```
Calling the main() procedure from the program
Starting main() function
Calling welcome() function
starting the welcome() function
************************
Welcome to the Multilayer Perceptron Neural Network
 trained using the backpropagation method.
Version 1.0, 03/25/2017, A.J. Maren
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******************
ending the welcome() function and returning to main
setting alpha, summedInput, maxNumIterations, and eta
alpha = 1.0 summedInput = 1
maxNumIterations = 2 eta = 0.5
Setting the debug xxx off statements to be false, so will print extra
calling the obtainNeuralNetworkSizeSpecs() function
Starting the obtainNeuralNetworkSizeSpecs() function
This network is set up to run the X-OR problem.
The numbers of nodes in the input, hidden, and output layers have been set to
2 each.
The number of input nodes is 2
The number of hidden nodes is 2
The number of output nodes is 2
Ending the obtainNeuralNetworkSizeSpecs() function
Flow-of-control trace: Back in main
I = number of nodes in input layer is 2
H = number of nodes in hidden layer is 2
O = number of nodes in output layer is 2
Initialize the trainingDataList to (0,0,0,0,0
Calling initializeWeightArray for input-to-hidden weights
starting the initializeWeightArray() function
initialize wt00 by calling initializeWeight() function
starting the InitializeWeight() function
weight = -0.8743401897182881
ending the InitializWeight() function
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initialize wt01 by calling initializeWeight() function
starting the InitializeWeight() function
weight = 0.021852356655784666
ending the InitializWeight() function
initialize wt02 by calling initializeWeight() function
starting the InitializeWeight() function
weight = -0.5701931574863277
ending the InitializWeight() function
initialize wt03 by calling initializeWeight() function
starting the InitializeWeight() function
weight = 0.994528119889952
ending the InitializWeight() function
  Inside initializeWeightArray
    The weights just initialized are:
     weight00 = -0.8743,
      weight01 = 0.0219,
     weight10 = -0.5702,
      weight11 = 0.9945,
    The weightArray just established is: [[-0.87434019 -0.57019316]
 [ 0.02185236  0.99452812]]
    Within this array:
       weight00 = -0.8743
                           weight10 = -0.5702
       weight01 = 0.0219 weight11 = 0.9945
  Returning to calling procedure
ending the initializeWeightArray() function and returning to main
Calling initializeWeightArray for Hidden-to-Output weights
starting the initializeWeightArray() function
initialize wt00 by calling initializeWeight() function
starting the InitializeWeight() function
weight = -0.48096945530919366
ending the InitializWeight() function
initialize wt01 by calling initializeWeight() function
starting the InitializeWeight() function
weight = -0.941926260530972
ending the InitializWeight() function
initialize wt02 by calling initializeWeight() function
starting the InitializeWeight() function
weight = 0.09592486977987913
ending the InitializWeight() function
initialize wt03 by calling initializeWeight() function
starting the InitializeWeight() function
weight = -0.24989251579461058
ending the InitializWeight() function
```

```
Inside initializeWeightArray
    The weights just initialized are:
     weight00 = -0.4810,
      weight01 = -0.9419,
      weight10 = 0.0959,
      weight11 = -0.2499,
    The weightArray just established is: [[-0.48096946 \quad 0.09592487]]
 [-0.94192626 -0.24989252]]
    Within this array:
       weight00 = -0.4810
                           weight10 = 0.0959
       weight01 = -0.9419 weight11 = -0.2499
  Returning to calling procedure
ending the initializeWeightArray() function and returning to main
calling initializeBiasWeightArray() function for Hidden nodes
starting initializeBiasWeightArray() function
initialize biasWeight() by calling initializeWeight() function
starting the InitializeWeight() function
weight = 0.8312551787243612
ending the InitializWeight() function
initialize biasWeight1 by calling initializeWeight() function
starting the InitializeWeight() function
weight = -0.5426447109691015
ending the InitializWeight() function
biasWeight0 = 0.8312551787243612
biasWeight1 = -0.5426447109691015
The biasWeightArray = [0.83125518 - 0.54264471]
ending initializeBiasWeightArray() function
returning to main
calling initializeBiasWeightArray() function for Output nodes
starting initializeBiasWeightArray() function
initialize biasWeight() by calling initializeWeight() function
starting the InitializeWeight() function
weight = 0.4909309716780714
ending the InitializWeight() function
initialize biasWeight1 by calling initializeWeight() function
starting the InitializeWeight() function
weight = 0.0042846390183264305
ending the InitializWeight() function
biasWeight0 = 0.4909309716780714
biasWeight1 = 0.0042846390183264305
```

The biasWeightArray = [0.49093097 0.00428464] ending initializeBiasWeightArray() function returning to main

The initial weights for this neural network are:

Input-to-Hidden Hidden-to-Output w(0,0) = -0.8743 w(1,0) = -0.5702 v(0,0) = -0.4810 v(1,0) = 0.0959 w(0,1) = 0.0219 w(1,1) = 0.9945 v(0,1) = -0.9419 v(1,1) = -0.2499

Bias at Hidden Layer Bias at Output Layer b(hidden,0) = 0.8313 b(output,0) = 0.4909 b(hidden,1) = -0.5426 b(output,1) = 0.0043

Establishing parameters - epsilon, iteration counter, SSE InitialTotal

Initialize SSE_InitialArray to zeros calling the computeSSE_Values() function to initialize the SSE_InitialArray starting the computeSSE_Values() function

Compute a single feed-forward pass and obtain the Actual Outputs for zeroth data set

Starting ComputeSingleFeedforwardPass() function The inputs transferred in are:

