

Capstone Project-2 BIKE SHARING DEMAND PREDICTION

TEAM MEMBERS

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- □ BUSINESS UNDERSTANDING
- □ DATA SUMMARY
- ☐ FEATURE ANALYSIS
- ☐ EXPLORATORY DATA ANALYSIS
- □ DATA PREPROCESSING
- ☐ IMPLEMENTING ALGORITHMS
- ☐ CHALLENGES
- □ CONCLUSIONS



BUSINESS UNDERSTANDING

- ➤ Bike rentals have became a popular service in recent years and it seems people are using it more often. With relatively cheaper rates and ease of pick up and drop at own convenience is what making this business thrive.
- > Mostly used by people having no personal vehicles and also to avoid congested public transport which that's why they prefer rental bikes.
- > Therefore, the business to strive and profit more, it has to be always ready and supply no. of bikes at different locations, to fulfil the demand.
- > Our project goal is a pre planned set of bike count values that can be a handy solution to meet all demands.



DATA SUMMARY

	Θ	1	2	3	4	5	6	7	8	9
Date	01/12/2017	01/12/2017	01/12/2017	01/12/2017	01/12/2017	01/12/2017	01/12/2017	01/12/2017	01/12/2017	01/12/2017
Rented Bike Count	254	204	173	107	78	100	181	460	930	490
Hour	0	1	2	3	4	5	6	7	8	9
Temperature(°C)	-5.2	-5.5	-6.0	-6.2	-6.0	-6.4	-6.6	-7.4	-7.6	-6.5
Humidity(%)	37	38	39	40	36	37	35	38	37	27
Wind speed (m/s)	2.2	0.8	1.0	0.9	2.3	1.5	1.3	0.9	1.1	0.5
Visibility (10m)	2000	2000	2000	2000	2000	2000	2000	2000	2000	1928
Dew point temperature(°C)	-17.6	-17.6	-17.7	-17.6	-18.6	-18.7	-19.5	-19.3	-19.8	-22.4
Solar Radiation (MJ/m2)	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.01	0.23
Rainfall(mm)	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Snowfall (cm)	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Seasons	Winter									
Holiday	No Holiday	No Holiday	No Holiday	No Holiday	No Holiday	No Holiday	No Holiday	No Holiday	No Holiday	No Holiday
Functioning Day	Yes									

- > This Dataset contains 8760 lines and 14 columns.
- ➤ Three categorical features 'Seasons', 'Holiday', & 'Functioning Day'.
- **➤** One Datetime features 'Date'.
- > We have some numerical type variables such as temperature, humidity, wind, visibility, dew point temp, solar radiation, rainfall, snowfall which tells the environment conditions at that particular hour of the day.

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FEATURE 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 %
- > Wind Speed 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
- > Functioning Day

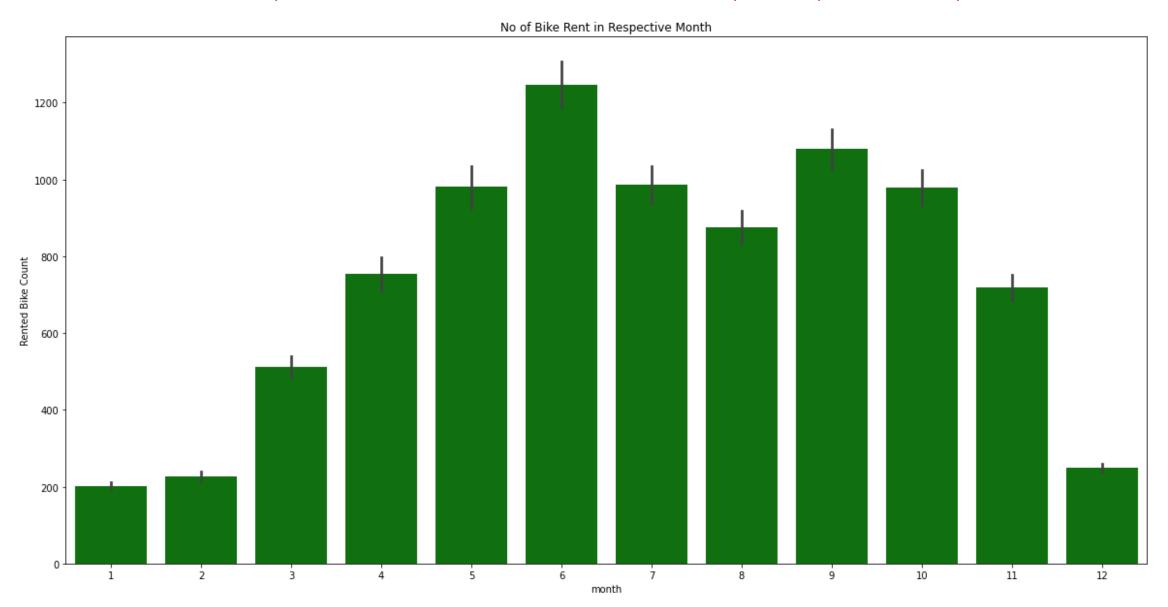
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INSIGHTS FROM OUR DATASET

