

# Seoul Bike Sharing Demand

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# Ins and outs of the problem

- Prediction of the needs of the clients in bike rentals
- Work on the efficiency of the companies to answer the customers needs without failing in having items



# A very documented dataset to work on

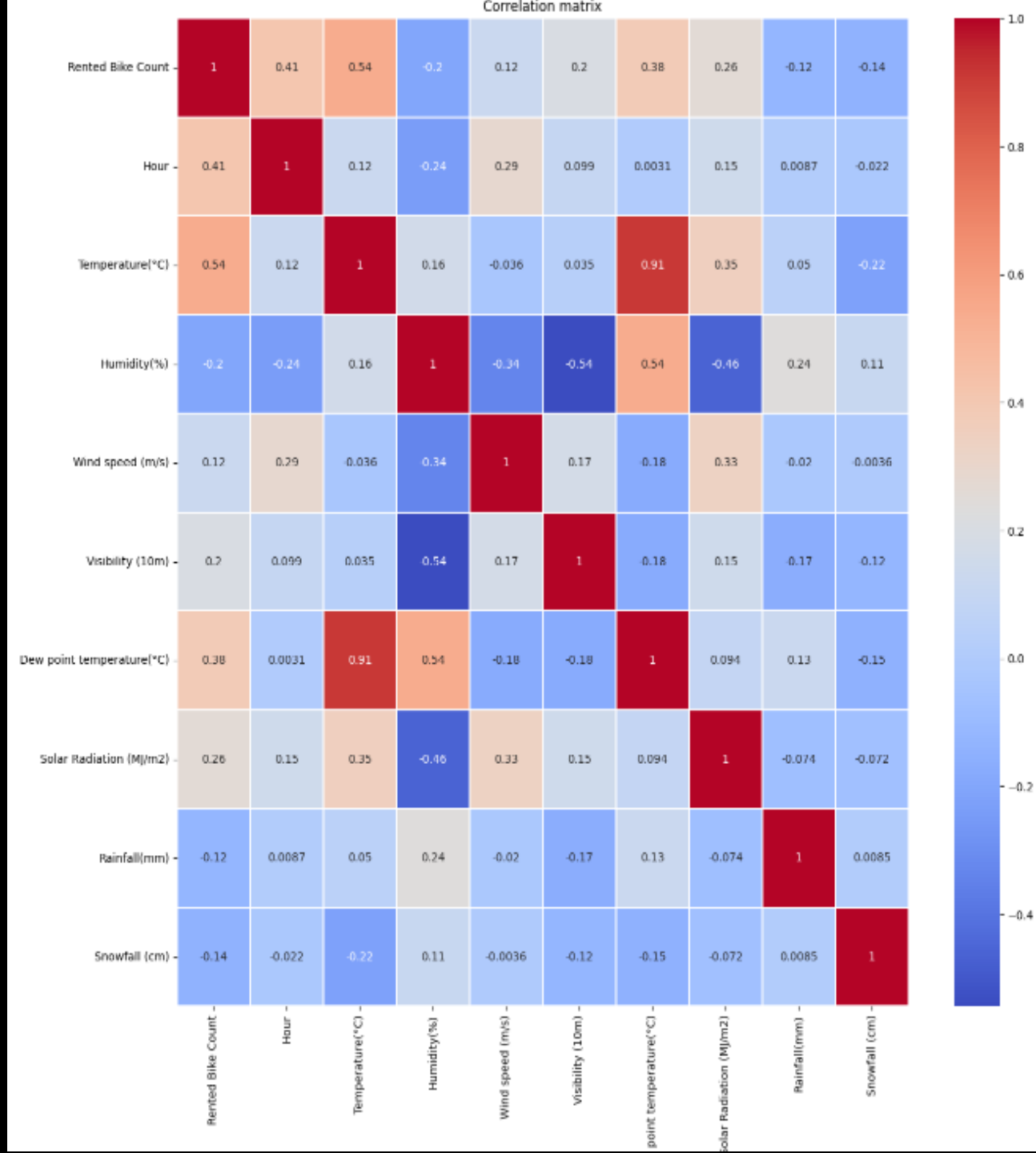
	Date	Rented Bike Count	Hour	Temperature(°C)	Humidity(%)	Wind speed (m/s)	Visibility (10m)	Dew point temperature(°C)	Solar Radiation (MJ/m2)	Rainfall(mm)	Snowfall (cm)	Seasons	Holiday	Functioning Day
0	01/12/2017	254	0	-5.2	37	2.2	2000	-17.6	0.0	0.0	0.0	Winter	No Holiday	Yes
1	01/12/2017	204	1	-5.5	38	0.8	2000	-17.6	0.0	0.0	0.0	Winter	No Holiday	Yes
2	01/12/2017	173	2	-6.0	39	1.0	2000	-17.7	0.0	0.0	0.0	Winter	No Holiday	Yes
3	01/12/2017	107	3	-6.2	40	0.9	2000	-17.6	0.0	0.0	0.0	Winter	No Holiday	Yes
4	01/12/2017	78	4	-6.0	36	2.3	2000	-18.6	0.0	0.0	0.0	Winter	No Holiday	Yes
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
8755	30/11/2018	1003	19	4.2	34	2.6	1894	-10.3	0.0	0.0	0.0	Autumn	No Holiday	Yes
8756	30/11/2018	764	20	3.4	37	2.3	2000	-9.9	0.0	0.0	0.0	Autumn	No Holiday	Yes
8757	30/11/2018	694	21	2.6	39	0.3	1968	-9.9	0.0	0.0	0.0	Autumn	No Holiday	Yes
8758	30/11/2018	712	22	2.1	41	1.0	1859	-9.8	0.0	0.0	0.0	Autumn	No Holiday	Yes
8759	30/11/2018	584	23	1.9	43	1.3	1909	-9.3	0.0	0.0	0.0	Autumn	No Holiday	Yes



