ML Assignment - 2

Name - JIT DAS Roll - 2206095 Section - IT-06

1. Use logistic regression to find decision boundary For the given database. Set your learning rate to 0.1. What is the cost function value and learning parameter value after convergence?

Ans: The cost function value in learning rate 0.1 is 0.2261.

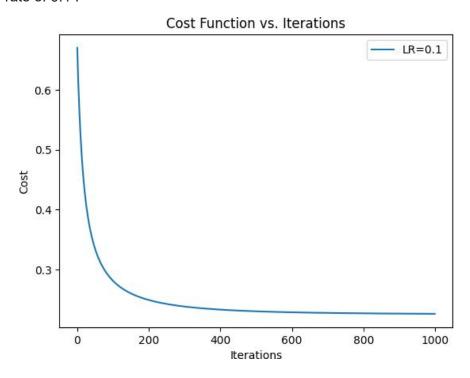
The learning parameter value after convergence is [[0.38193936]

[2.35449296]

[-2.53913828]]

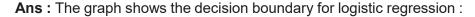
2. Plot cost function v/s iteration graph for the model trained in question?

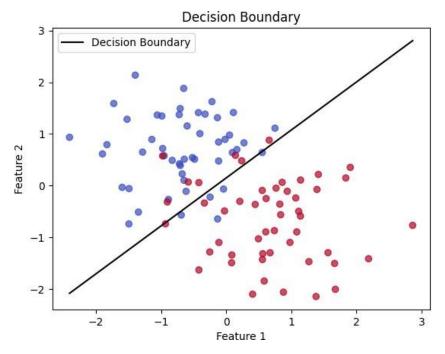
Ans: The graph shows the cost function vs. iterations for logistic regression with a learning rate of 0.1:



- The cost starts high (~0.7) and rapidly decreases in the early iterations.
- Around 200-300 iterations, the cost stabilizes, indicating convergence.
- After 600+ iterations, the cost remains almost constant, meaning further training does not significantly improve the model.

3. Plot the given dataset on a graph, use different colours for different classes and also show the decision boundary you obtained in question 1. Do not use scatter plot.

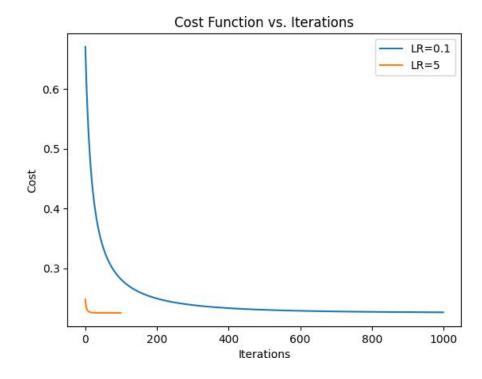




- The blue points represent one class, while the red points represent the other.
- The black line is the decision boundary, which separates the two classes based on the logistic regression model.
- Some points are misclassified (i.e., a few red points are on the blue side and vice versa), but overall, the boundary effectively separates the two groups.
- 4. Train your model with a learning rate of 0.1 and 5. Plot the cost-function v/s iteration curve for both learning rates on the same graph. For this task, only train your model for 100 iterations.

Ans: The graph compares the cost function vs. iterations for two different learning rates: 0.1 (blue) and 5 (orange):

- For LR = 0.1 (blue), the cost function decreases smoothly and gradually converges.
- For LR = 5 (orange), the cost function starts much lower and converges very quickly within a few iterations.
- However, a very high learning rate (like 5) may risk overshooting the optimal solution or causing instability.



5. Find the confusion matrix for your training dataset. Using the confusion matrix to calculate the accuracy, precision, recall, F1-score.

Ans:

• Confusion Matrix : [[44 5] [7 43]]

• Accuracy: 0.8788

• Precision: 0.8958

Recall: 0.8600

• F1-score: 0.8776

The model performs well with a high accuracy and F1-score. Some false positives (5) and false negatives (7) exist, but the model is generally reliable.

Github link: https://github.com/JITDAS-HUB/Logistic-Regression