BFS

```
import queue as Q
from RMP import dict gn
start='Arad'
goal='Bucharest'
result=''
def BFS(city,cityq,visitedq):
     global result
      if city == start:
            result=result+' '+city
      for eachcity in dict_gn[city].keys():
            if eachcity == goal:
                 result=result+' '+eachcity
                 return
            if eachcity not in cityq.queue and eachcity not in
visitedq.queue:
                 cityq.put(eachcity)
                 result=result+' '+eachcity
     visitedq.put(city)
     BFS(cityq.get(),cityq,visitedq)
def main():
     cityq=Q.Queue()
     visitedq=Q.Queue()
     BFS(start,cityq,visitedq)
     print('BFS Traversal from', start, 'to', goal, 'is:')
     print(result)
main()
```

IDFS

```
import queue as Q
from RMP import dict_gn

start='Arad'
goal='Bucharest'
result=''

def DLS(city, visitedstack, startlimit, endlimit):
    global result
    found=0
    result = result + city + ''
    visitedstack.append(city)
    if city == goal:
        return 1
    if startlimit==endlimit:
        return 0
    for eachcity in dict gn[city].keys():
```

```
if eachcity not in visitedstack:
     found=DLS(eachcity, visitedstack, startlimit+1, endlimit)
                if found:
                     return found
def IDDFS(city, visitedstack, endlist):
     global result
     for i in range (0, endlimit):
          print('Searching at limit: ',i)
          found=DLS(city, visitedstack, 0, i)
          if found:
                print('Found')
               break
          else:
                print('Not Found')
                print(result)
                print('----')
                result=''
                visitedstack=[]
def main():
     visitedstack=[]
     IDDFS(start, visitedstack, 9)
     print('IDDFS Traversal from', start, 'to', goal, 'is:')
     print(result)
main()
```

A* Search

```
from RMP import dict_gn
from RMP import dict_hm
import queue as Q

start = 'Bucharest'
goal = 'Bucharest'
result = ' '

def get_fn(citystr):
    cities=citystr.split(',')
    hm=gn=0
    for ctr in range(0,len(cities)-1):
        gn=gn+dict_gn[cities[ctr]][cities[ctr+1]]
    hm=dict_hm[cities[len(cities)-1]]
    return (hm+gn)

def expand(cityq):
```

```
global result
     tot, citystr, thiscity=cityq.get()
     if thiscity==goal:
          result=citystr+'::'+str(tot)
          return
     for cty in dict gn[thiscity]:
     cityq.put((get fn(citystr+","+cty),citystr+","+cty,cty))
     expand(cityq)
def main():
     cityq=Q.PriorityQueue()
     thiscity=start
     cityq.put((get_fn(start), start, thiscity))
     expand(cityq)
     print("The A* path with the total is: ")
     print(result)
main()
```

RBFS

```
from RMP import dict gn
from RMP import dict hm
import queue as Q
start = 'Arad'
goal = 'Sibiu'
result = ' '
def get fn(citystr):
     cities=citystr.split(',')
     hm=qn=0
     for ctr in range(0,len(cities)-1):
          gn=gn+dict gn[cities[ctr]][cities[ctr+1]]
     hm=dict hm[cities[len(cities)-1]]
     return (hm+gn)
def printout(cityq):
     for i in range(0,cityq.qsize()):
          print(cityq.queue[i])
def expand(cityq):
     global result
     tot, citystr, thiscity = cityq.get()
     print('---', tot, citystr, thiscity, '---')
     nexttot = 999
     if not cityq.empty():
```

```
nexttot, nextcitystr, nextthiscity=cityq.queue[0]
          if thiscity==goal and tot<nexttot:
               result = citystr + '::' + str(tot)
               return
          print('Expanded city----',thiscity)
          print('Second best f(n)-----', nexttot)
          tempq=Q.PriorityQueue()
          for cty in dict gn[thiscity]:
               tempq.put((get fn(citystr + "," + cty), citystr +
"," + cty,cty))
          for ctr in range (1,3):
               ctrtot,ctrcitystr,ctrthiscity = tempq.get()
               if ctrtot < nexttot:</pre>
                    cityq.put((ctrtot,ctrcitystr,ctrthiscity))
               else:
                    cityq.put((ctrtot,citystr,thiscity))
                    break
          printout(cityq)
          expand(cityq)
def main():
     cityq=Q.PriorityQueue()
     thiscity=start
     cityq.put((999,'NA','NA'))
     cityq.put((get fn(start), start, thiscity))
     expand(cityq)
     print('This RBFS path with the total is: ')
     print(result)
main()
```

