SAGNIK BASU

113EC0199

Simulate a 2input XOR gate using MLP with back-propagation algorithm. Use a MLP with 2 neural layers. The second layer should have only one neuron. Tune the number of neurons in the first layer starting with 2neurons to achieve the result. Repeat the experiment for 3input XOR gate also.

MATLAB CODE:

1) For two input xor gate

```
% multi layer perceptron
% XOR input for x1 and x2
input = [0 \ 0 \ ; 0 \ 1 \ ; 1 \ 0 \ ; 1 \ 1];
% Desired output of XOR
output = [0;1;1;0];
% Initialize the bias
bias = [-1 - 1 - 1];
% Learning coefficient
coeff = 0.7;
% Number of learning iterations
iterations = 1000;
% Calculate weights randomly using seed.
%rand('state',sum(100*clock));
weights = 5 + 2.*rand(3,3);
for i = 1:iterations
  out = zeros(4,1);
  numIn = length (input(:,1));
  for j = 1:numIn
```

```
% Hidden laver
        H1 = bias(1,1)*weights(1,1)+input(j,1)*weights(1,2)+input(j,2)*weights(1,3);
        % Send data through sigmoid function 1/1+e^-x
        % Note that sigma is a different m file
        % that I created to run this operation
        x2(1) = sigma(H1);
        H2 = bias(1,2)*weights(2,1) + input(j,1)*weights(2,2) + input(j,2)*weights(2,3);
        x2(2) = sigma(H2);
        % Output layer
        x3 1 = bias(1,3)*weights(3,1) + x2(1)*weights(3,2) + x2(2)*weights(3,3);
        out(j) = sigma(x3 1);
        error out=0;
        mse iter=50;
        for I=1:mse iter
             error in=0;
             for j2=1:numIn
                 out mse=sigma(bias(1,1)*weights(1,1)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,1)*weights(1,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+input(j2,2)+in
input(j2,2)*weights(1,3));
                 error in=error in+output(j2)-out mse;
             error out=error out+error in*error in;
        mseo(i)=error out/mse iter;
        % Adjust delta values of weights
        % For output layer:
        % delta(wi) = xi*delta,
        % delta = (1-actual output)*(desired output - actual output)
        delta3 1 = out(i)*(1-out(i))*(output(i)-out(i));
        % Propagate the delta backwards into hidden layers
        delta2 1 = x2(1)*(1-x2(1))*weights(3,2)*delta3 1;
        delta2_2 = x2(2)*(1-x2(2))*weights(3,3)*delta3_1;
        % Add weight changes to original weights
        % And use the new weights to repeat process.
        % delta weight = coeff*x*delta
        for k = 1:3
            if k == 1 \% Bias cases
                weights(1,k) = weights(1,k) + coeff*bias(1,1)*delta2 1;
                weights(2,k) = weights(2,k) + coeff*bias(1,2)*delta2_2;
                weights(3,k) = weights(3,k) + coeff*bias(1,3)*delta3 1;
            else % When k=2 or 3 input cases to neurons
                weights(1,k) = weights(1,k) + coeff*input(j,1)*delta2 1;
                weights(2,k) = weights(2,k) + coeff*input(j,2)*delta2 2;
                weights(3,k) = weights(3,k) + coeff*x2(k-1)*delta3 1;
            end
        end
    end
end
%% Plotting in matlab
p = [0\ 0\ 1\ 1;\ 0\ 1\ 0\ 1];
t = [0 \ 1 \ 1 \ 0];
%plotpv(p,t);
%plotpc(weights,bias)
```

