

Capstone Project-2

# Seoul Bike Sharing Demand Prediction

ML SUPERVISED REGRESSION

SAMEER THETE

- **Problem Statement**
- **Data Summary**
- **Data Analysis**
- **Analysis Details**
- **Feature Selection**
- **Data Preparation**
- **Implementing Various Regression Algorithms**
- **Challenges**
- **Conclusions**

## Problem Statements

- **Prediction of bike count required at each hour.**
- **Reduce waiting time of public.**

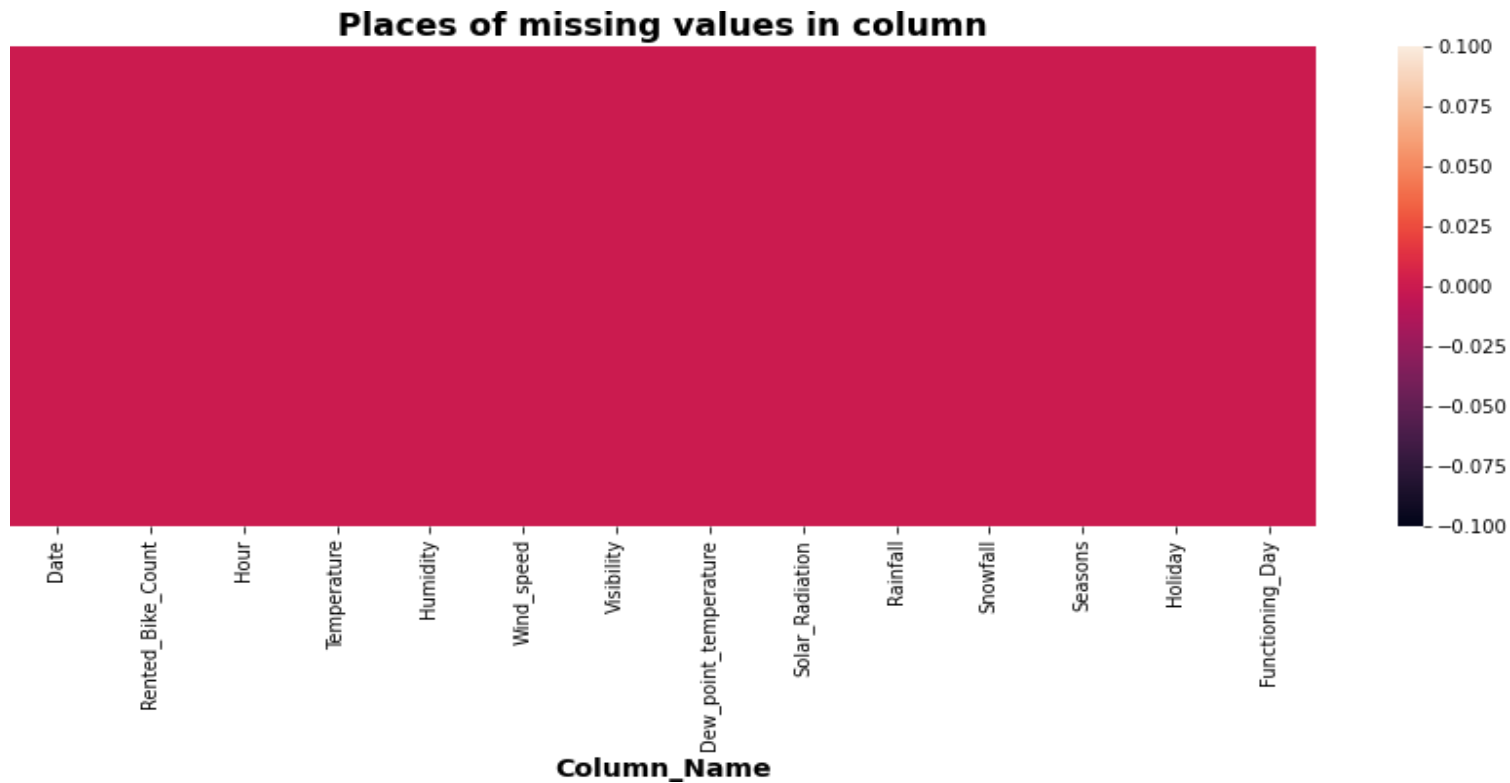
## Data Summary

- **Date : Year-Month-Day**
- **Rented Bike Count - Count of bikes rented at each hour**
- **Hour - Hour of the day**
- **Temperature - Temperature in Celsius**
- **Humidity - %**
- **Windspeed - m/s**
- **Visibility - 10m**
- **Dew point temperature -Celsius**
- **Solar radiation -MJ/m2**
- **Rainfall -mm**
- **Snowfall -cm**
- **Seasons -Winter, Spring, Summer, Autumn**
- **Holiday -Holiday/No Holiday**
- **Functional Day - NoFunc(Non Functional Hrs),Fun(Functional Hrs)**

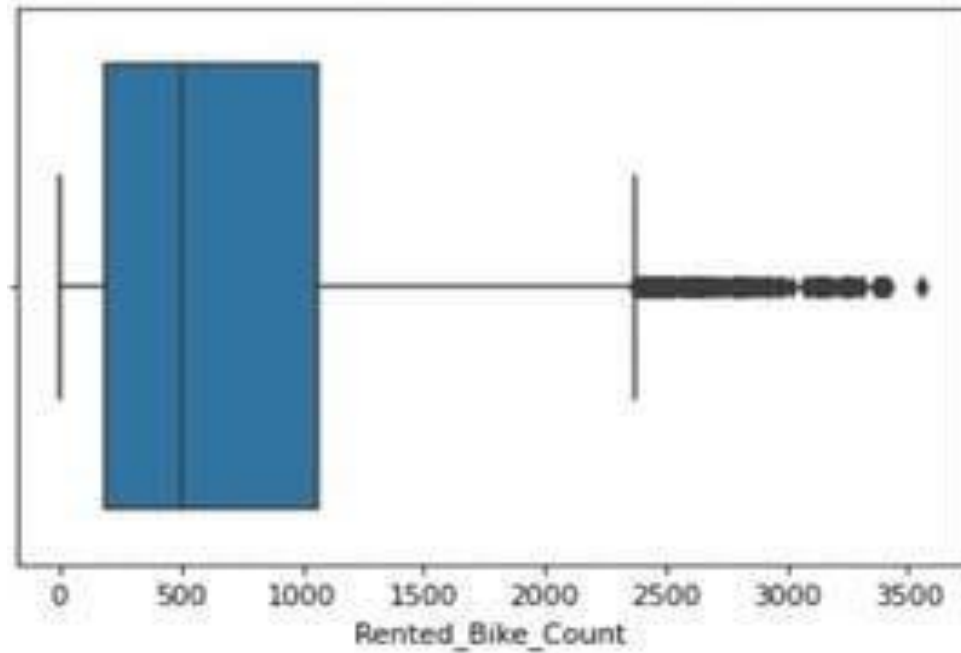
## Basic Data Exploration

- The dataset has 8760 rows and 14 features(columns).
- Three categorical features 'Seasons', 'Holiday', & 'Functioning Day'.
- One Datetime[ns] features 'Date'.
- Outliers present only in dependent variable.
- No Missing Values.
- No Duplicated values.
- No null values.

