COMP9417: Homework Set #2

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Question 1

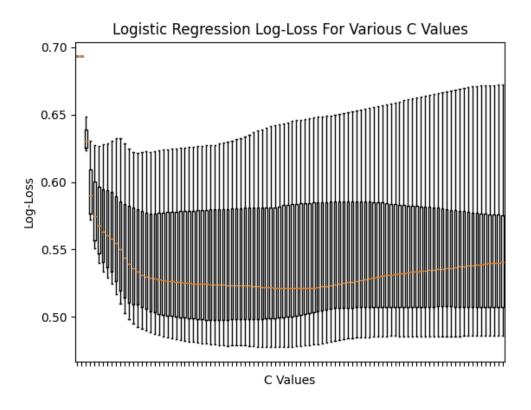
a

The possible values for both y_i and \tilde{y}_i are binary. Even though they have different values $(y_i \in \{0,1\})$ and $\tilde{y}_i \in \{-1,1\}$), the objective of each logistic regression implementation is to divide the dataset into 2 classifications. Because of this, the actual value that each classification has will not affect the parameters that the regression is attempting to optimise $((\hat{\beta}_0, \hat{\beta}))$ and (\hat{w}, \hat{c}) . Therefore the solutions for the parameters being minimised by each regression will be the same.

C is a hyper-parameter that adjusts the sensitivity that the model has to its coefficients. Compared with the standard LASSO parameter λ , C is a multiple of the Loss function whereas λ is a multiple of the Penalty.

b

Boxplot of testing accuracy for each value of C:



The value of C returning best results: 0.187947474747472

The testing accuracy of this model: 76%

From GridSearchCV:

The value of C returning best results: 0.0122191919191918

The testing accuracy of this model: 75.2%

In our answer for b , we determined the "best" value of C as the value which corresponded to the average lowest log-loss value across all folds. The value from the GridSearchCV are different because, by default, it determines the "best" value of C as the one which corresponds to the average highest score across all folds 1 .

We can modify the *GridSearchCV* class by providing our own metric for *scoring*. The following code is a scorer that uses the smallest *log-loss* value as its scoring metric.

 $^{^{1}}$ See *scoring* parameter in documentation. If the estimator provided exposes a *score* method and a value for *scoring* is not provided, then *score* is used to determine the "best" value of C