```
# import numpy as np
# import pandas as pd
# import matplotlib.pyplot as plt
# import sklearn.neural network
# xs=np.array([
      0,0,
#
      0,1,
      1,0,
      1,1
# ]).reshape(4,2)
# ys=np.array([0,1,1,0])#([0, 1, 1, 0])
# ys.reshape(4,)
model=sklearn.neural network.MLPClassifier(activation="relu", max iter=
10000, learning rate="constant", hidden layer sizes=(4)) #hidden layer
akta and neuron 3 ta aro ->hidden layer sizes=(2, 2, 4))#3 hidden
layer and first layer and second has 2 neurons and third has 4 neurons
# model.fit(xs,ys)
# import matplotlib.pyplot as plt
# plt.scatter(xs[:,0], xs[:,1], c=ys, cmap=plt.cm.brg)
# plt.show()
# print("score", model.score(xs, vs))
# print("prediction", model.predict(xs))
# print ("expected", np.array([0,1,1,0]))
# XS
# import numpy as np
# import pandas as pd
# import matplotlib.pyplot as plt
# from sklearn.datasets import load iris
# from sklearn.model selection import train test split
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from ucimlrepo import fetch ucirepo
# fetch dataset
heart disease = fetch ucirepo(id=45)
# data (as pandas dataframes)
X = heart disease.data.features
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y = heart disease.data.targets
# metadata
# print(heart disease.metadata)
# # variable information
# print(heart disease.variables)
X.shape
y.shape
X.shape
y.shape
def sigmoid(x):
    return 1 / (1 + np.exp(-x))
def mean_squared_error(y_pred, y_true):
    return ((y_pred - y_true) ** 2).sum() / (2 * y_pred.size)
def accuracy(y pred, y true):
    if isinstance(y true, pd.DataFrame):
        y_true = y_true.values
    acc = y pred.argmax(axis=1) == y true.argmax(axis=1)
    return acc.mean()
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test size=20, random state=4)
\# y_train = np.reshape(y_train, (y_train.shape[0], 1))
# y test = np.reshape(y test, (y test.shape[0], 1))
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=20, random_state=4)
learning rate = 0.1
iterations = 10000
N = y_{train.size}
input size = 13
hidden size = 2
output size = 1
results = pd.DataFrame(columns=['mse', 'accuracy'])
np.random.seed(10)
W1 = np.random.normal(scale=0.5, size=(input size, hidden size))
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W2 = np.random.normal(scale=0.5, size=(hidden size, output size))
results list = []
for itr in range(iterations):
    Z1 = np.dot(X_train, W1)
    A1 = sigmoid(Z1)
    Z2 = np.dot(A1, W2)
    A2 = sigmoid(Z2)
    mse = mean_squared_error(A2, y_train)
    # acc = accuracy(A2.values, y_train.values)
    acc = accuracy(A2, y train)
    results_list.append({"mse": mse, "accuracy": acc})
    E1 = A2 - y_train
    dW1 = E1 * \overline{A}2 * (1 - A2)
    E2 = np.dot(dW1, W2.T)
    dW2 = E2 * A1 * (1 - A1)
    w2 \text{ update} = np.dot(A1.T, dW1) / N
    w1 update = np.dot(X train.T, dW2) / N
    W2 = W2 - learning rate * w2 update
    W1 = W1 - learning rate * w1 update
Z1 = np.dot(X test, W1)
A1 = sigmoid(Z1)
Z2 = np.dot(A1, W2)
A2 = sigmoid(Z2)
acc = accuracy(A2, y_test)
print("Accuracy is {}".format(acc))
ModuleNotFoundError
                                           Traceback (most recent call
last)
<ipython-input-1-b0a2d226f21b> in <cell line: 4>()
      2 import pandas as pd
```

```
3 import matplotlib.pvplot as plt
----> 4 from ucimlrepo import fetch ucirepo
      6 # fetch dataset
ModuleNotFoundError: No module named 'ucimlrepo'
NOTE: If your import is failing due to a missing package, you can
manually install dependencies using either !pip or !apt.
To view examples of installing some common dependencies, click the
"Open Examples" button below.
