



# Machine Learning Online | Assignment - III

## Logistic Regression | Classification Problem

Consider the log-likelihood function for logistic regression:

$$L(\theta) = y(i) \log h_{\theta}(x(i)) + (1-y(i)) \log(1-h_{\theta}(x(i)))$$

1. The files “logisticX.csv” and “logisticY.csv” in the datasets folder contain the inputs ( $x(i) \in \mathbb{R}^2$ ) and outputs ( $y(i) \in \{0,1\}$ ) respectively for a binary classification problem, with one training example per row. Implement Mini-Batch Gradient Descent method for optimizing  $L(\theta)$ , and apply it to fit a logistic regression model to the data. Initialize params  $\theta = \sim 0$  (the vector of all zeros). What are the coefficients  $\theta$  resulting from your fit? (Remember to include the intercept term.)

2. Plot the training data (your axes should be  $x_1$  and  $x_2$ , corresponding to the two coordinates of the inputs, and you should use a different symbol for each point plotted to indicate whether that example had label 1 or 0). Also plot on the same figure the decision boundary fit by logistic regression. (i.e., this should be a straight line showing the boundary separating the region where  $h(x) > 0.5$  from where  $h(x) \leq 0.5$ .)
3. **Bonus Problem** Read and Implement Newton's Method (given in Andrew NG CS229 Notes) for Part-A. Compare your results - Final Loss and Total Iterations taken before convergence.
4. Repeat Part-A and Part-B using Sci-kit Learn [Logistic Regression Class](#). Compute 5-fold Cross Validation Score on Training Set. There is no test set for the current problem.