Input0 = 0Input1 = 0

The initial weights for this neural network are:
Input-to-Hidden Hidden-to-Output

w(0,0) = -0.874 w(0,1) = -0.570 v(0,0) = -0.481 v(0,1) = 0.096 w(1,0) = 0.022 w(1,1) = 0.995 v(1,0) = -0.942 v(1,1) = -0.250

For hiddenActivation0 from input0, input1 = 0, 0

computing hiddenActivation0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function The activation is 0.6966202653352188 Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0 , 0 The summed neuron input is 0.8313

The activation (applied transfer function) for that neuron is 0.6966 ending computeSingleNeuronActivation() function and returning activation

For hiddenActivation1 from input0, input1 = 0, 0 computing hiddenActivation1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function

compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.3675725698433179
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0 , 0 The summed neuron input is -0.5426

The activation (applied transfer function) for that neuron is 0.3676 ending computeSingleNeuronActivation() function and returning activation

In computeSingleFeedforwardPass:
Input node values: 0 , 0
The activations for the hidden nodes are:
 Hidden0 = 0.6966 Hidden1 = 0.3676

computing the activation of output node0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.5476393656209758
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.6966202653352188 , 0.3675725698433179

The summed neuron input is 0.1911

The activation (applied transfer function) for that neuron is 0.5476 ending computeSingleNeuronActivation() function and returning activation

computing the activation of output node1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function The activation is 0.32218817918715204 Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.6966202653352188, 0.3675725698433179

The summed neuron input is -0.7437

The activation (applied transfer function) for that neuron is 0.3222 ending computeSingleNeuronActivation() function and returning activation

Computing the output neuron activations

Back in ComputeSingleFeedforwardPass (for hidden-to-output computations) The activations for the output nodes are:

```
Output0 = 0.5476 Output1 = 0.3222
creating the actualAllNodesOutputList
the hiddenActivation0 = 0.6966202653352188
the hiddenActivation1 = 0.3675725698433179
the outputActivation0 = 0.5476393656209758
the outputActivation1 = 0.32218817918715204
ending ComputeSingleFeedforwardPass() function and returning to main
  In computeSSE Values
  Actual Node Outputs for (0,0) training set:
      input0 = 0 input1 = 0
      actualOutput0 = 0.5476   actualOutput1 = 0.3222
      error0 = -0.5476 error1 = 0.6778
   Initial SSE for (0,0) = 0.7593
Ending a single feed-forward pass to obtain the Actual Outputs for zeroth
data set
Compute a single feed-forward pass and obtain the Actual Outputs for first
data set
Starting ComputeSingleFeedforwardPass() function
The inputs transferred in are:
Input0 = 0
Input1 = 1
The initial weights for this neural network are:
     Input-to-Hidden
                                Hidden-to-Output
w(0,0) = -0.874 \quad w(0,1) = -0.570
                                         v(0,0) = -0.481 v(0,1) = 0.096
w(1,0) = 0.022 \quad w(1,1) = 0.995
                                      v(1,0) = -0.942 v(1,1) = -0.250
  For hiddenActivation0 from input0, input1 = 0 , 1
computing hiddenActivation0 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.5648973422025058
Ending the computeTransferFnction() function and returning the activation
  In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1
    The summed neuron input is 0.2611
    The activation (applied transfer function) for that neuron is 0.5649
ending computeSingleNeuronActivation() function and returning activation
  For hiddenActivation1 from input0, input1 = 0, 1
computing hiddenActivation1 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
```

Starting the computeTransferFnction() function
The activation is 0.6110869378709286
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0, 1 The summed neuron input is 0.4519

The activation (applied transfer function) for that neuron is 0.6111 ending computeSingleNeuronActivation() function and returning activation

In computeSingleFeedforwardPass:
Input node values: 0 , 1
The activations for the hidden nodes are:
 Hidden0 = 0.5649 Hidden1 = 0.6111

computing the activation of output node0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.5690192991279437
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.5648973422025058, 0.6110869378709286

The summed neuron input is 0.2779

The activation (applied transfer function) for that neuron is 0.5690 ending computeSingleNeuronActivation() function and returning activation

computing the activation of output node1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.336146803794963
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.5648973422025058, 0.6110869378709286

The summed neuron input is -0.6805

The activation (applied transfer function) for that neuron is 0.3361 ending computeSingleNeuronActivation() function and returning activation

Computing the output neuron activations

Back in ComputeSingleFeedforwardPass (for hidden-to-output computations)
The activations for the output nodes are:
 Output0 = 0.5690 Output1 = 0.3361

```
creating the actualAllNodesOutputList
the hiddenActivation0 = 0.5648973422025058
the hiddenActivation1 = 0.6110869378709286
the outputActivation0 = 0.5690192991279437
the outputActivation1 = 0.336146803794963
ending ComputeSingleFeedforwardPass() function and returning to main
  Actual Node Outputs for (0,1) training set:
    input0 = 0 input1 = 1
    actualOutput0 = 0.5690 actualOutput1 = 0.3361
    error0 = 0.4310 error1 = -0.3361
  Initial SSE for (0,1) = 0.2987
Ending a single feed-forward pass to obtain the Actual Outputs for first data
set
Compute a single feed-forward pass and obtain the Actual Outputs for second
data set
Starting ComputeSingleFeedforwardPass() function
The inputs transferred in are:
Input0 = 1
Input1 = 0
The initial weights for this neural network are:
    Input-to-Hidden Hidden-to-Output
w(0,0) = -0.874 \quad w(0,1) = -0.570
                                       v(0,0) = -0.481 v(0,1) = 0.096
w(1,0) = 0.022 w(1,1) = 0.995 v(1,0) = -0.942 v(1,1) = -0.250
  For hiddenActivation0 from input0, input1 = 1, 0
computing hiddenActivation0 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.4892304131816207
Ending the computeTransferFnction() function and returning the activation
  In computeSingleNeuronActivation with input0, input 1 given as: 1, 0
    The summed neuron input is -0.0431
    The activation (applied transfer function) for that neuron is 0.4892
ending computeSingleNeuronActivation() function and returning activation
  For hiddenActivation1 from input0, input1 = 1, 0
computing hiddenActivation1 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.372666973421586
```