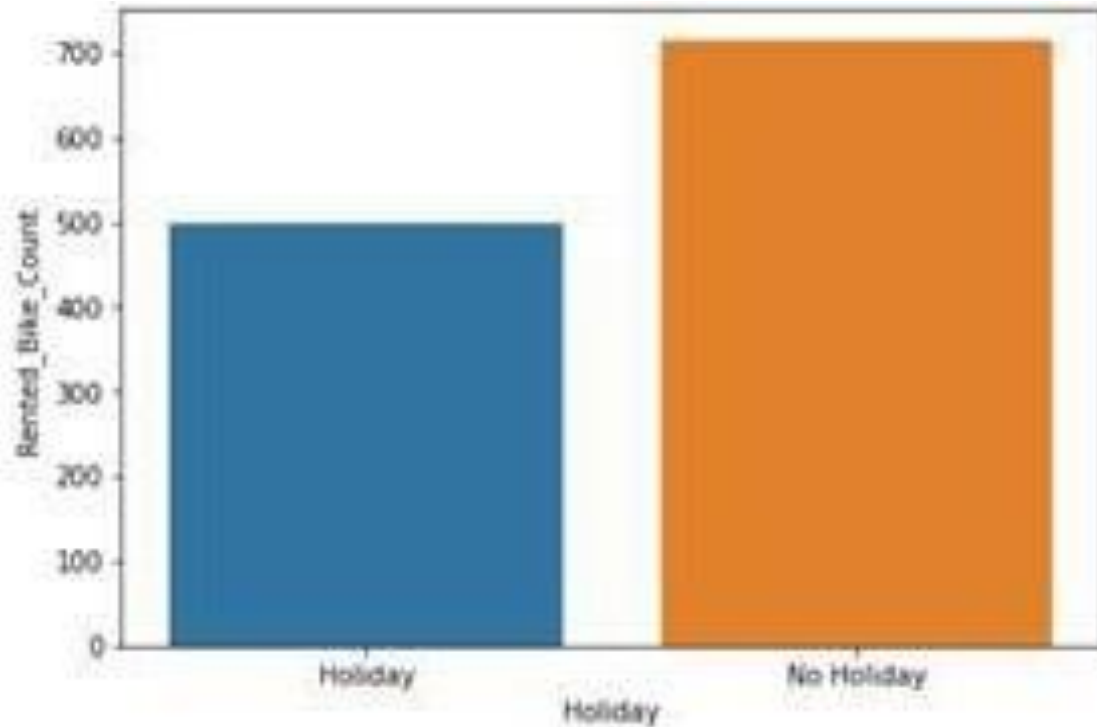
## Missing Values



Checking for the outlier in our dependent variable



Division on rented bike on holiday and non holiday days





## Distribution on rented bike according to different seasons

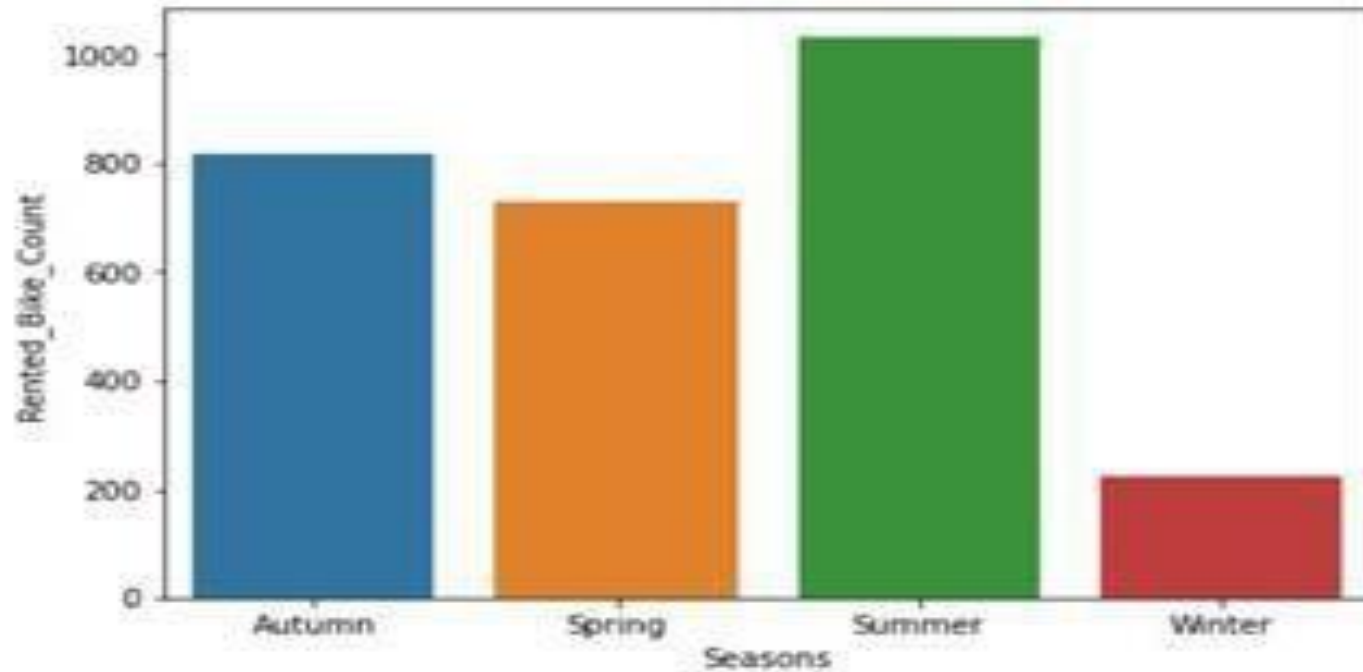
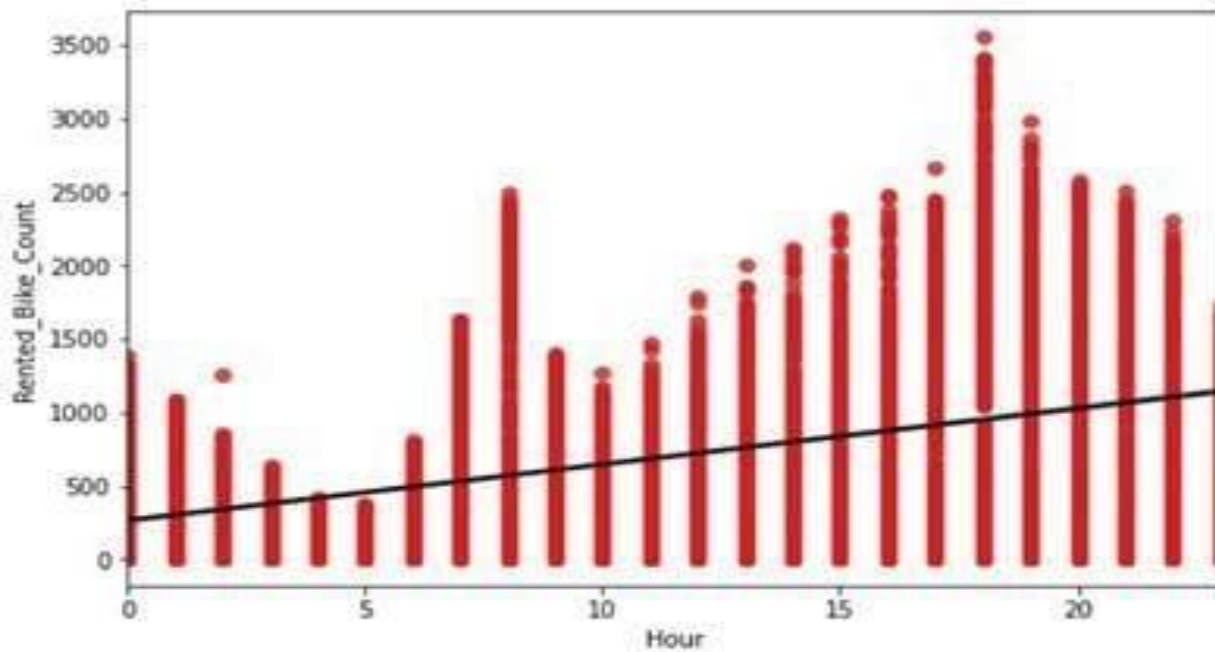
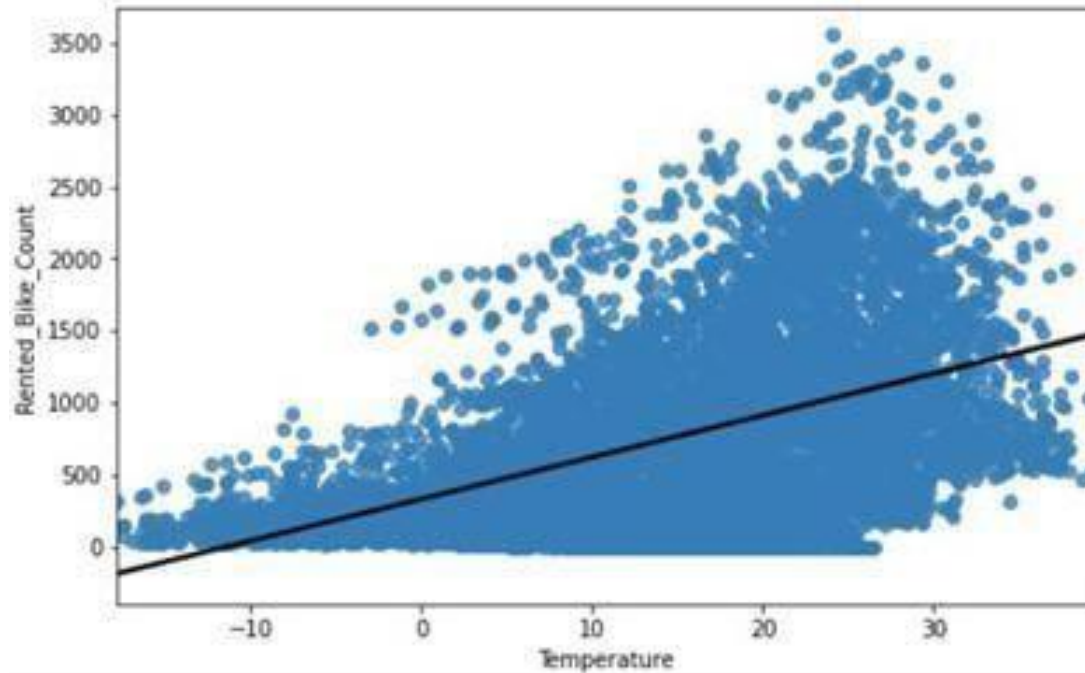


