BITS F464 - Machine Learning

Assignment-2A: Logistic Regression

Made by -

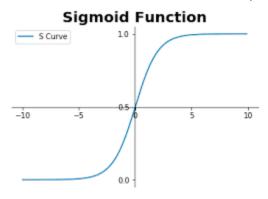
Kriti Jethlia (2018A7PS0223H) Tanay Gupta(2018AAPS0343H) Anusha Agarwal (2018A3PS0032H)

Model Description

• We have implemented Logistic regression for binary classification of a dataset.

$$y=rac{1}{1+e^{-(wx+b)}}$$

Here w is the weight of every feature and b is the bias. This y is the predicted y and the
function is a sigmoid function. It takes values between 0 and 1 therefore can be used in
probabilistic models. It is continuous and differentiable at all points.



- We need to find a w that maximizes the likelihood of training examples being predicted correctly to its corresponding class.
- We can find the optimal using 2 methods-
- Gradient Descent Algorithm- It is the process of minimizing a function by following the
 gradients of the cost function. We take eta step size and reach the global minima by
 reducing the error in every iteration. We walk in the negative direction of gradient.
 When the error between successive iterations comes close to 0, we stop. The following is
 the loss function in logistic regression.

$$J(\theta) = \frac{1}{m} \sum_{i=1}^{m} Cost(h_{\theta}(x^{(i)}), y^{(i)})$$

$$J(\theta) = \frac{1}{m} \left[\sum_{i=1}^{m} -y^{(i)} log(h_{\theta}(x^{(i)})) + (1 - y^{(i)}) log(1 - h_{\theta}(x^{(i)})) \right]$$

$$m = number\ of\ samples$$

- <u>Stochastic Gradient Descent Algorithm-</u> this is the same as gradient descent but here
 we use only 1 single value rather than the whole dataset to find the appropriate w. It is a
 faster algorithm.
- After we find the w, we predict the class using the sigmoid function. The threshold is set to be 0.5 If the function value is greater than the threshold, the class is 1 or else 0.
- According to the predicted values, we calculate accuracy, precision, recall and f score.

Results

a) Gradient Descent

Weights: [-7.06833619 -3.79093865 -4.78008557 -0.4997776]

Bias: 6.695597508493346 Loss: 16.671909166857183

Training parameters

Accuracy: 0.9907291666666667 Precision: 0.988628889354495 Recall: 0.9906661582497778 F score: 0.9896445566048753

Testing parameters

Accuracy: 0.9895631067961165 Precision: 0.9882641064116828 Recall: 0.9878896500746192 F score: 0.9880333310241675

b)Stochastic Gradient Descent

weights: [-1.18574627 -0.62340557 -0.65449352 -0.26792482]

bias: 0.5377976272539726 Loss: 0.14067225490084

Training parameters

Accuracy: 0.9726041666666667 Precision: 0.9813638167419462 Recall: 0.9568573465068745 F score: 0.96895166607547

Testing parameters

Accuracy: 0.9733009708737864 Precision: 0.977432589122159 Recall: 0.9614048882224535 F score: 0.9693020027189149

Attribute 1 is the most important feature

b)Plots of accuracy for 3 learning rates

Learning Rates = [0.001, 0.0001, 0.00001]