- > There are No Missing Values present.
- > There are No Duplicate values present.
- > There are No null values.
- > And finally we have 'rented bike count' variable which we need to predict for new observations.
- ➤ The dataset shows hourly rental data for one year (1 December 2017 to 31 November(2018)(365 days).we consider this as a single year data
- > So we convert the "date" column into 3 different column i.e "year", "month", "day".
- ➤ We change the name of some features for our convenience, they are as below 'Rented_Bike_Count', 'Hour', 'Temperature', 'Humidity', 'Wind_speed', 'Visibility', 'Dew_point_temperature', 'Solar_Radiation', 'Rainfall', 'Snowfall', 'Seasons', 'Holiday', 'Functioning_Day', 'month', 'weekdays_weekend'



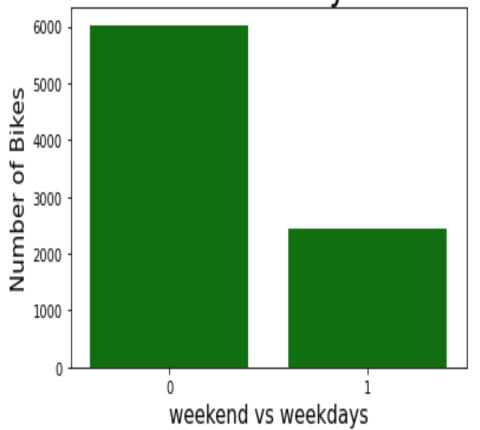
ANALYSIS OF MONTH VS RENTED BIKE

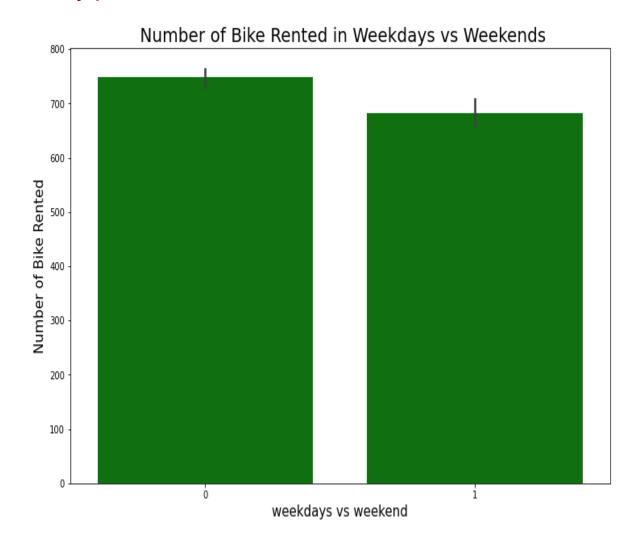




(Weekend vs weekdays)

Number of Weekend days vs Weekdays

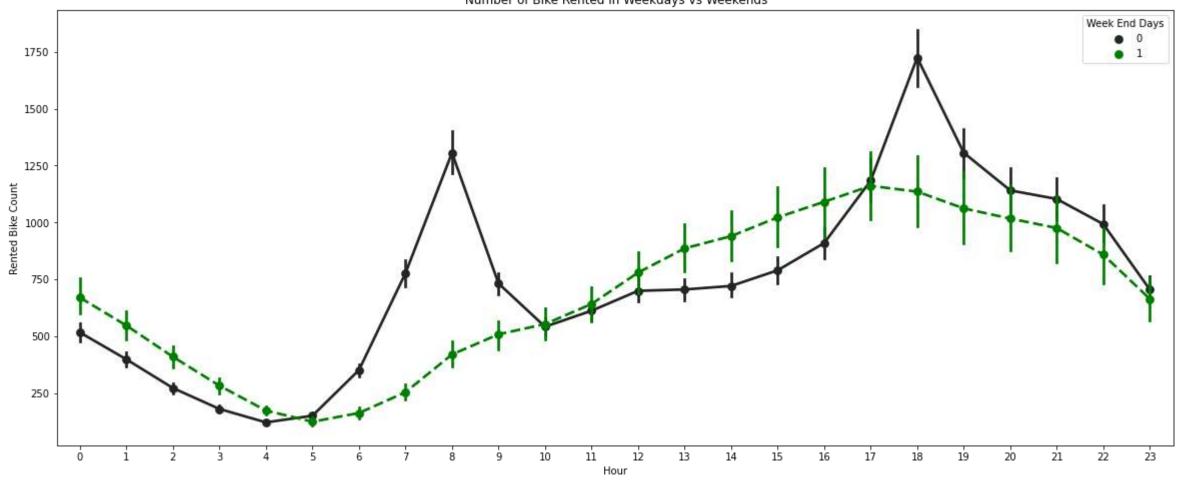






(HOUR)