In computeSingleNeuronActivation with input0, input 1 given as: 1 , 0 The summed neuron input is -0.5208

The activation (applied transfer function) for that neuron is 0.3727 ending computeSingleNeuronActivation() function and returning activation

In computeSingleFeedforwardPass:
Input node values: 1 , 0
The activations for the hidden nodes are:
 Hidden0 = 0.4892 Hidden1 = 0.3727

computing the activation of output node0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.5723325062488019
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.4892304131816207, 0.372666973421586

The summed neuron input is 0.2914

The activation (applied transfer function) for that neuron is 0.5723 ending computeSingleNeuronActivation() function and returning activation

computing the activation of output node1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.36594305801668164
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.4892304131816207 , 0.372666973421586

The summed neuron input is -0.5497

The activation (applied transfer function) for that neuron is 0.3659 ending computeSingleNeuronActivation() function and returning activation

Computing the output neuron activations

Back in ComputeSingleFeedforwardPass (for hidden-to-output computations)
The activations for the output nodes are:
 Output0 = 0.5723 Output1 = 0.3659

creating the actualAllNodesOutputList
the hiddenActivation0 = 0.4892304131816207

```
the hiddenActivation1 = 0.372666973421586
the outputActivation0 = 0.5723325062488019
the outputActivation1 = 0.36594305801668164
ending ComputeSingleFeedforwardPass() function and returning to main
  Actual Node Outputs for (1,0) training set:
     input0 = 1 input1 = 0
    actualOutput0 = 0.5723 actualOutput1 = 0.3659
error0 = 0.4277 error1 = -0.3659
  Initial SSE for (1,0) = 0.3168
Ending a single feed-forward pass to obtain the Actual Outputs for second
data set
Compute a single feed-forward pass and obtain the Actual Outputs for third
data set
Starting ComputeSingleFeedforwardPass() function
The inputs transferred in are:
Input0 = 1
Input1 = 1
The initial weights for this neural network are:
    Input-to-Hidden
                      Hidden-to-Output
w(0,0) = -0.874 w(0,1) = -0.570 v(0,0) = -0.481 v(0,1) = 0.096
w(1,0) = 0.022 \quad w(1,1) = 0.995
                                      v(1,0) = -0.942 v(1,1) = -0.250
  For hiddenActivation0 from input0, input1 = 1, 1
computing hiddenActivation0 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.35131176652597756
Ending the computeTransferFnction() function and returning the activation
  In computeSingleNeuronActivation with input0, input 1 given as: 1, 1
    The summed neuron input is -0.6133
    The activation (applied transfer function) for that neuron is 0.3513
ending computeSingleNeuronActivation() function and returning activation
  For hiddenActivation1 from input0, input1 = 1, 1
computing hiddenActivation1 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.6162675800048231
Ending the computeTransferFnction() function and returning the activation
```

In computeSingleNeuronActivation with input0, input 1 given as: 1, 1 The summed neuron input is 0.4737 The activation (applied transfer function) for that neuron is 0.6163 ending computeSingleNeuronActivation() function and returning activation In computeSingleFeedforwardPass: Input node values: 1 , 1 The activations for the hidden nodes are: $Hidden0 = 0.3513 \; Hidden1 = 0.6163$ computing the activation of output node0 by calling computeSingleNeuronActivation starting computeSingleNeuronActivation() function compute the activation by calling computeTransferFnctn() Starting the computeTransferFnction() function The activation is 0.5941326261750242 Ending the computeTransferFnction() function and returning the activation In computeSingleNeuronActivation with input0, input 1 given as: 0.35131176652597756 , 0.6162675800048231 The summed neuron input is 0.3811 The activation (applied transfer function) for that neuron is 0.5941 ending computeSingleNeuronActivation() function and returning activation computing the activation of output nodel by calling computeSingleNeuronActivation starting computeSingleNeuronActivation() function compute the activation by calling computeTransferFnctn() Starting the computeTransferFnction() function The activation is 0.3821043636347127 Ending the computeTransferFnction() function and returning the activation In computeSingleNeuronActivation with input0, input 1 given as: 0.35131176652597756 , 0.6162675800048231 The summed neuron input is -0.4806 The activation (applied transfer function) for that neuron is 0.3821 ending computeSingleNeuronActivation() function and returning activation Computing the output neuron activations Back in ComputeSingleFeedforwardPass (for hidden-to-output computations) The activations for the output nodes are: Output0 = 0.5941 Output1 = 0.3821 creating the actualAllNodesOutputList the hiddenActivation0 = 0.35131176652597756the hiddenActivation1 = 0.6162675800048231the outputActivation0 = 0.5941326261750242the outputActivation1 = 0.3821043636347127

```
ending ComputeSingleFeedforwardPass() function and returning to main
  Actual Node Outputs for (1,1) training set:
     input0 = 1 input1 = 1
     actualOutput0 = 0.5941 actualOutput1 = 0.3821
     error0 =
                    -0.5941 error1 =
  Initial SSE for (1,1) = 0.7348
Ending a single feed-forward pass to obtain the Actual Outputs for third data
set
Initializing array of SSE values
the SSE_InitialArray[0] = 0.7593377392113732 the SSE_InitialArray[1] = 0.7593377392113732
0.2987390382257382
the SSE InitialArray[2] = 0.3168138069220314 the SSE InitialArray[3] =
0.7347885949248943
  The initial total of the SSEs is 2.1097
ending the computeSSE Values() function and returning to main
The SSE Array = [0.7593377392113732, 0.2987390382257382, 0.3168138069220314,
0.7347885949248943, 2.1096791792840373]
The SSE InitialTotal = 2.1096791792840373
In main, SSE computations completed, Total of all SSEs = 2.1097
  For input nodes (0,0), SSE Array[0] = 0.7593
  For input nodes (0,1), SSE Array[1] = 0.2987
  For input nodes (1,0), SSE Array[2] = 0.3168
  For input nodes (1,1), SSE Array[3] = 0.7348
About to enter the while loop for 2 iterations
starting at top of while loop for first time
create random trainingDataList by calling obtainRandomXORTrainingValues()
function
starting obtainRandomXORTrainingValues() function
The trainingDataList = (0, 1, 1, 0, 1)
\verb|ending| obtain Random XOR Training Values()| function| and returning to main|\\
Iteration number 0
Randomly selected training data set number 1
The inputs and desired outputs for the X-OR problem from this data set are:
          Input0 = 0
                                  Input1 = 1
Desired Output0 = 1
                       Desired Output1 = 0
calling ComputeSingleFeedforwardPass to obtain actual outputs for training
Starting ComputeSingleFeedforwardPass() function
The inputs transferred in are:
Input0 = 0
```

```
Input1 = 1
```

