

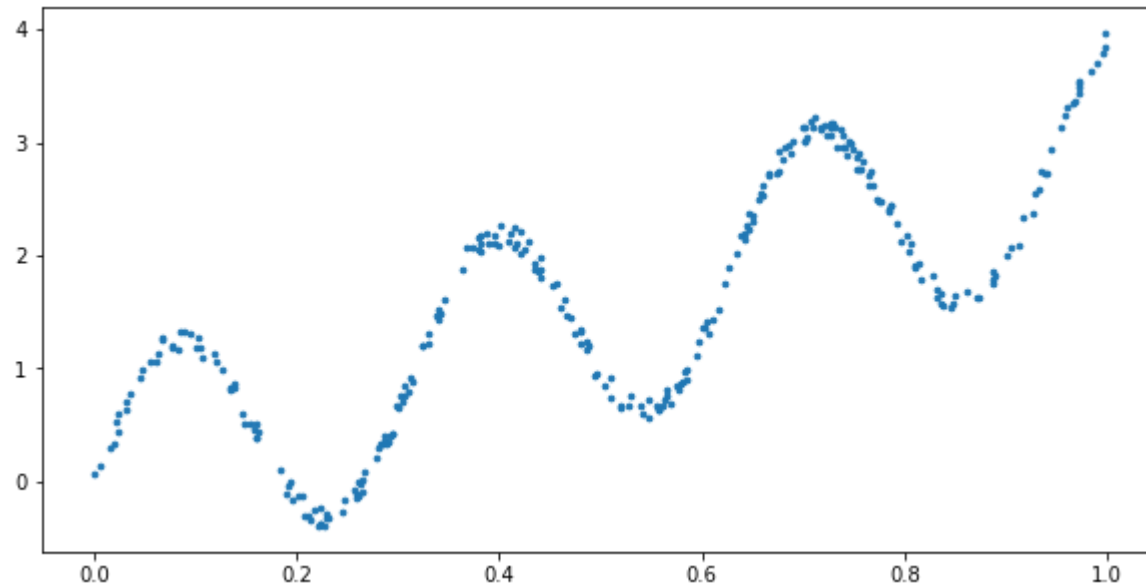
```
In [1]: import numpy as np
        from math import *
        import matplotlib.pyplot as plt
```

Q 1 and 2

```
In [8]: # Inputs
        n= 300
        x = [np.random.uniform(low =0, high = 1) for i in range(n)]
        # Vector Fields
        v = [np.random.uniform(low =-0.1, high = 0.1) for i in range(n)]
        #v
        mean = np.mean(x)
        std  = np.std(x)
        x_stand = []
        for i in range(len(x)):
            x_stand.append([1,(x[i]-mean)/std])
```

Q 3

```
In [9]: d = []  
for i in range(0,n):  
    d.append(sin(20*x[i]) + 3*x[i] + v[i])  
d = np.asarray(d)  
# d  
fig, ax = plt.subplots(figsize = (10,5))  
x = np.asarray(x)  
#print(x)  
plt.scatter(x,d, s = 7)  
plt.show()
```



```

In [250]: import numpy as np
          np.random.seed(100)

          # hyperbolic Activation Function
          def af(t):
              return tanh(t)

          # Derivative of tanh(v)
          def af_derivative(x):
              return 1-(tanh(x)**2)

          def mse(x, d, w, nLayers, nNodes):
              c = 0
              n = len(x)
              for i in range(0,n):
                  prev = x[i]
                  for j in range(nLayers):
                      u = []
                      p = []
                      for k in range(nNodes):
                          t = np.dot(w[j]['weights'][k], prev)
                          u.append(t)
                      for l in u:
                          p.append(af(l))
                      p.append(1)
                      prev = np.asarray(p)
                      #print(w[nLayers]["weights"][0])
                      y = np.dot(w[nLayers]["weights"][0], prev)
                      c += (d[i] - y)**2
              return c/n

          #np.random.seed(100)
          class NeuralNetwork:

              def __init__(self,x=[],y=[],numLayers=2,numNodes=2, numOutputs = 1, eta=0.001,maxIter=10000, ep = 0):
                  self.labels = y
                  self.nLayers = numLayers
                  self.nNodes = numNodes
                  self.numOutputs = numOutputs
                  self.eta = eta
                  self.temp_eta = eta
                  self.maxIt = maxIter

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        self.ep = ep
        #self.train()
        #self.g = len(self.data[0])*numLayers + numLayers*numNodes + numNodes*numOutputs
        #    newdata = []
        #    for i in range(len(x)):
        #        newdata.append(np.append(x[i],1))
        self.data = x
        self.weights = [{"weights":np.random.uniform(low=-2, high = 2, size = (self.nNodes, len(self.data[0
])))}}]
        for i in range(self.nLayers):
            self.weights.append({"weights":np.random.uniform(low=-2, high = 2, size = (self.nNodes,self.nNodes+1))})
        if self.nLayers >= 0:
            self.weights.append({"weights":np.random.uniform(low=-2, high = 2, size = (self.numOutputs,self.nNodes+1))})
        self.temp_weights = self.weights

    def train(self):
        print(self.weights)
        temp_weights = self.weights
        temp_eta = self.temp_eta
        temp2 = mse(self.data, self.labels, self.weights, self.nLayers+1, self.nNodes)
        print(temp2)
        e = 0
        obj = []
        epoch = []
        cos = 1000000000
        while cos >= self.ep and e < self.maxIt:
            prev = cos
            for i in range(len(self.data)):
                self.backprop(self.labels[i], self.data[i])
                for m in range(len(self.weights)):
                    for j in range(len(self.weights[m]['weights'])):
                        for k in range(len(self.weights[m]['weights'][j])):
                            self.weights[m]['weights'][j][k] -= self.eta *self.weights[m]['g'][j][k]

            cos = mse(self.data, self.labels, self.weights, self.nLayers+1,self.nNodes)
            #        epoch.append(e)
            #        obj.append(cos)
            #        e += 1

            if cos >= prev:

```

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        self.eta = 0.9*self.eta
        obj = []
        epoch = []
        e = 0
        obj.append(cos)
        epoch.append(e)
        cos = 100000000
        self.weights = temp_weights
        print(temp_weights)

        if self.eta <= 0.00001:
            self.eta = temp_eta
            obj = []
            epoch = []
            e = 0
            cos = 100000000
            self.weights = [{"weights":np.random.uniform(low =-2, high = 2, size = (self.nNodes, len(
self.data[0])))}]}
            for i in range(self.nLayers):
                self.weights.append({"weights":np.random.uniform(low =-2, high = 2, size = (self.nNodes, self.nNodes+1))})
            if self.nLayers >= 0:
                self.weights.append({"weights":np.random.uniform(low =-2, high = 2, size = (self.nNodes, self.nNodes+1))})

            elif cos < prev:
                epoch.append(e)
                obj.append(cos)
                e += 1
        return self.weights, obj, epoch

def feedforward(self,x=[]):
    prev = x
    r = []
    r.append(prev)
    t = []
    for j in range(self.nLayers+1):
        l = []
        s = []
        for m in range(self.nNodes):
            #print(self.weights[j]["weights"][m])
            s.append(np.matmul(self.weights[j]["weights"][m], prev))

```

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        for k in s:
            l.append(af(k))
        l.append(1)
        prev = np.asarray(1)
        t.append(s)
        r.append(1)

    s = []
    p = np.matmul(self.weights[self.nLayers+1]["weights"][0], 1)
    s.append(p)
    t.append(s)
    return t, r, p

def backprop(self, d, data):
    der = []
    #t, r = self.feedforward(data)
    # print('\n')
    # print("t = " + str(t))
    # print('\n')
    # print('r = ' + str(r))
    t, r, q = self.feedforward(data)

    for i in range(self.nLayers+1):
        a = []
        for m in range(self.nNodes):
            a.append(af_derivative(t[i][m]))
        der.append(a)

    diff = []
    s = []
    s.append(af_derivative(t[self.nLayers+1][0]))
    diff.append(d - q)
    der.append(s)

    for i in reversed(range(len(self.weights))):
        layer = self.weights[i]
        errors = []
        if i != len(self.weights)-1:
            for j in range(len(layer['weights'])):
                error = 0

                for k in range(len(self.weights[i + 1]['weights'])):
                    error += (self.weights[i + 1]['weights'][k][j] * self.weights[i + 1]['s'][k])

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        errors.append(error)
    else:
        errors.append(diff[0])
    layer['s'] = []
    for j in range(len(layer['weights'])):
        layer['s'].append(errors[j]*der[i][j])

# Finding the corresponding gradient vector

    for j in range(len(self.weights)):
        layer = self.weights[j]
        layer['g'] = []
        for k in range(len(self.weights[j]['weights'])):
            s = []
            for m in range(len(self.weights[j]['weights'][k])):
                s.append(((-(r[j][m])*self.weights[j]['s'][k])*2)/len(self.data))
            layer['g'].append(s)

    return 0.0
```

```

In [252]: nHiddenLayers = 0
nNodes = 24
nOut = 1
# weights = weights_initialization(x,nLayers,nNodes, nOut)
eta = 10
iter = 10000
ep = 0.005

w, obj ,epoch = NeuralNetwork(x_stand,d, nHiddenLayers, nNodes, nOut, eta, iter, ep).train()

[{'weights': array([[ 0.17361977, -0.88652246],
                    [-0.30192964,  1.37910453],
                    [-1.98112458, -1.51372352],
                    [ 0.68299634,  1.30341102],
                    [-1.45317364,  0.30037332],
                    [ 1.56528782, -1.16319151],
                    [-1.25868712, -1.56649244],
                    [-1.12121003,  1.91449514],
                    [ 1.2467326 , -1.31223595],
                    [ 1.26489899, -0.90370501],
                    [-0.27318327,  1.76011928],
                    [ 1.27059752, -0.6555522 ],
                    [-1.29835819, -0.50867181],
                    [-1.97724597, -0.99029459],
                    [ 1.18265003, -1.93898012],
                    [ 0.39537351,  0.41521816],
                    [-1.57940926, -0.47222622],
                    [-1.85409577,  1.56164625],
                    [ 1.92368343, -1.76023204],
                    [ 1.56218378,  0.307606  ],
                    [ 0.96991876,  0.52073575],
                    [ 0.32736877, -1.91824347],
                    [-1.15989369,  0.17873951],
                    [ 1.07646068, -0.99721908]])}, {'weights': array([[ -0.85641724,  1.40958035,  1.90002597,  1.53941317,
-0.56196862,
                    0.39543578, -0.58081755, -0.63923914, -1.28767604, -1.04922317,
                    -1.82055087,  0.02172572, -0.49499018,  0.3712216 ,  0.5197675 ,
                    -1.42959874,  1.7353652 ,  1.78551952,  0.40918663, -0.44893488,
                    -0.54724798, -1.18261889, -0.89293975, -1.01385648, -1.30556799]])}]
57.28091217862603

