

# Machine Learning Based Target Transformation using Regressor Classification for Heart Disease

## Envisage



Rincy Merlin Mathew, M. Shyamala Devi, Shermin Shamsudheen

**Abstract:** *In the modern scenario of technological growth, the life style of an individual varies with the economic status. The world population is prone towards chronic deadly diseases due to the variety of food habits. The usages of electronic equipments have raised the population to waste their quality time towards exercise. The lack of physical activity has symptoms towards bad quality of life. With this background information, this paper concentrates on predicting the type of heart disease by applying target transformation using various machine learning regression models. This paper uses the Heart disease data set extracted from UCI Machine Learning Repository. The anaconda Navigator IDE along with Spyder is used for implementing the Python code. Our contribution is folded in three ways. First, the data segregation is done and it is preprocessed to extract the relationship and dependency of each parameters. Second, the dataset is subjected to process to identify the target distribution of classes in the dependent variable. Third, the dataset is fitted to the Ridge regressor, Huber regressor, SGD regressor and PerceptronCV regressor by applying with and without target transformation. Fourth, dataset is feature scaled and then fitted to the Ridge regressor, Huber regressor, SGD regressor and PerceptronCV regressor by applying with and without target transformation. Fifth, the performance analysis is done by analyzing the Mean Absolute Error and R2 Score. Experimental results show that, the Perceptron regressor CV has the effectiveness with the mean absolute error of 1.00 and R2 score of 0.04 for the heart disease prediction.*

**Index Terms:** *Machine Learning, Regressor, Target transformation, MAE and R2 Score.*

## I. INTRODUCTION

The world population is now endured with the various deadly chronic diseases because of the lack of metabolic activities. The lifestyle of the individual drastically changes due to the advancement and usage of electronic devices. There is a lack of awareness among the people towards

healthy nutrition food habits. This leads to lack of exercises and physical activity for any individual. According to the survey, the disease affected patients are mostly suffered from cardio vascular disease and they are prone to extra infectious complications too. The health care industry is towards predicting the chronic disease in early stages and to treat them in advance. This methodology of diagnosing the chronic diseases in advance may help the physicians for treating the patients in advance and to reduce the serious complications. The heart disease may result in severe and sudden cardiac arrest at any time. This may lead to dangerous adverse health effect to the patients resulting to death. So the health care industry is spending their outcome towards predicting the heart disease in advance. Since the symptoms for patients affected with heart disease are enormous, this result in challenging issue for the researchers to find the exact clinical parameter that is responsible for the heart disease.

The paper is organized in such a way that Section 2 deals with the related works. Section 3 discuss about the proposed work followed by the implementation and Performance Analysis in Section 4. The paper is concluded with Section 5.

## II. RELATED WORK

### A. Literature Review

The target transformation for the multi target regression models are built for attaining the continuous multiple target outputs. It uses the multiple regression chains which leads to building the individual regression models that uses the previous prediction of the regression models. It also finds the multiple correlation chain for predicting the multiple target transformed output. It can also be subjected to the non parametric statistical tests [1].

The decision support system has to be more effective for diagnosing the heart disease. The neural network application is used in predicting the heart disease with the target transformation of the dependent target class variable. They have verified and used the various types of neural system with all the clinical parameters [2].

The neural network is extended to form the modular neural network for predicting the heart disease. It is also possible to perform the manual analysis method to predicting the heart disease. The same can be extended to be programmatic which is done by the segmental neural network which results in the intelligent expert structural model for analyzing the various patterns of heart disease [3]. The machine learning feature selection and feature extraction methods can be used for the prediction of any factor in different application can be learnt through this article [4] – [28].

Manuscript published on January 30, 2020.

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III. PROPOSED WORK

In our proposed work, the customer churn bank modelling data set is subjected to predict the customer churn analysis. Our implementation in this paper is folded in six ways.