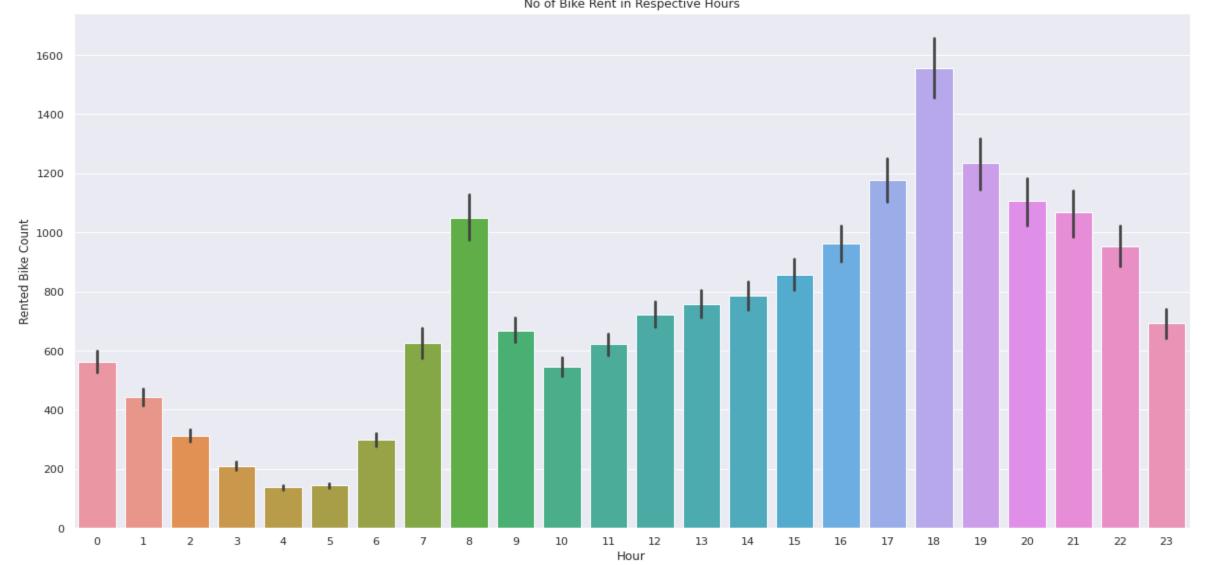






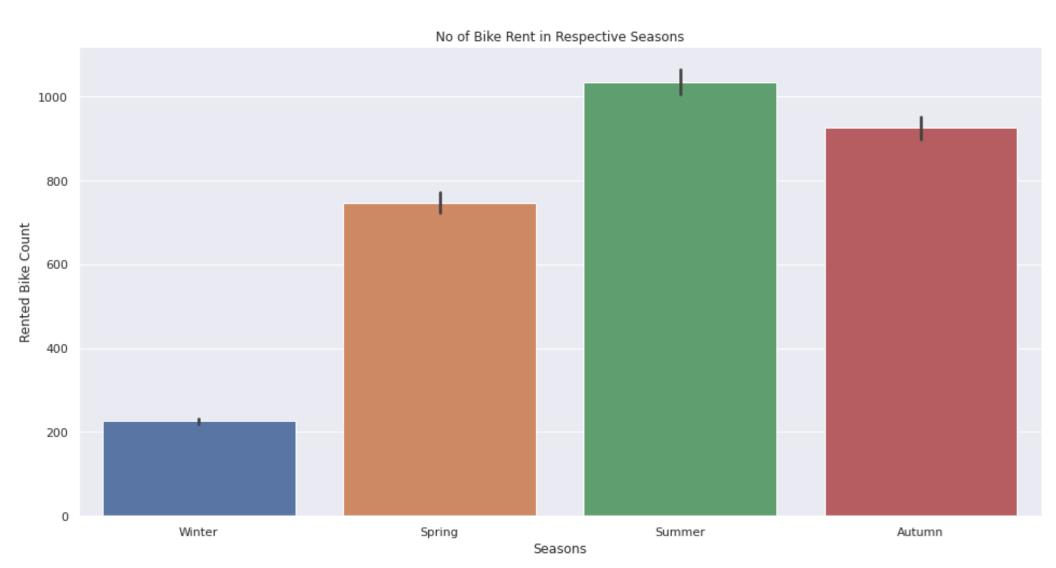
(HOUR)