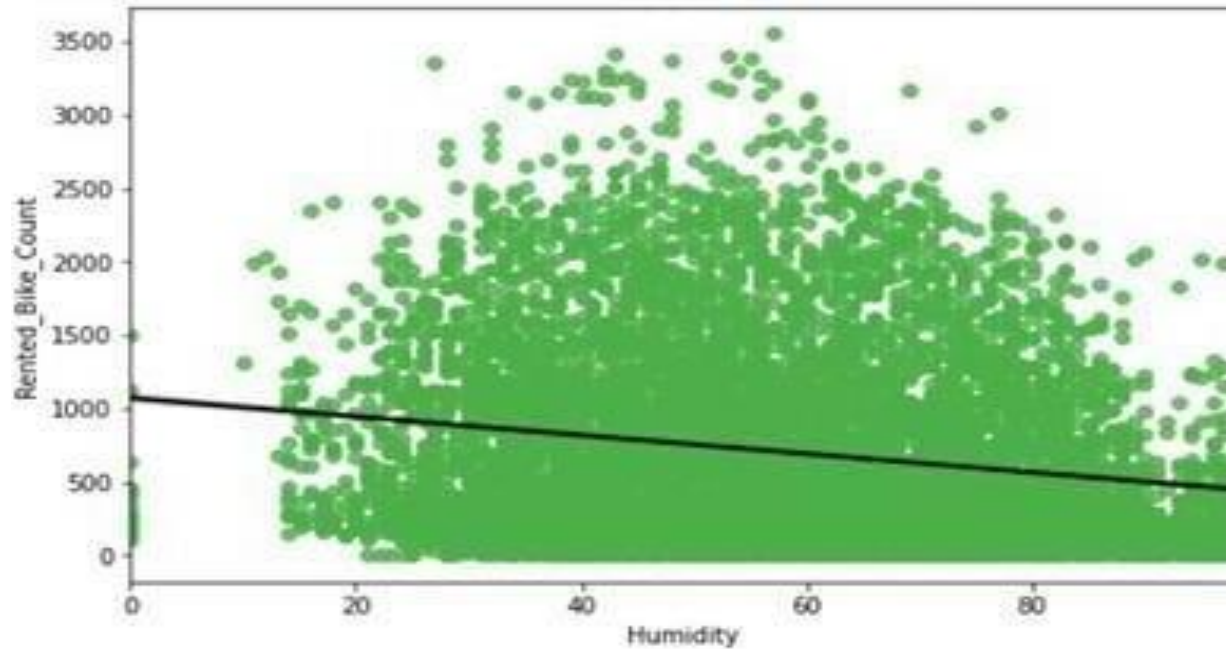
Chart showing distribution of Rented bike count per hour



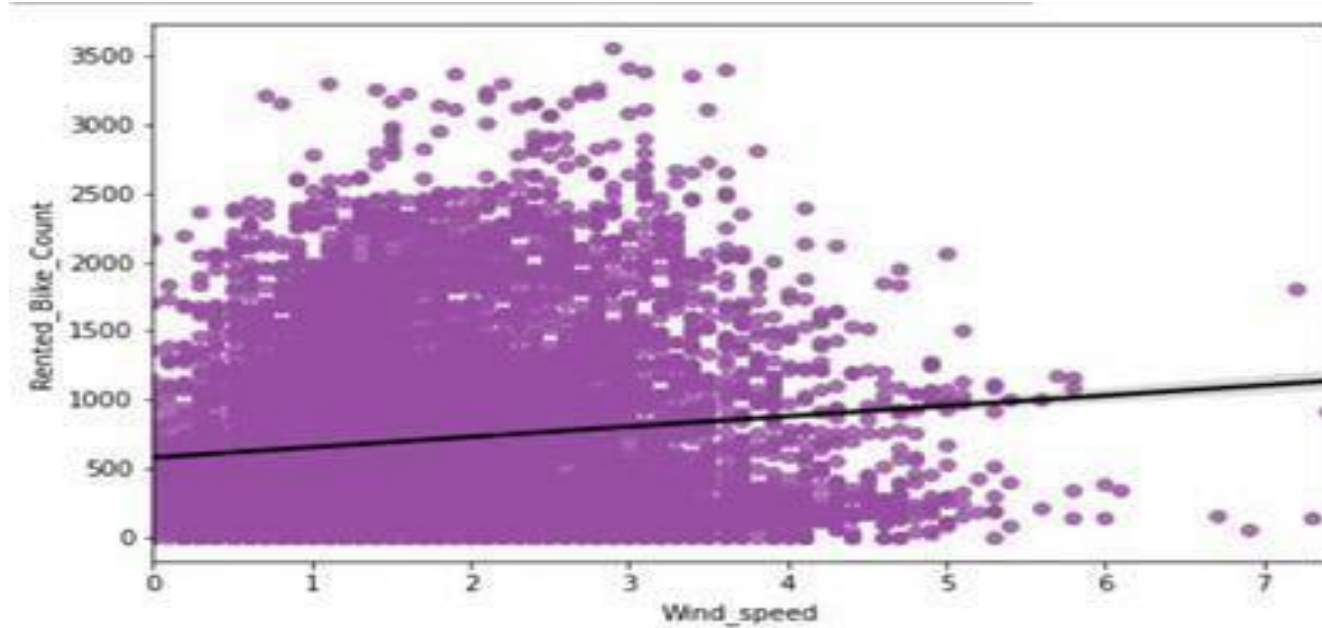
## Relation of our dependent variable with Temperature