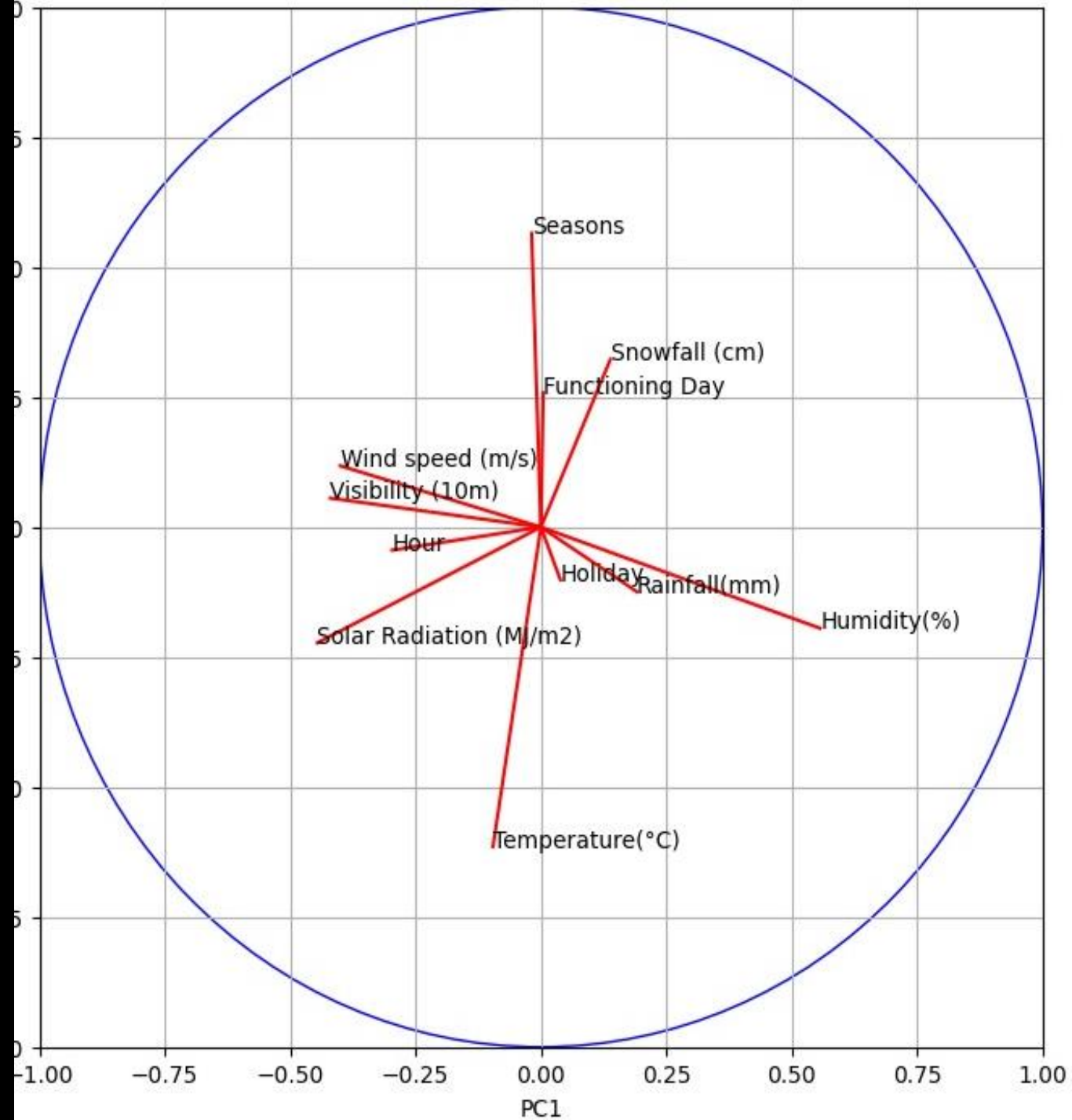
# Data preprocessing:

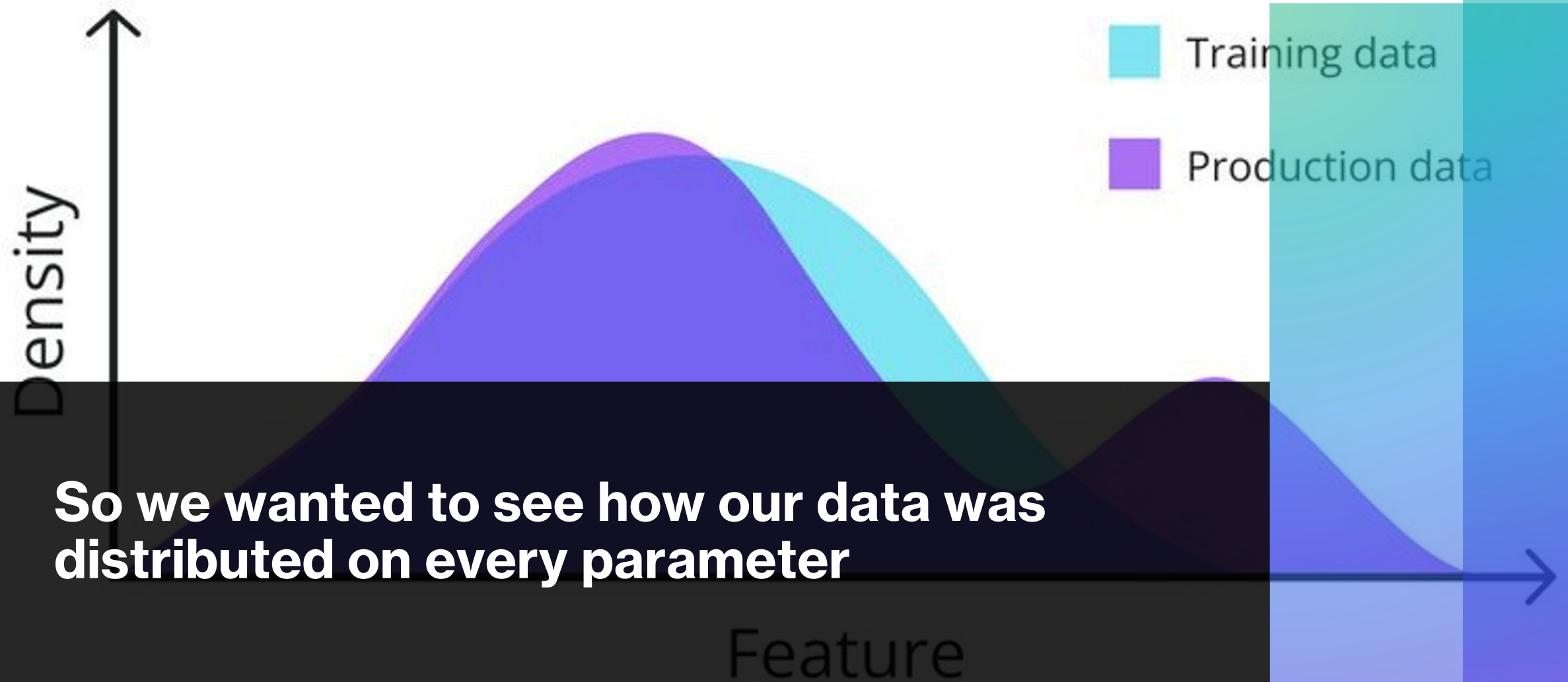
- No missing value
- Normalization of the data set
- Encoding the categorical variables
- Creation of a subset with only numerical values