No of Bike Rent in Respective Hours





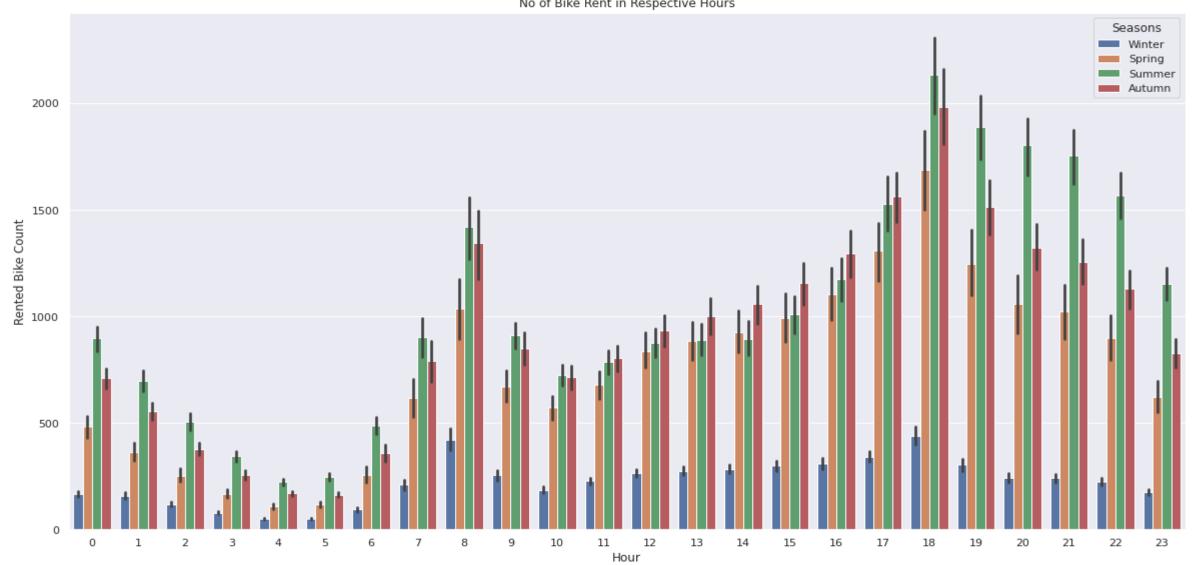
(SEASON)





(Hour and Season)

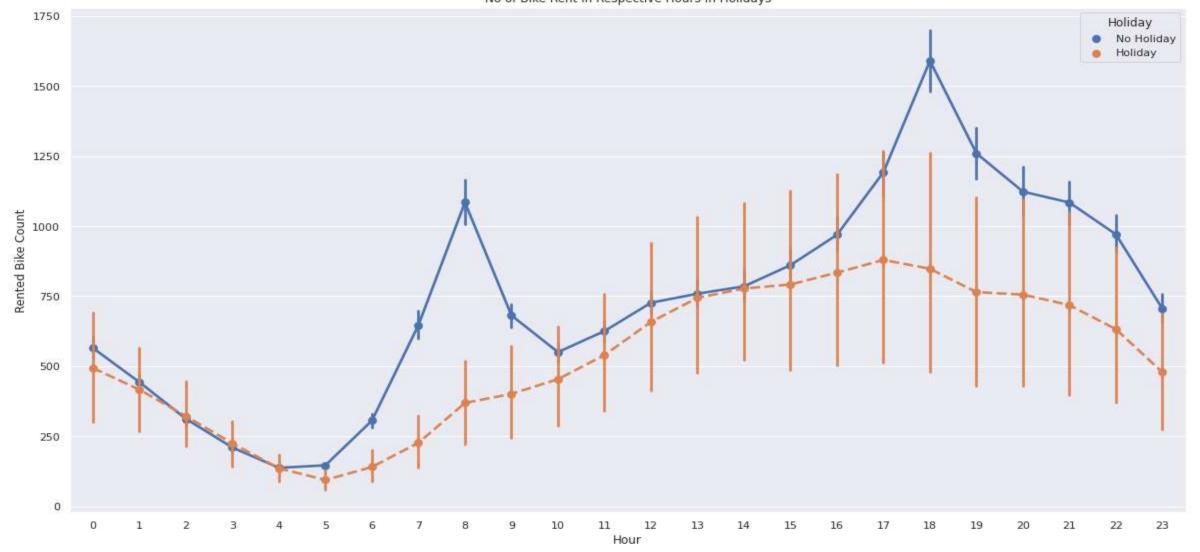






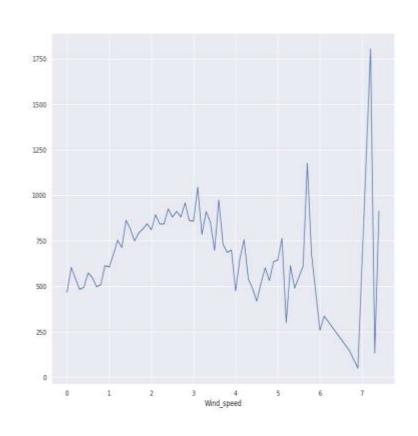
(HOUR AND HOLIDAY)

