02/11/20

EXPERIMENT NO. 4

1KS17CS030

AIM: Build an Artificial Neural Network by Implementing the Back Propagation algorithm & test the same using appropriate dataset

Dataset:

2.7810836	2.550537003	0
1.465489372	2.362125076	0
3.396561688	4.400293529	0
1.38807019	1.850220317	0
3.06407232	3.005305973	0
7.627531214	2.759262235	1
5.332441248	2.088626775	1
6.922596716	1.77106367	1
8.675418651	-0.242068655	1
7.673756466	3.508563011	1

ALGORITHM:

The stochastic gradient descent version of the BACKPROPAGATION algorithm for feed forward networks containing two layers of sigmoid units.

Step 1: Begins by constructing a network with the desired number of hidden & output units and initializing all network weights to small random values. For each training example, it applies the network to the example, calculates the error of the network output for this example, computes the gradient with respect to the error on this example, then updates all weights in the network. This gradient descent step is iterated until the network performs acceptably well.

Step 2: The gradient descent weight-update rule is similar to the delta training rule. The only difference is that the error $(t-o)$ in the delta rule is replaced by a more complex error term η_j.

Step3: Updates weights incrementally, following the presentation of each training example. This corresponds to a stochastic approximation to gradient descent. To obtain the true gradient of E one would sum the δ_j, x_{ji} values over all training examples before altering weight values.

Step 4: The weight-update loop in BACKPROPAGATION may be iterated thousands of times in a typical application. A variety of termination conditions can be used to halt the procedure.

One may choose to halt after a fixed number of iterations through the loop, or once the error on the training examples falls below some threshold.

PROGRAM:

```
from math import exp
from random import seed
from random import random

# Initialize a network
def initialize_network(n_inputs, n_hidden, n_outputs):
    network = list()
    hidden_layer = [{'weights':[random() for i in range(n_inputs+1)]}
    for i in range(n_hidden)]
    network.append(hidden_layer)
    output_layer = [{'weights': [random() for i in range(n_hidden+1)]}
    for i in range(n_outputs)]
    network.append(output_layer)
    return network.
```

```
#calculate neuron activation for an input
def activate(weights, inputs):
    activation = weights[-1]
    for i in range(len(weights)-1):
        activation += weights[i] * inputs[i]
    return activation.

#Transfer neuron activation.
def transfer(activation):
    return 1.0 / (1.0 + exp(-activation))

# Forward propagate input to a network output
def forward_propagate(network, row):
    inputs = row
    for layer in network:
        new_inputs = []
        for neuron in layer:
            activation = activate(neuron['weights'], inputs)
            neuron['output'] = transfer(activation)
            new_inputs.append(neuron['output'])
        inputs = new_inputs
    return inputs.

# calculate the derivative of an neuron output
def transfer_derivative(output):
    return output * (1.0 - output)
```

```
#Backpropagate error & store in neurons.
def backward_propagate_error(network, expected):
    for i in reversed(range(len(network))):
        layer = network[i]
        errors = list()
        if i != len(network)-1:
            for j in range(len(layer)):
                error = 0.0
                for neuron in network[i+1]:
                    error += (neuron['weights'][j] *
                              neuron['delta'])
                errors.append(error)
        else:
            for j in range(len(layer)):
                neuron = layer[j]
                errors.append(expected[j] - neuron['output'])
        for j in range(len(layer)):
            neuron = layer[j]
            neuron['delta'] = errors[j] * transfer_derivative(
                                neuron['output'])
#Update network weights with error
def update_weights(network, row, l_rate):
    for i in range(len(network)):
        inputs = row[:-1]
        if i != 0:
            inputs = [neuron['output'] for neuron in network[i-1]]
        for neuron in network[i]:
            for j in range(len(inputs)):
                neuron['weights'][j] += l_rate * neuron['delta'] * inputs[j]
            neuron['weights'][-1] += l_rate * neuron['delta']
```

```
# Train a network for a fixed number of epochs.
def train_network(network, train, l_rate, n_epoch, n_outputs):
    for epoch in range(n_epoch):
        sum_error = 0
        for row in train:
            outputs = forward_propagate(network, row)
            expected = [0 for i in range(n_outputs)]
            expected[row[-1]] = 1.
            sum_error += sum([(expected[i]-outputs[i])**2 for i
              in range(len(expected))])
            backward_propagate_error(network, expected)
            update_weights(network, row, l_rate)
        print('>epoch=%d, lrate=%.3f, error=%.3f' %
                (epoch, l_rate, sum_error))

# Test training backprop algorithm.
seed(1)
dataset = [[2.7810836, 2.550537003, 0],
           [1.465489372, 2.362125076, 0],
           [3.396561688, 4.400293529, 0],
           [1.38807019, 1.850220317, 0],
           [3.06407232, 3.005305973, 0],
           [7.627531214, 2.759262235, 1],
           [5.332441248, 2.088626775, 1],
           [6.922596716, 1.77106367, 1],
           [8.675418651, -0.242068655, 1],
           [7.673756466, 3.508563011, 1]]
```

```
n_inputs = len(dataset[0]) - 1
n_outputs = len(set([row[-1] for row in dataset]))
network = initialize_network(n_inputs, 2, n_outputs)
print(network)
train_network(network, dataset, 0.5, 20, n_outputs)
for layer in network:
    print(layer)
```

OUTPUT:

```
>epoch=0, lrate=0.500, error=6.350
>epoch=1, lrate=0.500, error=5.531
>epoch=2, lrate=0.500, error=5.221
>epoch=3, lrate=0.500, error=4.951
>epoch=4, lrate=0.500, error=4.519
>epoch=5, lrate=0.500, error=4.173
>epoch=6, lrate=0.500, error=3.835
>epoch=7, lrate=0.500, error=3.506
>epoch=8, lrate=0.500, error=3.192
>epoch=9, lrate=0.500, error=2.898
>epoch=10, lrate=0.500, error=2.626
>epoch=11, lrate=0.500, error=2.377
>epoch=12, lrate=0.500, error=2.153
>epoch=13, lrate=0.500, error=1.953
>epoch=14, lrate=0.500, error=1.774
>epoch=15, lrate=0.500, error=1.614
>epoch=16, lrate=0.500, error=1.472
>epoch=17, lrate=0.500, error=1.346
>epoch=18, lrate=0.500, error=1.233
>epoch=19, lrate=0.500, error=1.132
[{'weights': [-1.4688375095432327, 1.850887325439514, 1.0858178629550297], 'output': 0.029980305604426185, 'delta': -0.0059546604162323625}, {'weights': [0.37711098142462157, -0
.0625909894552989, 0.27651237026427l6], 'output': 0.9456229000211323, 'delta': 0.002627965285086383l}]
[{'weights': [2.515394649397849, -0.3391927502445985, -0.9671565426390275], 'output': 0.23648794202357587, 'delta': -0.04270059278364587}, {'weights': [-2.5584149848484263, 1.00
36422106209202, 0.423830864675827l5], 'output': 0.7790535202438367, 'delta': 0.038031325964373541]
```