

Assignment3 Logistic Regression

assignment_3.py X

C:\Users> sulim > Downloads > assignment_3.py > ...

```
1  import numpy as np
2  import pandas as pd
3  from sklearn.linear_model import LogisticRegression
4  from sklearn.preprocessing import StandardScaler
5  from sklearn.model_selection import train_test_split
6  from sklearn.metrics import accuracy_score
7  import matplotlib.pyplot as plt
8
9  class Logistic_Regression():
10
11     # Constructor
12     def __init__(self, dataset_name, learning_rate, no_of_iterations):
13         self.dataset_name = dataset_name
14         self.learning_rate = learning_rate
15         self.no_of_iterations = no_of_iterations
16         self.cost_fun = [] # Cost Array
17
18     # Training
19     def fit(self, X, Y):
20         self.m, self.n = X.shape
21         self.w = np.zeros(self.n)
22         self.b = 0
23         self.X = X
24         self.Y = Y
25
26         for i in range(self.no_of_iterations):
27             self.update_weights()
28             cost = self.compute_cost()
29             self.cost_fun.append(cost)
30
31         print("Final estimates of b and w are: ", self.b, self.w)
32
33     # Cost function
34     print('Initial loss\t:', self.cost_fun[0])
35     print('Final loss\t:', self.cost_fun[-1])
36     self.plot_cost_fun()
37
38     # Gradient Descent
39     def update_weights(self):
40
41         z = np.dot(self.X, self.w) + self.b
42         predictions = self.sigmoid(z)
43
44         # Differentiation
45         dw = (1/self.m) * np.dot(self.X.T, (predictions - self.Y))
46         db = (1/self.m) * np.sum(predictions - self.Y)
47
48         #update
49         self.w -= self.learning_rate * dw
50         self.b -= self.learning_rate * db
51
52     def sigmoid(self, z):
53         z = 1 / (1 + np.exp(-z))
54         return z
55
56     # Log Loss
57     def compute_cost(self):
58         z = np.dot(self.X, self.w) + self.b
59         predictions = self.sigmoid(z)
60         cost = -(1/self.m) * np.sum(self.Y * np.log(predictions) + (1 - self.Y) * np.log(1 - predictions))
61         return cost
62
63     def predict(self, X):
64         Y_pred = self.sigmoid(np.dot(X, self.w) + self.b)
65         Y_pred = np.where(Y_pred > 0.5, 1, 0)
66         return Y_pred
67
```

```

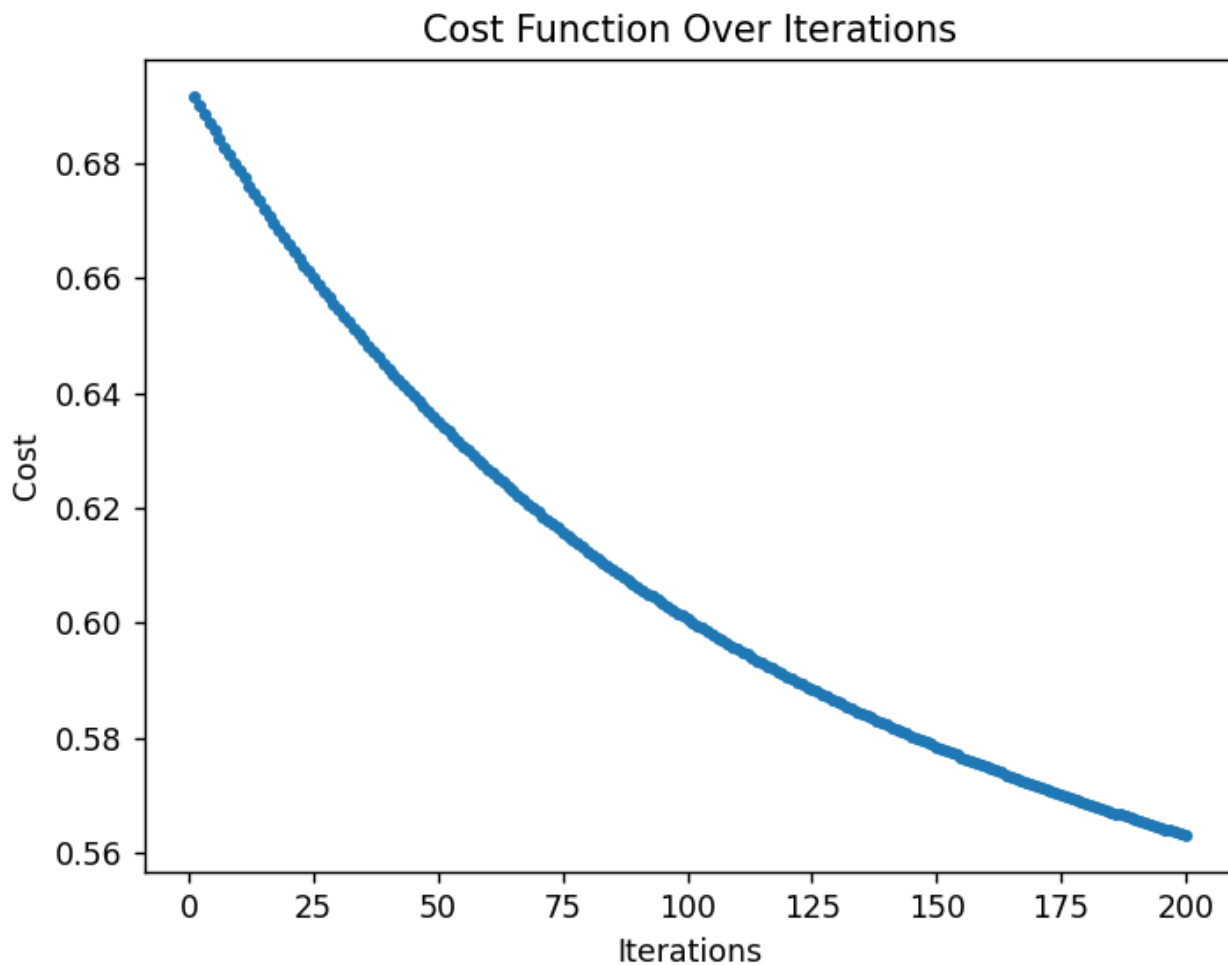
55
56 # Log Loss
57 def compute_cost(self):
58     z = np.dot(self.X, self.w) + self.b
59     predictions = self.sigmoid(z)