# Correlation Matrix



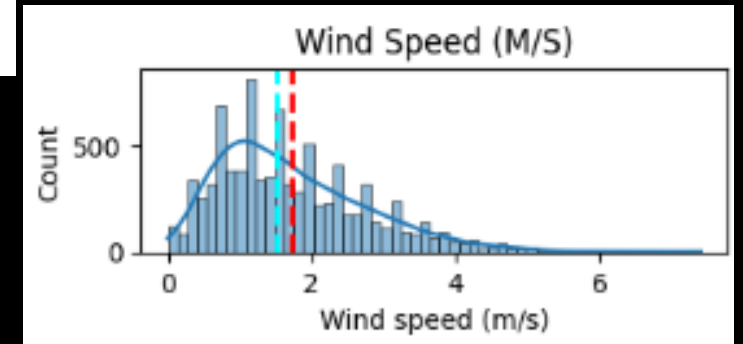
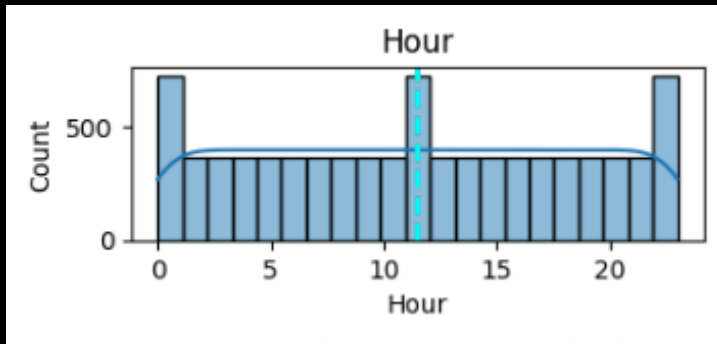
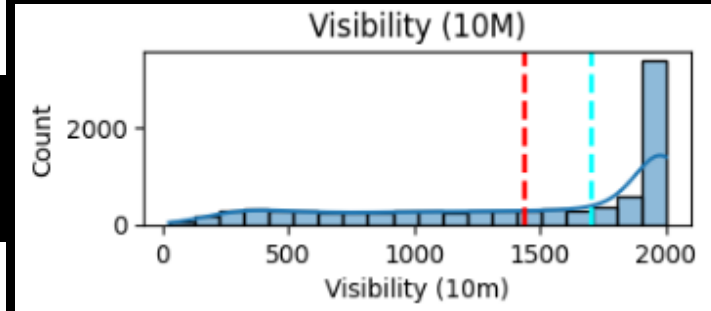
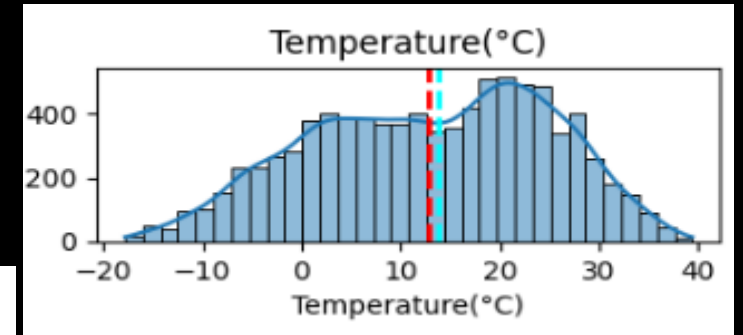
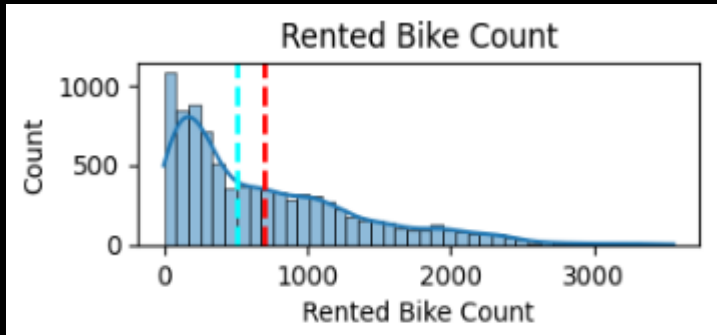
# Correlation circle





