

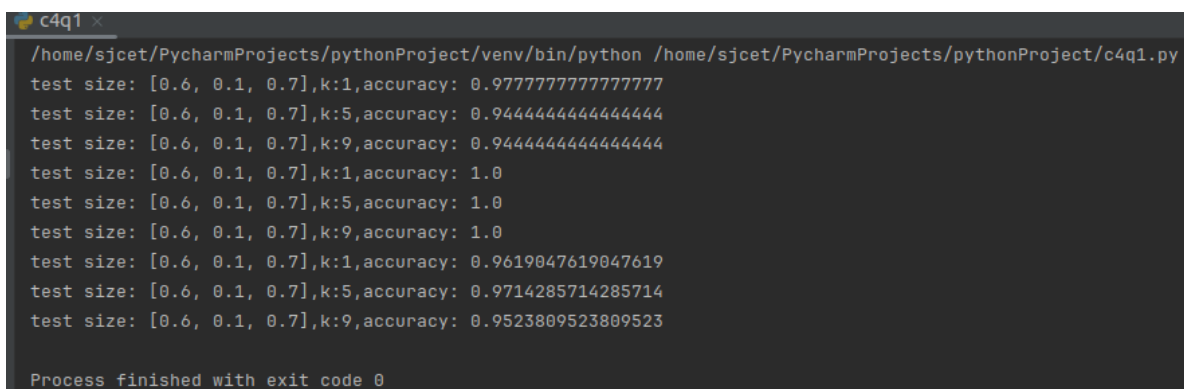
1. Using the iris data set implement the KNN algorithm. Take different values for Test and training data set .Also use different values for k. Also find the accuracy level.

Code

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
iris_data = pd.read_csv("iris.csv")
X = iris_data.iloc[:, :-1].values
y = iris_data.iloc[:, -1].values

test_sizes = [0.6, 0.1, 0.7]
k_values = [1,5,9]
for test_size in test_sizes:
    for k in k_values:
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=test_size,
random_state=42)
        knn = KNeighborsClassifier(n_neighbors=k)
        knn.fit(X_train, y_train)
        y_pred = knn.predict(X_test)
        accuracy = accuracy_score(y_test, y_pred)
        print(f'test size: {test_sizes},k:{k},accuracy: {accuracy}')
```

Output



```
c4q1 x
/home/sjctet/PycharmProjects/pythonProject/venv/bin/python /home/sjctet/PycharmProjects/pythonProject/c4q1.py
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.9777777777777777
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.9444444444444444
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.9444444444444444
test size: [0.6, 0.1, 0.7],k:1,accuracy: 1.0
test size: [0.6, 0.1, 0.7],k:5,accuracy: 1.0
test size: [0.6, 0.1, 0.7],k:9,accuracy: 1.0
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.9619047619047619
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.9714285714285714
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.9523809523809523

Process finished with exit code 0
```

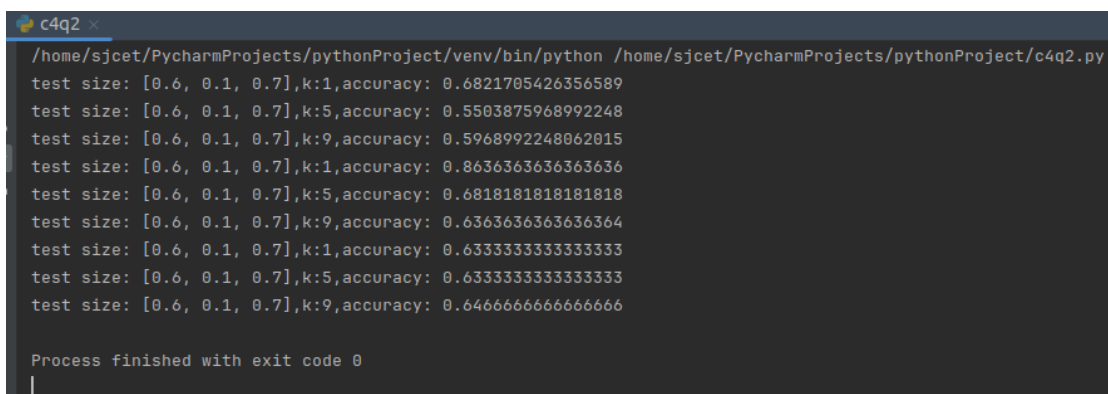
2. Download another data set suitable for the KNN and implement the KNN algorithm. Take different values for Test and training data set .Also use different values for k.

Code

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
data = pd.read_csv("glass.csv")
X = data.iloc[:, :-1].values
y = data.iloc[:, -1].values

test_sizes = [0.6, 0.1, 0.7]
k_values = [1,5,9]
for test_size in test_sizes:
    for k in k_values:
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=test_size,
random_state=42)
        knn = KNeighborsClassifier(n_neighbors=k)
        knn.fit(X_train, y_train)
        y_pred = knn.predict(X_test)
        accuracy = accuracy_score(y_test, y_pred)
        print(f'test size: {test_sizes},k:{k},accuracy: {accuracy}')
```

