

1b2.

```
>> %Get the time from Part a
>> time = [tout(1)];
>> for i = 1:10
time = [time; tout(i*5 + 1)];
end
>> time = time'
time =
0 1.0000 2.0000 3.0000 4.0000 5.0000 6.0000 7.0000 8.0000 9.0000
10.0000

>> %Get the output from Part a
>> x = [xout(1,:)];
>> for i = 1:10
x = [x; xout(i*5 + 1,1) xout(i*5 + 1,2)];
end
>> x = x'
x =
-1.0000 -0.0144 0.2457 0.1804 0.0652 -0.0133 -0.0361 -0.0236 -0.0050 0.0047
0.0055
1.0000 1.5969 1.3539 1.0824 0.9581 0.9379 0.9653 0.9961 1.0100 1.0093
1.0036

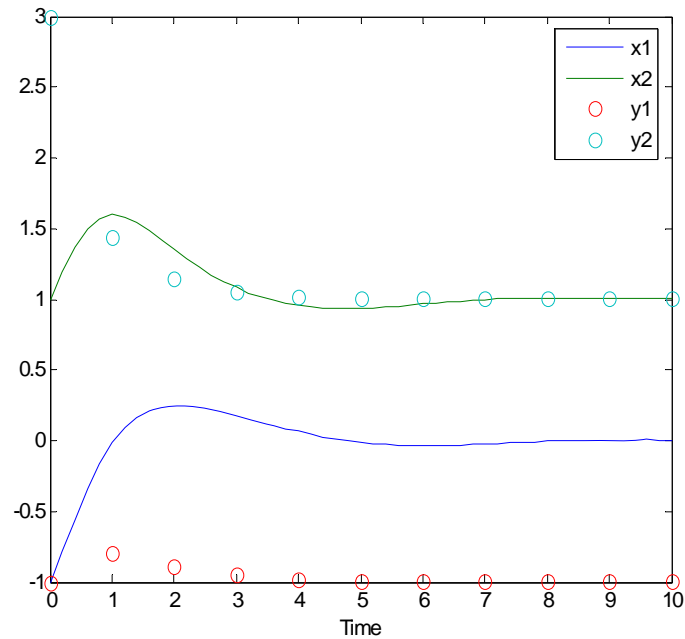
>> I = eye(2);
>> E = [0 1; -1 0];
>> x0 = [-1;1];
>> %Assumptions: One layer, transfer function = purelin (derivate = 1),
&
>> %learning rate = 0.1

>> %Backpropagation algorithm to reduce mean square error from textbook
>> %Let w1[0] = 10 and w2[0] = 10
>> w = [10;10];
>> weights = [];
>> out = [];
>> for t = 0:1:10
y = (2*I - ( inv( I + (exp(w(1)*t)*E) ) * inv( I + (exp(w(2)*t)*E) ) ))
* x0;
out = [out y];
error = x(:,1) - y;
s = -2 * [1 0; 0 1] * error;
w = w - .1*s*(t)';
weights = [weights w];
end

>> w
w =
21.0000
-1.0000

1b3.
>> out = [];
>> for t = 0:1:10
y = (2*I - ( ( I + inv(exp(w(1)*t)*E) ) * inv( I + (exp(w(2)*t)*E) ) ))
* x0;
out = [out y];
end
```

```
>> plot(tout,xout,time,out,'o')
>> xlabel('Time')
```



```
>> sum = 0;
>> for t = 1:11
sum = sum + (x(t)-out(t))^2;
end
>> mean_square_error = sum/22
mean_square_error =
    0.4227
```

>> %y(t) is represented by 'o' in the diagram. They do not match up with x(t) but demonstrate a similar curvature. As $t \rightarrow \infty$, y(t) goes towards the initial conditions over time. With the weight of $y1 = 21$ and the weight of $y2 = -1$, the mean square error is 0.4227.