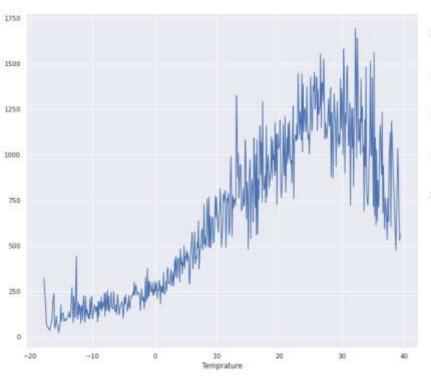


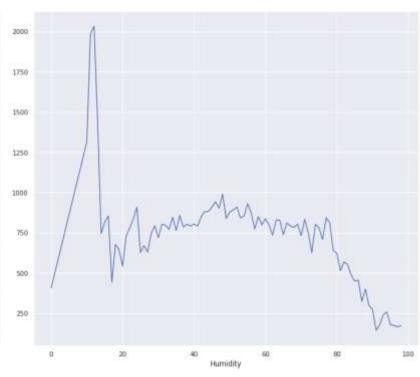




NUMERICAL VS RENTED BIKE COUNT

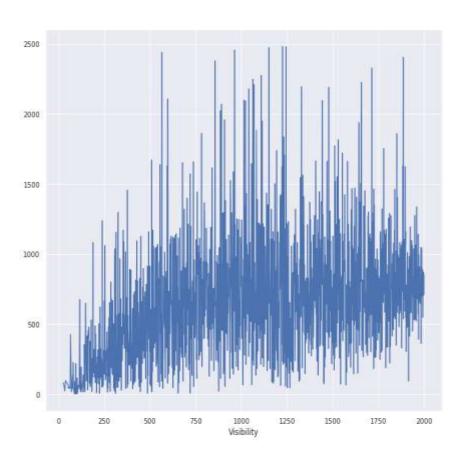


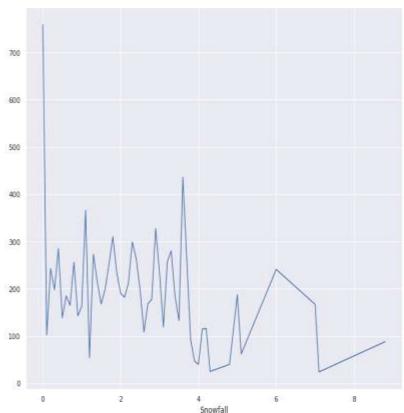


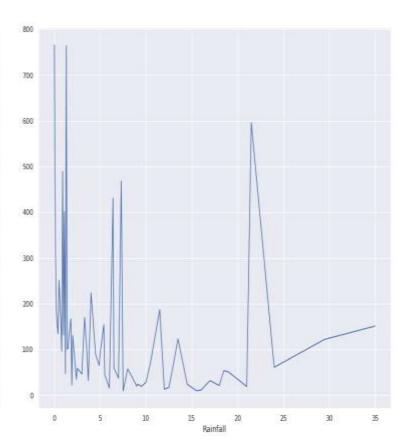




NUMERICAL VS RENTED BIKE COUNT

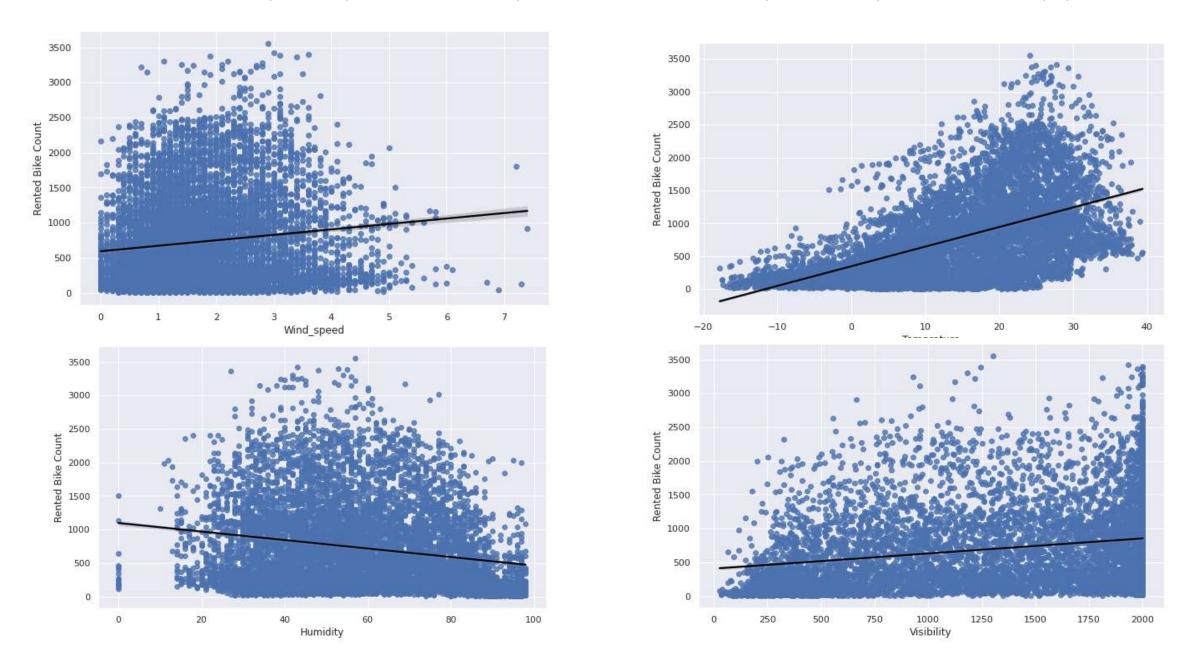






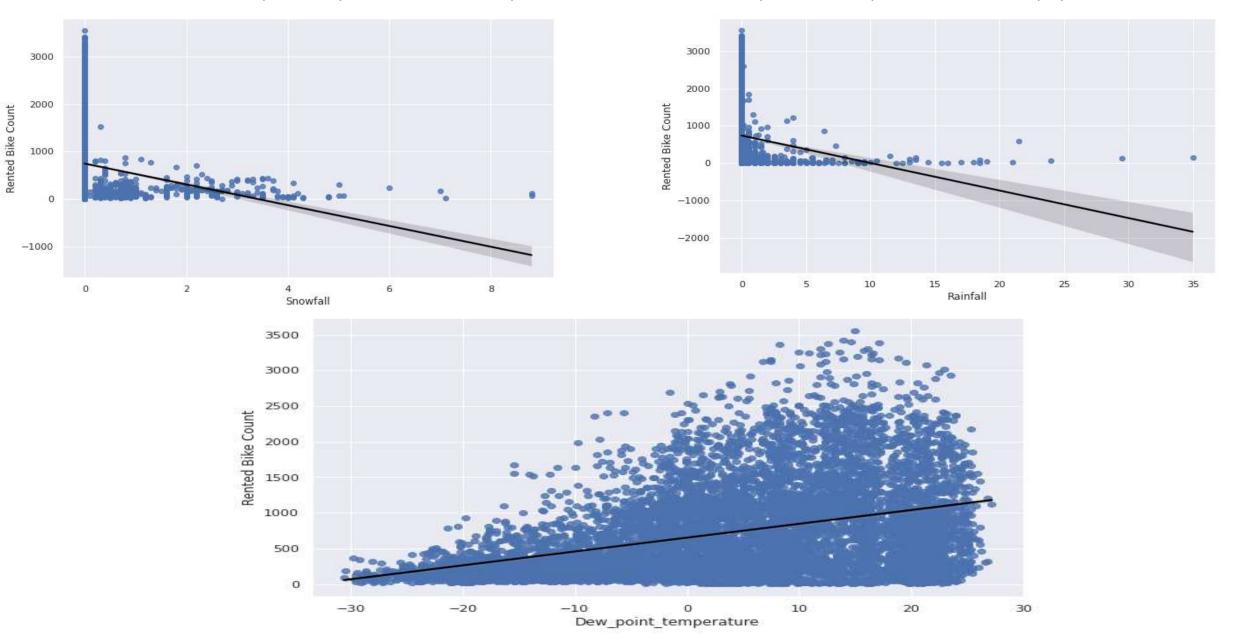
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REGRESSION PLOT FOR NUMERICAL VARIABLE





REGRESSION PLOT FOR NUMERICAL VARIABLE



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- 0.2

CORRELATION HEATMAP

Rented Bike Count	1	0.56	0.2	0.13	0.21	0.4	0.27	0.13	0.15
Temprature	0.56	1	0.17	0.038	0.028	0.91	0.35	0.052	0.22
Humidity	0.2	0.17	1	0.34	0.55	0.54	0.46	0.24	0.11
Wind_speed	0.13	0.038	0.34	1	0.18	0.18	0.33	0.025	0.0038
Visibility	0.21	0.028	0.55	0.18	1	0.18	0.15	0.17	0.12
Dew_point_temperature	0.4	0.91	0.54	0.18	0.18	1	0.099	0.13	0.15
Solar Radiation (MJ/m2)	0.27	0.35	0.46	0.33	0.15	0.099	1	0.074	0.073
Rainfall	0.13	0.052	0.24	0.025	0.17	0.13	0.074	1	0.0086
Snowfall	0.15	0.22	0.11	0.0038	0.12	0.15	0.073	0.0086	1
	Rented Bike Count	Emprature	Humidity	Wind_speed	Visibility	Dew_point_temperature	Solar Radiation (MJ/m2)	Rainfall	Snowfall



