

IMPORTING LIBRARIES

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
```

GENERATING 200 POINTS UNIFORMLY SPACED IN THE INTERVAL [-1,1]

```
In [2]: step = (2/200)
X = np.arange(-1,1,step)
X
```

```
Out[2]: array([-1.0000000e+00, -9.9000000e-01, -9.8000000e-01, -9.7000000e-01,
-9.6000000e-01, -9.5000000e-01, -9.4000000e-01, -9.3000000e-01,
-9.2000000e-01, -9.1000000e-01, -9.0000000e-01, -8.9000000e-01,
-8.8000000e-01, -8.7000000e-01, -8.6000000e-01, -8.5000000e-01,
-8.4000000e-01, -8.3000000e-01, -8.2000000e-01, -8.1000000e-01,
-8.0000000e-01, -7.9000000e-01, -7.8000000e-01, -7.7000000e-01,
-7.6000000e-01, -7.5000000e-01, -7.4000000e-01, -7.3000000e-01,
-7.2000000e-01, -7.1000000e-01, -7.0000000e-01, -6.9000000e-01,
-6.8000000e-01, -6.7000000e-01, -6.6000000e-01, -6.5000000e-01,
-6.4000000e-01, -6.3000000e-01, -6.2000000e-01, -6.1000000e-01,
-6.0000000e-01, -5.9000000e-01, -5.8000000e-01, -5.7000000e-01,
-5.6000000e-01, -5.5000000e-01, -5.4000000e-01, -5.3000000e-01,
-5.2000000e-01, -5.1000000e-01, -5.0000000e-01, -4.9000000e-01,
-4.8000000e-01, -4.7000000e-01, -4.6000000e-01, -4.5000000e-01,
-4.4000000e-01, -4.3000000e-01, -4.2000000e-01, -4.1000000e-01,
-4.0000000e-01, -3.9000000e-01, -3.8000000e-01, -3.7000000e-01,
-3.6000000e-01, -3.5000000e-01, -3.4000000e-01, -3.3000000e-01,
-3.2000000e-01, -3.1000000e-01, -3.0000000e-01, -2.9000000e-01,
-2.8000000e-01, -2.7000000e-01, -2.6000000e-01, -2.5000000e-01,
-2.4000000e-01, -2.3000000e-01, -2.2000000e-01, -2.1000000e-01,
-2.0000000e-01, -1.9000000e-01, -1.8000000e-01, -1.7000000e-01,
-1.6000000e-01, -1.5000000e-01, -1.4000000e-01, -1.3000000e-01,
-1.2000000e-01, -1.1000000e-01, -1.0000000e-01, -9.0000000e-02,
-8.0000000e-02, -7.0000000e-02, -6.0000000e-02, -5.0000000e-02,
-4.0000000e-02, -3.0000000e-02, -2.0000000e-02, -1.0000000e-02,
 8.8817842e-16, 1.0000000e-02, 2.0000000e-02, 3.0000000e-02,
 4.0000000e-02, 5.0000000e-02, 6.0000000e-02, 7.0000000e-02,
 8.0000000e-02, 9.0000000e-02, 1.0000000e-01, 1.1000000e-01,
 1.2000000e-01, 1.3000000e-01, 1.4000000e-01, 1.5000000e-01,
 1.6000000e-01, 1.7000000e-01, 1.8000000e-01, 1.9000000e-01,
 2.0000000e-01, 2.1000000e-01, 2.2000000e-01, 2.3000000e-01,
 2.4000000e-01, 2.5000000e-01, 2.6000000e-01, 2.7000000e-01,
 2.8000000e-01, 2.9000000e-01, 3.0000000e-01, 3.1000000e-01,
 3.2000000e-01, 3.3000000e-01, 3.4000000e-01, 3.5000000e-01,
 3.6000000e-01, 3.7000000e-01, 3.8000000e-01, 3.9000000e-01,
 4.0000000e-01, 4.1000000e-01, 4.2000000e-01, 4.3000000e-01,
 4.4000000e-01, 4.5000000e-01, 4.6000000e-01, 4.7000000e-01,
 4.8000000e-01, 4.9000000e-01, 5.0000000e-01, 5.1000000e-01,
 5.2000000e-01, 5.3000000e-01, 5.4000000e-01, 5.5000000e-01,
 5.6000000e-01, 5.7000000e-01, 5.8000000e-01, 5.9000000e-01,
 6.0000000e-01, 6.1000000e-01, 6.2000000e-01, 6.3000000e-01,
 6.4000000e-01, 6.5000000e-01, 6.6000000e-01, 6.7000000e-01,
```