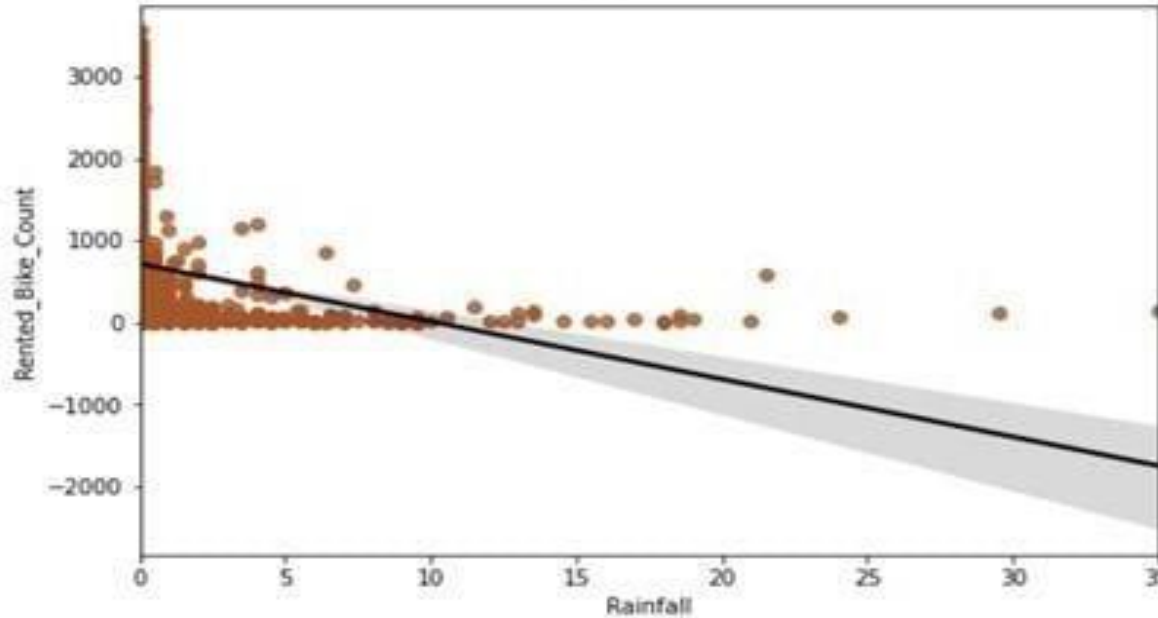
## Relation of our dependent variable with Humidity



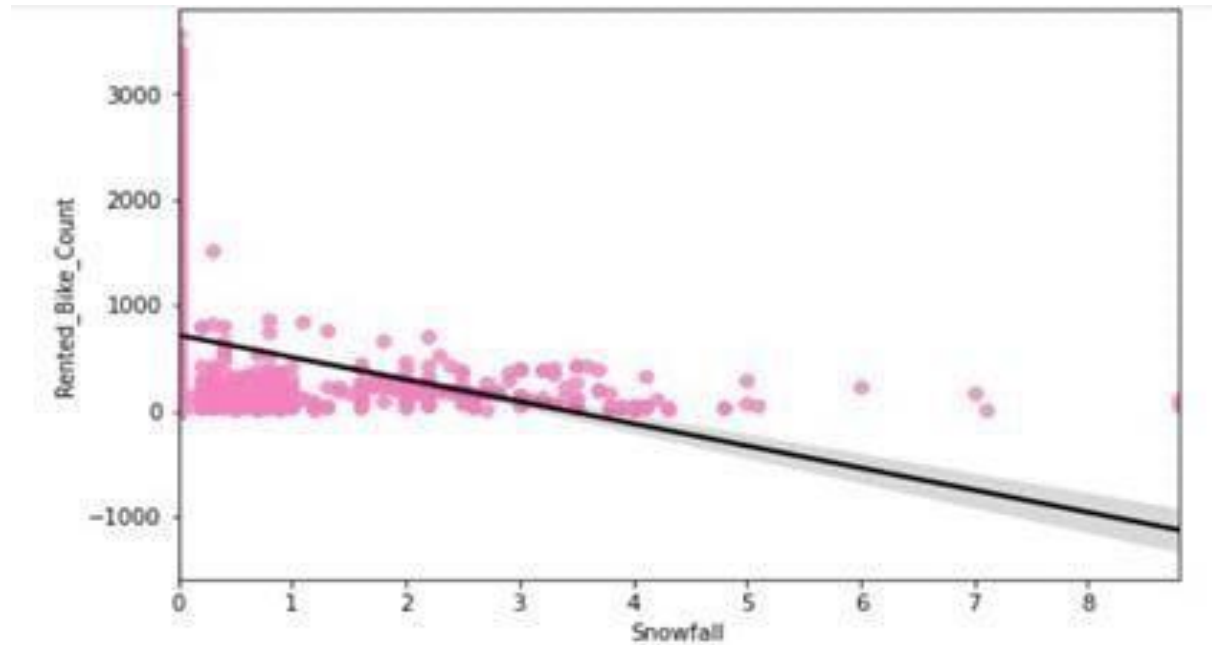
Relation of our dependent variable with wind speed



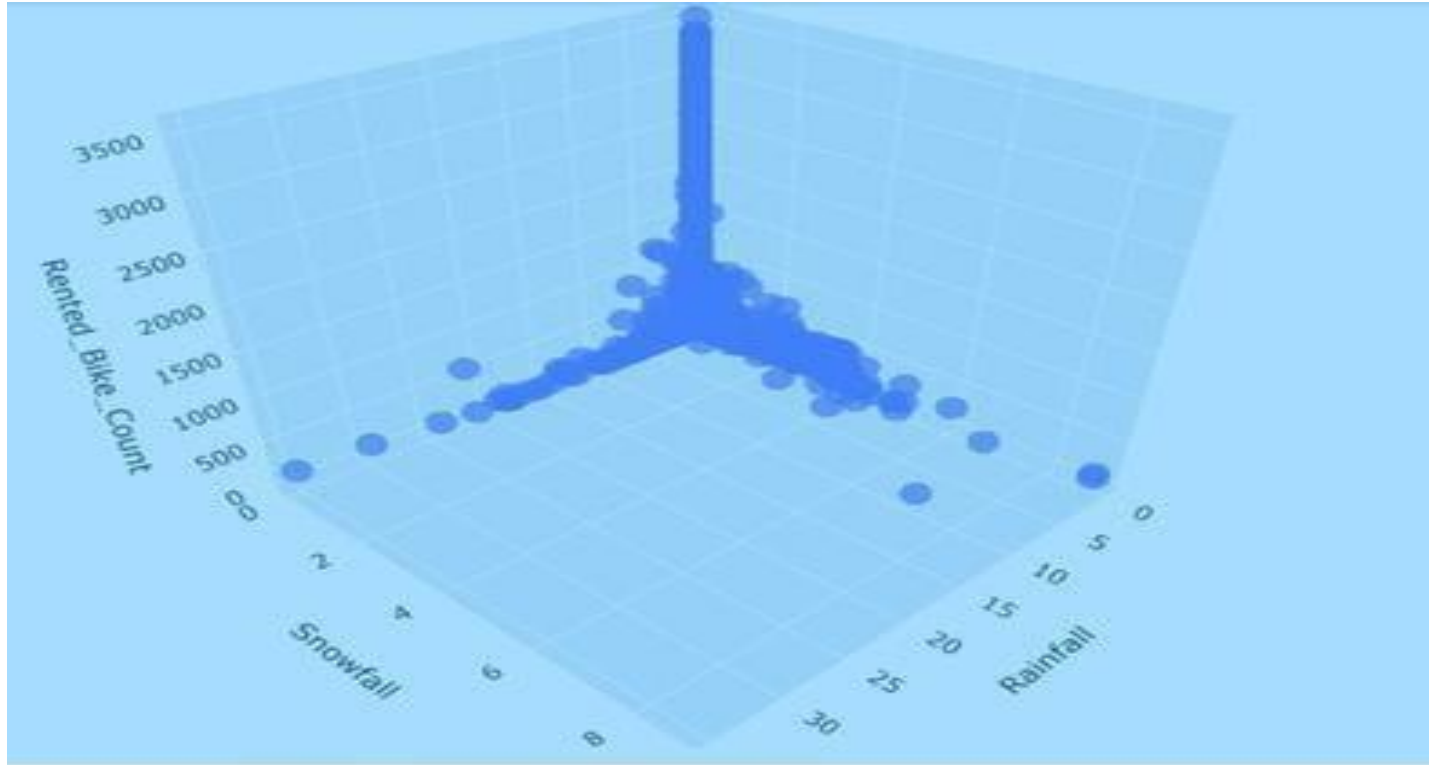
## Relation of our dependent variable with Rainfall



