

```
In [26]: df.head(1)
```

Out[26]:

label message Length

0 ham\tGo until jurong point crazy.. Available only in bugis n great world... 89

```
In [28]:
         class PreProcessText(object):
             def __init__(self):
                  pass
             def __remove_punctuation(self, text):
                  Takes a String
                  return : Return a String
                  .....
                  message = []
                  for x in text:
                      if x in string.punctuation:
                          pass
                      else:
                          message.append(x)
                  message = ''.join(message)
                  return message
             def __remove_stopwords(self, text):
                  Takes a String
                  return List
                  0.00
                  words= []
                  for x in text.split():
                      if x.lower() in stopwords.words('english'):
                          pass
                      else:
                          words.append(x)
                  return words
             def token_words(self,text=''):
                  Takes String
                  Return Token also called list of words that is used to
                  Train the Model
                  message = self.__remove_punctuation(text)
                  words = self.__remove_stopwords(message)
                  return words
```

```
In [47]:
         import nltk
          nltk.download()
          showing info https://raw.githubusercontent.com/nltk/nltk data/gh-pages/index.xm
          1 (https://raw.githubusercontent.com/nltk/nltk data/gh-pages/index.xml)
Out[47]: True
In [45]: mess = 'Sample message! Notice: it has punctuation.'
          obj = PreProcessText()
          words = obj.token_words(mess)
          print(words)
          ['Sample', 'message', 'Notice', 'punctuation']
In [50]:
         bow_transformer = CountVectorizer(analyzer=obj.token_words).fit(df["message"])
In [51]: messages bow = bow transformer.transform(df["message"])
In [52]:
         print("Shape of sparese matrix {}".format(messages_bow.shape))
          Shape of sparese matrix (5574, 2961)
In [53]: | tfidf transformer = TfidfTransformer().fit(messages bow)
In [54]: | messages_tfidf = tfidf_transformer.transform(messages_bow)
In [55]:
         from sklearn.naive bayes import MultinomialNB
In [56]: | model = MultinomialNB().fit(messages tfidf,df["label"])
         all predictions = model.predict(messages tfidf)
In [57]:
          pred = pd.DataFrame(data=all predictions)
          pred.head(6)
Out[57]:
                                                0
          0
                                         ham\tSorry
             ham\tl cant pick the phone right now. Pls send...
             ham\tl cant pick the phone right now. Pls send...
             ham\tl cant pick the phone right now. Pls send...
          3
           4
                                        ham\tSorry
           5
                                         ham\tSorry
```

```
In [58]: df["label"].head(6)
Out[58]: 0
                                     ham\tGo until jurong point
                             ham\t0k lar... Joking wif u oni...
              spam\tFree entry in 2 a wkly comp to win FA Cu...
         2
         3
              ham\tU dun say so early hor... U c already the...
                          ham\tNah I don't think he goes to usf
              spam\tFreeMsg Hey there darling it's been 3 we...
         Name: label, dtype: object
In [60]: from sklearn.preprocessing import MinMaxScaler
In [61]: from sklearn.preprocessing import StandardScaler
In [64]: from sklearn.decomposition import PCA
In [65]:
         pca=PCA(n_components=2)
In [75]: X = df['label']
         y = df['message']
In [76]: from sklearn.model selection import train test split
         X train, X test, y train, y test = train test split(X, y, test size = 0.2, random
In [80]: df['message'] = df['message'].astype(object)
 In [ ]:
```