The initial weights for this neural network are: Input-to-Hidden Hidden-to-Output w(0,0) = -0.874 w(0,1) = -0.570 v(0,0) = -0.481 v(0,1) = 0.096 w(1,0) = 0.022 w(1,1) = 0.995 v(1,0) = -0.942 v(1,1) = -0.250 $w(1,0) = 0.022 \quad w(1,1) = 0.995$ For hiddenActivation0 from input0, input1 = 0, 1 computing hiddenActivation0 by calling computeSingleNeuronActivation starting computeSingleNeuronActivation() function compute the activation by calling computeTransferFnctn() Starting the computeTransferFnction() function The activation is 0.5648973422025058 Ending the computeTransferFnction() function and returning the activation In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1 The summed neuron input is 0.2611 The activation (applied transfer function) for that neuron is 0.5649 ending computeSingleNeuronActivation() function and returning activation For hiddenActivation1 from input0, input1 = 0, 1 computing hiddenActivation1 by calling computeSingleNeuronActivation starting computeSingleNeuronActivation() function compute the activation by calling computeTransferFnctn() Starting the computeTransferFnction() function The activation is 0.6110869378709286 Ending the computeTransferFnction() function and returning the activation In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1 The summed neuron input is 0.4519 The activation (applied transfer function) for that neuron is 0.6111 ending computeSingleNeuronActivation() function and returning activation In computeSingleFeedforwardPass: Input node values: 0 , 1 The activations for the hidden nodes are: $Hidden0 = 0.5649 \; Hidden1 = 0.6111$

The activations for the hidden nodes are:
 Hidden0 = 0.5649 Hidden1 = 0.6111

computing the activation of output node0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.5690192991279437
Ending the computeTransferFnction() function and returning the activation

```
In computeSingleNeuronActivation with input0, input 1 given as:
0.5648973422025058 , 0.6110869378709286
    The summed neuron input is 0.2779
    The activation (applied transfer function) for that neuron is 0.5690
ending computeSingleNeuronActivation() function and returning activation
computing the activation of output nodel by calling
computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.336146803794963
Ending the computeTransferFnction() function and returning the activation
  In computeSingleNeuronActivation with input0, input 1 given as:
0.5648973422025058 , 0.6110869378709286
    The summed neuron input is -0.6805
    The activation (applied transfer function) for that neuron is 0.3361
ending computeSingleNeuronActivation() function and returning activation
  Computing the output neuron activations
  Back in ComputeSingleFeedforwardPass (for hidden-to-output computations)
  The activations for the output nodes are:
    Output0 = 0.5690 Output1 = 0.3361
creating the actualAllNodesOutputList
the hiddenActivation0 = 0.5648973422025058
the hiddenActivation1 = 0.6110869378709286
the outputActivation0 = 0.5690192991279437
the outputActivation1 = 0.336146803794963
ending ComputeSingleFeedforwardPass() function and returning to main
In main; have just completed a feedfoward pass with training set inputs 0 1
  The activations (actual outputs) for the two hidden neurons are:
    actualHiddenOutput0 = 0.5649
    actualHiddenOutput1 = 0.6111
  The activations (actual outputs) for the two output neurons are:
    actualOutput0 = 0.5690
    actualOutput1 = 0.3361'
  Initial SSE (before backpropagation) = 0.298739
  Corresponding SSE (from initial SSE determination) = 0.298739
Calling BackpropagateOutputToHidden() function
Starting the BackpropagateOutputToHidden() function
Performs the backpropagation of weight changes onto hidden-to-output weigts
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
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```
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Calling the PrintAndTraceBackpropagateOutputToHidden() function
Starting PrintAndTraceBackpropagateOutputToHidden() function
In Print and Trace for Backpropagation: Hidden to Output Weights
 Assuming alpha = 1
 The hidden node activations are:
   Hidden node 0: 0.5649 Hidden node 1: 0.6111
 The output node activations are:
   Output node 0:
                   0.569
                            Output node 1: 0.336
 The transfer function derivatives are:
   Deriv-F(0): 0.245 Deriv-F(1):
                                             0.223
The computed values for the deltas are:
              eta * error * trFncDeriv * hidden
 deltaVWt00 = 0.50 * 0.4310 * 0.2452 * 0.5649
 deltaVWt01 = 0.50 * -0.3361 * 0.2232 * 0.5649
 deltaVWt10 = 0.50 * 0.4310 * 0.2452
                                         * 0.6111
 deltaVWt11 = 0.50 * -0.3361 * 0.2232 * 0.6111
Values for the hidden-to-output connection weights:
          Old:
                  New:
                           eta*Delta:
                          0.0299
        -0.4810 -0.4511
[0,0]:
[0,1]:
       -0.9419 -0.9631 -0.0212
[1,0]: 0.0959 0.1282 0.0323
[1,1]: -0.2499 -0.2728 -0.0229
Ending PrintAndTraceBackpropagateOutputToHidden() function
Back in the BackpropagateOutputToHidden() function
Ending the BackpropagateOutputToHidden() function
Calling BackpropagateBiasOutputWeights() function
Starting the BackpropagateBiasOutputWeights() function
Calling the computeTransferFnctnDeriv() function for outputNode0
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Calling the computeTransferFnctnDeriv() function for outputNode1
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
```

