

In [24]:

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#Q1
import numpy as np
import matplotlib.pyplot as plt

def f(x):
    w = np.array([1, -1, -12, 15, 5])
    M = np.size(w) - 1
    return np.sum([x**i*w[M-i] for i in range(0, M+1)], axis=0)

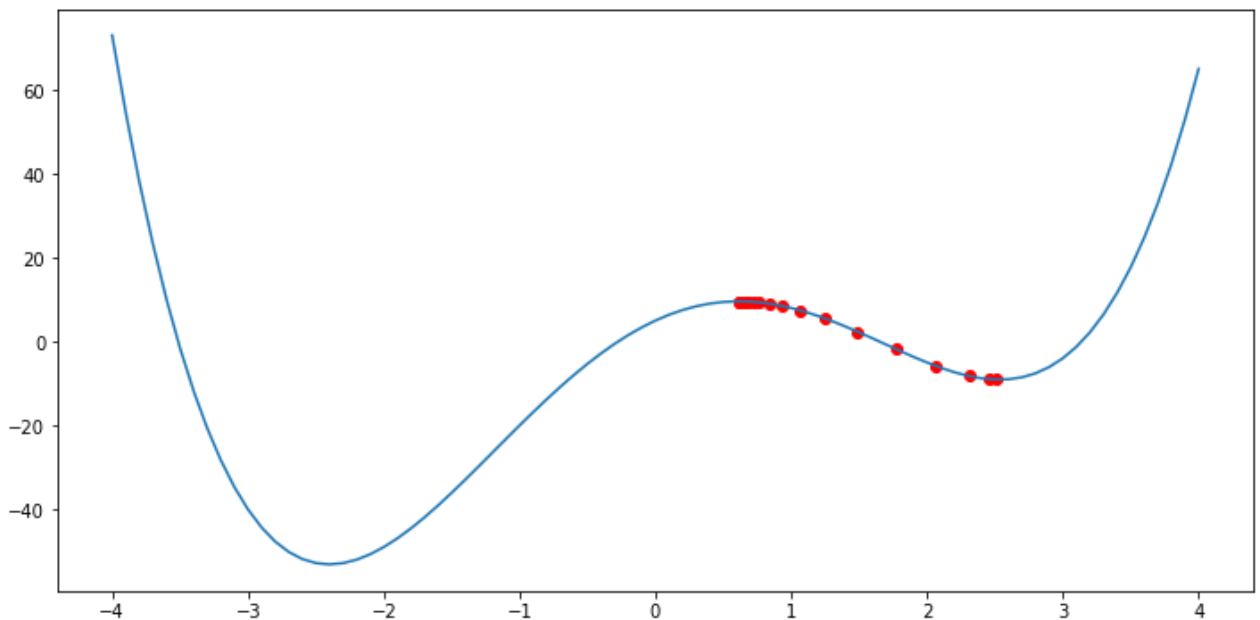
def g(x):
    w = np.array([1, -1, -12, 15, 5])
    M = np.size(w) - 1
    return np.sum([i*x**(i-1)*w[M-i] for i in range(0, M+1)], axis=0)

alpha= 0.02
x = 0.62
x_hist= np.array(x)
fx_hist= np.array(f(x))
for i in range(20):
    x = x - alpha*g(x)
    x_hist = np.append(x_hist, x)
    fx_hist= np.append(fx_hist, f(x))

print('x = ',x,'f(x)=',f(x))
fig = plt.figure(figsize=(12,6))
ax = plt.subplot (1,1,1)
delta = 0.1
x_ = np.arange(-4, 4+ delta, delta)
ax.plot(x_, f(x_))
#ax.plot(x, g(x))
ax.scatter(x_hist, fx_hist, c='r')
plt.show()

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x = 2.5104174088324025 f(x)= -9.073558171240812



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In [26]: #finding a root close to x0
from scipy.optimize import fsolve
from scipy.optimize import minimize
x0=0.7
root=fsolve(g,x0)#gradient is zero at this point
print(root)
#using scipy to find the minimum
minimum=minimize(f,x0)
print(minimum)
```

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[0.61654501]
      fun: -9.083837308515939
     hess_inv: array([[0.02625738]])
        jac: array([-7.62939453e-06])
    message: 'Optimization terminated successfully.'
       nfev: 24
        nit: 3
       njev: 8
      status: 0
     success: True
         x: array([2.53385792])
```

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In [27]: #Q2
import numpy as np
import tensorflow as tf
from tensorflow import keras
import matplotlib.pyplot as plt
from tensorflow.keras.datasets import cifar10 , mnist

(x_train ,y_train), (x_test,y_test) = cifar10.load_data( )
# ( x_train , y_train ) , ( x_test , y_test ) = mnist . load_data ( )
print ( " x_train => ", x_train.shape)
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Ntr = x_train.shape[0]
Nte = x_test.shape[0]
Din = 3072 # CIFAR10
# Din = 784 # MINIST
x_train = x_train[range(Ntr) , :]
x_test = x_test[range(Nte) , :]
y_train = y_train[range(Ntr)]
y_test = y_test[range(Nte)]

K=len(np.unique(y_train))

y_train = tf.keras.utils.to_categorical(y_train,num_classes=K)
y_test = tf.keras.utils.to_categorical(y_test,num_classes=K)

x_train = np.reshape(x_train,(Ntr,Din))
x_test = np.reshape(x_test,(Nte,Din))
x_train = x_train.astype(np.float32)
x_test = x_test.astype(np.float32)
x_train/=255.
x_test/=255.
```

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x_train => (50000, 32, 32, 3)
```

```
In [28]: # Utility function for displaying
def display(y_train, y_test, y_train_pred, y_test_pred, loss_history, w, showim = True)
    plt.plot(loss_history)
```