## Relation of our dependent variable with Snowfall

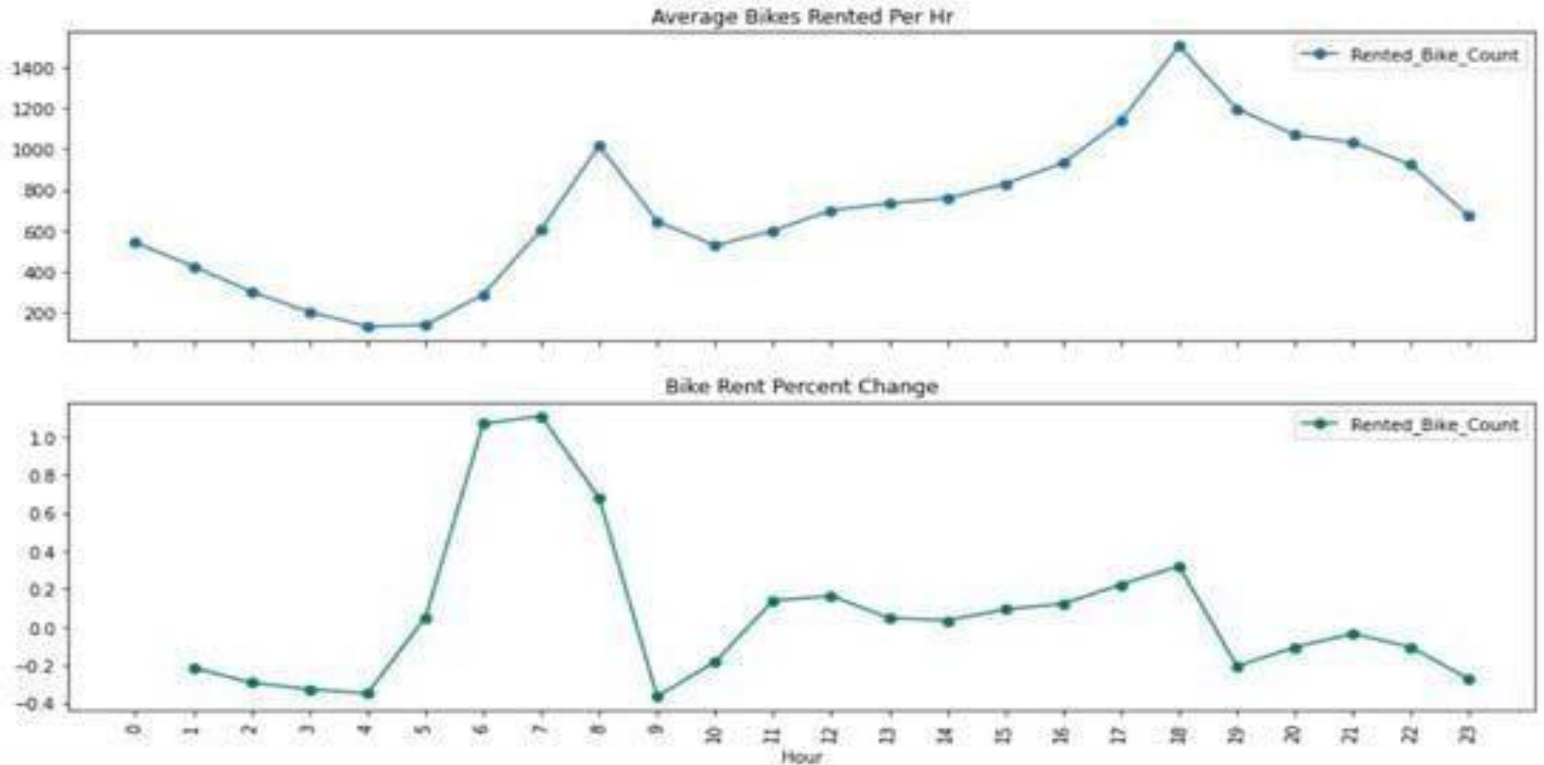


3-d plot showing relation between Snowfall , Rainfall  
and Rented bike count

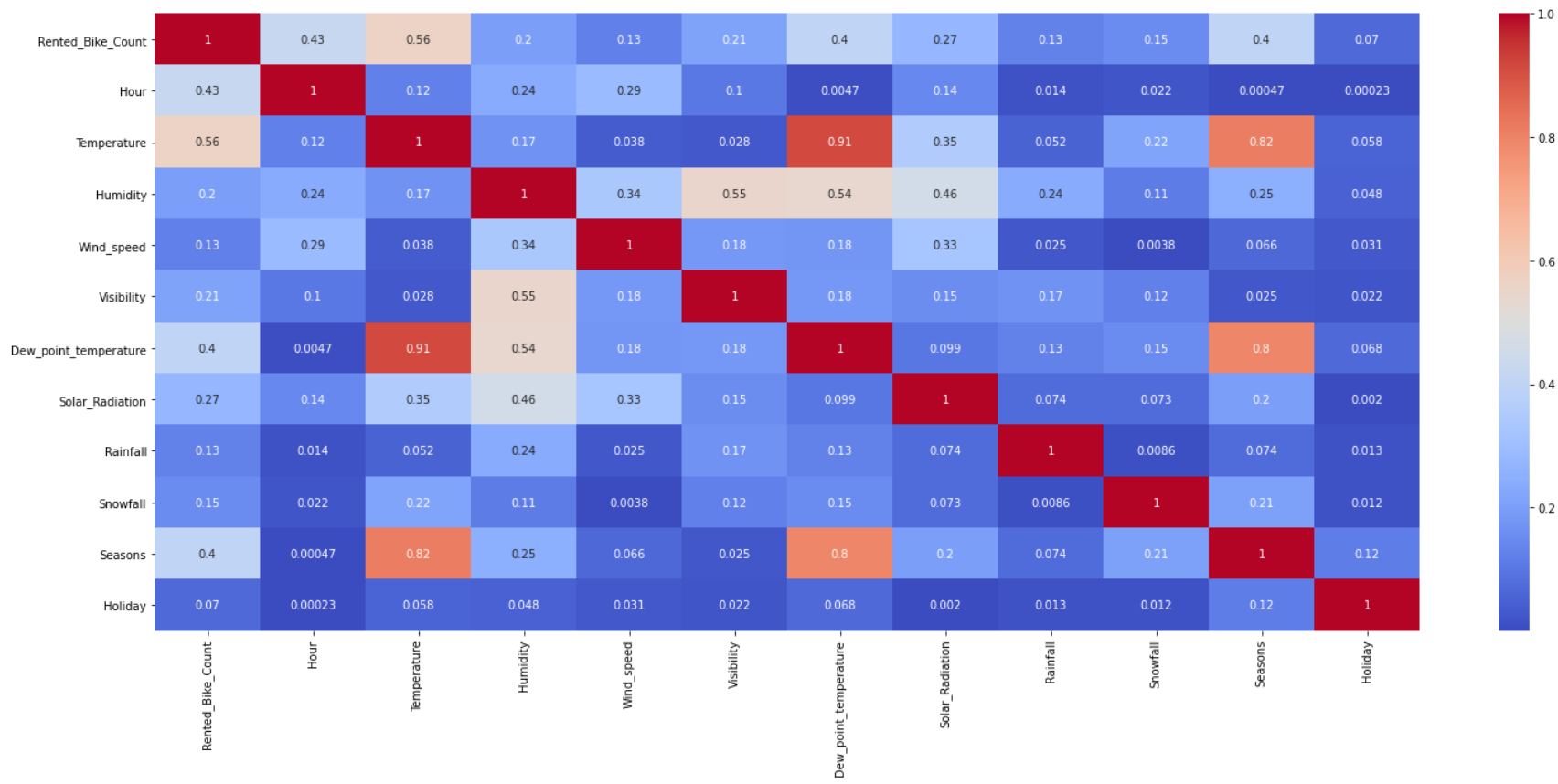




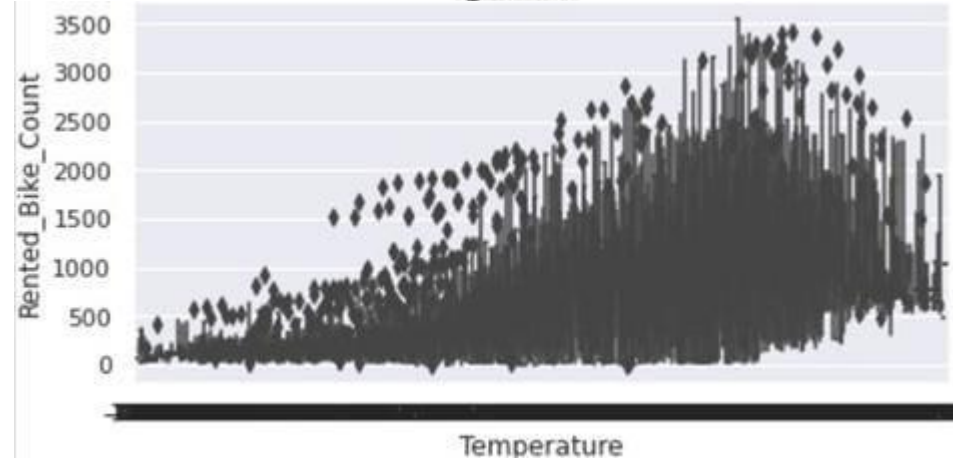
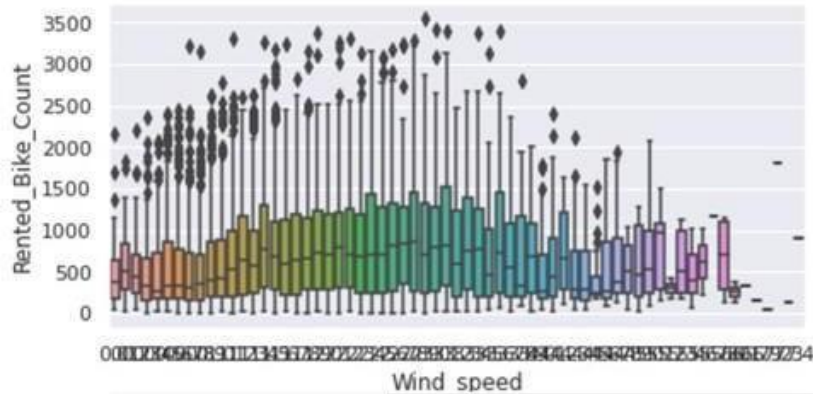
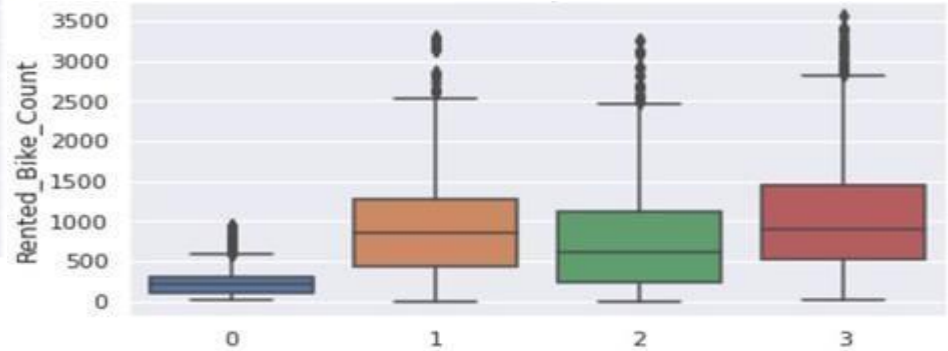
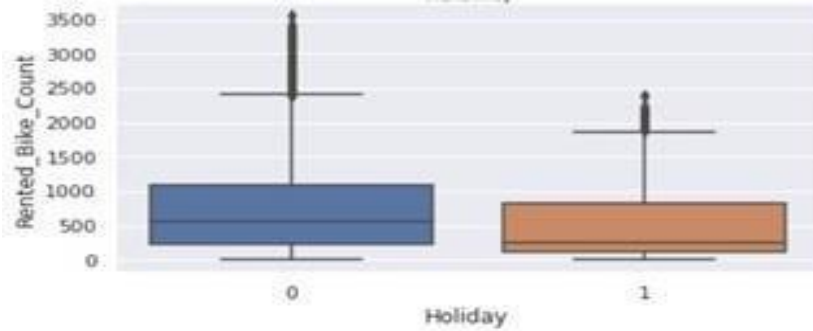
## Per hour distribution