The newBiasOutputWeightArray = [0.54377704 - 0.0332213]

Ending the BackpropagateBiasOutputWeights() function Calling BackpropagateHiddenToInput() function Starting the BackpropagateHiddenToInput() function This Backpropagates the weight changes onto the input-to-hidden weights Calling the computeTransferFnctnDeriv() function for hiddenNode0 Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function Calling the computeTransferFnctnDeriv() function for hiddenNode1 Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function Calling the computeTransferFnctnDeriv() function for outputNode0 Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function Calling the computeTransferFnctnDeriv() function for outputNode1 Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function Calling the PrintAndTraceBackpropagateHiddenToInput() function Starting PrintAndTraceBackpropagateHiddenToInput() function Traces the backpropagation of the input-to-hidden weights In Print and Trace for Backpropagation: Input to Hidden Weights Assuming alpha = 1The hidden node activations are: Hidden node 0: 0.5649 Hidden node 1: 0.6111 The output node activations are: Output node 0: 0.569 Output node 1: The transfer function derivatives at the hidden nodes are: Deriv-F(0): 0.246 Deriv-F(1): The computed values for the deltas are: eta * error * trFncDeriv * input given H deltaWWt00 = 0.50 * 0.4310 * 0.2458 * 0.0000 * 0.1093 deltaWWt01 = 0.50 * -0.3361 * 0.2377 * 0.0000 * 0.1253deltaWWt10 = 0.50 * 0.4310 * 0.2458 * 1.0000 * 0.1093