- (i) First, the data segregation is done and it is preprocessed to extract the relationship and dependency of each parameters.
- (ii) Second, the dataset is subjected to process to identify the target distribution of classes in the dependent variable.
- (iii) Third, the dataset is fitted to the Ridge regressor, Huber regressor, SGD regressor and PerceptronCV regressor by applying with and without target transformation.
- (iv) Fourth, dataset is feature scaled and then fitted to the Ridge regressor, Huber regressor, SGD regressor and PerceptronCV regressor by applying with and without target transformation.
- (v) Fifth, the performance analysis is done by analyzing the Mean Absolute Error and R2 Score.

A. System Architecture

The overall block diagram of the proposed work is shown in Fig. 1.

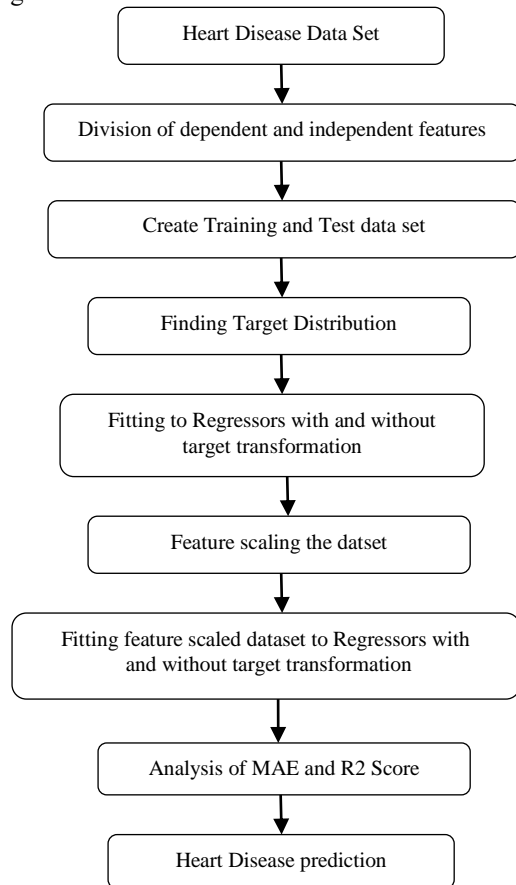


Fig. 1 Overall Design of Work

IV. IMPLEMENTATION AND PERFORMANCE ANALYSIS

A. Target Transformation Heart Disease Dataset

The Heart Disease dataset extracted from UCL ML Repository is used for implementation with 13 independent attribute and 1 diagnosis dependent attribute. The dataset consists of 779 individual's data. The attribute are shown below.

|     |  |
|-----|--|
| 1.  | Age  |
| 2.  | Sex  |
| 3.  | Chest-pain type                                  |
| 4.  | Resting Blood Pressure                           |
| 5.  | Serum Cholestrol                                 |
| 6.  | Fasting Blood Sugar                              |
| 7.  | Resting ECG                                      |
| 8.  | Max heart rate                                   |
| 9.  | Angina   |
| 10. | ST depression                                    |
| 11. | Peak exercise ST segment                         |
| 12. | Number of major vessels                          |
| 13. | Thal   |
| 14. | Diagnosis of heart disease - Dependent Attribute |

The dependency of each parameters is shown in Fig. 2.