Output



```
c4q2 x
/home/sjcet/PycharmProjects/pythonProject/venv/bin/python /home/sjcet/PycharmProjects/pythonProject/c4q2.py
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.6821705426356589
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.5503875968992248
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.5968992248062015
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.8636363636363636
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.6818181818181818
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.6363636363636364
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.6333333333333333
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.6333333333333333
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.6466666666666666

Process finished with exit code 0
|
```

3. Using iris data set, implement naive bayes classification for different naive Bayes classification algorithms.(i) gaussian (ii) bernoulli etc)

- Find out the accuracy level w.r.t to each algorithm
- Display the no:of mislabeled classification from test data set
- List out the class labels of the mismatching records

Code

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
dataset = pd.read_csv('iris.csv')
X = dataset.iloc[:, :4].values
y = dataset['variety'].values
dataset.head(5)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.5)
from sklearn.naive_bayes import GaussianNB

classifier = GaussianNB ()
classifier.fit(X_train, y_train)
y_pred = classifier.predict(X_test)
print(y_pred)

from sklearn.metrics import confusion_matrix
cm =confusion_matrix(y_test, y_pred)
print(cm)

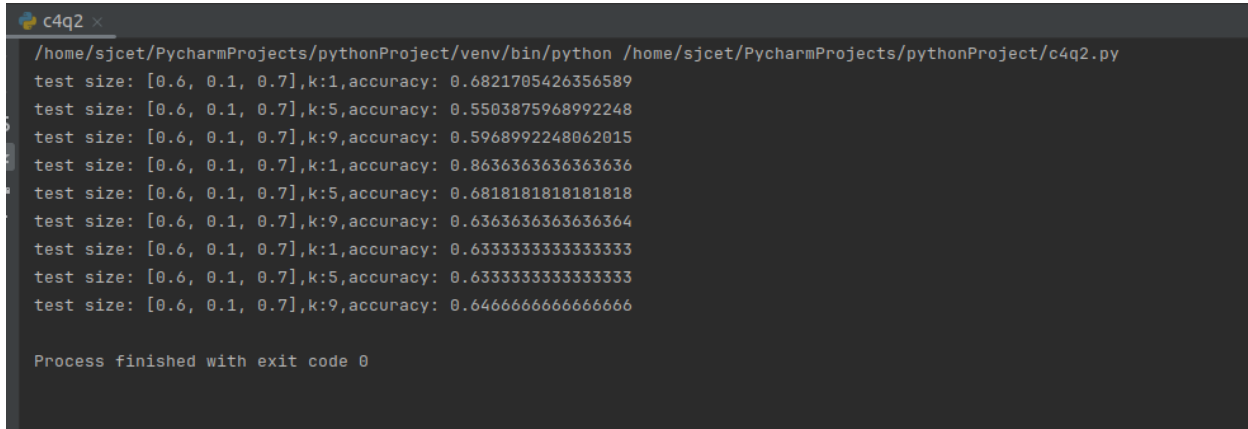
from sklearn.metrics import accuracy_score
print ("Accuracy: ", accuracy_score (y_test, y_pred))
df = pd.DataFrame({'Real Values':y_test, 'Predicted Values':y_pred})
print(df)
print()
from sklearn.naive_bayes import BernoulliNB
classif = BernoulliNB ()
classif.fit(X_train, y_train)
y_pred = classif.predict(X_test)
print(y_pred)

from sklearn.metrics import confusion_matrix
cmx =confusion_matrix(y_test, y_pred)
```