deltaWWt11 = 0.50 * -0.3361 * 0.2377 * 1.0000 * 0.1253

```
Values for the input-to-hidden connection weights:
          Old:
                  New: eta*Delta:
        -0.8743 -0.8743 0.0000
[0,0]:
[0,1]: 0.0219 0.0219 0.0000
[1,0]: -0.5702 -0.5678 0.0024
[1,1]: 0.9945 0.9980 0.0034
Ending the PrintAndTraceBackpropagateHiddenToInput() function
Ending the BackpropagateHiddenToInput() function
Calling BackpropagateBiasHiddenWeights() function
Starting the BackpropagateBiasHiddenWeights() function
This backpropagates weight changes onto the bias-to-hidden connection weights
Calling the computeTransferFnctnDeriv() function for outputNode0
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Calling the computeTransferFnctnDeriv() function for outputNode1
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Calling the computeTransferFnctnDeriv() function for hiddenNode0
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Calling the computeTransferFnctnDeriv() function for hiddenNode1
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
The newBiasHiddenWeightArray = [0.83369106 - 0.5392125]
Ending the BackpropagateBiasHiddenWeights() function
The newBiasWeightArray = [[0.5437770357373254, -0.033221298627139594],
[0.8336910603185961, -0.5392124998204]]
    The weights before backpropagation are:
        Input-to-Hidden
                                                  Hidden-to-Output
   w(0,0) = -0.874 \quad w(1,0) = -0.570
                                           v(0,0) = -0.481 \quad v(1,0) =
0.096
   w(0,1) = 0.022 \quad w(1,1) = 0.995
                                          v(0,1) = -0.942 v(1,1) = -0.250
   The weights after backpropagation are:
        Input-to-Hidden
                                                 Hidden-to-Output
   w(0,0) = -0.874 w(1,0) = -0.568 v(0,0) = -0.451 v(1,0) =
0.128
```

```
w(0,1) = 0.022 w(1,1) = 0.998 v(0,1) = -0.963 v(1,1) = -0.273
Calling the ComputeSingleFeedforwardPass() function
Starting ComputeSingleFeedforwardPass() function
The inputs transferred in are:
Input0 = 0
Input1 = 1
The initial weights for this neural network are:
    Input-to-Hidden
                      Hidden-to-Output
w(0,0) = -0.874 w(0,1) = -0.568 v(0,0) = -0.451 v(0,1) = 0.128
w(1,0) = 0.022 \quad w(1,1) = 0.998
                                    v(1,0) = -0.963 v(1,1) = -0.273
 For hiddenActivation0 from input0, input1 = 0, 1
computing hiddenActivation0 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.5654959585571423
Ending the computeTransferFnction() function and returning the activation
 In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1
   The summed neuron input is 0.2635
   The activation (applied transfer function) for that neuron is 0.5655
ending computeSingleNeuronActivation() function and returning activation
 For hiddenActivation1 from input0, input1 = 0, 1
computing hiddenActivation1 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.6119023244302164
Ending the computeTransferFnction() function and returning the activation
 In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1
    The summed neuron input is 0.4553
    The activation (applied transfer function) for that neuron is 0.6119
ending computeSingleNeuronActivation() function and returning activation
 In computeSingleFeedforwardPass:
 Input node values: 0 , 1
 The activations for the hidden nodes are:
   Hidden0 = 0.5655 \ Hidden1 = 0.6119
computing the activation of output node0 by calling
computeSingleNeuronActivation
```

```
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.5779304340535262
Ending the computeTransferFnction() function and returning the activation
  In computeSingleNeuronActivation with input0, input 1 given as:
0.5654959585571423 , 0.6119023244302164
    The summed neuron input is 0.3143
    The activation (applied transfer function) for that neuron is 0.5779
ending computeSingleNeuronActivation() function and returning activation
computing the activation of output node1 by calling
computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.33019875492840167
Ending the computeTransferFnction() function and returning the activation
 In computeSingleNeuronActivation with input0, input 1 given as:
0.5654959585571423 , 0.6119023244302164
    The summed neuron input is -0.7073
    The activation (applied transfer function) for that neuron is 0.3302
ending computeSingleNeuronActivation() function and returning activation
  Computing the output neuron activations
  Back in ComputeSingleFeedforwardPass (for hidden-to-output computations)
  The activations for the output nodes are:
   Output0 = 0.5779 Output1 = 0.3302
creating the actualAllNodesOutputList
the hiddenActivation0 = 0.5654959585571423
the hiddenActivation1 = 0.6119023244302164
the outputActivation0 = 0.5779304340535262
the outputActivation1 = 0.33019875492840167
ending ComputeSingleFeedforwardPass() function and returning to main
In main; have just completed a single step of backpropagation with inputs 0 1
    The new SSE (after backpropagation) = 0.287174
    Error(0) = 0.4221, Error(1) = -0.3302
    SSE(0) = 0.1781,
                       SSE(1) =
                                  0.1090
  The difference in initial and the resulting SSEs is: 0.0116
   The training has resulted in improving the total SSEs
  The previous SSE Total was 2.1097
  The new SSE Total was 2.0981
    For node 0: Desired Output = 1 New Output = 0.5779
```

```
For node 1: Desired Output = 0 New Output = 0.3302
   Error(0) = 0.4221, Error(1) = -0.3302
    SSEO(0) = 0.1781,
                        SSE(1) = 0.1090
 Delta in the SSEs is 0.0116
SSE improvement
Iteration number 0
starting at the top of the while loop again
create random trainingDataList by calling obtainRandomXORTrainingValues()
function
starting obtainRandomXORTrainingValues() function
The trainingDataList = (0, 1, 1, 0, 1)
ending obtainRandomXORTrainingValues() function and returning to main
Iteration number 1
Randomly selected training data set number 1
The inputs and desired outputs for the X-OR problem from this data set are:
         Input0 = 0
                                 Input1 = 1
Desired Output0 = 1    Desired Output1 = 0
calling ComputeSingleFeedforwardPass to obtain actual outputs for training
Starting ComputeSingleFeedforwardPass() function
The inputs transferred in are:
Input0 = 0
Input1 = 1
The initial weights for this neural network are:
    Input-to-Hidden
                               Hidden-to-Output
w(0,0) = -0.874 \quad w(0,1) = -0.568
                                        v(0,0) = -0.451 v(0,1) = 0.128
w(1,0) = 0.022 \quad w(1,1) = 0.998
                                      v(1,0) = -0.963 v(1,1) = -0.273
 For hiddenActivation0 from input0, input1 = 0, 1
computing hiddenActivation0 by calling computeSingleNeuronActivation
starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()
Starting the computeTransferFnction() function
The activation is 0.5654959585571423
Ending the computeTransferFnction() function and returning the activation
 In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1
   The summed neuron input is 0.2635
    The activation (applied transfer function) for that neuron is 0.5655
ending computeSingleNeuronActivation() function and returning activation
```

For hiddenActivation1 from input0, input1 = 0, 1

computing hiddenActivation1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.6119023244302164
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1 The summed neuron input is 0.4553

The activation (applied transfer function) for that neuron is 0.6119 ending computeSingleNeuronActivation() function and returning activation

In computeSingleFeedforwardPass:
Input node values: 0 , 1
The activations for the hidden nodes are:
 Hidden0 = 0.5655 Hidden1 = 0.6119

computing the activation of output node0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.5779304340535262
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.5654959585571423 , 0.6119023244302164

The summed neuron input is 0.3143

The activation (applied transfer function) for that neuron is 0.5779 ending computeSingleNeuronActivation() function and returning activation

computing the activation of output nodel by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function The activation is 0.33019875492840167 Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.5654959585571423 , 0.6119023244302164

The summed neuron input is -0.7073

The activation (applied transfer function) for that neuron is 0.3302 ending computeSingleNeuronActivation() function and returning activation