## Correlation between different factors



## Outliers present in our important independent features



## Linear Regression

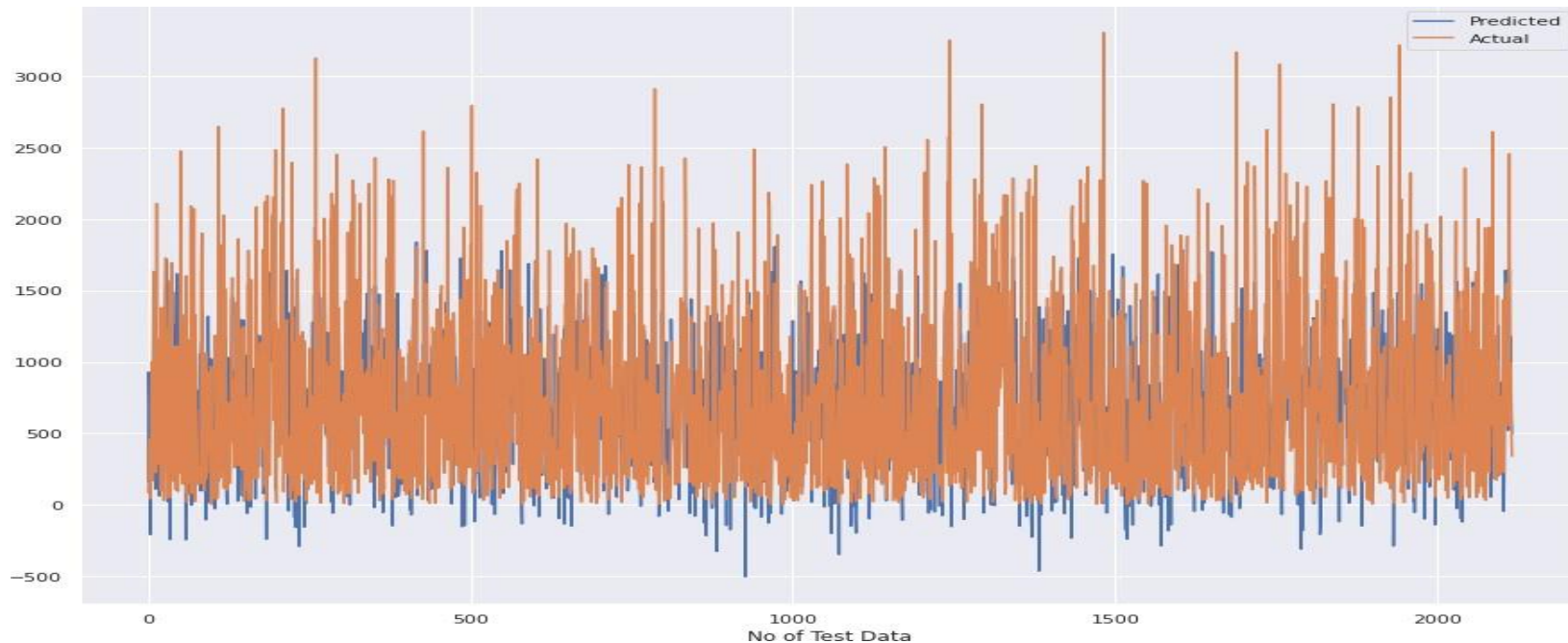
MSE : 198793.5341180045

RMSE : 445.8626852720515

MAE : 333.68919457334323

Adjusted R2 : 0.5049660638596776

r2\_score: 0.5073055437091121



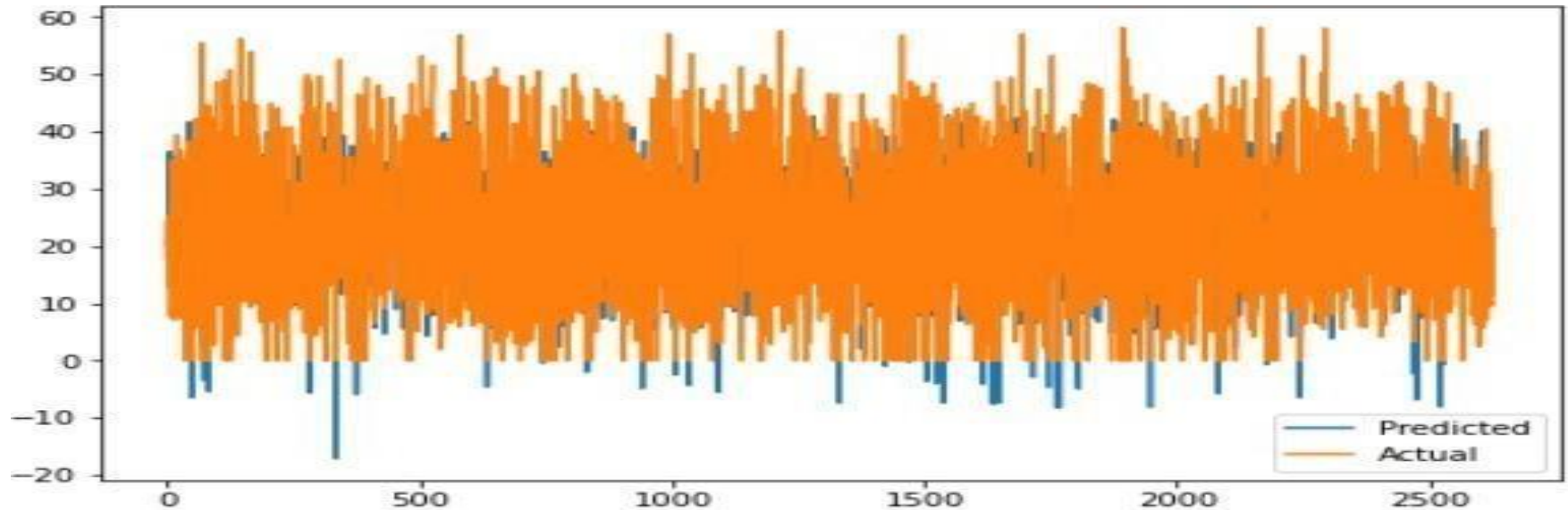
## Lasso Regression

MSE : 198793.663747306

RMSE : 445.86283064111325

r2\_score 0.5073052224328767

MAE : 333.68926336070683



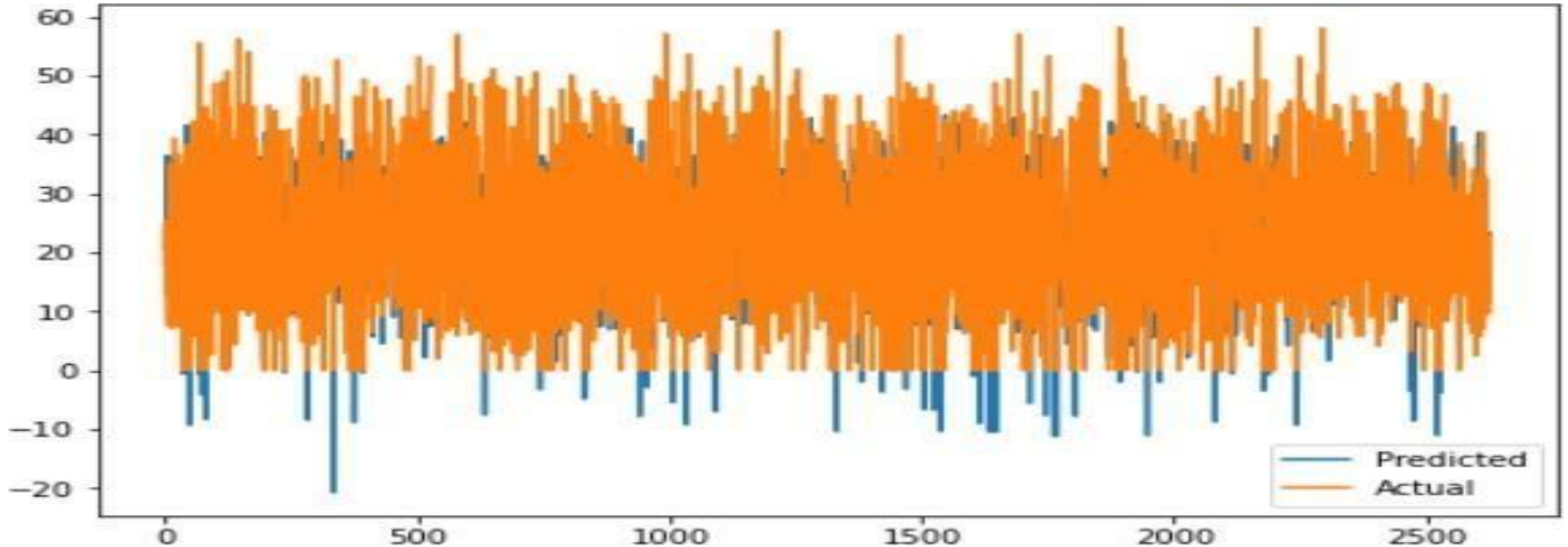
# Ridge Regression

MSE : 198890.40226455292

RMSE : 445.97130206388044

MAE : 333.7678564764892

r2\_score 0.5070654634720594

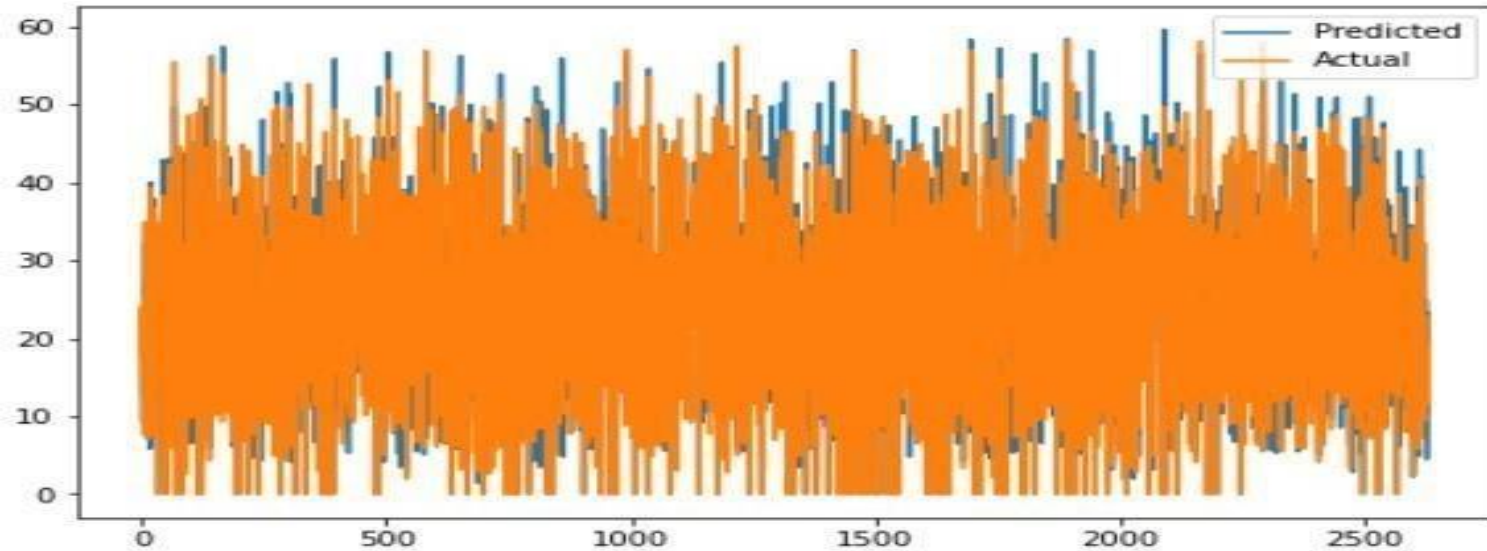