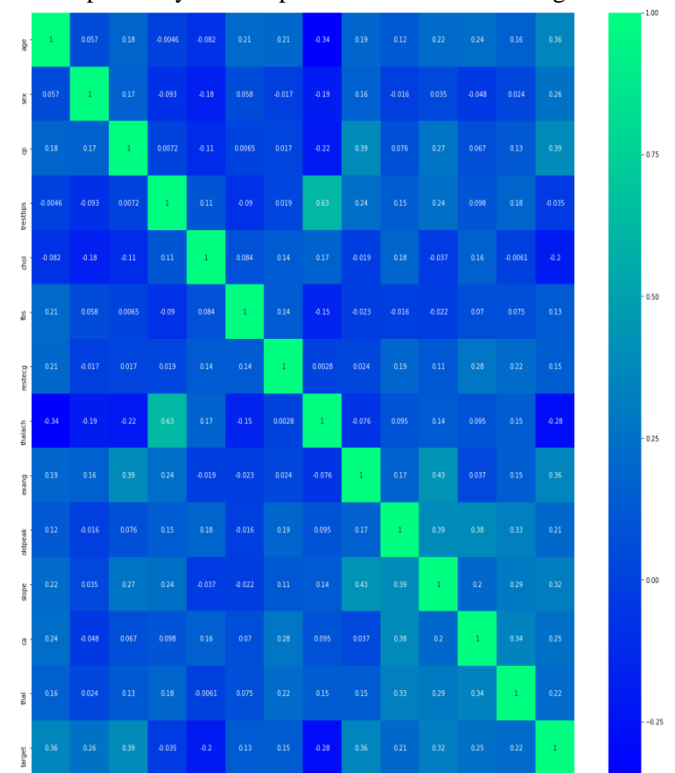


Fig. 2 Correlation Matrix of Heart Disease dataset

The classes of the dependent attribute is shown in Fig. 3.

| Chest Pain Format | Description        |
|-------------------|--------------------|
| 1                 | Typical Angina     |
| 2                 | Atypical Angina    |
| 3                 | Non - Anginal Pain |
| 4                 | Asymptotic         |

Fig. 3. Pain Type Class Levels

The expansion of the dataset attributes shown in Fig. 4.



| S.No | Attribute                  | Description  |
|------|----------------------------|--|
| 1.   | Age                        | Age of the individual.   |
| 2.   | Sex                        | Gender of the individual with the following format: 1 = male 0 = female.   |
| 3.   | Chest-pain type            | Type of chest-pain as in Table. 2.   |
| 4.   | Resting Blood Pressure     | Resting blood pressure value in mmHg (unit)  |
| 5.   | Serum Cholestrol           | serum cholestrol in mg/dl (unit)   |
| 6.   | Fasting Blood Sugar        | Fasting blood sugar value of an individual with 120mg/dl. If fasting blood sugar > 120mg/dl then : 1 (true) else : 0 (false) |
| 7.   | Resting ECG                | 0 = normal<br>1 = having ST-T wave abnormality<br>2 = left ventricular hypertrophy   |
| 8.   | Max heart rate             | Max Heart Rate   |
| 9.   | Angina                     | 1 = yes 0 = no   |
| 10.  | ST depression              | Value (integer or float).  |
| 11.  | Peak exercise ST segment   | 1 = Upsloping<br>2 = Flat<br>3 = Downsloping   |
| 12.  | Number of major vessels    | Values (0-3) colored by flourosopy   |
| 13.  | Thal                       | displays the Thalassemia :<br>3 = Normal 6 = Fixed Defect 7 = Reversible Defect  |
| 14.  | Diagnosis of heart disease | 0 = Absence<br>1,2,3,4 = Present.  |

Fig. 4. Heart Disease Dataset Design

The division of the target heart disease class type variable is shown in Fig.5.

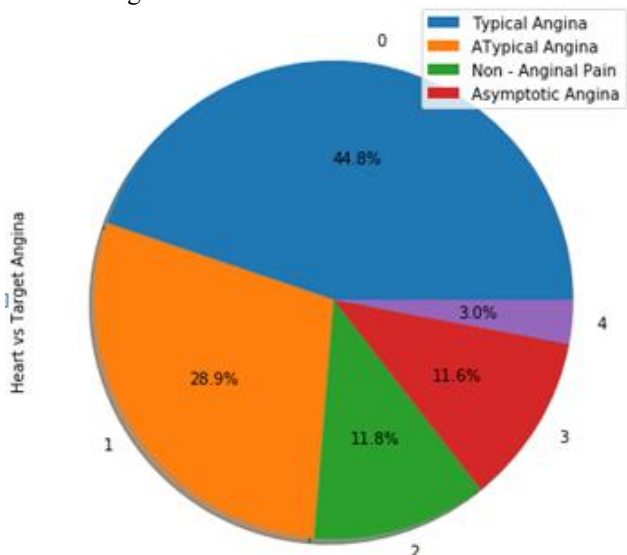


Fig. 5. Target Class Distribution

The dataset is fitted to the Ridge regressor, Huber regressor, SGD regressor and PerceptronCV regressor by applying with and without target transformation. The dataset is feature scaled and then fitted to the Ridge regressor, Huber regressor, SGD regressor and PerceptronCV regressor by applying with and without target transformation. is shown in the Fig. 6 – Fig 14.