60     cost = -(1/self.m) * np.sum(self.Y * np.log(predictions) + (1 - self.Y) * np.log(1 - predictions))
61     return cost
62
63 def predict(self, X):
64     Y_pred = self.sigmoid(np.dot(X, self.w) + self.b)
65     Y_pred = np.where(Y_pred > 0.5, 1, 0)
66     return Y_pred
67
68 def load_dataset(self):
69     heart_dataset = pd.read_csv(self.dataset_name)
70     heart_dataset.famhist.replace(('Present', 'Absent'), (0, 1), inplace=True)
71     X = heart_dataset.drop(columns=['row.names', 'chd'], axis=1)
72     y = heart_dataset["chd"]
73     X = X.values
74     y = y.values
75     return X, y
76
77 def sklearn_LR(self, X_train, X_test, Y_train, Y_test):
78     lr = LogisticRegression()
79     lr.fit(X_train, Y_train)
80     y_pred = lr.predict(X_test)
81     testing_data_accuracy = accuracy_score(Y_test, y_pred)
82     print('sklearn classifier: Accuracy score of the testing data : ', testing_data_accuracy)
83
84 def plot_cost_fun(self):
85     plt.plot(range(1, self.no_of_iterations + 1), self.cost_fun, marker='.')
86     plt.title('Cost Function Over Iterations')
87     plt.xlabel('Iterations')
88     plt.ylabel('Cost')
89     plt.show()
90
91
92 if __name__ == "__main__":
93     dataset_name = "c:\\Users\\sulim\\Downloads\\Heart.csv"
94     learning_rate = 0.01
95     no_of_iterations = 200
96     classifier = LogisticRegression(dataset_name, learning_rate, no_of_iterations)
97     X, y = classifier.load_dataset()
98     scaler = StandardScaler()
99     scaler.fit(X)
100     X = scaler.transform(X)
101     X_train, X_test, Y_train, Y_test = train_test_split(X, y, test_size=0.2, random_state=2)
102
103     classifier.fit(X_train, Y_train)
104     y_pred = classifier.predict(X_test)
105     testing_data_accuracy = accuracy_score(Y_test, y_pred)
106     print('Accuracy score of the testing data : ', testing_data_accuracy)
107
108     classifier.sklearn_LR(X_train, X_test, Y_train, Y_test)
109
110     input_data = [124, 4.00, 12.42, 31.29, 1, 54, 23.23, 2.06, 42]
111     input_data_as_numpy_array = np.asarray(input_data)
112     input_data_reshaped = input_data_as_numpy_array.reshape(1, -1)
113     standardized_data = scaler.transform(input_data_reshaped)
114
115     prediction = classifier.predict([standardized_data])
116
117     if prediction[0] == 0:
118         print('The person is not coronary heart disease')
119     else:
120         print('The person has coronary heart disease')

