```
1 import numpy as np
 2 import math
 3 import matplotlib.pyplot as plt
 5 np.random.seed(2)
 6 \times = \text{np.random.uniform(size=}(1,300))
 7 x = x - np.mean(x)
 8 v = np.random.uniform(low = -1/10, high = 1/10, size=(
   1,300))
 9 d_{true} = np.sin(20*x) + 3*x + v
10
11 plt.scatter(x,d_true)
12 plt.show()
13 # 1 input
14 # N=24 hidden neurons and 1 output neuron
15 # 3N+1 weights
16 # actFun final = tanh v
17 step = 4*10**(-2)
18 N = 24
19 \text{ corr} = 0
20 np.random.seed(3)
21 bias_hidden = np.random.uniform(size=(1,N))
22 wIn = np.random.uniform(size=(1,N))
23 wOut = np.random.uniform(size=(1,N))
24 bias_out = np.random.uniform()
25 \text{ miscl} = 300
26 MSE=[100]
27 corr=0
28 epoch=0
29
30 while (MSE[-1]>0.02):
31
       epoch+=1
32
       errors=[]
33
       res=[]
       for el,ris in zip(x.T,d_true[0]):
34
35
           #forward
           a = (el * wIn) + bias_hidden
36
37
           d = np.tanh(a)
38
           d w = d * wOut
39
           b = d_w.sum() + bias_out
40
           f x = b
```

```
File - C:\Users\davib\Dropbox (Politecnico Di Torino Studenti)\PC\Desktop\UIC\Neural Networks\Scripts\04-6629655130-Bartole
41
             res.append(f_x)
42
43
             #backpropagation
             diff = (ris - f_x)
44
             wOut = wOut + step * (2 * diff) * (d)
45
46
             wIn = wIn + step * (2 * diff) * wOut * (1 - (
    d**2)) * el
47
             bias_hidden = bias_hidden + step * (2 * diff
    ) * wOut * (1 - (d**2))
48
             bias_out = bias_out + step * (2 * diff)
49
             errors.append(diff**2)
50
        MSE.append(sum(errors)/len(errors))
51
52
        if (MSE[-1]>MSE[-2]):
53
54
             step = step - 0.1*step
55
56 plt.scatter(x,d_true)
57 plt.scatter(x,res)
58 plt.legend(['true','learned'])
59 plt.show()
60 plt.plot([i for i in range(epoch)], MSE[1:])
61 plt.title('MSE vs number of epochs')
62 plt.show()
```