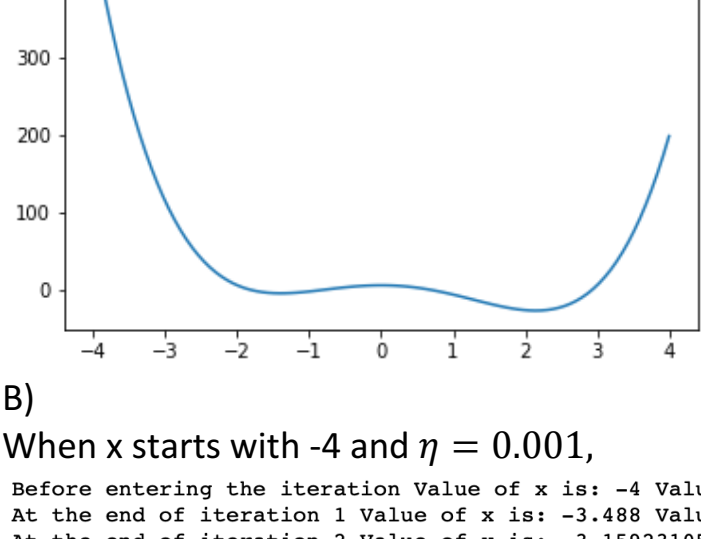


Assignment 4

1)  
A)  
Answer:  
Value of x at local minimum = -1.397180  
Value of x at global minimum = 2.147180



B)  
When x starts with -4 and  $\eta = 0.001$ ,  
Before entering the iteration Value of x is: -4 Value of f is: 454  
At the end of iteration 1 Value of x is: -3.488 Value of f is: 240.907412201472  
At the end of iteration 2 Value of x is: -2.159231053824 Value of f is: 148.52441854620668  
At the end of iteration 3 Value of x is: -2.922916422502639 Value of f is: 99.4029877988203  
At the end of iteration 4 Value of x is: -2.7420316758639505 Value of f is: 70.07121494417243  
At the end of iteration 5 Value of x is: -2.5977950740777596 Value of f is: 51.16573699678769  
At the end of iteration 6 Value of x is: -2.479400344271616 Value of f is: 38.296442311327496

After running 1200 iterations,  
Before entering the iteration Value of x is: -4 Value of f is: 454  
At the end of iteration 1 Value of x is: -3.488 Value of f is: 240.907412201472  
At the end of iteration 2 Value of x is: -3.159231053824 Value of f is: 148.52441854620668  
At the end of iteration 3 Value of x is: -2.922916422502639 Value of f is: 99.4029877988203  
At the end of iteration 4 Value of x is: -2.7420316758639505 Value of f is: 70.07121494417243  
At the end of iteration 5 Value of x is: -2.5977950740777596 Value of f is: 51.16573699678769  
At the end of iteration 6 Value of x is: -2.479400344271616 Value of f is: 38.296442311327496  
At the end of iteration 1195 Value of x is: -1.3971808598447308 Value of f is: -4.34895772410303  
At the end of iteration 1196 Value of x is: -1.3971808598447308 Value of f is: -4.34895772410303  
At the end of iteration 1197 Value of x is: -1.3971808598447308 Value of f is: -4.34895772410303  
At the end of iteration 1198 Value of x is: -1.3971808598447308 Value of f is: -4.34895772410303  
At the end of iteration 1199 Value of x is: -1.3971808598447308 Value of f is: -4.34895772410303  
At the end of iteration 1200 Value of x is: -1.3971808598447308 Value of f is: -4.34895772410303

The value of x has converged to the local minimum.