```
print(cmx)

from sklearn.metrics import accuracy_score
print ("Accuracy: ", accuracy_score (y_test, y_pred))
fd = pd.DataFrame({'Real Values':y_test, 'Predicted Values':y_pred})
print(fd)
```

Output



```
c4q2 x
/home/sjcet/PycharmProjects/pythonProject/venv/bin/python /home/sjcet/PycharmProjects/pythonProject/c4q2.py
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.6821705426356589
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.5503875968992248
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.5968992248062015
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.8636363636363636
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.6818181818181818
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.6363636363636364
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.6333333333333333
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.6333333333333333
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.6466666666666666

Process finished with exit code 0
```

4. Use car details CSV file and implement decision tree algorithm

- Find out the accuracy level.
- Display the no: of mislabelled classification from test data set
- List out the class labels of the mismatching records

Code

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score

data = pd.read_csv('car.csv')
print(data.head())
col_names = ['buying', 'maint', 'doors', 'persons', 'lug_boot', 'safety',
'class']
data.columns = col_names
print(col_names)

data['class'],class_names = pd.factorize(data['class'])
data['buying'],_ = pd.factorize(data['buying'])
data['maint'],_ = pd.factorize(data['maint'])
```

```

data['doors'],_ = pd.factorize(data['doors'])
data['persons'],_ = pd.factorize(data['persons'])
data['lug_boot'],_ = pd.factorize(data['lug_boot'])
data['safety'],_ = pd.factorize(data['safety'])

print(data.head())
X = data.iloc[:, :-1]
y = data.iloc[:, -1]

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)

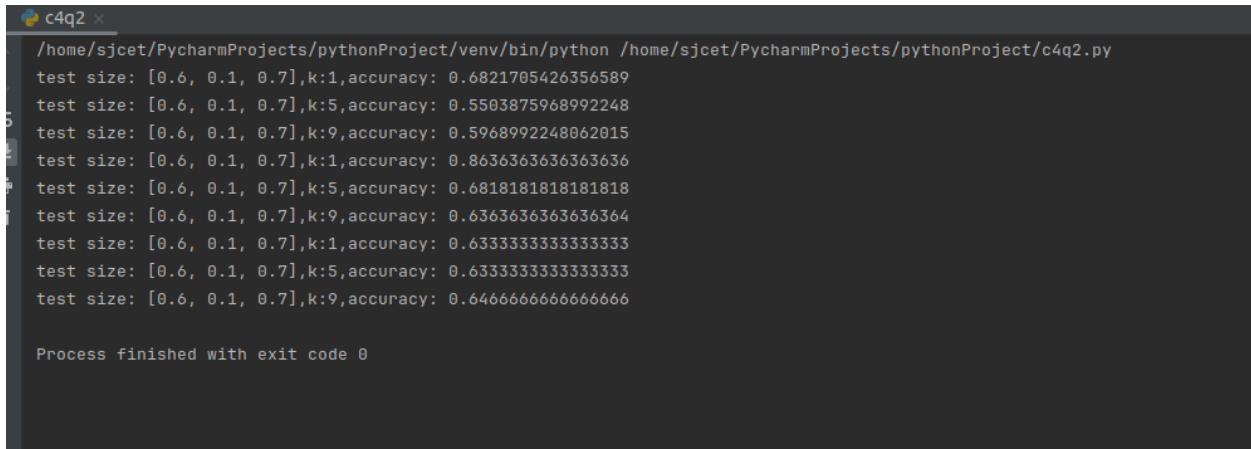
tree1 = DecisionTreeClassifier()

tree1.fit(X_train, y_train)
y_pred = tree1.predict(X_test)
# how did our model perform?
count_misclassified = (y_test != y_pred).sum()
print('Misclassified samples count:', count_misclassified)

accuracy = accuracy_score (y_test, y_pred)
print("Accuracy:", accuracy)

