Ex No. 9	Invalous and of the alaman and invalous the
Date:	Implementation of Backpropagation Algorithm

### Aim

To build an artificial neural network by implementing backpropagation algorithm.

### **Definitions**

#### **Neural Network**

Artificial neural networks, usually simply called neural networks or neural nets, are computing systems inspired by the biological neural networks that constitute animal brains. An ANN is based on a collection of connected units or nodes called artificial neurons, which loosely model the neurons in a biological brain.

# **Backpropagation Algorithm**

**Backpropagation** is the essence of neural network training. It is the method of fine-tuning the weights of a neural network based on the error rate obtained in the previous epoch (i.e., iteration). Proper tuning of the weights allows you to reduce error rates and make the model reliable by increasing its generalization. Backpropagation in neural network is a short form for "backward propagation of errors." It is a standard method of training artificial neural networks. This method helps calculate the gradient of a loss function with respect to all the weights in the network.

## **Procedure**

Open PyCharm Community Edition.

Go to File menu → New Project → Specify the project name → Press "Create" button.

Right Click on Project name  $\rightarrow$  New  $\rightarrow$  Python File  $\rightarrow$  Specify the file name  $\rightarrow$  Press Enter.

Type the following codes. Right click on file name or coding window → Select "Run" to view the result.

## Backprop.py

```
# Import Libraries
import numpy as np
import pandas as pd
from sklearn.datasets import load_iris
from sklearn.model selection import train test split
import matplotlib.pyplot as plt
# Load dataset
data = load iris()
# Get features and target
X=data.data
y=data.target
# Get dummy variable
y = pd.get\_dummies(y).values
print(y[:3])
#Split data into train and test data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=20, random_state=4)
# Initialize variables
learning rate = 0.1
iterations = 5000
N = y_{train.size}
# number of input features
input size = 4
# number of hidden layers neurons
hidden size = 2
# number of neurons at the output layer
output\_size = 3
results = pd.DataFrame(columns=["mse", "accuracy"])
# Initialize weights
```

```
np.random.seed(10)
# initializing weight for the hidden layer
W1 = np.random.normal(scale=0.5, size=(input size, hidden size))
# initializing weight for the output layer
W2 = np.random.normal(scale=0.5, size=(hidden_size, output_size))
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):
  acc = y_pred.argmax(axis=1) == y_true.argmax(axis=1)
  return acc.mean()
for itr in range(iterations):
  # feedforward propagation
  # on hidden layer
  Z1 = np.dot(X_train, W1)
  A1 = sigmoid(Z1)
  # on output layer
  Z2 = np.dot(A1, W2)
  A2 = sigmoid(Z2)
  # Calculating error
  mse = mean_squared_error(A2, y_train)
  acc = accuracy(A2, y_train)
  results = results.append({"mse": mse, "accuracy": acc}, ignore_index=True)
  # backpropagation
  E1 = A2 - y train
  dW1 = E1 * A2 * (1 - A2)
  E2 = np.dot(dW1, W2.T)
  dW2 = E2 * A1 * (1 - A1)
  # weight updates
  W2\_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
print(results.mse.plot(title="Mean Squared Error"))
plt.show()
print(results.accuracy.plot(title="Accuracy"))
plt.show()
```

```
# feedforward

Z1 = \text{np.dot}(X_{\text{test}}, W1)

A1 = \text{sigmoid}(Z1)

Z2 = \text{np.dot}(A1, W2)

A2 = \text{sigmoid}(Z2)
```

acc = accuracy(A2, y\_test)
print("Accuracy: {}".format(acc))

# Output

C:\Users\2mca1\PycharmProjects\sumaiya\venv\Scripts\python.exe C:\Users/2mca1\PycharmProjects\sumaiya\backprop.py

 $[[1\ 0\ 0]]$ 

 $[1\ 0\ 0]$ 

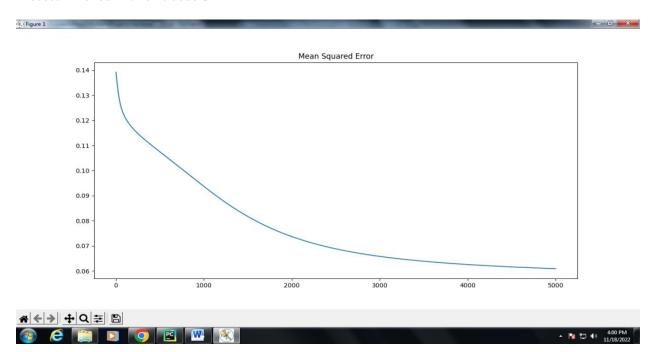
 $[1\ 0\ 0]]$ 

AxesSubplot(0.125,0.11;0.775x0.77)

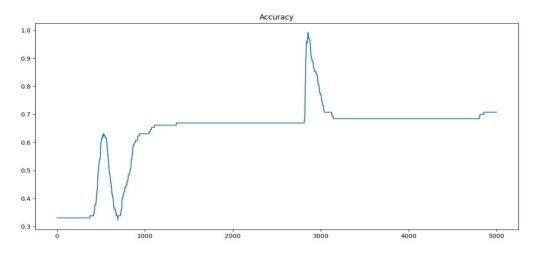
AxesSubplot(0.125,0.11;0.775x0.77)

Accuracy: 0.8

Process finished with exit code 0



4. Figure 1





# Result

Thus	, an artificial	neural	network	hac l	heen	constructed	hv	imn	lementing	hack	nronac	ration	aloo	orithm
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