C)  
When x starts with 4 and we do 1200 iterations  
Before entering the iteration Value of x is: 4 Value of f is: 198  
At the end of iteration 1 Value of x is: 3.68 Value of f is: 110.61233152  
At the end of iteration 2 Value of x is: 3.450896164 Value of f is: 94.53629857986431  
At the end of iteration 3 Value of x is: 3.276396901609702 Value of f is: 37.31076130762675  
At the end of iteration 4 Value of x is: 3.138067975365072 Value of f is: 19.971643395808066  
At the end of iteration 5 Value of x is: 3.0222601730040535 Value of f is: 8.32266113072949  
At the end of iteration 6 Value of x is: 2.932268937523244 Value of f is: 0.17057478693808548  
At the end of iteration 1195 Value of x is: 2.1471808598447315 Value of f is: -26.611979775899698  
At the end of iteration 1196 Value of x is: 2.1471808598447315 Value of f is: -26.611979775899698  
At the end of iteration 1197 Value of x is: 2.1471808598447315 Value of f is: -26.611979775899698  
At the end of iteration 1198 Value of x is: 2.1471808598447315 Value of f is: -26.611979775899698  
At the end of iteration 1199 Value of x is: 2.1471808598447315 Value of f is: -26.611979775899698  
At the end of iteration 1200 Value of x is: 2.1471808598447315 Value of f is: -26.611979775899698

D)  
When x starts with -4 and  $\eta = 0.01$   
Before entering the iteration Value of x is: -4 Value of f is: 454  
At the end of iteration 1 Value of x is: 1.12 Value of f is: -8.71561728  
At the end of iteration 2 Value of x is: 1.35146997 Value of f is: -14.187220487602376  
At the end of iteration 3 Value of x is: 1.981299140655707 Value of f is: -19.554256180837104  
At the end of iteration 4 Value of x is: 1.800149502820235 Value of f is: -23.55150883046352  
At the end of iteration 5 Value of x is: 1.959648783032446 Value of f is: -25.464284221395847  
At the end of iteration 6 Value of x is: 2.0585082124451446 Value of f is: -26.3830081197323108  
At the end of iteration 1195 Value of x is: 2.147180859844728 Value of f is: -26.61197977589969  
At the end of iteration 1196 Value of x is: 2.147180859844728 Value of f is: -26.61197977589969  
At the end of iteration 1197 Value of x is: 2.147180859844728 Value of f is: -26.61197977589969  
At the end of iteration 1198 Value of x is: 2.147180859844728 Value of f is: -26.61197977589969  
At the end of iteration 1199 Value of x is: 2.147180859844728 Value of f is: -26.61197977589969  
At the end of iteration 1200 Value of x is: 2.147180859844728 Value of f is: -26.61197977589969

We see that x has converged to global minimum

E)  
We can see that x values bounces all over the place never to converge at a single point. This is because the learning rate is too high  
Before entering the iteration Value of x is: -4 Value of f is: 454  
At the end of iteration 1 Value of x is: 47.2 Value of f is: 9699505.955200002  
At the end of iteration 2 Value of x is: -82626.05440000004 Value of f is: 9.32187574662132e+19  
At the end of iteration 3 Value of x is: 451278842347294.25 Value of f is: 8.294875771953886e+58  
At the end of iteration 4 Value of x is: -7.35232383267769e+43 Value of f is: 5.846261165795431e+175  
At the end of iteration 5 Value of x is: 3.179542992304716e+131 Value of f is: nan  
At the end of iteration 6 Value of x is: -inf Value of f is: nan

2)  
A)  
The pseudo code implements stochastic gradient descent (SGD)

To complete one epoch there should be 500 iterations,  
To complete 100 epochs there should be 100 \* 500 =  
50000 iterations

B)

i)

$$\Delta V_{23} = -\eta \frac{d}{dV_{23}} \left[ \frac{1}{2} (x_1 - y_1)^2 + 7(x_2 - y_2)^2 \right]$$
$$\Delta V_{23} = -\eta \left[ 7(x_2 - y_2) \left( -\frac{dy_2}{dV_{23}} \right) \right]$$
$$= -\eta \left[ 7(x_2 - y_2) \frac{d}{dV_{23}} (v^T z_3 + v_0) \right]$$
$$\Delta V_{23} = 7\eta (x_2 - y_2) z_3$$

ii)