```

Output



```

c4q2 x
/home/sjcet/PycharmProjects/pythonProject/venv/bin/python /home/sjcet/PycharmProjects/pythonProject/c4q2.py
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.6821705426356589
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.5503875968992248
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.5968992248062015
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.8636363636363636
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.6818181818181818
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.6363636363636364
test size: [0.6, 0.1, 0.7],k:1,accuracy: 0.6333333333333333
test size: [0.6, 0.1, 0.7],k:5,accuracy: 0.6333333333333333
test size: [0.6, 0.1, 0.7],k:9,accuracy: 0.6466666666666666

Process finished with exit code 0

```

5. Implement Simple and multiple linear regression for the data sets 'student_score.csv' and 'company_data .csv' respectively

Code

```

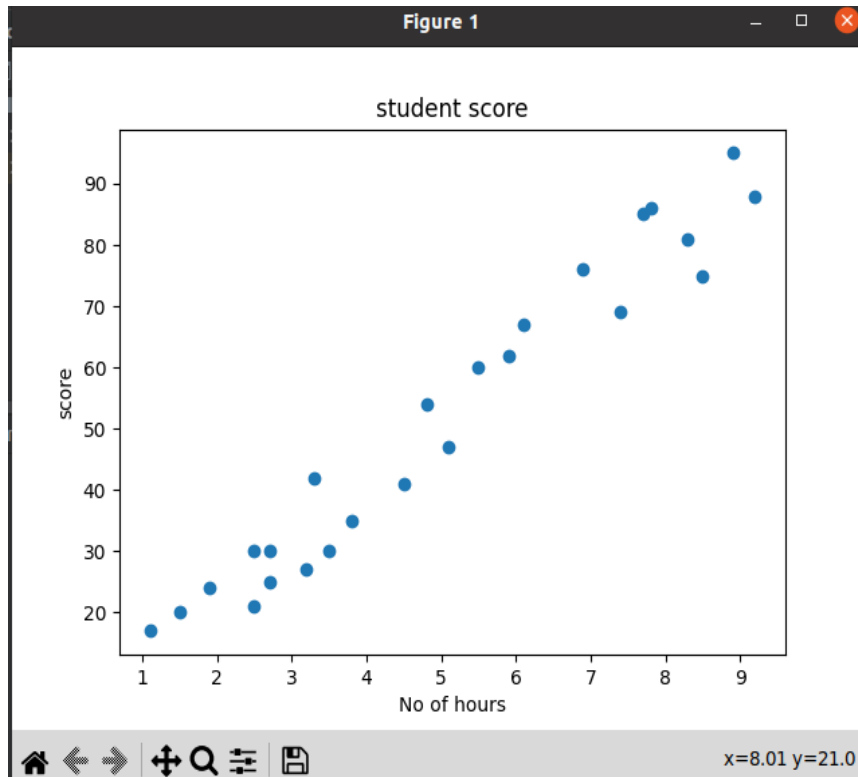
import pandas as pd
import numpy as np
from sklearn.metrics import accuracy_score

```

```

student = pd.read_csv('student_scores.csv')
print(student.head())
student.describe()
student.info()
import matplotlib.pyplot as plt
Xax=student.iloc[:, 0]
yax=student.iloc[:, 1]
plt.scatter(Xax,yax)
plt.xlabel("No of hours")
plt.ylabel("score")
plt.title("student score")
plt.show()
X = student.iloc[:, :-1]
y = student.iloc[:, 1]
print('X values:')
print(X)
print(' y values :')
print(y)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.3)
print(X_train)
from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(X_train, y_train)
print('INTERCEPT=', regressor.intercept_)
print('CO_EFFICIENT=', regressor.coef_)
y_pred = regressor.predict(X_test)
for(i,j) in zip(y_test, y_pred):
    if(i!=j):
        print("Actual value :", i, "predicted value :",j)
        print("number of mislabelled points from test data set :", (y_test != y_pred).sum())
        from sklearn import metrics
        print("mean absolute error :", metrics.mean_absolute_error(y_test, y_pred))
        print("mean squared error :", metrics.mean_squared_error(y_test, y_pred))
        print("root mean squared error :", np.sqrt(metrics.mean_squared_error(y_test,
y_pred)))

```



Multiple Regression

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn import metrics
```

```
advertising = pd.read_csv('Company_data.csv')
advertising.head()
advertising.describe()
advertising.info()
```