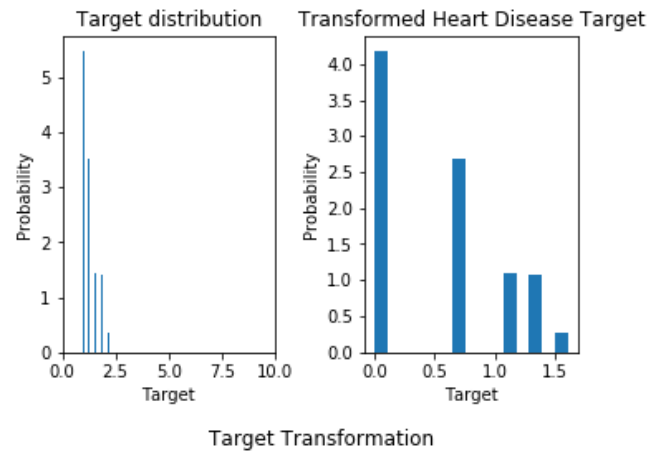


Fig. 6. Division of Target Class with Target transformation

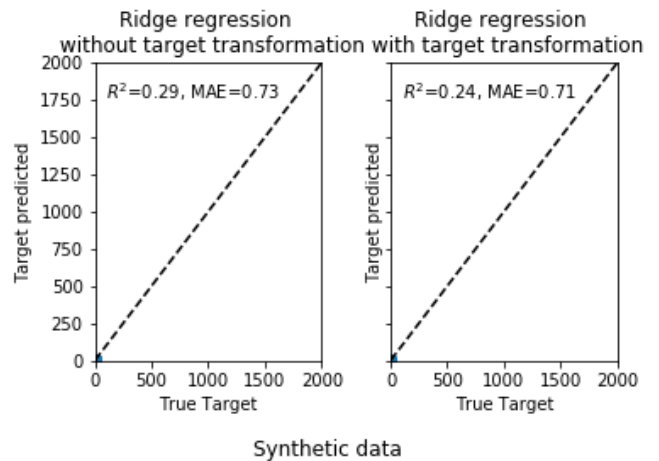


Fig. 7. Ridge Regression with Target transformation

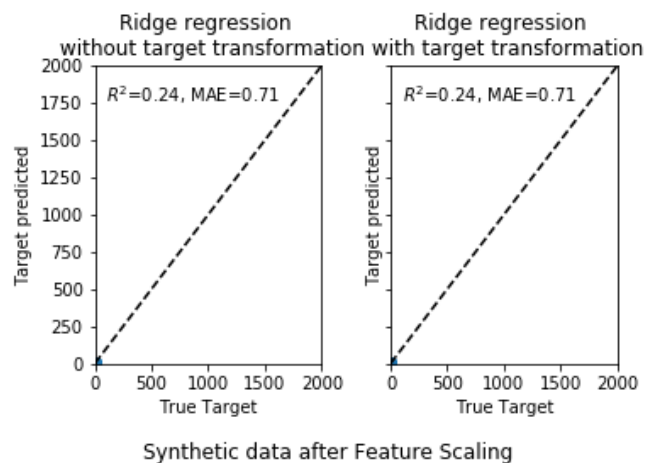


Fig. 8. Feature Scaled Ridge Regression with Target transformation

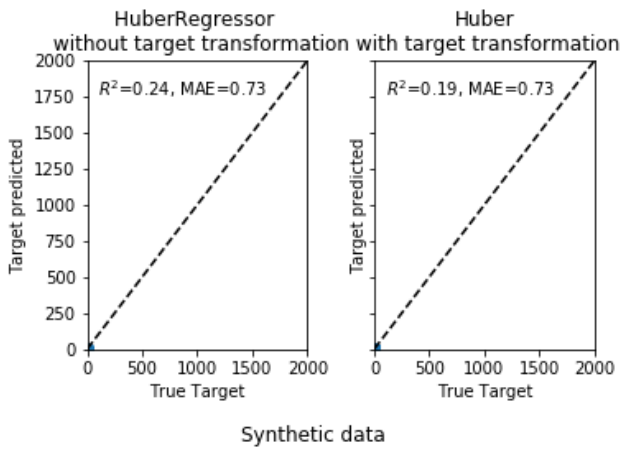


Fig. 9. Huber Regressor with Target transformation

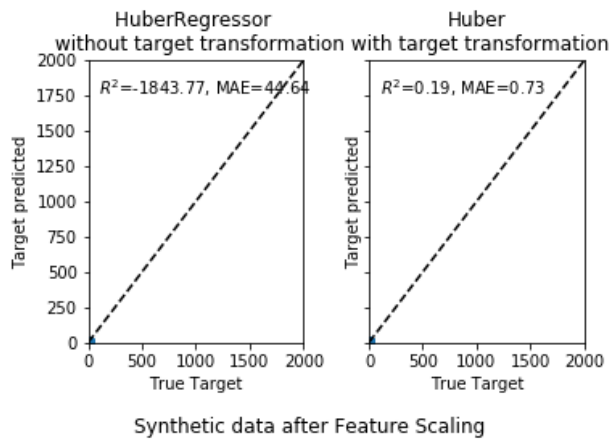


Fig. 10. Feature Scaled Huber Regressor with Target transformation

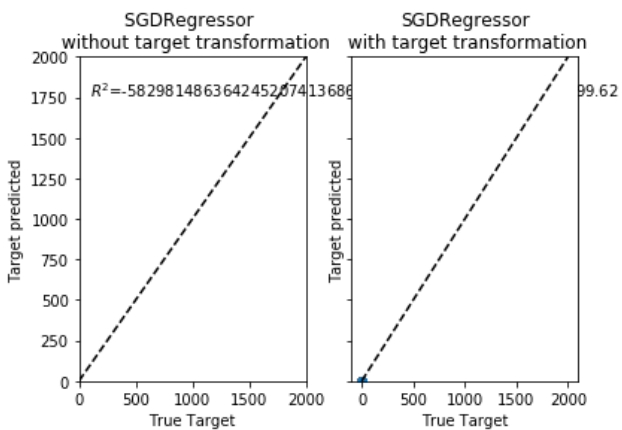


Fig. 11. SGD Regressor with Target transformation

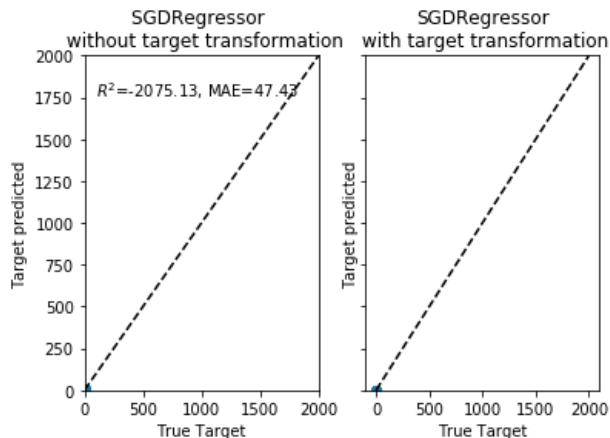


Fig. 12. Feature Scaled SGD Regressor with Target transformation

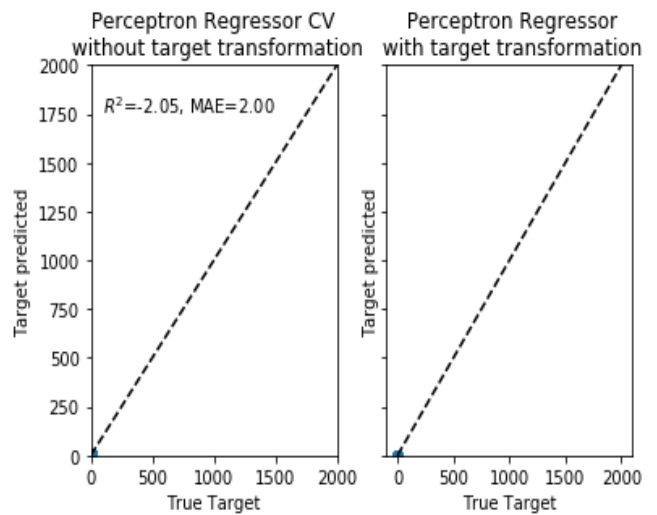


Fig. 13. Perceptron Regressor with Target transformation

The 2 component PCA reduced dataset is fitted to various solvers of logistic regression classifiers like liblinear, lbfgs, newtoncg, sag, saga and warn and their performance is analyzed with the confusion matrix and is shown in Fig. 12.

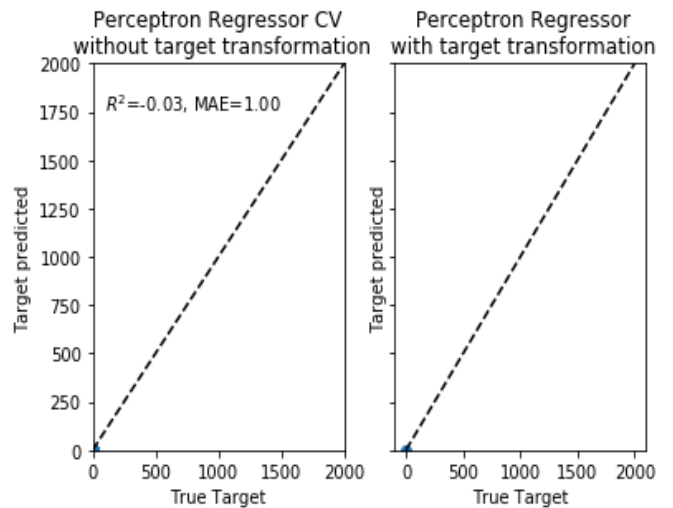


Fig. 14. Feature Scaled Perceptron with Target transformation

The performance analysis for the dataset with and without feature scaling along with the target transformation is shown in Table. 1 – Table 4.