```

Best inputs from the question:

learning_rate = 0.01

no_of_iterations = 200



```
PS C:\Users\sulim> python -u "c:\Users\sulim\Downloads\assignment_3.py"
```

```
Final estimates of b and w are: -0.2322948376833924 [ 0.08459305  0.21284346  0.16957313  0.13797504 -0.168603   0.07177381  
0.02002443  0.02174723  0.22796475]
```

```
Initial loss : 0.6916306519720461
```

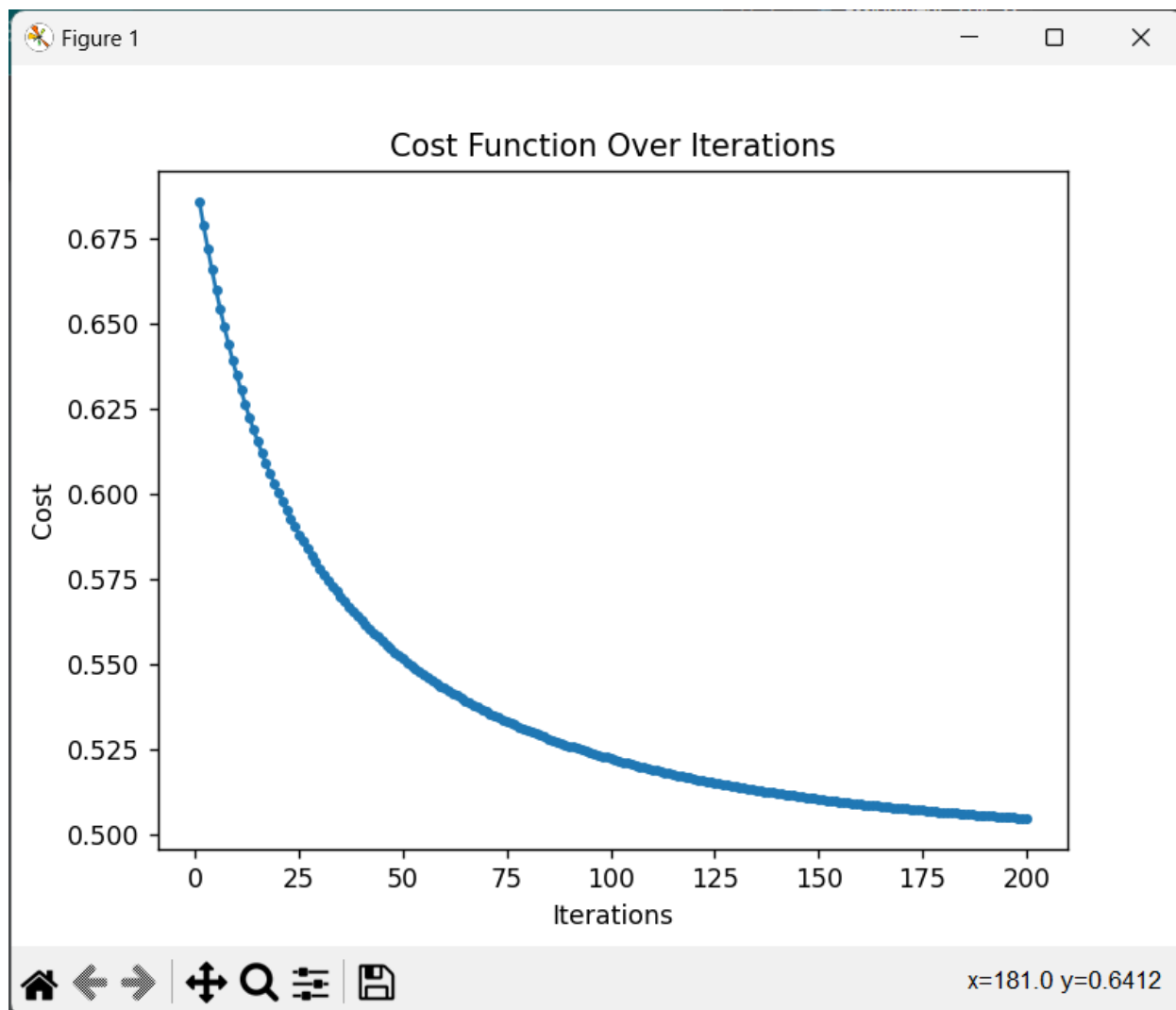
```
Final loss : 0.5630774601264
```

```
Accuracy score of the testing data : 0.6559139784946236
```

```
sklearn classifier: Accuracy score of the testing data : 0.6451612903225806
```

