

# Lab 02 - Regression

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INDEX NO. : 210518H

## Final Rank/Score

Rank	Participant	Score	Rank	Time
10	Isara Liyanage	5.73902	3	19h
11	Chethmi-210417X	5.75954	10	8h
12	Janindu-210306G	5.76470	8	21m
13	Nisith - 210144G	5.76979	7	2d
14	Lakshan-210353V	5.78666	12	8h
15	Auliš Arinili	5.79937	11	13d
16	<b>Kumudh-210518H</b>	5.80448	31	6h
<p> Your Best Entry! Your submission scored 6.15379, which is not an improvement of your previous score. Keep trying!</p>				
17	Ravindu-210685N	5.80819	7	2h
18	Rupanshu Kapoor	5.81079	25	2d
19	Sanjana K.Y.C. - 210572P	5.82801	11	20h

For this report I will do a comparison between the first submission I made, and the final submission made using various evaluation metrics to understand how these two models perform against each other. Given below are the scores obtained by the two models for the submissions made to the Kaggle Competition.

### Model 01(First Submission)

2	Kumudh-210518H	5.87589	1	8h
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### Model 02(Final Submission)

14	Kumudh-210518H	5.80448	31	4h
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As you can see from above the score improves from 5.87589 to 5.80448 when we go from model 01 to 02. This is on the unseen test.csv. We will be evaluating on the train.csv dataset. First, we will see how these models work. The differences are highlighted in yellow for the second model.

### **Model 01**

- Missing values were handled using a prediction model rather than imputation. Used Linear Regressor as the model for predicting missing values.
- Encoding done using Ordinal encoder.
- No Feature Scaling.
- No Feature Selection.
- No Dimensionality Reduction.
- The ML model was made using ensemble methods. Created a Stacked Model using XGBRegressor, RandomForestRegressor and ExtraTreesRegressor as base models and Linear Regressor as the meta model.
- No Hyper parameter tuning.

### **Model 02**

- Missing values were handled using a prediction model rather than imputation. Used **Random Forest Classifier** as the model for predicting missing values.
- Encoding done using Ordinal encoder.
- **Feature Scaling was done.**
- No Feature Selection.
- No Dimensionality Reduction.
- The ML model was made using ensemble methods. Created a **Blended Model** using XGBRegressor, RandomForestRegressor and ExtraTreesRegressor as base models and **Lasso Regressor** as the meta model.
- **Hyper parameter tuning was done for the base models.**

Now that we have an understanding of the two models, we will delve onto the evaluation of the performance of the two models on the train.csv dataset.

	<b>Model 01</b>	<b>Model 02</b>
<b>Root Mean Square Error</b>	5.497825695324197	5.386956791860444
<b>Mean Absolute Error</b>	4.292745969987085	4.143820939580833
<b>Mean Square Error</b>	30.22608737616699	29.01930347737137
<b>R-Squared</b>	0.8410428074185596	0.8510758765918485
<b>Mean Absolute Percentage Error</b>	0.0641511156662835	0.06144337899991974
<b>Explained Variance Score</b>	0.8410428074185596	0.8512308082536888
<b>Max Error</b>	37.26656870944318	40.89042940836921
<b>Mean Squared Log Error</b>	0.00742650810532961	0.0073521119987292665

### **Root Mean Square Error (RMSE)**

*Model 01: 5.497826    Model 02: 5.386957*

RMSE is calculated by squaring the errors and then taking the root to compensate for negative value errors. Lower values indicate better performance. Model 02 has a slightly lower RMSE, suggesting it is better at predicting the target variable.

### Mean Absolute Error (MAE)

*Model 01: 4. 292746    Model 02: 4. 143821*

MAE represents the average magnitude of the errors in predictions. Again absolute values taken to compensate for negative errors. Like RMSE, lower values are desirable. Model 02 again shows a lower MAE.

### Mean Squared Error (MSE)

*Model 01: 30. 226087    Model 02: 29. 019303*

MSE measures the average of the squares of the errors. Lower values indicate better performance, and Model 02 has a lower MSE.

### R-Squared (R2)

*Model 01: 0. 841043    Model 02: 0. 851076*

R-squared represents the proportion of the variance in the dependent variable that is predictable from the independent variables. Values closer to 1 indicate a better fit. Model 02 has a higher R2, suggesting it explains more variance in the data.

### Mean Absolute Percentage Error (MAPE)

*Model 01: 0. 064151    Model 02: 0. 061443*

MAPE expresses errors as a percentage of the actual values. Lower values indicate better accuracy. Model 02 shows a slightly lower MAPE.

### Explained Variance Score

*Model 01: 0. 841043    Model 02: 0. 851231*

EVS indicates the proportion of the variance in the dependent variable that is explained by the independent variables. Model 02 has a higher EVS which is desirable.

### Max Error

*Model 01: 37. 266568    Model 02: 40. 890429*

Max Error represents the maximum residual error. Lower values are preferred. Model 01 has a lower max error. Unhandled Outliers contribute to this error.

### Mean Squared Log Error (MSLE)

*Model 01: 0. 007426    Model 02: 0. 007352*

MSLE measures the mean of the squared differences between the natural logarithm of the predicted values and the natural logarithm of the observed values. Lower values indicate better performance, and both models have similar MSLE.

As you can see above in all the evaluation metrics Model 02 outperforms Model 01. This is understandable because Model 02 was built by improving possible issues that I identified in Model 01. Possible reasons could be the use of Classifier instead of a Regressor for predicting missing values, Scaling the processed dataset and hyper parameter tuning. Another reason could be that Model 01 is overfitting therefore performing poorly on unseen data. This issue is solved in Model 02 by the use of regularization from Lasso regressor as the meta model.