\# x = pd.DataFrame(iris.data, columns=['SL', 'SW', 'PL', 'PW'])
# y = pd.DataFrame(iris.target , columns=['Target'])
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
# %matplotlib inline
df=pd.read csv('/content/drive/MyDrive/versity/Thesis/data/diabetes.cs
v')
df.head(10)
df.isnull().sum()
#display dataset randomly
df.sample(10)
#shape of the dataset
df.shape
df.columns
"""# Split the data frame into X and y"""
target name='Outcome'
y=df['Outcome']
X=df.drop(target name,axis=1)
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X.head()
"""# Feature scaling tecniques
#standard scaler
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
scaler.fit(X)
SSX=scaler.transform(X)
#scaler = StandardScaler()
#scaler.fit(X train)
#X train = scaler.transform(X train)
#X test = scaler.transform(X test)
"""# Train Test split
from sklearn.model selection import train test split
X train, X test, y train, y test=train test split(SSX, y, test size=0.25, ra
ndom state=42)
X train.shape, y train.shape
X_test.shape,y_test.shape
#Linear Regression
from sklearn.linear model import LinearRegression
Linearregression=LinearRegression()
Linearregression.fit(X train,y train)
linear predict=Linearregression.predict(X test)
df=pd.DataFrame({'Actual':y test, 'Predicted':linear predict}).round(0)
df
#check accuracy mean
from sklearn.metrics import mean absolute error
mae for linear=mean absolute error(y test,linear predict)
print(f"mean absolute error {mae_for_linear}")
# from sklearn.metrics import accuracy score
# accuracy = accuracy_score(y_test , linear_predict)
# Logistic Regression
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from sklearn.linear model import LogisticRegression
lr=LogisticRegression()
lr.fit(X train,y train)
lr predict=lr.predict(X test)
df=pd.DataFrame({'Actual':y_test,'Predicted':lr predict}).round(0)
from sklearn.metrics import mean absolute error
mae_for_log=mean_absolute_error(y_test,lr_predict)
print(f"mean absolute error {mae for log}")
# KNN Classification
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
range k = range(1, 15)
scores = \{\}
scores list = []
for k in range k:
  classifier=KNeighborsClassifier(n neighbors=k)
  classifier.fit(X train,y train)
  knn predict=classifier.predict(X test)
  scores[k]=metrics.accuracy_score(y test,knn predict)
  scores list.append(metrics.accuracy score(y test,knn predict))
result=metrics.confusion matrix(y test,knn predict)
print("Confusion Matrix:")
print(result)
result1 = metrics.classification report(y test , knn predict)
print("Classification Report:",)
"""**Now, we will be plotting the relationship between the values of K
and the corresponding testing
accuracy. It will be done using matplotlib library.**
plt.plot(range k,scores list)
plt.xlabel("Values of K")
plt.ylabel("Accuracy")
classifier = KNeighborsClassifier(n neighbors = 2)
classifier.fit(X train , y train)
knn predict=classifier.predict(X test)
df=pd.DataFrame({'Actual':y_test,'Predicted':knn predict}).round(0)
df
from sklearn.metrics import mean_absolute_error
knn predict mean=mean absolute error(y test,knn predict)
```

```
print(f"mean absolute error{knn predict mean}")
#naive bayes
from sklearn.naive bayes import GaussianNB
nb=GaussianNB()
nb.fit(X train,y train)
nb predict=nb.predict(X test)
df=pd.DataFrame({'Actual':y test,'Predicted':nb predict}).round(0)
df
from sklearn.metrics import mean_absolute error
nb predict mean=mean absolute error(y test,nb predict)
print(f"mean absolute error{nb predict mean}")
import pandas as pd
import numpy as np
from sklearn import datasets
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches
import sklearn.metrics as sm
%matplotlib inline
iris = datasets.load iris()
print(iris.target names)
x = pd.DataFrame(iris.data , columns=['SL','SW','PL','PW'])
y = pd.DataFrame(iris.target , columns=['Target'])
iris k mean model = KMeans(n clusters =3) # we set K = 3
iris_k_mean_model.fit(x)
predictedY=iris k mean model.predict(x)
sm.accuracy score(predictedY , y['Target'])
sm.confusion matrix(predictedY , y['Target'])
```