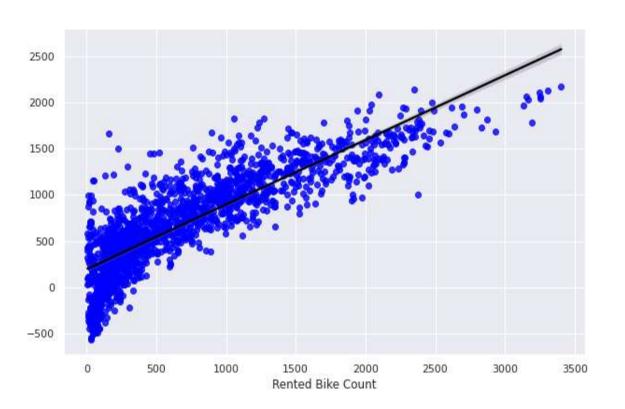
MODEL BUILDING

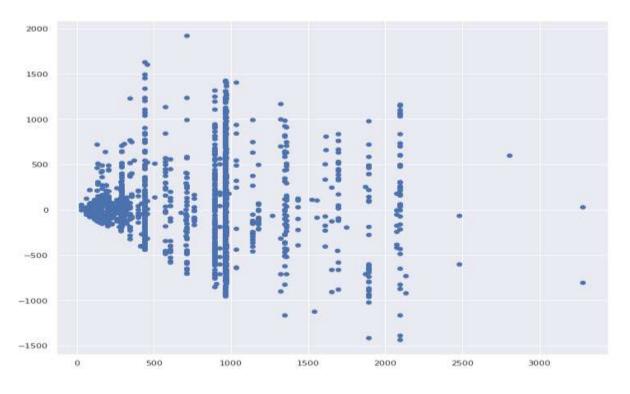
- > LINEAR REGRESSION
- > LASSO REGRESSION
- > RIDGE REGRESSION
- > DECISION TREES REGRESSOR
- > RANDOM FOREST REGRESSOR
- > GRADIENT BOOSTED REGRESSOR
- > GRADIENT BOOSTING REGRESSOR WITH GRIDSEARCHCV

DECISION TREE

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LINEAR REGRESSION





MSE: 123688.20499160145

RMSE: 123688.20499160145

MAE : 266.2024881866509

R2: 0.713929384414175

Adjusted R2 : 0.7050387071473394

MSE: 203680.2100098523

RMSE: 451.3094393094967

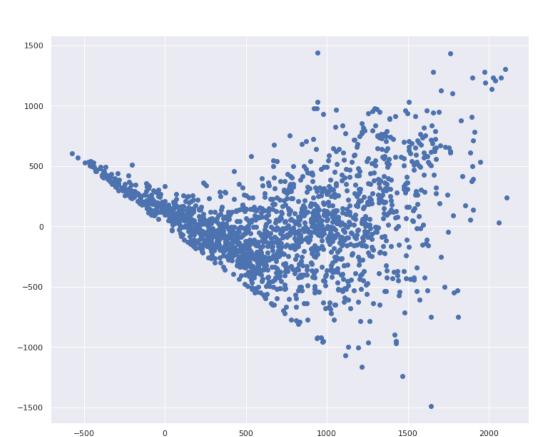
MAE : 314.06006792277816

R2: 0.5289209422666861

Adjusted R2 : 0.5142804596680273

LASSO REGRESSION

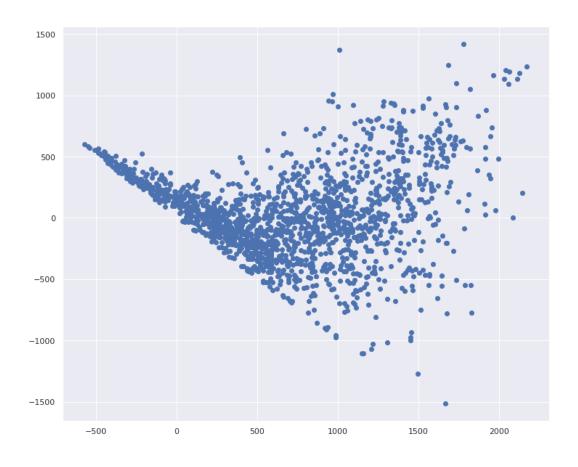




MSE: 126977.54006523405 RMSE: 356.3390801823932 MAE: 268.42480406043046 R2: 0.7063216896509918

Adjusted R2: 0.6971945758010227

RIDGE REGRESSION



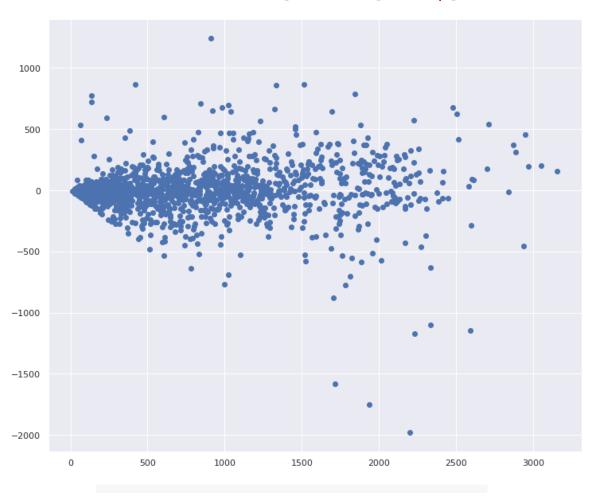
MSE: 123730.48290489083 RMSE: 351.7534405018533 MAE: 266.21508579231676 R2: 0.7138316025061814