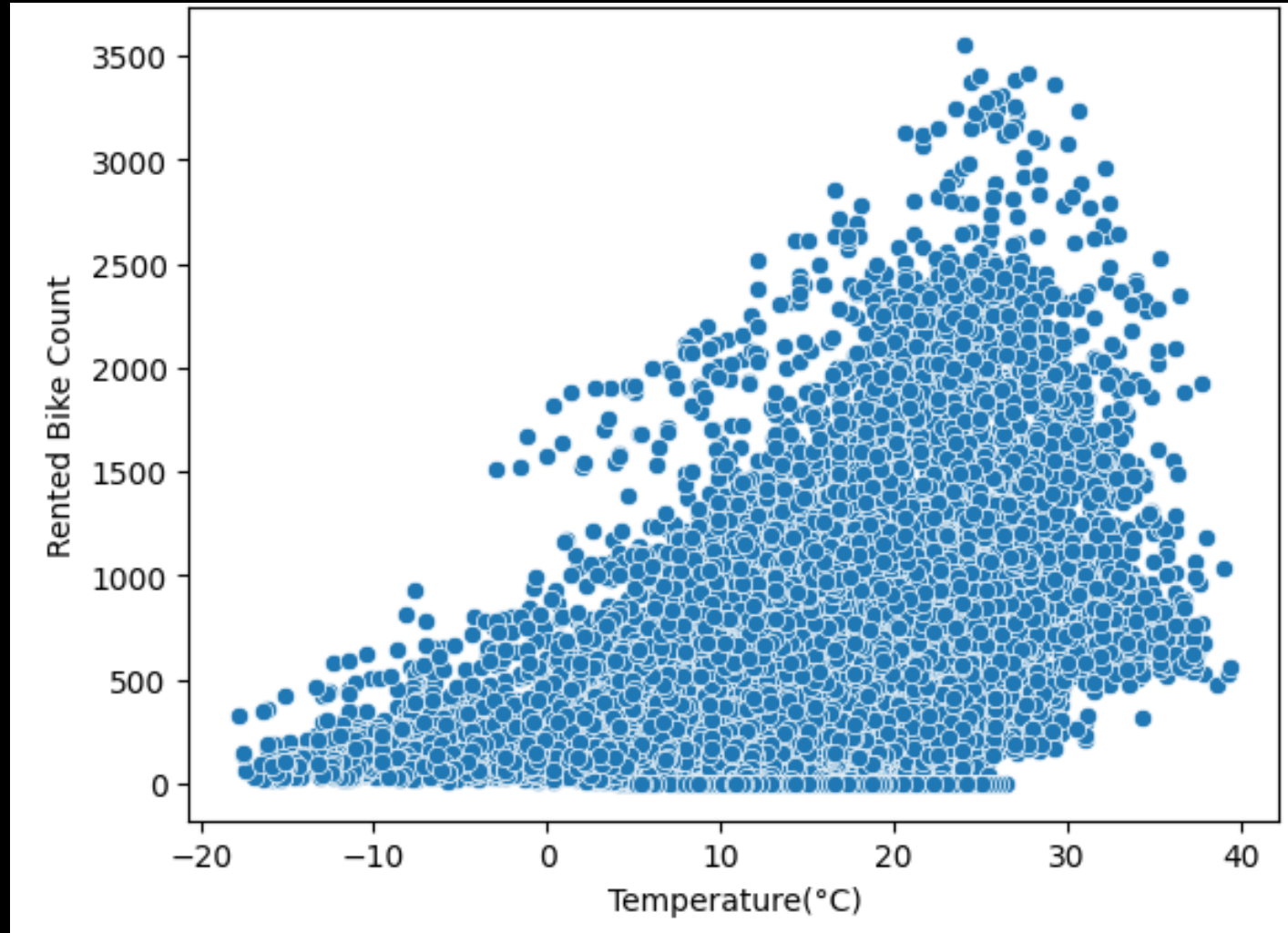
**So we wanted to see how our data was distributed on every parameter**