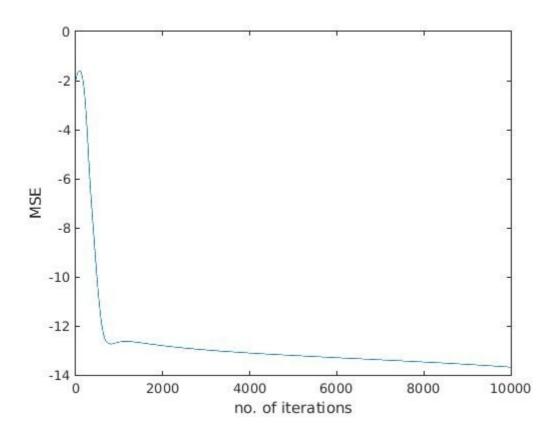
2) Three input xor gate

```
% multi layer perceptron
% XOR input for x1 and x2
input = [0\ 0\ 0;\ 0\ 1\ 0;\ 1\ 0\ 0;\ 1\ 1\ 0;\ 1\ 0\ 1;\ 1\ 1;\ 0\ 1\ 1;\ 0\ 0\ 1];
% Desired output of XOR
output = [0;1;1;0;0;1;0;1];
% Initialize the bias
bias = [-1 -1 -1 -1];
% Learning coefficient
coeff = 0.7;
% Number of learning iterations
iterations = 2000;
% Calculate weights randomly using seed.
%rand('state',sum(100*clock));
weightsb = -1 + 2.*rand(1,4);
weights11= -1 + 2.*rand(1,4);
weights12 = -1 + 2.*rand(1,4);
weights13 = -1 + 2.* rand(1,4);
weights21= -1 + 2.*rand(1,1);
weights22 = -1 + 2.* rand(1,1);
weights23= -1 + 2.*rand(1,1);
for i = 1:iterations
 out = zeros(4,1);
 numIn = length (input(:,1));
 for j = 1:numIn
   % Hidden layer
   H1 = bias(1,1)*weightsb(1,1)+input(j,1)*weights11(1,1)+
input(j,2)*weights11(1,2)+input(j,3)*weights11(1,3);
   % Send data through sigmoid function 1/1+e^-x
   % Note that sigma is a different m file
   % that I created to run this operation
   x2(1) = sigma(H1);
   H2 = bias(1,2)*weightsb(1,2) + input(j,1)*weights12(1,1) +
input(j,2)*weights12(1,2)+input(j,3)*weights12(1,3);
   x2(2) = sigma(H2);
   H3 = bias(1,3)*weightsb(1,3) + input(j,1)*weights13(1,1) +
input(j,2)*weights13(1,2)+input(j,3)*weights13(1,3);
   x2(3) = sigma(H3);
```

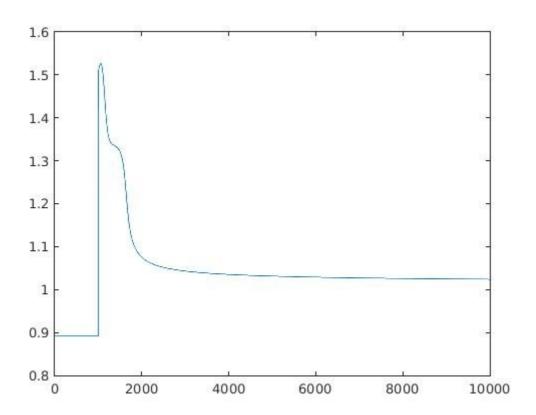
```
% Output layer
   x3 1 = bias(1,4)*weightsb(1,4) + x2(1)*weights21+x2(2)*weights22+x2(3)*weights23;
   out(j) = sigma(x3 1);
   error out=0;
   mse iter=50;
   %for I=1:mse_iter
      test=xor test(50);
     for I=1:50
       error in=0;
        out mse=sigma(bias(1,3)*weightsb(1,4)+test(1,1)*weights21+
test(I,2)*weights22+test(I,3)*weights23);
        error in=error in+output(j2)-out mse;
      error out=error out+error in*error in;
   end
   %mse(i)=error out/8;
   % Adjust delta values of weights
   % For output layer:
   % delta(wi) = xi*delta,
   % delta = (1-actual output)*(desired output - actual output)
   delta3 1 = out(j)*(1-out(j))*(output(j)-out(j));
   % Propagate the delta backwards into hidden layers
   delta2 1 = x2(1)*(1-x2(1))*weights21*delta3 1;
   delta2_2 = x2(2)*(1-x2(2))*weights22*delta3_1;
   delta2 3 = x2(3)*(1-x2(3))*weights23*delta3 1;
   % Add weight changes to original weights
   % And use the new weights to repeat process.
   % delta weight = coeff*x*delta
   % for k = 1:3
    % if k == 1 % Bias cases
       weightsb(1,1) = weightsb(1,1) + coeff*bias(1,1)*delta2 1;
       weightsb(1,2) = weightsb(1,2) + coeff*bias(1,2)*delta2 2;
       weightsb(1,3) = weightsb(1,3) + coeff*bias(1,3)*delta2 3;
       weightsb(1,4) = weightsb(1,4) + coeff*bias(1,4)*delta3_1;
    % else % When k=2 or 3 input cases to neurons
       for k=1:4
       weights11(1,k) = weights11(1,k) + coeff*input(j,1)*delta2 1;
       weights12(1,k) = weights12(1,k) + coeff*input(j,2)*delta2^{-}2;
       weights13(1,k) = weights13(1,k) + coeff*input(j,3)*delta2 3;
       weights21= weights21 + coeff*x2(1)*delta3 1;
       weights22 = weights22 + coeff*x2(2)*delta3 1;
       weights23 = weights23 + coeff*x2(3)*delta3 1;
      % weights14(4,k) = weights11(3,k) + coeff*bias(1,3)*delta3 1;
     %end
   % end
 end
```

end

OUTPUT:



1) For two input xor gate



2) For three input xor gate