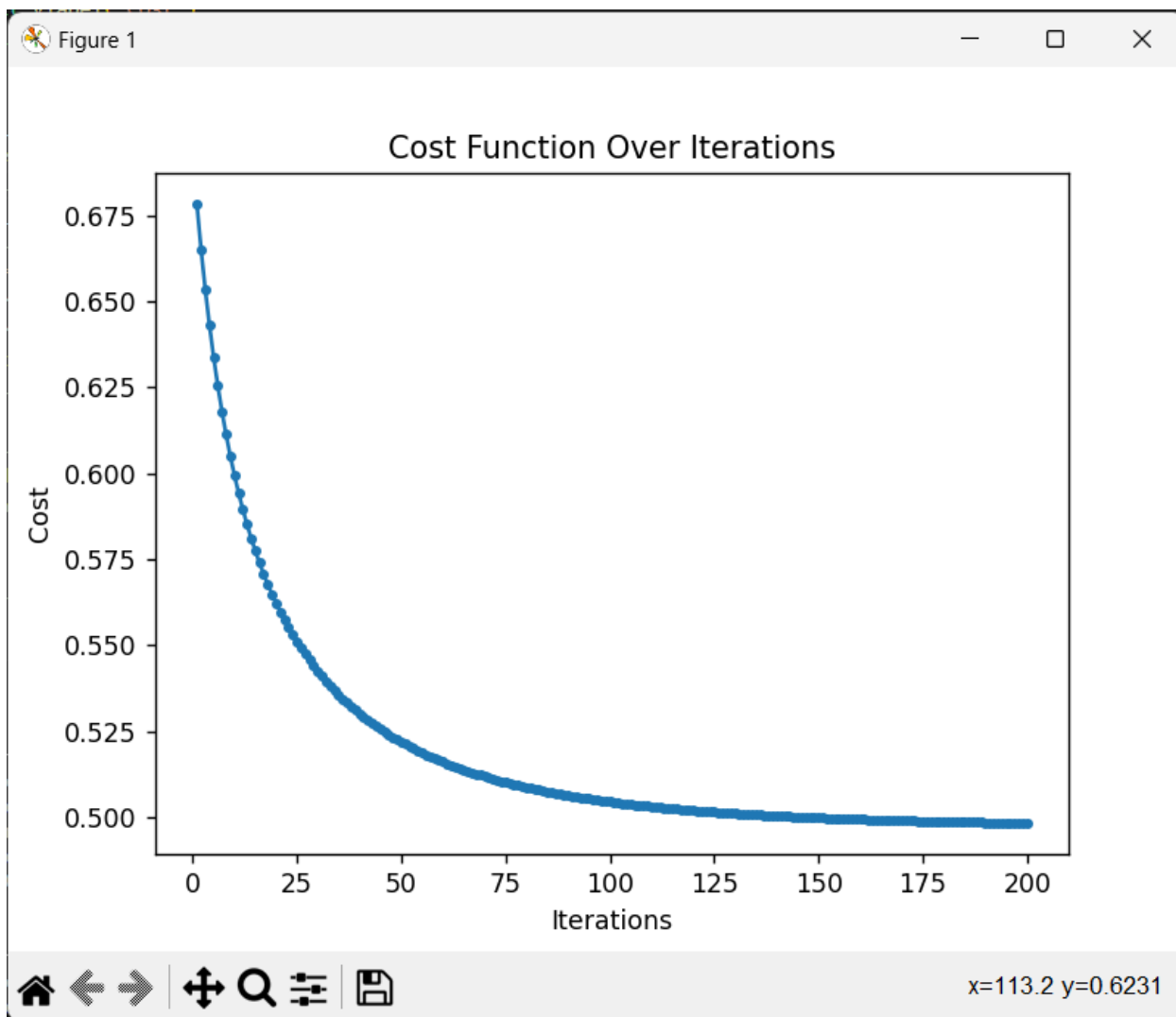
```
The person has coronary heart disease
```