Computing the output neuron activations

Back in ComputeSingleFeedforwardPass (for hidden-to-output computations)

```
The activations for the output nodes are:
   Output0 = 0.5779 Output1 = 0.3302
creating the actualAllNodesOutputList
the hiddenActivation0 = 0.5654959585571423
the hiddenActivation1 = 0.6119023244302164
the outputActivation0 = 0.5779304340535262
the outputActivation1 = 0.33019875492840167
ending ComputeSingleFeedforwardPass() function and returning to main
In main; have just completed a feedfoward pass with training set inputs 0 1
 The activations (actual outputs) for the two hidden neurons are:
   actualHiddenOutput0 = 0.5655
   actualHiddenOutput1 = 0.6119
 The activations (actual outputs) for the two output neurons are:
   actualOutput0 = 0.5779
    actualOutput1 = 0.3302'
 Initial SSE (before backpropagation) = 0.287174
 Corresponding SSE (from initial SSE determination) = 0.287174
Calling BackpropagateOutputToHidden() function
Starting the BackpropagateOutputToHidden() function
Performs the backpropagation of weight changes onto hidden-to-output weigts
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Starting the computeTransferFnctnDeriv() function
This computes the derivate of the transfer function
Ending the computeTransferFnctnDeriv() function
Calling the PrintAndTraceBackpropagateOutputToHidden() function
Starting PrintAndTraceBackpropagateOutputToHidden() function
In Print and Trace for Backpropagation: Hidden to Output Weights
 Assuming alpha = 1
 The hidden node activations are:
   Hidden node 0:
                    0.5655 Hidden node 1: 0.6119
 The output node activations are:
   Output node 0: 0.578 Output node 1: 0.330
 The transfer function derivatives are:
   Deriv-F(0):
                    0.244 Deriv-F(1):
The computed values for the deltas are:
              eta * error * trFncDeriv * hidden
 deltaVWt00 = 0.50 * 0.4221 * 0.2439 * 0.5655
 deltaVWt01 = 0.50 * -0.3302 * 0.2212 * 0.5655
 deltaVWt10 = 0.50 * 0.4221 * 0.2439 * 0.6119
 deltaVWt11 = 0.50 * -0.3302 * 0.2212 * 0.6119
```

Values for the hidden-to-output connection weights:

Old: New: eta*Delta:
[0,0]: -0.4511 -0.4220 0.0291
[0,1]: -0.9631 -0.9838 -0.0206
[1,0]: 0.1282 0.1597 0.0315
[1,1]: -0.2728 -0.2952 -0.0223

Ending PrintAndTraceBackpropagateOutputToHidden() function

Back in the BackpropagateOutputToHidden() function

Ending the BackpropagateOutputToHidden() function

Calling BackpropagateBiasOutputWeights() function

Starting the BackpropagateBiasOutputWeights() function

Calling the computeTransferFnctnDeriv() function for outputNode0

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the computeTransferFnctnDeriv() function for outputNode1

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

The newBiasOutputWeightArray = [0.54240802 - 0.03222998]

Ending the BackpropagateBiasOutputWeights() function

Calling BackpropagateHiddenToInput() function

Starting the BackpropagateHiddenToInput() function
This Backpropagates the weight changes onto the input-to-hidden weights

Calling the computeTransferFnctnDeriv() function for hiddenNode0

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the computeTransferFnctnDeriv() function for hiddenNode1

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the computeTransferFnctnDeriv() function for outputNode0

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the compute Transfer Fnctn Deriv() function for output Node 1 $\,$

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the PrintAndTraceBackpropagateHiddenToInput() function

Starting PrintAndTraceBackpropagateHiddenToInput() function Traces the backpropagation of the input-to-hidden weights

In Print and Trace for Backpropagation: Input to Hidden Weights Assuming alpha = 1

The hidden node activations are:

Hidden node 0: 0.5655 Hidden node 1: 0.6119

The output node activations are:

Output node 0: 0.578 Output node 1:

The transfer function derivatives at the hidden nodes are: Deriv-F(0): 0.246 Deriv-F(1):

The computed values for the deltas are:

eta * error * trFncDeriv * input * SumTerm for

given H

deltaWWt00 = 0.50 * 0.4221 * 0.2457 * 0.0000 * 0.1276 deltaWWt01 = 0.50 * -0.3302 * 0.2375 * 0.0000 * 0.1442deltaWWt10 = 0.50 * 0.4221 * 0.2457 * 1.0000 * 0.1276deltaWWt11 = 0.50 * -0.3302 * 0.2375 * 1.0000 * 0.1442

Values for the input-to-hidden connection weights:

New: Old: eta*Delta:

-0.8743 0.0000 [0,0]: -0.8743 [0,1]: 0.0219 0.0219 0.0000 [1,0]: -0.5678 -0.5648 0.0029[1,1]: 0.9980 1.0019 0.0039

Ending the PrintAndTraceBackpropagateHiddenToInput() function

Ending the BackpropagateHiddenToInput() function

Calling BackpropagateBiasHiddenWeights() function

Starting the BackpropagateBiasHiddenWeights() function This backpropagates weight changes onto the bias-to-hidden connection weights

Calling the computeTransferFnctnDeriv() function for outputNode0

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the computeTransferFnctnDeriv() function for outputNode1

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the computeTransferFnctnDeriv() function for hiddenNode0

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

Calling the computeTransferFnctnDeriv() function for hiddenNode1

Starting the computeTransferFnctnDeriv() function This computes the derivate of the transfer function Ending the computeTransferFnctnDeriv() function

The newBiasHiddenWeightArray = [0.83419032 - 0.53871162]

Ending the BackpropagateBiasHiddenWeights() function

The newBiasWeightArray = [[0.5424080209906543, -0.03222998368408356],[0.8341903238759356, -0.5387116193191024]]

