

# Report

## ML Assignment 4

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### [IPYNB LINK](#)

- **Batch Gradient Descent (BGC)**

**Approach:** Used scaled features list to iterate over all samples the no. of times taken as argument and updated coefficients each time.

- **Stochastic Gradient Descent (SGD)**

**Approach:** It takes the no. of iterations as argument and in each iteration a random index is selected from the training set and coefficients are updated after each iteration. The error calculated using a separate function is returned.

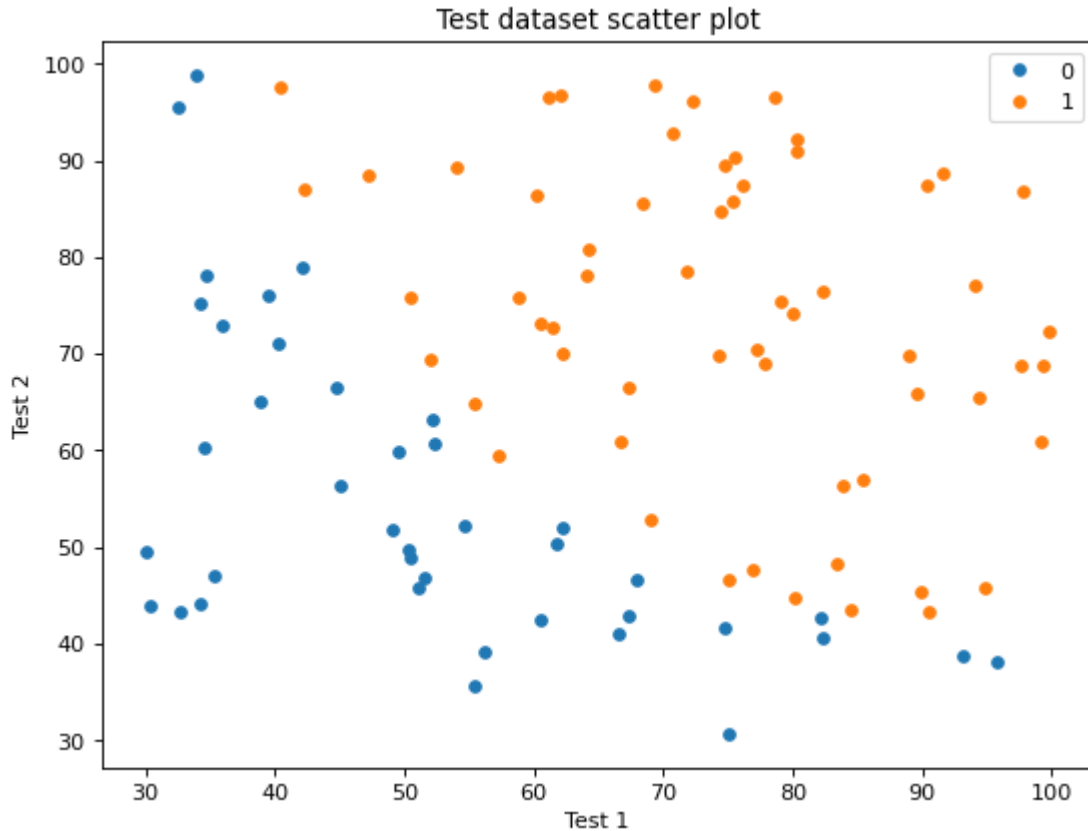
- **Mini-Batch Gradient Descent (MBGD)**

**Approach:** It takes the learning rate( $\alpha$ ), no. of iterations( $k$ ) and batch size( $b$ ) as argument and in each iteration a batch index( $i$ ) is selected from the range of number of batches. Then in the nested loop the next step of gradient descent is calculated using the ( $b$ ) no of samples starting from  $i*b$  index and coefficients are updated after each iteration. The error calculated using a separate function is returned.

# Q1

Classification model that estimates an applicant's probability of getting admission to an institution based on the scores from those two examinations.

## Dataset scatter plot:



**a**-No feature scaling, No higher order terms, No regularisation

Algorithm	$\alpha$ (Learning Rate)	k(No. of Iterations)	% Error
BGD	0.01	100000	10.0
SGD	0.00001	10000	20.0
MBGD	0.000001	100 (Batch Size=15)	20.0

***With feature scaling, No higher order terms, No regularisation***

Algorithm	$\alpha$ (Learning Rate)	k(No. of Iterations)	% Error
BGD	0.000000000000000001	400	6.66
SGD	0.000000000000000001	300	13.33
MBGD	0.0001	1000 (Batch Size=15)	10.0

**b-*With feature scaling, with higher order terms, No regularisation***

Algorithm	$\alpha$ (Learning Rate)	k(No. of Iterations)	% Error
BGD	0.000000000000000001	60	10.0
SGD	0.000000000000000001	3000	10.0
MBGD	0.000000000000000001	6000 (Batch Size=5)	6.66

**C-*With feature scaling, with higher order terms, with regularisation***

Algorithm	$\alpha$ (Learning Rate)	k(No. of Iterations)	% Error
BGD	0.000000000000000001	100	33.33
SGD	0.000000000000000001	500	10.0
MBGD	0.000000000000000001	50 (Batch Size=10)	10.0

(Note: In regularisation, lambda was taken as 1.)