DECISION TREE

```
import numpy as np
import pandas as pd
from sklearn import tree
from sklearn.preprocessing import LabelEncoder

# Loading the data
PlayTennis = pd.read_csv('PlayTennis.csv')

Le = LabelEncoder()

PlayTennis['outlook'] = Le.fit_transform(PlayTennis['outlook'])
PlayTennis['temp'] = Le.fit_transform(PlayTennis['temp'])
PlayTennis['humidity'] =
Le.fit_transform(PlayTennis['humidity'])
PlayTennis['windy'] = Le.fit_transform(PlayTennis['windy'])
```

```
PlayTennis['play'] = Le.fit_transform(PlayTennis['play'])
print(PlayTennis)
y = PlayTennis['play']
X = PlayTennis.drop(['play'], axis=1)

clf = tree.DecisionTreeClassifier(criterion='entropy')
clf = clf.fit(X, y)

tree.plot_tree(clf)

o = int(input("What is the outlook?: "))
t = int(input("What is the temperatur?: "))
h = int(input("What is the humidity?: "))
w = int(input("How windy it is?: "))

y_predict = clf.predict([[o,t,h,w]])
v = 'PLAY' if y_predict[0] == 1 else "not play"
print("As per situation one should",v)
```

FEED FORWARD BACKPROPAGATION NEURAL NETWORK

```
import numpy as np
X = np.array(([2, 9], [1, 5], [3, 6]), dtype=float)
y= np.array(([92], [86], [89]), dtype=float)
X = X/np.amax(X,axis=0)
y = y/np.amax(y)
def sigmoid (x):
    return ((1/1 + np.exp(-x)))
def derivatives_sigmoid(x):
     return x * (1 - x)
epoch=7000
#epoch = 2
lr=0.1
inputlayer neurons = 2
hiddenlayer neurons = 3
output neurons = 1
\#wh = 2 \times 3
wh=np.random.uniform(size=(inputlayer neurons,
hiddenlayer neurons))
\#bh = 1 \times 3
bh=np.random.uniform(size=(1, hiddenlayer neurons))
#wout = 3 \times 1
```

```
wout=np.random.uniform(size=(hiddenlayer neurons,
output neurons))
\#bout = 1 \times 1
bout=np.random.uniform(size=(1,output neurons))
for i in range (epoch) :
     #between input layer and hidden layer
     #dot product of X and wh
     hinp1=np.dot(X,wh)
     hinp=hinp1 + bh
     hlayer act = sigmoid(hinp)
     #between hidden layer and output
     #dot product of outputs of hidden layer and wout
     outinp1=np.dot(hlayer act, wout)
     outinp= outinp1+ bout
     output = sigmoid(outinp)
     #Backpropogation of error
     EO = y-output
     if i == 0:
          print('Error',EO)
     outgrad = derivatives sigmoid(output)
     d output = EO* outgrad
     EH = d output.dot(wout.T)
     hiddengrad = derivatives sigmoid(hlayer act)
     #weight adjustments
     d hiddenlayer = EH * hiddengrad
     wout += hlayer act.T.dot(d output) *lr
     bout += np.sum(d output, axis=0,keepdims=True) *lr
     wh += X.T.dot(d hiddenlayer)*lr
print('Error', EO)
print("Input: \n" + str(X))
print("Actual Output: \n" + str(y))
print("Prdeicted Output: \n" ,output)
```

SVM

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC, LinearSVC
from sklearn.metrics import confusion_matrix
from sklearn.metrics import*# classification_report
```