```

6.8000000e-01, 6.9000000e-01, 7.0000000e-01, 7.1000000e-01,
7.2000000e-01, 7.3000000e-01, 7.4000000e-01, 7.5000000e-01,
7.6000000e-01, 7.7000000e-01, 7.8000000e-01, 7.9000000e-01,
8.0000000e-01, 8.1000000e-01, 8.2000000e-01, 8.3000000e-01,
8.4000000e-01, 8.5000000e-01, 8.6000000e-01, 8.7000000e-01,
8.8000000e-01, 8.9000000e-01, 9.0000000e-01, 9.1000000e-01,
9.2000000e-01, 9.3000000e-01, 9.4000000e-01, 9.5000000e-01,
9.6000000e-01, 9.7000000e-01, 9.8000000e-01, 9.9000000e-01])

```

CREATING Y AS GIVEN IN THE QUESTION

```

In [3]: Y=[]
for i in X:
    if((i>=-0.5 and i<0.1) or (i>=0.5)):
        Y.append(1)
    else:
        Y.append(-1)
Y=np.array(Y)
Y

```

```

Out[3]: array([-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1])

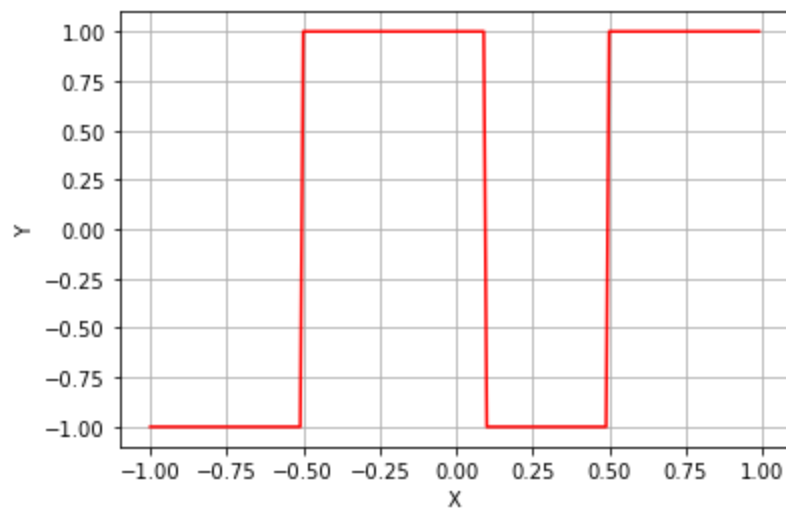
```

PLOTTING X & Y

```

In [4]: plt.plot(X,Y,color='red')
plt.xlabel("X")
plt.ylabel("Y")
plt.grid()
plt.show()

```

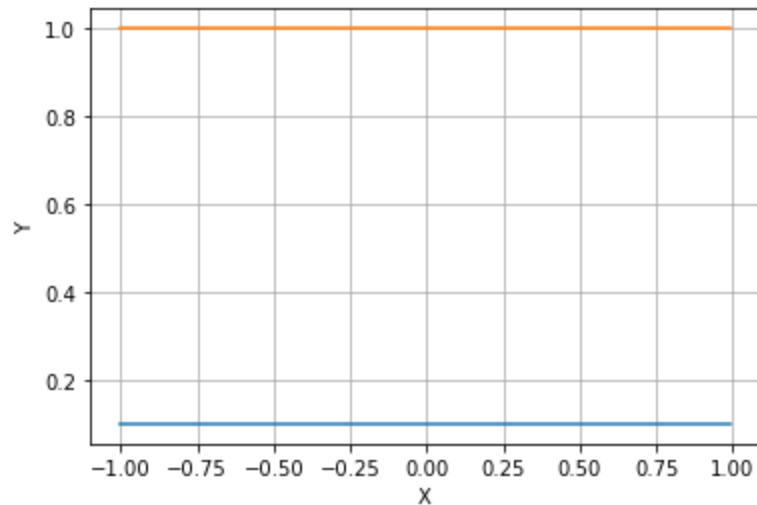


Answer Of Q2.b

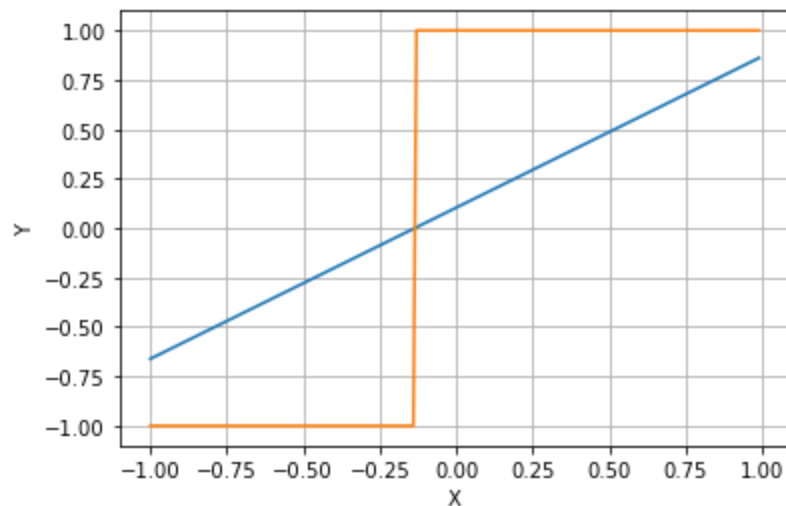
In [5]:

```
errors=[]
for degree in range(0,9):
    poly=PolynomialFeatures(degree)
    Xdf=pd.DataFrame(X)
    Xpoly=poly.fit_transform(Xdf)
    lr.fit(Xpoly,Y)
    print("Degree : ",end="")
    print(degree)
    print("Train Accuracy : ",end="")
    score = lr.score(Xpoly,Y)
    print(score)
    Ypred=lr.predict(Xpoly)
    if(degree==3):
        deg3pol=Ypred
        RMS=(1/200)*np.sum(np.square(Y-Ypred))
        print("Error(RMS) : ",end="")
        print(RMS)
        errors.append(RMS)
        plt.plot(X,Ypred)
        plt.plot(X,np.sign(Ypred))
        plt.xlabel("X")
        plt.ylabel("Y")
        plt.grid()
        plt.show()
```

Degree : 0
Train Accuracy : 0.0
Error(RMS) : 0.9900000000000003



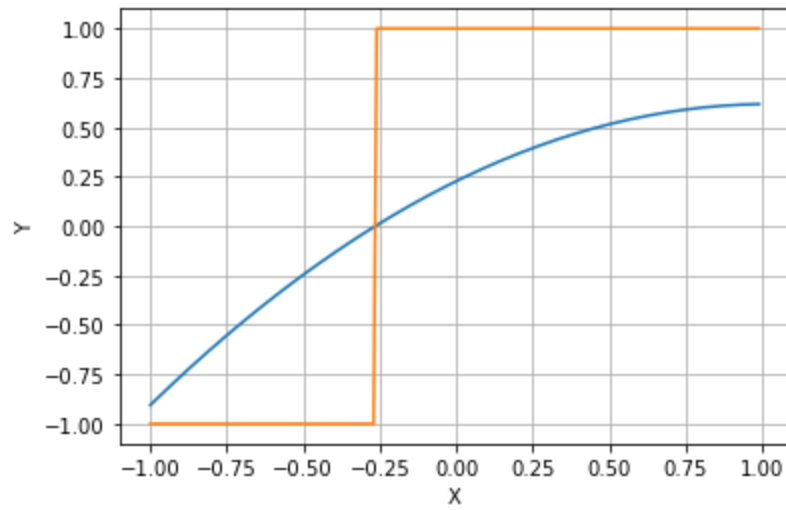
Degree : 1
Train Accuracy : 0.19705038080497495
Error(RMS) : 0.7949201230030751



Degree : 2

Train Accuracy : 0.20942692784241324

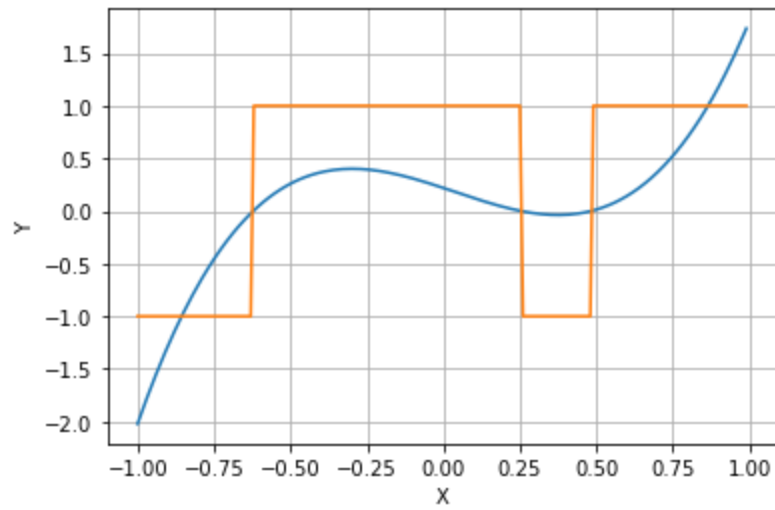
Error(RMS) : 0.7826673414360111



Degree : 3

Train Accuracy : 0.40069842880320705

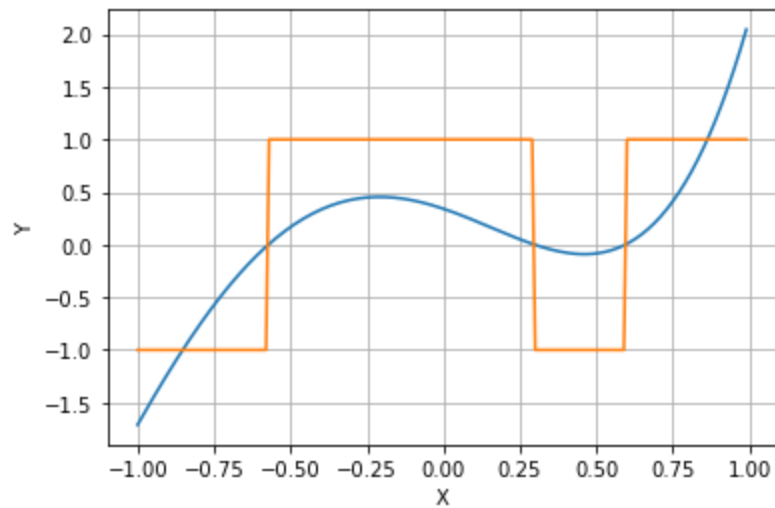
Error(RMS) : 0.5933085554848252



Degree : 4

Train Accuracy : 0.4126551059785555

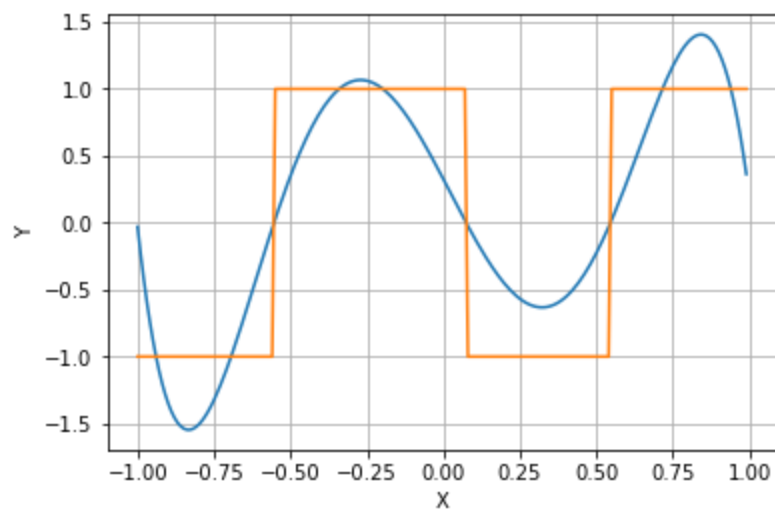