$$\Delta W_{hj} = -\eta \frac{\partial E}{\partial W_{hj}}$$
$$= -\eta \sum \frac{\partial E}{\partial y_i^+} \frac{dy_i^+}{dz_h^+} \frac{\partial z_h^+}{\partial W_{hj}}$$
$$\frac{\partial z_h^+}{\partial W_{hj}} = z_h^+ (1 - z_h^+) x_j^+$$
$$\frac{dy_i^+}{dz_h^+} = V_h$$
$$\frac{\partial E}{\partial y_i^+} = \frac{d}{dy_i^+} \left[ \frac{1}{2} \left[ 3(x_1 - y_1)^2 + 7(x_2 - y_2)^2 \right] \right]$$
$$\frac{\partial E}{\partial y_1^+} = -3(x_1 - y_1) \quad \frac{\partial E}{\partial y_2^+} = -7(x_2 - y_2)$$
$$\Delta W_{hj} = -\eta \left[ \left( \frac{\partial E}{\partial y_1^+} \frac{dy_1^+}{dz_h^+} \frac{\partial z_h^+}{\partial W_{hj}} \right) + \left( \frac{\partial E}{\partial y_2^+} \frac{dy_2^+}{dz_h^+} \frac{\partial z_h^+}{\partial W_{hj}} \right) \right] x_j^+$$
$$= -\eta \left[ \left[ -3(x_1 - y_1) V_h z_h^+ (1 - z_h^+) \right] + \left[ -7(x_2 - y_2) V_h z_h^+ (1 - z_h^+) \right] \right] x_j^+$$
$$= \eta \left[ \left[ 3(x_1 - y_1) V_h z_h^+ (1 - z_h^+) \right] + \left[ 7(x_2 - y_2) V_h z_h^+ (1 - z_h^+) \right] \right] x_j^+$$

3)  
A)  
Neural Net CB, Neural NetRZeroOne and NeuralNetCK cannot produce negative output  
So only NeuralNetRK can produce negative outcome since it uses a linear function

B)  
NeuralNetCK is the only one Neural Net which can ensure the sum of output to be 1.

C)  
Neural NetCK is the best option here, since the sum of p1,p2,p3 which are the output is one. For this case where k=3 and p1,p2,p3 are the probabilities of class cat, face and tree which is similar to the output of neural net CK

D)  
Neural NetRZeroOne is the best option here because of the following reasons,  
i) The output is always 0 or 1  
ii) All classification algorithms return probabilities we only need 0 or 1

4)  
A)  
We can explain this with an example, Suppose almond = 1, anise =2, creostate = 3 and fishy = 4 then the neural net model would consider or treat fishy as a value which is double of anise which is not true, this can lead to a wrong answer, whereas in the case of random forest. Since neural network is an algorithm that learns from the weight of a variable its not ok.  
For our example,  
A simplified random forest algorithm can be,  
If odor =1 do  
Almond stuff  
Else if odor ==2 do  
Anise stuff.....

Here there is no problem but when you consider an algorithm which learns by using weight as a parameter like  
Prediction = weight \* odor+.....  
The weight matters, so its not correct to use label encoding but its useful to use one hot encoding.

B)  
i)  
Transformed Data Set (with optimal way of hot encoding)

	Z1	Z2	Z3	Z4	Z5	Label
X1	0	0	0	1	1	0
X2	0	0	1	0	0	0

Where z1 = 1 if odor = almond, z2 = 1 if odor = anise, z3 =1 of odor = fishy and z5 =1 if stalkshape= tapering.

Transformed Data Set (with hot encoding on every attribute)

	Z1	Z2	Z3	Z4	Z5	Z6	Label
X1	0	0	0	1	1	0	0
X2	0	0	1	0	0	1	0

ii) On using one hot encoding there is a problem of losing the hierarchy or order ie high > medium > low. So its better to use label encoding which retains the order.

iii) It is fine to use (0,1) instead of one hot encoding because it produces or represents the same dataset.  
For example consider a dataset of coin flips

	Results
X1	Head
X2	Tail
X3	Head

On using just (1,0) with 1 = head and 0 = tail, we get

	Result
X1	1
X2	0
X3	1

And now applying one hot encoding we generate two variables namely d1 and d2 where d1 = 1 when result is head and d2 =1 when result is tail.

	D1	D2
X1	1	0
X2	0	1
X3	1	0

Here the column d2 is redundant and can be eliminated because only the column d1 is enough to represent the dataset. Since X1 and X3 have the same value and only X2 is different its is mandatory for X2 to have the value which is not X1 or X3. But when we consider three valued attributes the value of X2 can be either of the remaining two values not chosen by X1 or X3.

So in the case of three valued attributes one hot encoding is compulsory.