The weights before backpropagation are:

Input-to-Hidden Hidden-to-Output w(0,0) = -0.874 w(1,0) = -0.568 $v(0,0) = -0.451 \quad v(1,0) =$ 0.128

 $w(0,1) = 0.022 \quad w(1,1) = 0.998$ v(0,1) = -0.963 v(1,1) = -0.273

The weights after backpropagation are:

Input-to-Hidden Hidden-to-Output w(0,0) = -0.874 w(1,0) = -0.565 v(0,0) = -0.422 v(1,0) =0.160 w(0,1) = 0.022 w(1,1) = 1.002 v(0,1) = -0.984 v(1,1) = -0.295

Calling the ComputeSingleFeedforwardPass() function

Starting ComputeSingleFeedforwardPass() function The inputs transferred in are:

Input0 = 0Input1 = 1

Input-to-Hidden

The initial weights for this neural network are:

Hidden-to-Output w(0,0) = -0.874 w(0,1) = -0.565 v(0,0) = -0.422 v(0,1) = 0.160w(1,0) = 0.022 w(1,1) = 1.002v(1,0) = -0.984 v(1,1) = -0.295

For hiddenActivation0 from input0, input1 = 0, 1

computing hiddenActivation0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function The activation is 0.5662170147588406 Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1 The summed neuron input is 0.2664

The activation (applied transfer function) for that neuron is 0.5662 ending computeSingleNeuronActivation() function and returning activation

For hiddenActivation1 from input0, input1 = 0 , 1 computing hiddenActivation1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function The activation is 0.6128359345509735 Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0 , 1 The summed neuron input is 0.4592

The activation (applied transfer function) for that neuron is 0.6128 ending computeSingleNeuronActivation() function and returning activation

In computeSingleFeedforwardPass:
Input node values: 0 , 1
The activations for the hidden nodes are:
 Hidden0 = 0.5662 Hidden1 = 0.6128

computing the activation of output node0 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.586584615580249
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.5662170147588406, 0.6128359345509735

The summed neuron input is 0.3499

The activation (applied transfer function) for that neuron is 0.5866 ending computeSingleNeuronActivation() function and returning activation

computing the activation of output node1 by calling computeSingleNeuronActivation

starting computeSingleNeuronActivation() function
compute the activation by calling computeTransferFnctn()

Starting the computeTransferFnction() function
The activation is 0.32440084935542585
Ending the computeTransferFnction() function and returning the activation

In computeSingleNeuronActivation with input0, input 1 given as: 0.5662170147588406, 0.6128359345509735The summed neuron input is -0.7336

```
The activation (applied transfer function) for that neuron is 0.3244 ending computeSingleNeuronActivation() function and returning activation
```

Computing the output neuron activations Back in ComputeSingleFeedforwardPass (for hidden-to-output computations) The activations for the output nodes are: Output0 = 0.5866 Output1 = 0.3244creating the actualAllNodesOutputList the hiddenActivation0 = 0.5662170147588406the hiddenActivation1 = 0.6128359345509735the outputActivation0 = 0.586584615580249the outputActivation1 = 0.32440084935542585ending ComputeSingleFeedforwardPass() function and returning to main In main; have just completed a single step of backpropagation with inputs 0 1 The new SSE (after backpropagation) = 0.276148Error(0) = 0.4134, Error(1) = -0.3244SSE(0) = 0.1709,SSE(1) = 0.1052The difference in initial and the resulting SSEs is: 0.0110 The training has resulted in improving the total SSEs The previous SSE Total was 2.0981 The new SSE Total was 2.0871 For node 0: Desired Output = 1 New Output = 0.5866 For node 1: Desired Output = 0 New Output = 0.3244 Error(0) = 0.4134, Error(1) = -0.3244SSEO(0) = 0.1709, SSE(1) = 0.1052Delta in the SSEs is 0.0110 SSE improvement Iteration number 1 Out of while loop Initial Total SSE = 2.1097 Final Total SSE = 2.0871Delta in the SSEs is 0.0226 SSE total improvement The initial weights for this neural network are: Input-to-Hidden Hidden-to-Output w(0,0) = -0.874 w(0,1) = -0.570 v(0,0) = -0.481 v(0,1) = 0.096 w(1,0) = 0.022 w(1,1) = 0.995 v(1,0) = -0.942 v(1,1) = -0.250Input-to-Hidden Hidden-to-Output v(0,0) = -0.481 v(0,1) = 0.096The final weights for this neural network are: Input-to-Hidden Hidden-to-Output w(0,0) = -0.874 w(0,1) = -0.565 v(0,0) = -0.422 v(0,1) = 0.160 w(1,0) = 0.022 w(1,1) = 1.002 v(1,0) = -0.984 v(1,1) = -0.295 $v(0,0) = -0.422 \quad v(0,1) = 0.160$ The SSE values at the beginning of training were: SSE Initial[0] = 0.7593SSE Initial[1] = 0.2987SSE Initial[2] = 0.3168

```
SSE Initial[3] = 0.7348
```

The total of the SSE values at the beginning of training is 2.1097

The SSE values at the end of training were:

SSE[0] = 0.7593 SSE[1] = 0.2761 SSE[2] = 0.3168 SSE[3] = 0.7348

The total of the SSE values at the end of training is 2.0871

Values for the new outputs compared with previous, given only a partial backpropagation training:

Old: New: nu*Delta:
Output 0: Desired = 1 Old actual = 0.5779 Newactual 0.5866
Output 1: Desired = 0 Old actual = 0.3302 Newactual 0.3244

ending main() function