```
from sklearn.metrics import accuracy score
#import seaborn as sns
df = pd.read csv('diabetes.csv')
#df.head()
x = df.drop('Outcome', axis=1)
y = df['Outcome']
x train = x.iloc[:600]
x test = x.iloc[600:]
y train = y[:600]
y test = y[600:]
classifier = SVC(kernel="linear")
classifier.fit(x train, y train)
y_pred = classifier.predict(x test)
print('Confusion Martix:')
print(confusion matrix(y test, y pred))
print('Accuracy Score')
print(accuracy score(y test, y pred))
print('Classification Report:')
print(classification report(y test, y pred))
```

ADABOOST

```
import pandas as pd
import warnings
from sklearn.model_selection import train_test_split
from sklearn.ensemble import AdaBoostClassifier
from sklearn.metrics import confusion_matrix, accuracy_score,
classification_report

warnings.filterwarnings("ignore")

data = pd.read_csv("apples_and_oranges.csv")
print(data)

#test_size = 0.2 =>20% test, 80% training

training_set, test_set = train_test_split(data, test_size =0.2,
random_state = 1)
X_train = training_set.iloc[:,0:2].values
Y_train = training_set.iloc[:,2].values
```

```
X_test = test_set.iloc[:,0:2].values
Y_test = test_set.iloc[:,2].values

#base_estimator: it is a week learner used to train the model.
It uses DecisionTreeClassifier as default weak learner for
training purpose. You can also specify different machine
learning algorithms.
#N_estimators: Number of weak learners to train iteratively.
#learning_rate: It contributes to the weights of weak learners.
It uses 1 as a default value.
Adaboost = AdaBoostClassifier(n_estimators=100, learning_rate =
1, random_state = 1)
adaboost.fit(X_train,Y_train)
Y_pred =adaboost.predict(X_test)
test_set["Predictions"] = Y_pred
print(Y_pred)
```

NAÏVE BAYES'

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.naive bayes import MultinomialNB, GaussianNB
from sklearn.metrics import confusion matrix
from sklearn.metrics import classification report
from sklearn.metrics import accuracy score
from sklearn.preprocessing import LabelEncoder
df = pd.read csv('disease.csv')
le = LabelEncoder()
df['Sore Throat'] = le.fit transform(df['Sore Throat'])
df['fever'] = le.fit transform(df['fever'])
df['Swollen Glands'] = le.fit transform(df['Swollen Glands'])
df['congestion'] = le.fit transform(df['congestion'])
df['headache'] = le.fit transform(df['headache'])
df['diagnosis'] = le.fit transform(df['diagnosis'])
x = df.drop('diagnosis',axis=1)
y = df['diagnosis']
x train, x test, y train, y test =
train test split(x,y,test size=0.2)
classifier = MultinomialNB()
classifier.fit(x train, y train)
```

```
y_pred = classifier.predict(x_test)
print('Confusion Matrix:')
print(confusion_matrix(y_test, y_pred))
print('classification Report:')
print(classification_report(y_test, y_pred))
```

KNN

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import confusion matrix
from sklearn.metrics import classification report
plt.style.use('ggplot')
df = pd.read csv('diabetes.csv')
#df.head()
#df.shape
#df.dtypes
X = df.drop('Outcome',axis=1).values
y = df['Outcome'].values
x train, x test, y train, y test = train test split(X, y,
test size=0.4, random state=42, shuffle=True)
neighbors = np.arange(1, 9)
train accuracy = np.empty(len(neighbors))
test accuray = np.empty(len(neighbors))
for i, k in enumerate (neighbors):
    knn = KNeighborsClassifier(n neighbors=k) #Setup a KNN
classifier with k neighbours
    knn.fit(x train,y train)
    train accuracy[i] = knn.score(x train, y train)
    test accuray[i] = knn.score(x test, y test)
knn = KNeighborsClassifier(n neighbors=7)
knn.fit(x train, y train)
s = knn.score(x test, y test)
print(s)
y pred = knn.predict(x test)
confusion matrix(y test, y pred)
```

```
print(classification_report(y_test, y_pred))

plt.title('KNN Vari')
plt.plot(neighbors, test_accuray, label='Testing Accuracy')
plt.plot(neighbors, train_accuracy, label='Training Accuracy')
plt.legend()
plt.xlabel('Number of neighbours')
plt.ylabel('Accuracy')
plt.show()
```

-BY LAXMAN