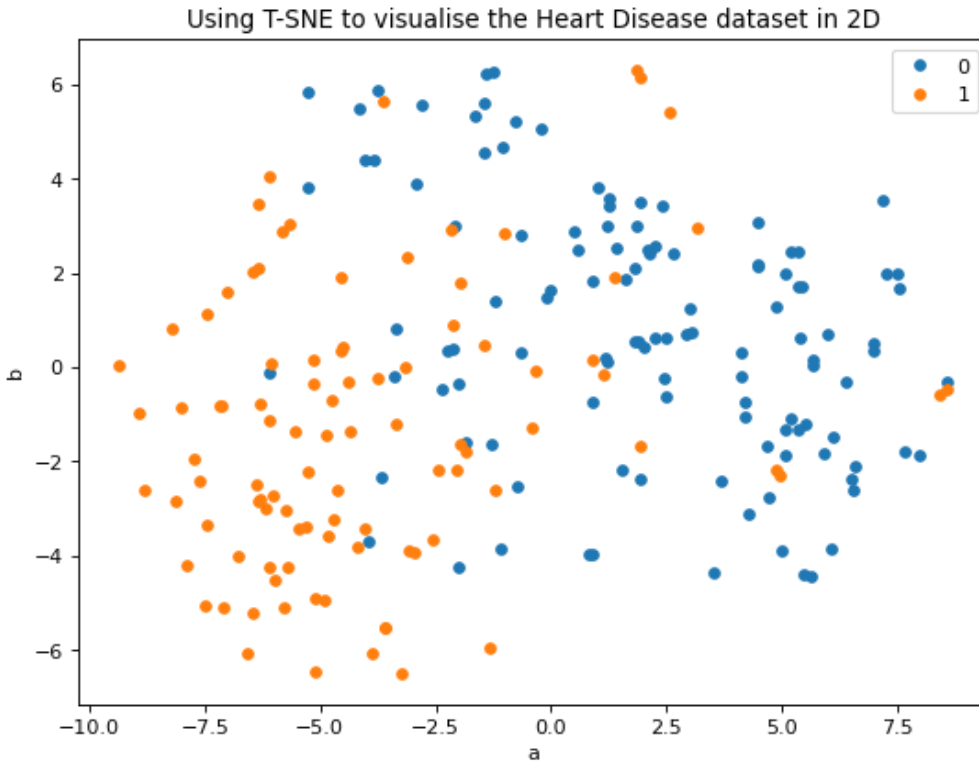
**Analysis:** There were 70 points in the training dataset and 30 in the testing dataset. Since the dataset was very small, it is observed that the results are very inconsistent. In stochastic and mini-batch gradient descent we have an element of randomisation which combined with the small sample size caused the results to vary by a large amount. The % error values in the tables above are taken to be the best possible after lots of trial and errors over a range of  $\alpha$ (Learning Rate) and k(No. of Iterations).

The `Predict()` function takes the two test scores as an argument and returns 1 and 0 to indicate admission or not.

## Q2

Classifier using logistic regression on Cleveland Medical data set for heart disease diagnosis

### Dataset scatter plot:



Since, the dataset has 14 features, here t-sne was used to visualize it in 2D plot.

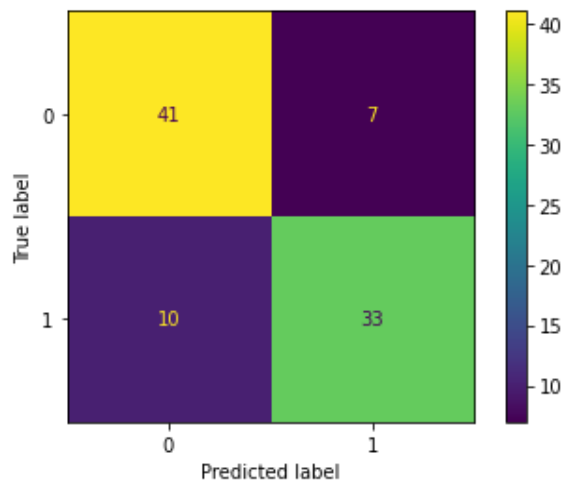
No. of features taken	$\alpha$ (Learning Rate)	k(No. of Iterations)	% Error
4(reduced with PCA)	0.001	10000	17.58
13	0.001	10000	17.58

Note: PCA was used to reduce the dataset to 4 features. LDA was not used because in it the no. of components are always less than no. of classes of labels. Error was observed to remain the same when all 13 features were used in logistic regression.

## Analysis:

- Batch Gradient Descent was used to calculate the error %.
- The classifier() function uses the same algorithm to return the prediction labels on the test set.
- The predicted list and the correct labels were used to create the confusion matrix which shows that 74 out of 91 were classified correctly while there were 7 false positives and 10 false negatives.
- The single\_classifier() function takes as argument the features of a single patient and predicts heart disease.

Confusion matrix for 4 features case:



- Specificity=  $41/(41+7)$  = 0.85
- Sensitivity=  $33/(33+10)$  = 0.77
- Precision=  $33/(33+7)$  = 0.82
- Negative Predictive value=  $41/(41+10)$  = 0.80