learning rate = 0.01



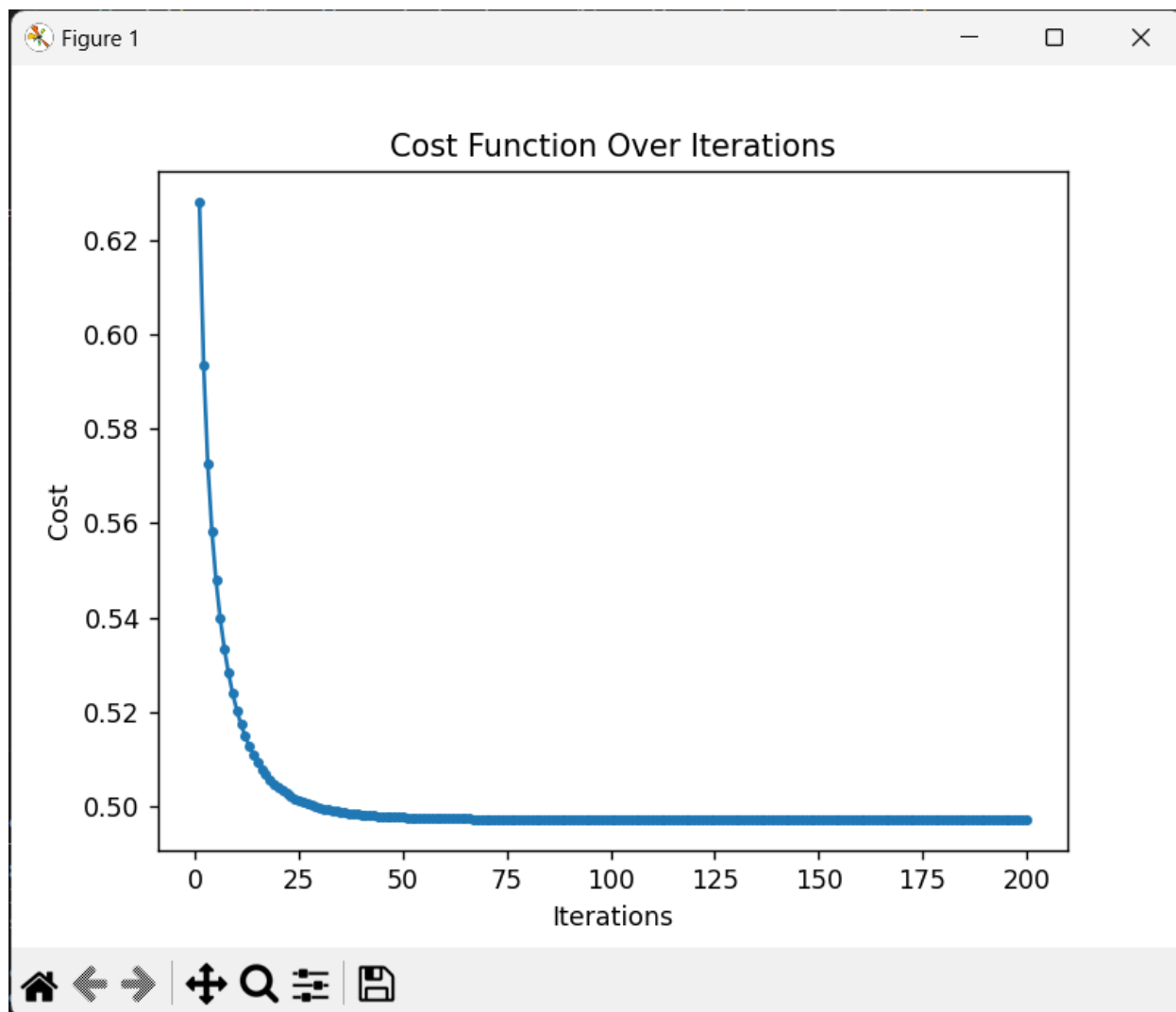
```
Final estimates of b and w are: [-0.6336274079860791 [ 0.09369933  0.45793394  0.3304137  0.18732974 -0.35489961  0.25657642
-0.12704601  0.00642576  0.46937039]
Initial loss : 0.6856541361532598
Final loss : 0.5045538057185711
Accuracy score of the testing data : 0.6559139784946236
sklearn classifier: Accuracy score of the testing data : 0.6451612903225806
The person has coronary heart disease
```