# Some distribution graphics we found

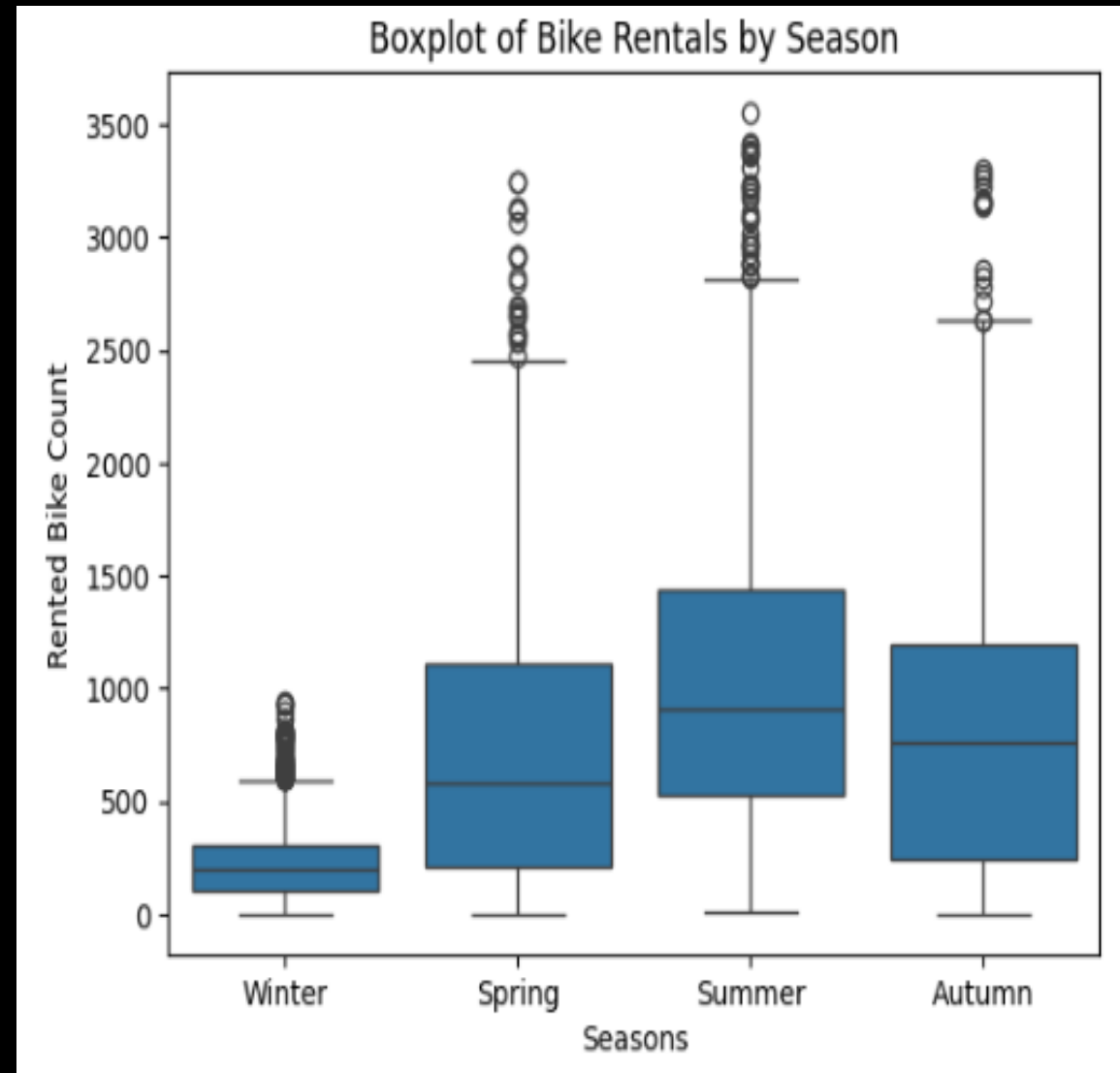


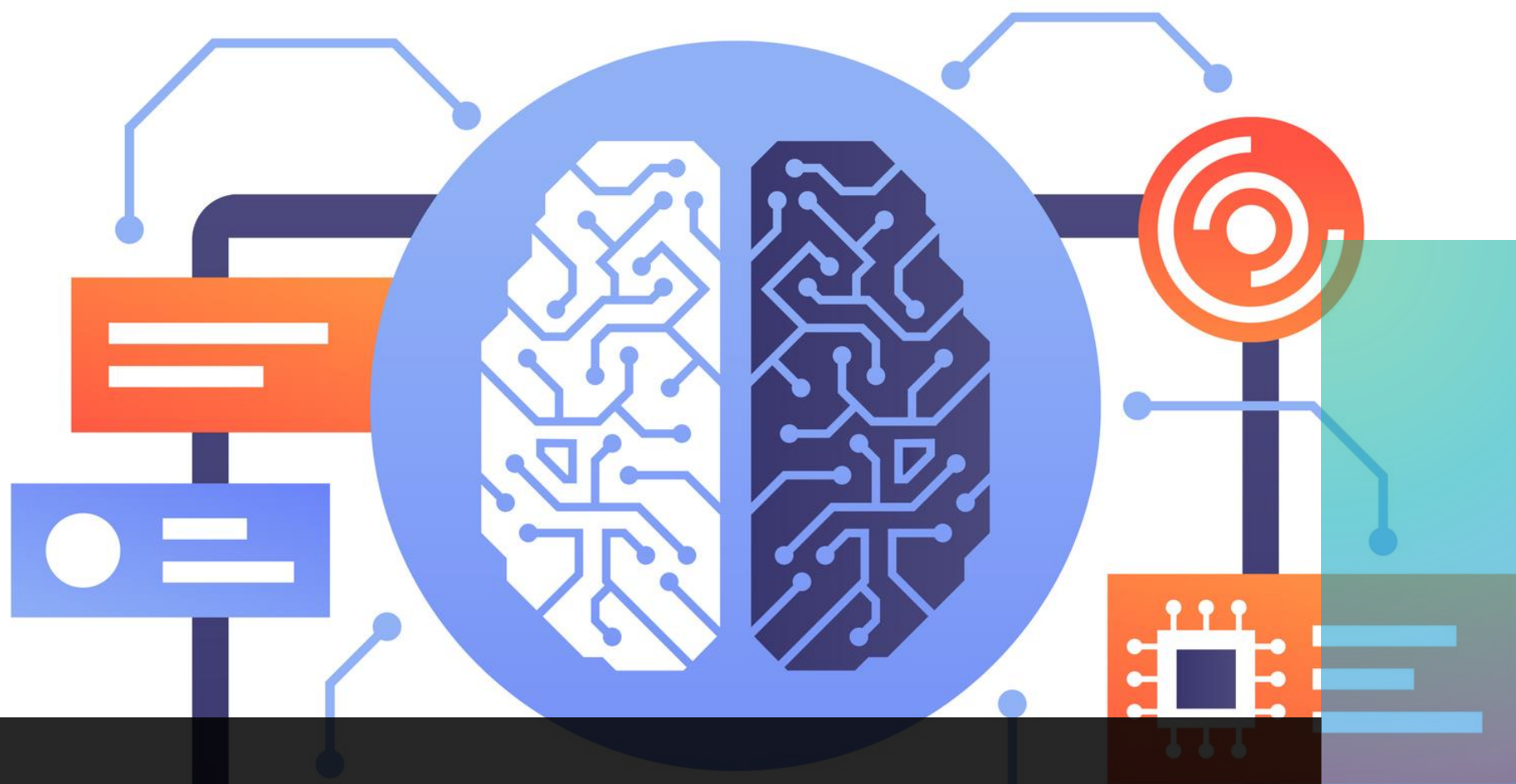


# How impactful the temperature is on bike location



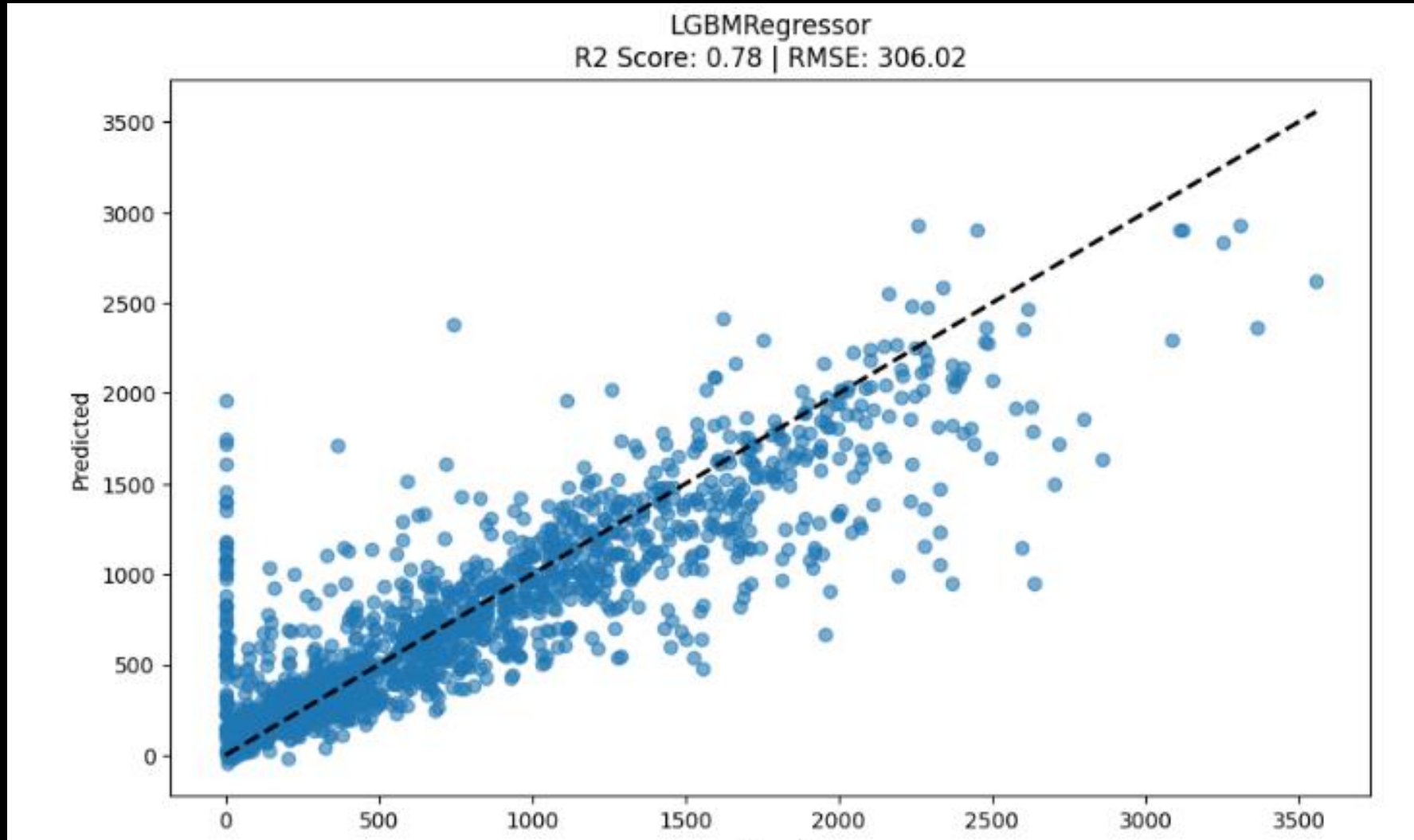