Error(RMS) : 0.5814714450812302



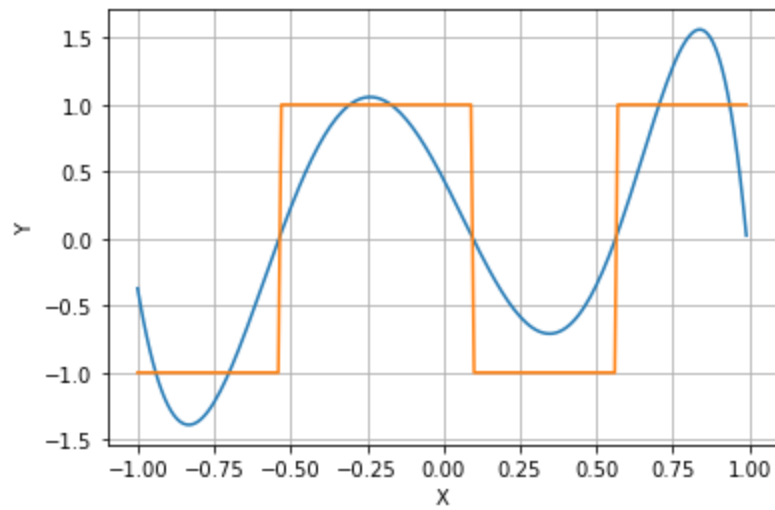
Degree : 5

Train Accuracy : 0.7141288136135671

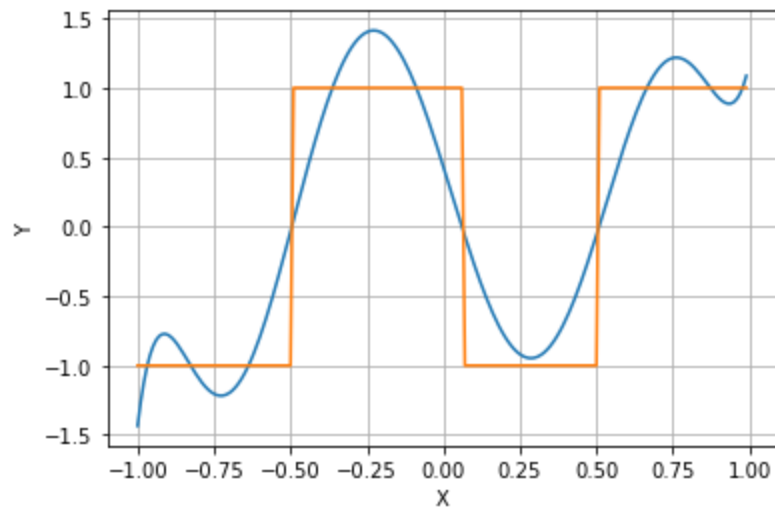
Error(RMS) : 0.2830124745225687



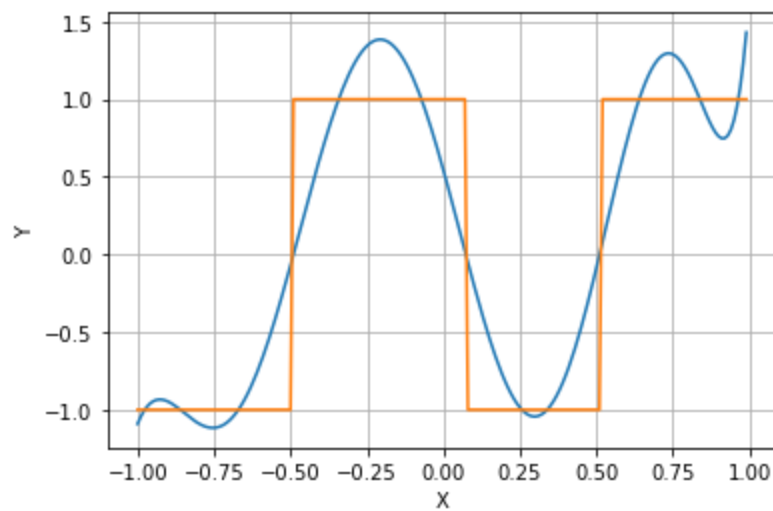
Degree : 6
 Train Accuracy : 0.7252590694544849
 Error(RMS) : 0.27199352124006004



Degree : 7
 Train Accuracy : 0.8259180227766211
 Error(RMS) : 0.1723411574511452



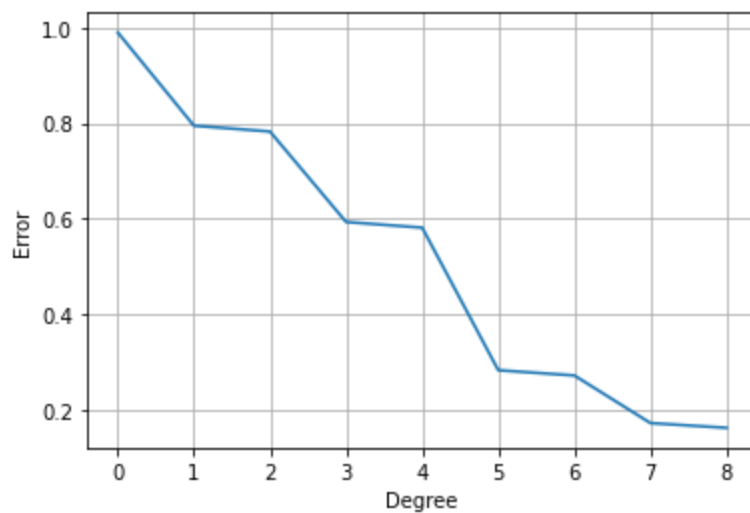
Degree : 8
 Train Accuracy : 0.8359627493186512
 Error(RMS) : 0.1623968781745354



Answer Of Q2.a

PLOTTING ERROR WITH RESPECT TO DEGREE