Table. 1 Analysis of Regressors without Transformation

| Regressor  | No Feature Scaling without Target |          |
|------------|-----------------------------------|----------|
|            | MAE                               | R2 Score |
| Ridge      | 0.73                              | 0.29     |
| Huber      | 0.73                              | 0.24     |
| SGD        | 5.89                              | 0.96     |
| Perceptron | 2.05                              | 2.00     |

Table. 2 Analysis of Regressors with Transformation

| Regressor  | No Feature Scaling with Target |          |
|------------|--------------------------------|----------|
|            | MAE                            | R2 Score |
| Ridge      | 0.71                           | 0.24     |
| Huber      | 0.73                           | 0.19     |
| SGD        | 5.86                           | 0.99     |
| Perceptron | 2.02                           | 2.01     |

**Table. 3 Analysis of Regressors with Feature Scaling and without Transformation**

| Regressor  | With Feature Scaling without Target |          |
|------------|-------------------------------------|----------|
|            | MAE                                 | R2 Score |
| Ridge      | 0.71                                | 0.24     |
| Huber      | 44.64                               | 18.43    |
| SGD        | 2.07                                | 47.43    |
| Perceptron | 1.00                                | 0.03     |

**Table. 4 Analysis of Regressors with Feature Scaling and Transformation**

| Regressor  | With Feature Scaling with Target |          |
|------------|----------------------------------|----------|
|            | MAE                              | R2 Score |
| Ridge      | 0.71                             | 0.24     |
| Huber      | 0.73                             | 0.19     |
| SGD        | 2.04                             | 47.42    |
| Perceptron | 1.00                             | 0.04     |

**V. CONCLUSION**

This paper attempts to analyze the performance of the regressors with and without applying feature scaling and target transformation. The metrics like mean absolute error and r2 score is analyzed for the dataset before and after applying feature scaling. Experimental results show that, the Perceptron regressor CV has the effectiveness with the mean absolute error of 1.00 and R2 score of 0.04 for the heart disease prediction.

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