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# For displaying the weights matrix w as an image. 32*32*3 assumption is there
if showim:
    f, axarr = plt.subplots(2, 5)
    f.set_size_inches(16, 6)
    for i in range(10):
        img = w[:, i].reshape(32, 32, 3) # CIFAR10
        # img = w1[:, i].reshape(28, 28) # MNIST
        img = (img - np.amin(img))/(np.amax(img) - np.amin(img))
        axarr[i//5, i%5].imshow(img)
    plt.show()

train_acc = np.mean(np.abs(np.argmax(y_train, axis=1) == np.argmax(y_train_pred, axis=1)))
print("train_acc = ", train_acc)

test_acc = np.mean(np.abs(np.argmax(y_test, axis=1) == np.argmax(y_test_pred, axis=1)))
print("test_acc = ", test_acc)

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In [34]:

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std = 1e-5
w = std*np.random.randn(Din, K)
b = np.zeros(K)
lr = 1e-3
lr_decay=0.1
epochs = 11
batch_size = 100
loss_hist = []
rng = np.random.default_rng(seed = 0)

for e in range(epochs):
    indices = np.arange(Ntr)
    rng.shuffle(indices)

    for batch in range(Ntr//batch_size):
        batch_indices = indices[batch*batch_size:(batch+1)*batch_size]
        x = x_train[batch_indices]
        y = y_train[batch_indices]

        #forward pass
        y_pred = x@w + b
        loss = 1./batch_size*np.square(y_pred-y).sum()
        loss_hist.append(loss)

        #backward pass
        dy_pred = 1./batch_size*2.0*(y_pred - y)
        dw = x.T @ dy_pred
        db = dy_pred.sum(axis = 0)*1
        w = w - lr*dw
        b = b - lr*db

    if e % 5 == 0:
        print("Iteration %d / %d: loss %f"%(e, epochs,loss))
    if e % 10 == 0:
        lr *= lr_decay

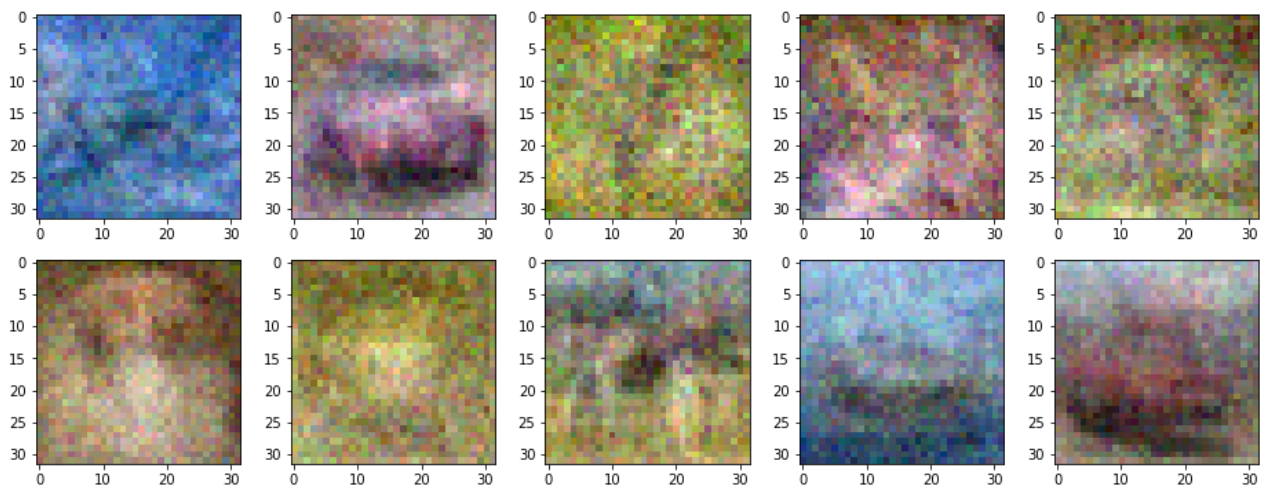
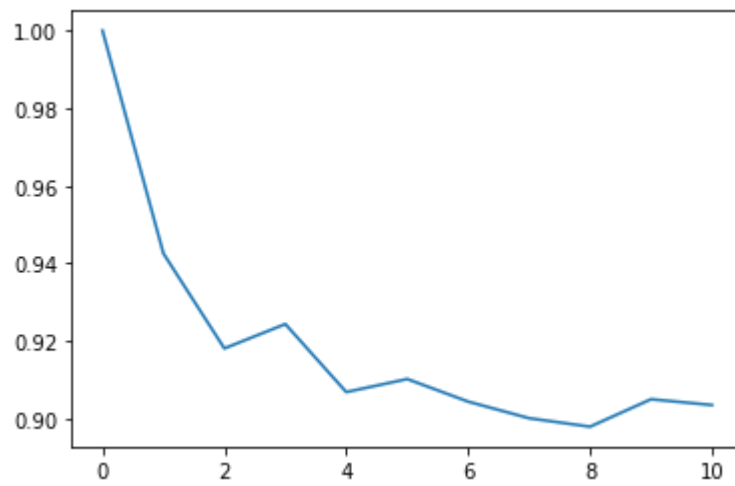
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Iteration 0 / 11: loss 1.000111
Iteration 5 / 11: loss 0.910138
Iteration 10 / 11: loss 0.903427

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In [35]: y_train_pred=x_train.dot(w)+b  
y_test_pred=x_test.dot(w)+b  
display(y_train,y_test,y_train_pred,y_test_pred,loss_hist,w,showim = True)
```



```
train_acc = 0.10528  
test_acc = 0.104
```

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In [9]:
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