```
In [6]: plt.plot(errors)
plt.xlabel("Degree")
plt.ylabel("Error")
plt.grid()
plt.show()
```



Yes, the error rate decreases when we increase the degree of the polynomial as shown in the figure above.

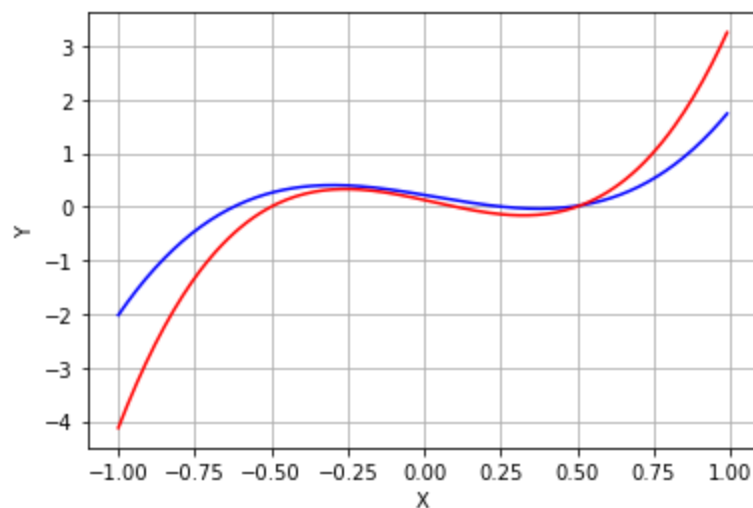
Answer Of Q2.c