learning rate = 0.05



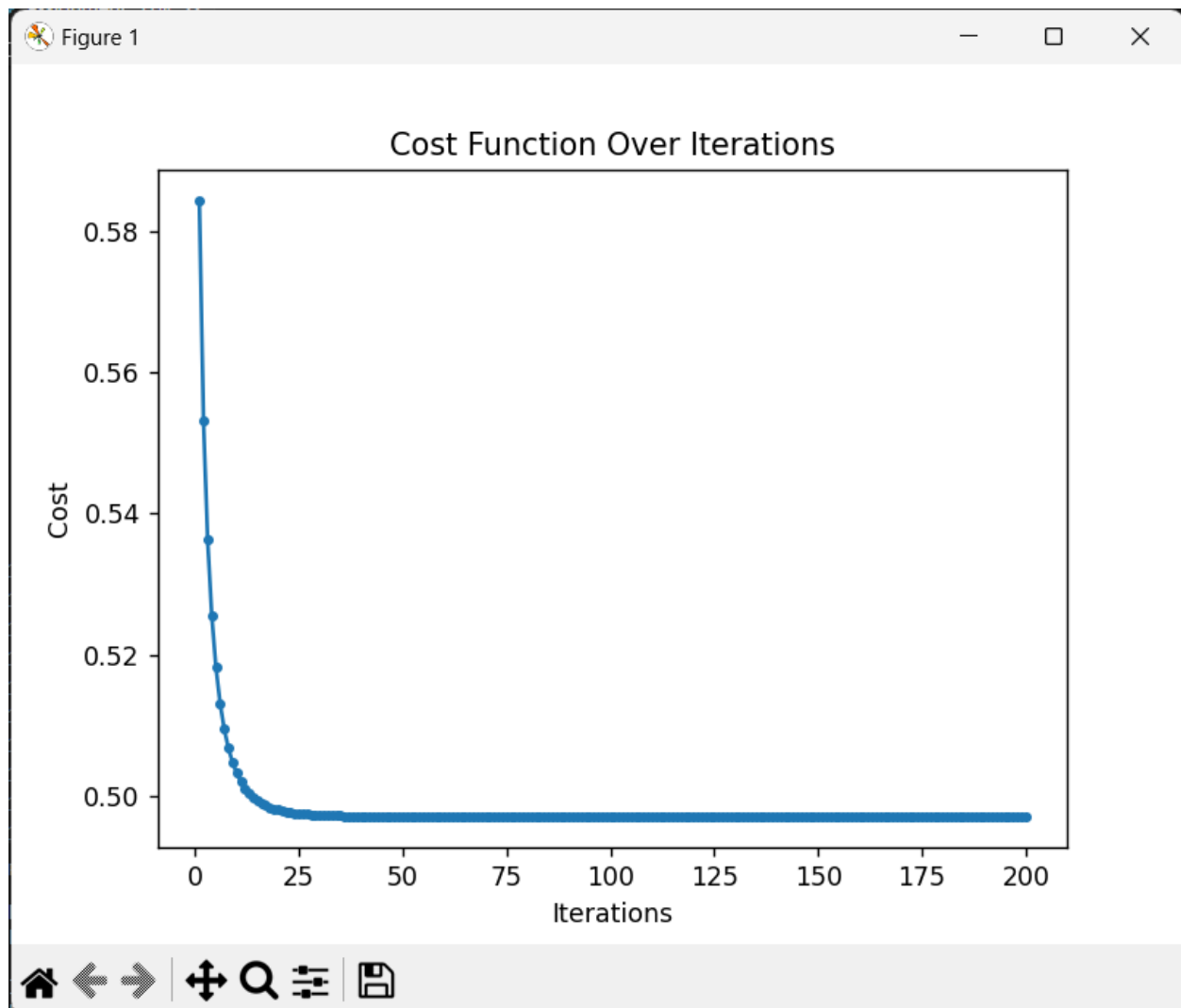
```
Final estimates of b and w are: -0.853475051264791 [ 0.05821073  0.52707969  0.40063137  0.27720767 -0.41194364  0.43391011
-0.29914796  0.00879877  0.66069812]
Initial loss   : 0.6281958211874307
Final loss     : 0.49707485591133244
Accuracy score of the testing data : 0.6559139784946236
sklearn classifier: Accuracy score of the testing data : 0.6451612903225806
The person has coronary heart disease
```

learning rate = 0.5



```
Final estimates of b and w are: [-0.7734509602241763 [ 7.92998250e-02  5.21037559e-01  3.78212746e-01  2.11533465e-01
-3.99405503e-01  3.58784004e-01 -2.12066373e-01  2.41819935e-04
 5.76643936e-01]
Initial loss : 0.6783849683017921
Final loss   : 0.4982401711142727
Accuracy score of the testing data : 0.6559139784946236
sklearn classifier: Accuracy score of the testing data : 0.6451612903225806
The person has coronary heart disease
```

learning rate = 0.1



```
Final estimates of b and w are: -0.8535507577795765 [ 0.05817623  0.52728662  0.40021322  0.2836978 -0.412031  0.43433799
-0.30378521  0.00878373  0.65776339]
Initial loss : 0.584308274533989
Final loss   : 0.497073785464207
Accuracy score of the testing data : 0.6559139784946236
sklearn classifier: Accuracy score of the testing data : 0.6451612903225806
The person has coronary heart disease
```

learning rate = 1