```
X = advertising.iloc[:, :-1]
y = advertising.iloc[:, -1]
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)
reg = LinearRegression()
reg.fit(X_train, y_train)
print("Name: Aswathy Chandran")
```

```

print("Reg No: SJC22MCA-2016")
print("Batch: 22-24")
print()
print('Intercept is:', reg.intercept_)
print('Co Efficients are:', reg.coef_)
y_pred = reg.predict(X_test)

for(i,j) in zip(y_test, y_pred):
    if(i!=j):
        print('Actual value: ', i, 'Predicted value: ', j)
print('No: of mislabeled points: ', (y_test != y_pred).sum())

print("Mean Absolute error :", metrics.mean_absolute_error(y_test,y_pred))
print("Mean Squared error :", metrics.mean_squared_error(y_test,y_pred))
print("Root Mean Squared error :",
np.sqrt(metrics.mean_squared_error(y_test,y_pred)))

```

Output

```

c4q5.1 x
/home/sjcet/PycharmProjects/pythonProject/venv/bin/python /home/sjcet/PycharmProjects/pythonProj
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0    TV          200 non-null    float64
1    Radio       200 non-null    float64
2    Newspaper   200 non-null    float64
3    Sales       200 non-null    float64
dtypes: float64(4)
memory usage: 6.4 KB
Intercept is: 4.595239802802677
Co Efficients are: [ 0.05531693  0.10970665 -0.00310274]
Actual value: 18.0 Predicted value: 17.653802883971252
Actual value: 16.4 Predicted value: 16.027847183360677
Actual value: 17.6 Predicted value: 14.798182300894544
Actual value: 20.9 Predicted value: 18.416426729943744
Actual value: 6.6 Predicted value: 9.13730221106121
Actual value: 16.0 Predicted value: 14.976821474646476
Actual value: 19.0 Predicted value: 19.35204885824649
Actual value: 10.4 Predicted value: 10.899984833058513
Actual value: 17.1 Predicted value: 15.996460812848703
Actual value: 21.4 Predicted value: 23.768229583143025
Actual value: 17.4 Predicted value: 21.480619438395
Actual value: 6.9 Predicted value: 6.227759490257272
Actual value: 17.8 Predicted value: 21.099296282499623
Actual value: 16.1 Predicted value: 21.53282418251726
Actual value: 5.5 Predicted value: 7.953357076434626
Actual value: 17.5 Predicted value: 18.21406937276536
Actual value: 18.4 Predicted value: 17.8100933108416

```


c4q5.1 x

```
Actual value: 18.3 Predicted value: 18.845788861100214
Actual value: 12.3 Predicted value: 11.419690810882065
Actual value: 17.9 Predicted value: 16.879936929127737
Actual value: 5.3 Predicted value: 5.284344270569222
Actual value: 10.8 Predicted value: 11.271206151993773
Actual value: 5.9 Predicted value: 5.898441894760172
Actual value: 15.2 Predicted value: 15.162130478668427
Actual value: 7.0 Predicted value: 7.859454626082876
Actual value: 12.6 Predicted value: 8.812901470170448
Actual value: 12.6 Predicted value: 12.425760127908593
Actual value: 14.0 Predicted value: 12.533011075010783
Actual value: 10.6 Predicted value: 10.633054015197729
Actual value: 25.4 Predicted value: 25.10228663027957
Actual value: 20.7 Predicted value: 21.51822690433077
Actual value: 1.6 Predicted value: 8.951351238891473
Actual value: 19.7 Predicted value: 20.60377784140941
Actual value: 16.8 Predicted value: 20.080225674786874
Actual value: 14.6 Predicted value: 15.306150770427468
Actual value: 14.7 Predicted value: 14.00512827346257
Actual value: 10.5 Predicted value: 10.12933010756786
Actual value: 9.2 Predicted value: 9.558016021599364
Actual value: 17.2 Predicted value: 17.189390665472853
Actual value: 23.2 Predicted value: 22.361197475208147
Actual value: 13.2 Predicted value: 10.043523628059852
No: of mislabeled points: 60
Mean Absolute error : 1.2451499063516045
Mean Squared error : 3.4927724151539223
Root Mean Squared error : 1.868896041826276
```

Process finished with exit code 0