```



```
In [253]: len(epoch)
```

```
Out[253]: 4006
```

```
In [254]: cos = mse(x_stand, d, w, 1,24)
cos
```

```
Out[254]: 0.0049992814326523475
```

Q 4

```
In [255]: def feedforward(data, weights, nLayers, nNodes):
            n = []
            for i in range(len(data)):
                #print(i)
                prev = data[i]

                for j in range(nLayers):
                    l = []
                    s = []
                    for m in range(nNodes):
                        s.append(np.dot(weights[j]["weights"][m], prev))

                    for k in s:
                        l.append(tanh(k))
                    prev = np.asarray(l)

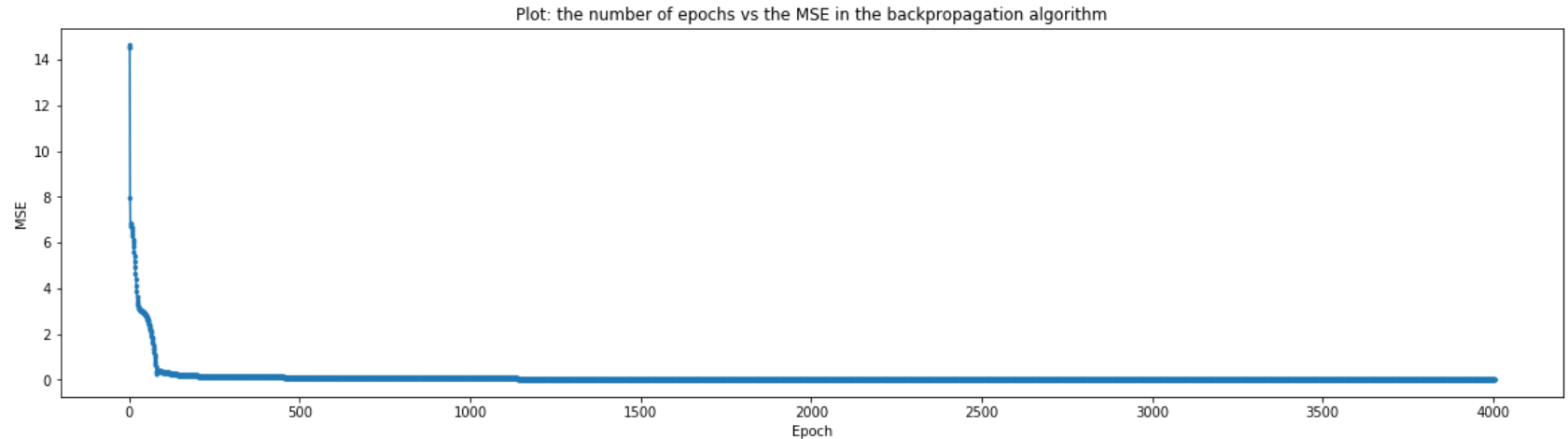
                l.append(1)
                p = np.dot(weights[nLayers]["weights"][0], l)
                #print(p)

                n.append(p)
            return n
```

```
In [256]: y = feedforward(x_stand, w, 1, 24)
```

Plot: the number of epochs vs the MSE in the backpropagation algorithm.

```
In [257]: fig, ax = plt.subplots(figsize = (20,5))  
plt.scatter(epoch,obj, s = 7)  
plt.plot(epoch,obj)  
plt.xlabel("Epoch")  
plt.ylabel("MSE")  
plt.title('Plot: the number of epochs vs the MSE in the backpropagation algorithm')  
plt.show()
```



Q 5

```
In [258]: fig, ax = plt.subplots(figsize = (15,5))  
plt.scatter(x,d, s = 7)  
plt.scatter(x,y, c = 'r', s = 15)  
plt.xlabel("x")  
plt.show()
```

