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 \text{ 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

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

- 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

- 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

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 misslabelled 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))
```