```
In [7]: def Q2_c_fun(x, c):
ans=[]
for i in x:
ans.append(c*(i+0.5)*(i-0.1)*(i-0.5))
return ans
```

```
In [8]: Cubic_Polynomial = Q2_c_fun(X, 5)
```

Plotting the polynomials

```
In [9]: plt.plot(X,deg3pol,color="blue")
plt.plot(X,Cubic_Polynomial,color='red')
plt.xlabel("X")
plt.ylabel("Y")
plt.grid()
plt.show()
```

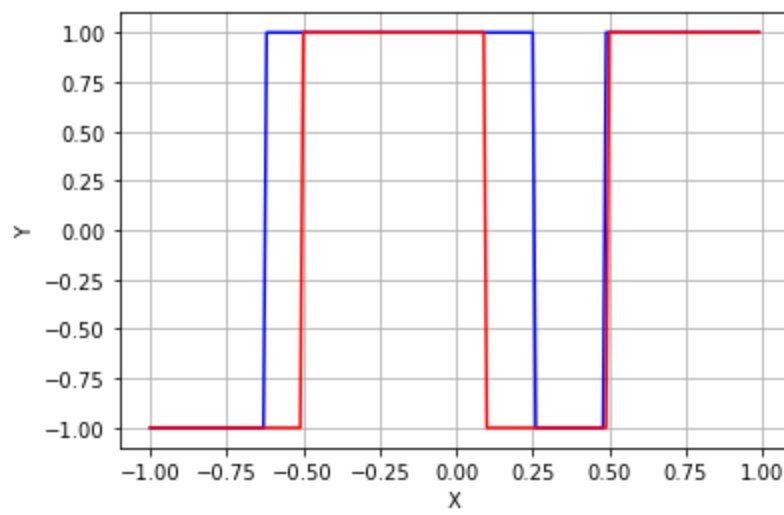


```
In [10]: np.sign(Cubic_Polynomial)
```

```
Out[10]: array([-1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1.,
        -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1.,
        -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1.,
        -1., -1., -1., -1., -1., -1., -1., -1., -1., -1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.,  1., -1., -1., -1., -1., -1., -1.,
        -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1.,
        -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1., -1.,
        -1., -1., -1., -1., -1., -1.,  1.,  1.,  1.,  1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,  1.,
         1.,  1.,  1.,  1.,  1.] )
```

Plotting the classifiers

```
In [11]: plt.plot(X,np.sign(deg3pol),color="blue")
plt.plot(X,np.sign(Cubic_Polynomial),color='red')
plt.xlabel("X")
plt.ylabel("Y")
plt.grid()
plt.show()
```



We can see that using the cubic polynomial and the given classifier we can perfectly classify the dataset. But we can not do that using degree 3 polynomial classifier.

In []: