Date

Expt. No. 4

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Program 4

Build an Artificial Neural Network by implementing the Backpropogation Algorithm and test the same using appropriate data sets.

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

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

def activate(weights, inputs):
    activation = weights[-1]
    for i in range(len(weights)-1):
        activation += weights[i] * inputs[i]
    return activation

def transfer(activation):
    return 1.0/(1.0+exp(-activation))
```

Teacher's Signature

```
def forward_propogate(network, row):
    input = 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

def transfer_derivative(output):
    return output * (1.0 - output)

def backward_propogate_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'])
```

Teacher's Signature _____

```
    for j in range(len(layer)):
        neuron = layer[j]
        neuron['delta'] = errors[j] * transfer_derivative(neuron
                                                ['output'])

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']

def train_network(network, l_rate, n_epoch, n_outputs):
    for epoch in range(n_epoch):
        sum_error = 0
        for row in train:
            outputs = forward_propogate(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_propogate_error(network, expected)
            epdate_weights(network, row, l_rate)
        print('>epoch = %d, lrate = %.3f, error = %.3f'
```

Teacher's Signature _____

```
% (epoch, l_rate, sum_error))

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)
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 = 5.021
>epoch =3 , lrate = 0.500 , error = 4.951
>epoch =4 , lrate = 0.500 , error = 4.579
>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 = 1.473
>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, 1.8508873254391511, 1.0858178029550297], 'output': 0.029980305604442185, 'delta': -0.005954660416233625}, {'weights': [0.3771109814246215?, -0.06259099055289, 0.2765123702642165], 'output': 0.9456229000211323, 'delta': 0.002627965285086388?}]

[{'weights': [2.5539493978402, -0.3391275024459855, -0.9675654639027502], 'output': 0.23648794202357587, 'delta': -0.042700592788145877}, {'weights': [-2.5584144886232, 1.0036422106209202,

0.42282864675827715], 'output': 0.7790535200438367, 'delta':
0.08801325964377354}]