1

Code:

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn import datasets, linear model, metrics
data boston= datasets.load boston(return X y=False)
X = data boston.data
y = data boston.target
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 state=1)
```

```
reg = linear model.LinearRegression()
reg.fit(X train, y train)
print('Coefficients are: \n', reg.coef )
print('Variance score is: {}'.format(reg.score(X test,
y test)))
from sklearn.linear model import LinearRegression
y pred = lm.predict(X test)
plt.scatter(y test,y pred )
```

```
plt.xlabel("Prices: ")
plt.ylabel("Predicted prices:")
plt.title("Prices vs Predicted prices")
plt.show() #Ideally, the scatter plot should create a
plt.style.use('fivethirtyeight')
plt.scatter(reg.predict(X train), reg.predict(X train) -
y train,
            color="red", s=10, label='Train data')
plt.scatter(reg.predict(X test), reg.predict(X test) -
y test,
            color="blue", s=10, label='Test data')
```

```
## plotting line for zero residual error
plt.hlines(y=0, xmin=0, xmax=50, linewidth=2)

## plotting legend
plt.legend(loc='upper right')

## plot title
plt.title("Residual errors")

## function to show plot
plt.show()
```

2.

Code:

```
import pandas
import matplotlib.pyplot as pl
```

```
from sklearn.cluster import KMeans
from sklearn.decomposition import PCA
variables=pandas.read csv('Customers.csv')
Y=variables[['Annual Income']]
X=variables[['Spending Score']]
Nc=range(1,20)
kmeans=[KMeans(n clusters=i) for i in Nc]
score=[kmeans[i].fit(Y).score(Y)for i in
range(len(kmeans))]
pl.plot(Nc, score)
pl.xlabel('Number of Clusters')
pl.ylabel('Score')
pl.show()
pca=PCA(n components=1).fit(Y)
pca d=pca.transform(Y)
pca c=pca.transform(X)
kmeans=KMeans(n clusters=5)
```

```
kmeansoutput=kmeans.fit(Y)
pl.figure('5 Cluster K-Means')
pl.scatter(pca c[:, 0], pca d[:, 0],
c=kmeansoutput.labels )
pl.xlabel('Annual Income')
pl.ylabel('Spending Score')
pl.title('5 Cluster K-Means')
pl.show()
```

Code: Using Linear Kernel

```
import sklearn
from sklearn import svm
from sklearn import datasets, metrics
from sklearn.cross validation import train test split
import matplotlib.pyplot as plt
from matplotlib import style
style.use("ggplot")
breastcancer data=datasets.load breast cancer()
x=breastcancer data.data
y=breastcancer data.target
clf=svm.SVC(kernel='linear',C=1.0)
```

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_si
ze=0.2)
model= clf.fit(x_train,y_train)
print(model.score(x_test,y_test))
```

Code: Using rbf Kernel

```
import sklearn
from sklearn import svm
from sklearn import datasets, metrics
from sklearn.cross_validation import train_test_split
import matplotlib.pyplot as plt
from matplotlib import style
style.use("ggplot")

#Loading the dataset
breastcancer_data=datasets.load_breast_cancer()
```

```
#getting the data and response of the dataset
x=breastcancer_data.data
y=breastcancer_data.target

clf=svm.SVC(kernel='rbf',C=1.0)

x_train,x_test,y_train,y_test=train_test_split(x,y,test_si
ze=0.2)
model= clf.fit(x_train,y_train)

print(model.score(x_test,y_test))
```

4.

```
import nltk
from nltk.corpus import wordnet as wn
import re, collections
from nltk.book import FreqDist
```

```
file=open("got.txt")
t=file.read()
from nltk.stem import WordNetLemmatizer
lemmetizer=WordNetLemmatizer()
print(lemmetizer.lemmatize(t))
from nltk.tokenize import word tokenize
from nltk.tag import pos tag
s=nltk.pos tag(nltk.word tokenize(t))
print(s)
file without verbs = [word for word, tag in s if tag !=
'VBG' and tag != 'VBZ' and tag!='VBN']
z=' '.join(file without verbs) # z is the file
print(z)
s1=nltk.pos tag(nltk.word tokenize(z))
```

```
print(s1)
fdist=FreqDist(z)
print(fdist)
q=fdist.most common(5)
print(q)
def tokens (text):
    11 11 11
    return re.findall('[a-z]+', text.lower())
WORD COUNTS = collections.Counter(tokens(z))
print (WORD COUNTS)
print (WORD COUNTS.most common(5))
```

```
##go through the original file
file=open("got.txt")
t=file.read()
```