Adjusted R2: 0.7049378863135033

RANDOM FOREST

RANDOM FOREST WITH TUNING







MSE : 37965.17489125812 RMSE : 194.84654190223165 MAE : 115.0099291199055

R2 : 0.9121927515020274

Adjusted R2: 0.9094638242178125

MSE : 36490.735509509745

RMSE : 191.02548392690895 MAE : 121.89040756054341

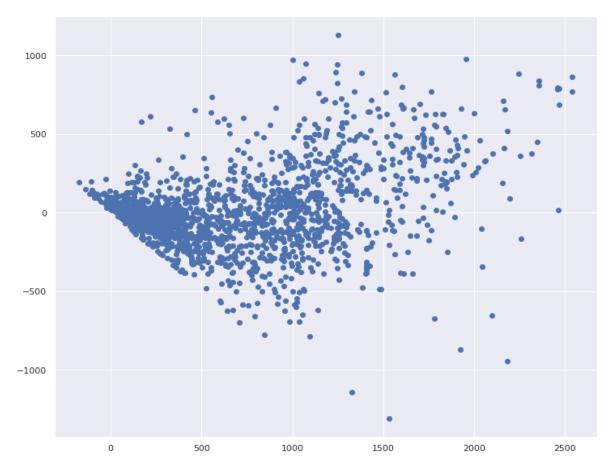
R2: 0.9156028889650893

Adjusted R2 : 0.9129799440151927

GRADIENT BOOSTING

GRADIENT BOOSTING WITH TUNING





-500-1000 -15001000 1500 3000

MSE : 67552.40626356455 RMSE : 259.9084574683259 MAE : 185.0604581428772

R2: 0.8437623179555898

Adjusted R2 : 0.8389066678737708

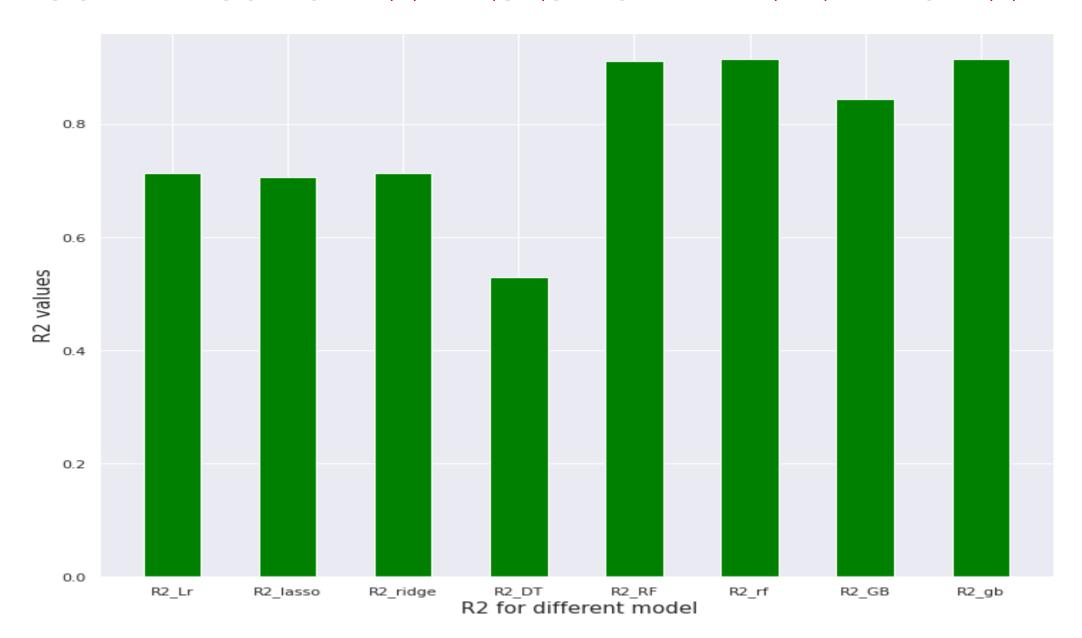
MSE : 36490.735509509745 RMSE : 191.02548392690895 MAE : 121.89040756054341

R2: 0.9156028889650893

Adjusted R2 : 0.9129799440151927



COMPARISON OF R2 VALUES FOR DIFFERENT MODEL





CONCLUSION

- > 'Hour' of the day holds the most important feature.
- > Bike rental count is mostly correlated with the time of the day as it is peak at 10 am morning and 8 pm at evening.
- > We observed that bike rental count is high during working days than non working day.
- > We see that people generally prefer to bike at moderate to high temperatures, and when little windy
- ➤ It is observed that highest number bike rentals counts in Autumn & Summer seasons & the lowest in winter season. We observed that the highest number of bike rentals on a clear day and the lowest on a snowy or rainy day. We observed that with increasing humidity, the number of bike rental counts decreases.



THANK YOU