## Decision Tree

MSE : 111943.4251299008

r2\_score:0.7225568466076131

RMSE : 334.579475057722



193.50543221539914

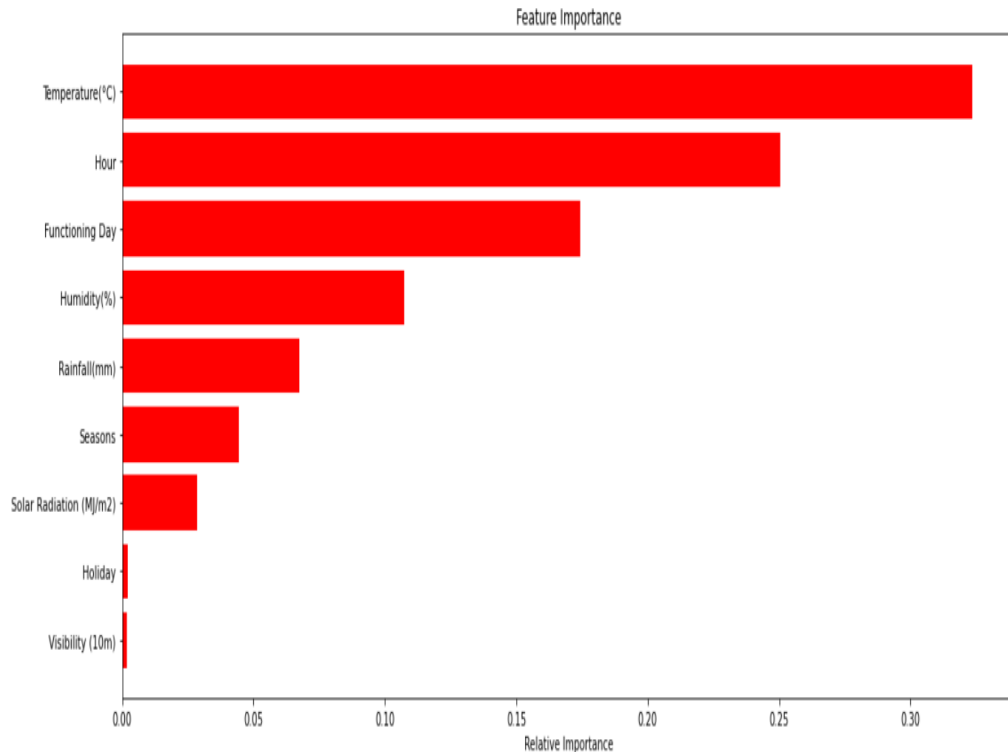
Adjusted R2 : 0.7212394527168611

MAE :



# Gradient Boosting Machine

MAE : 174.081134728031  
MSE : 67935.3191486026  
RMSE : 260.6440468313109  
Adjusted R2 : 0.830828056906927  
r2\_score 0.831627546241016



# Random Forest

MSE : 60132.13303353803

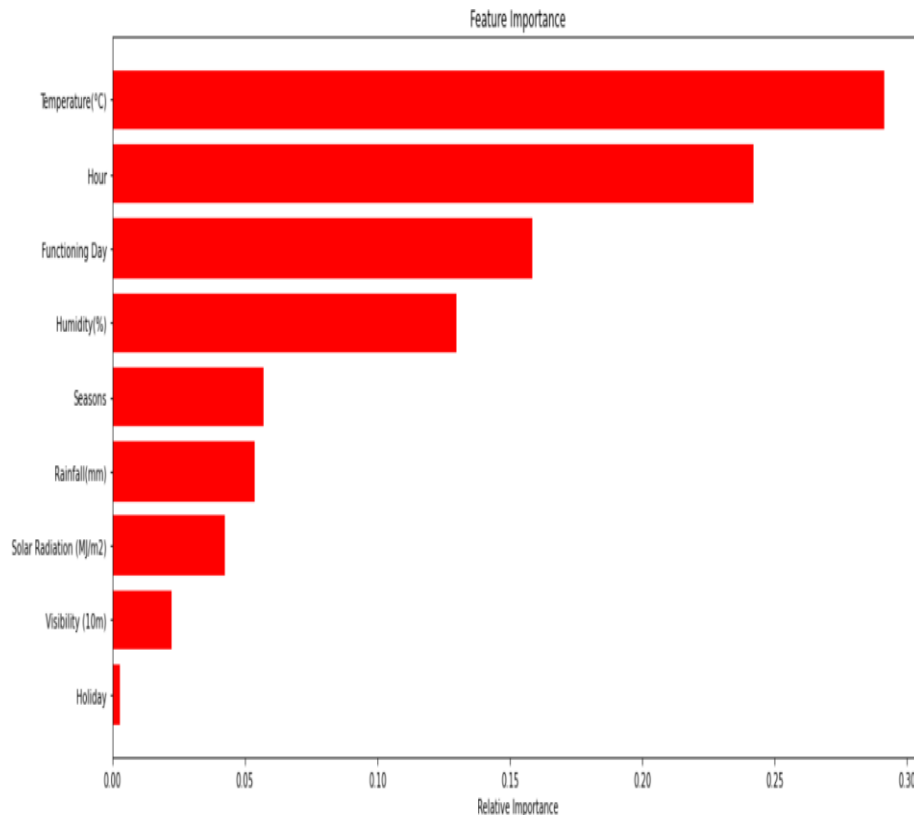
RMSE : 245.21854137388965

MAE : 150.1287009919697

Adjusted R2 : 0.8502594833570604

r2\_score:0.8509671417532936

## Feature Importance



# XGBoost

MSE : 54287.031544213925

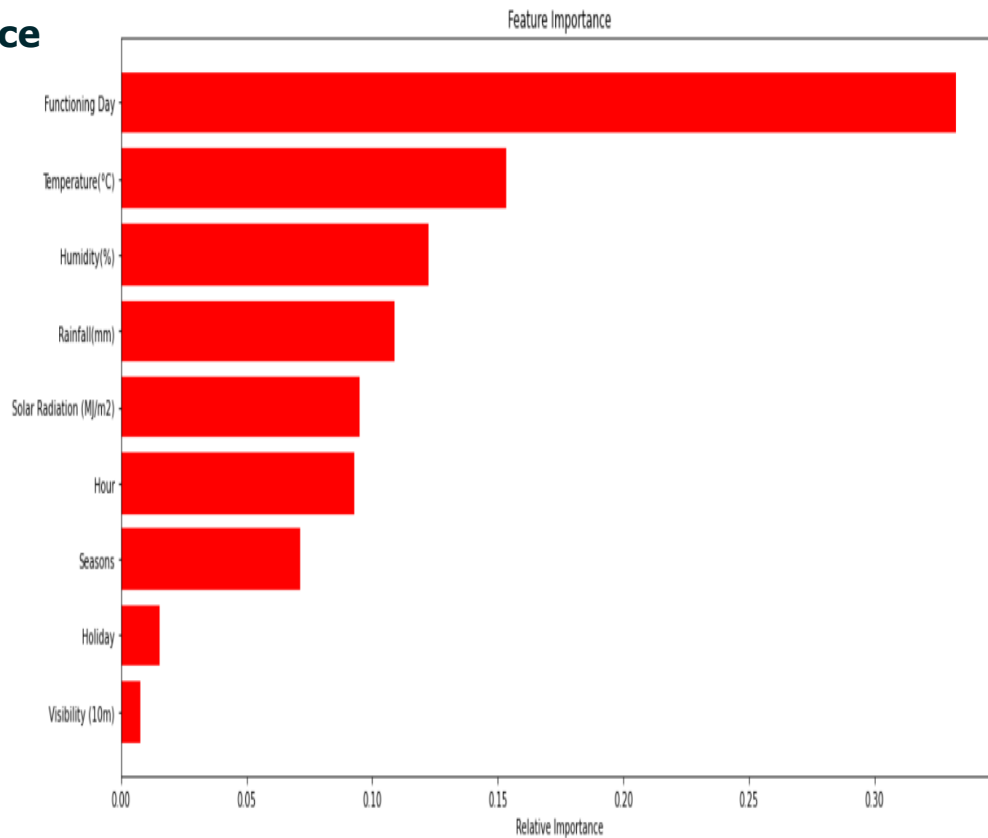
RMSE : 232.9957758076612

MAE : 143.48340080681663

adj\_r2: 0.8657453657658387

r2 :0.8662260483087465

## Feature Importance



## Challenges

- **Large Dataset to handle.**
- **Needs to plot lot of Graphs to analyse.**
- **Carefully handled Feature selection part as it affects the R2 score.**
- **Carefully tuned Hyperparameters as it affects the R2 score.**

## Conclusion

- **The Rented Bike Count has been increased from 2017 to 2018.**
- **No overfitting** is seen.
- **XGBoost Regressor** gives the **highest R2 score** of **96.6%** for Train Set and **89.4%** for Test set.
- **Feature Importance** value for Random Forest, Gradient Boost, and XGBoost **are different.**
- **We can deploy this model.**

**THANK YOU**