# Impact of seasons on bike rentals





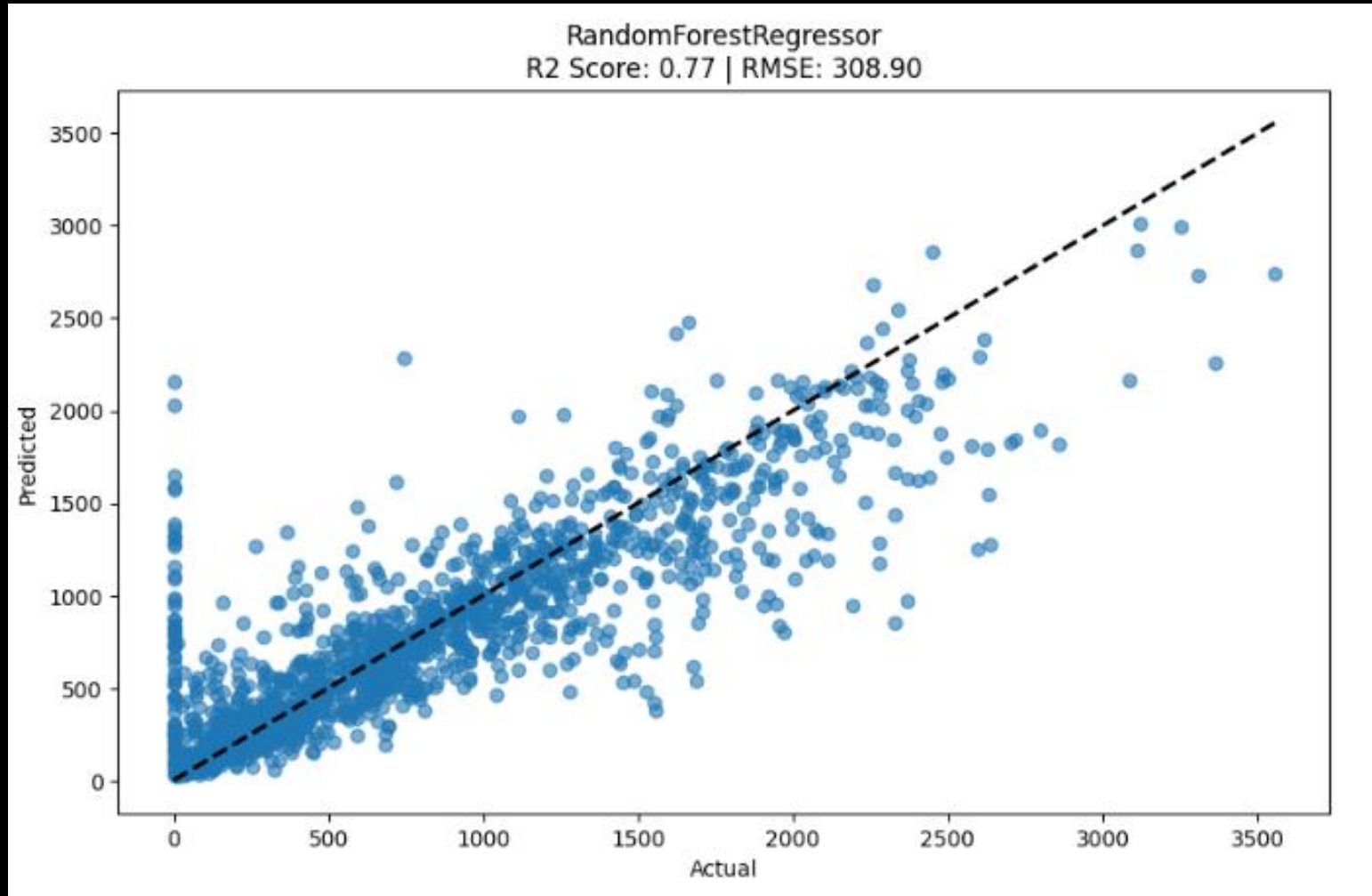
**After analyzing how the data was distributed, we started to use different models**

# The best models



- LGBM Regressor Model, score 0,78

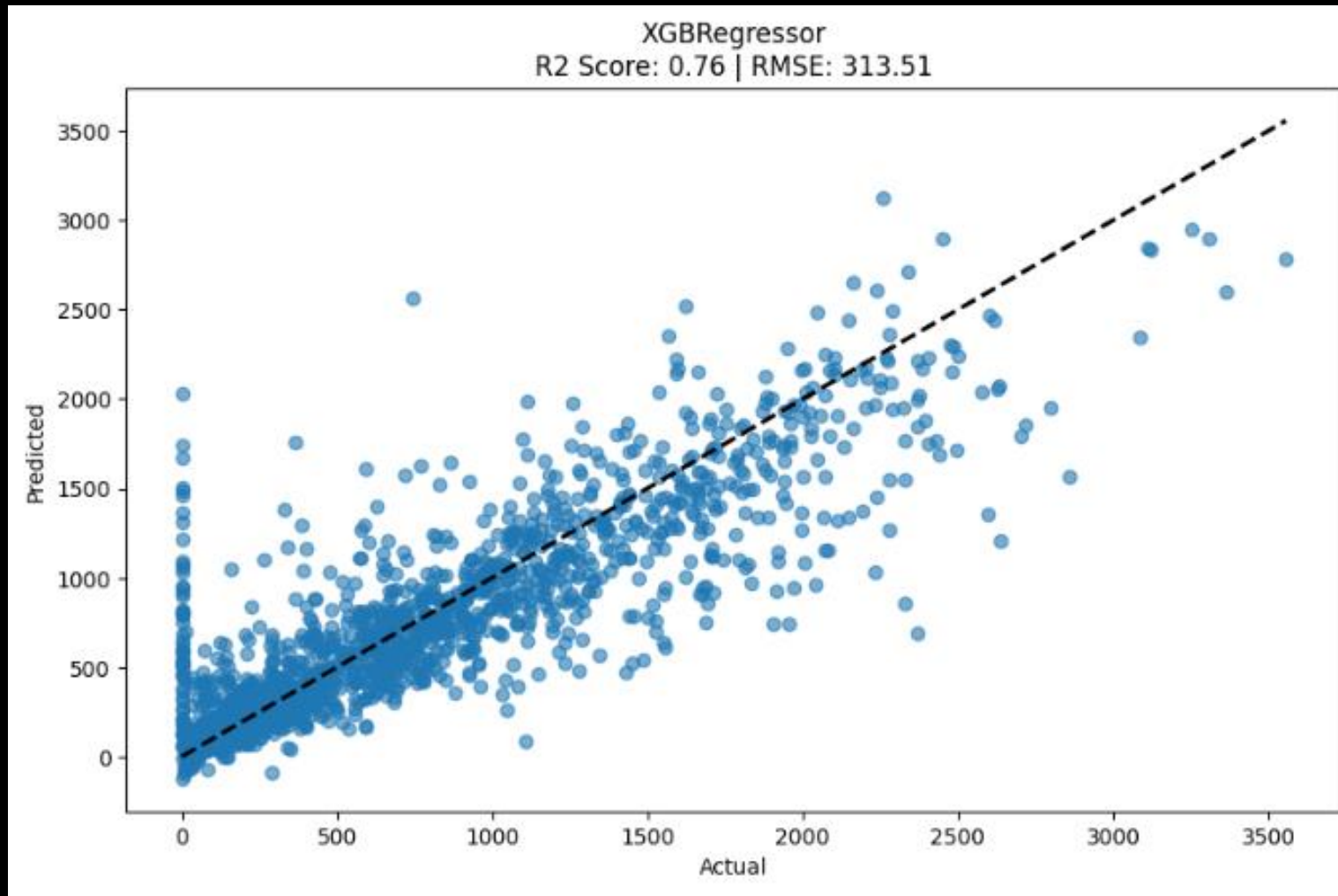
# The best models



RandomForest Regressor Model, score 0,77

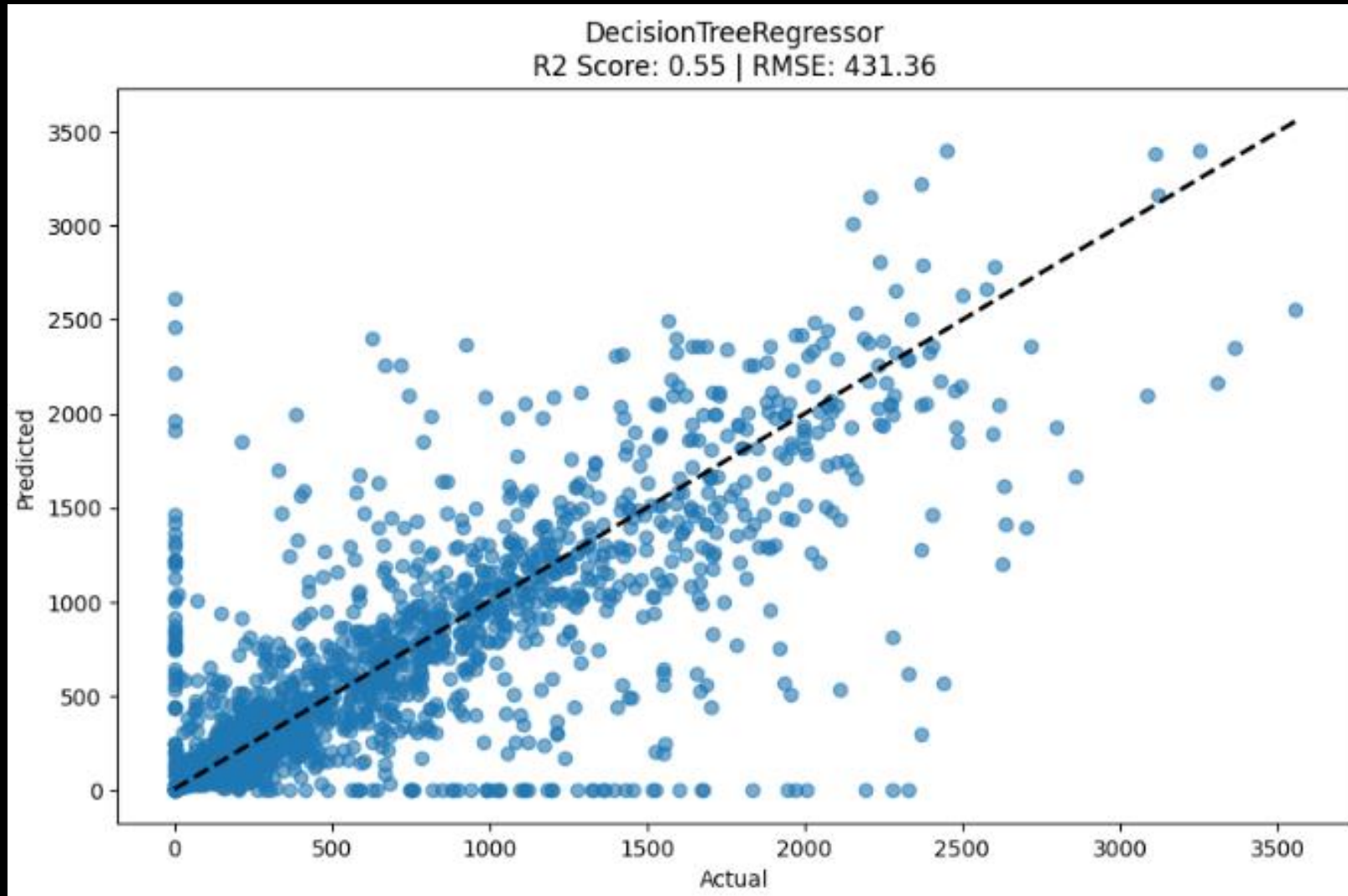


# The best models



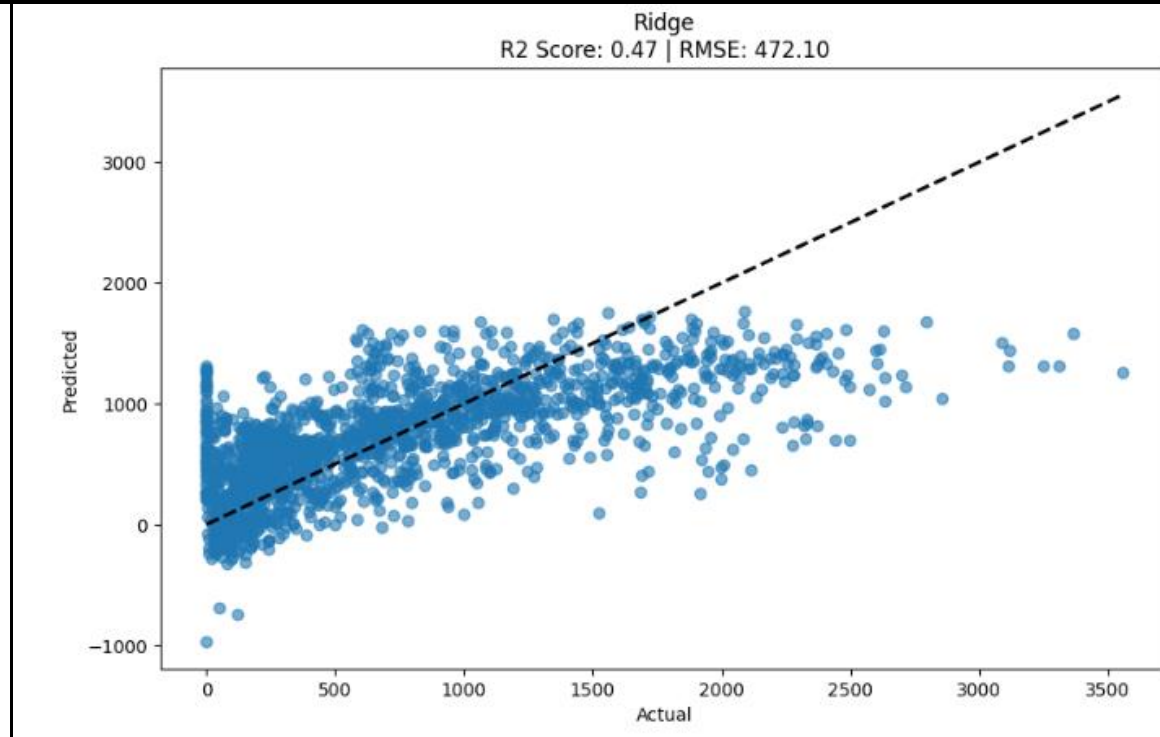
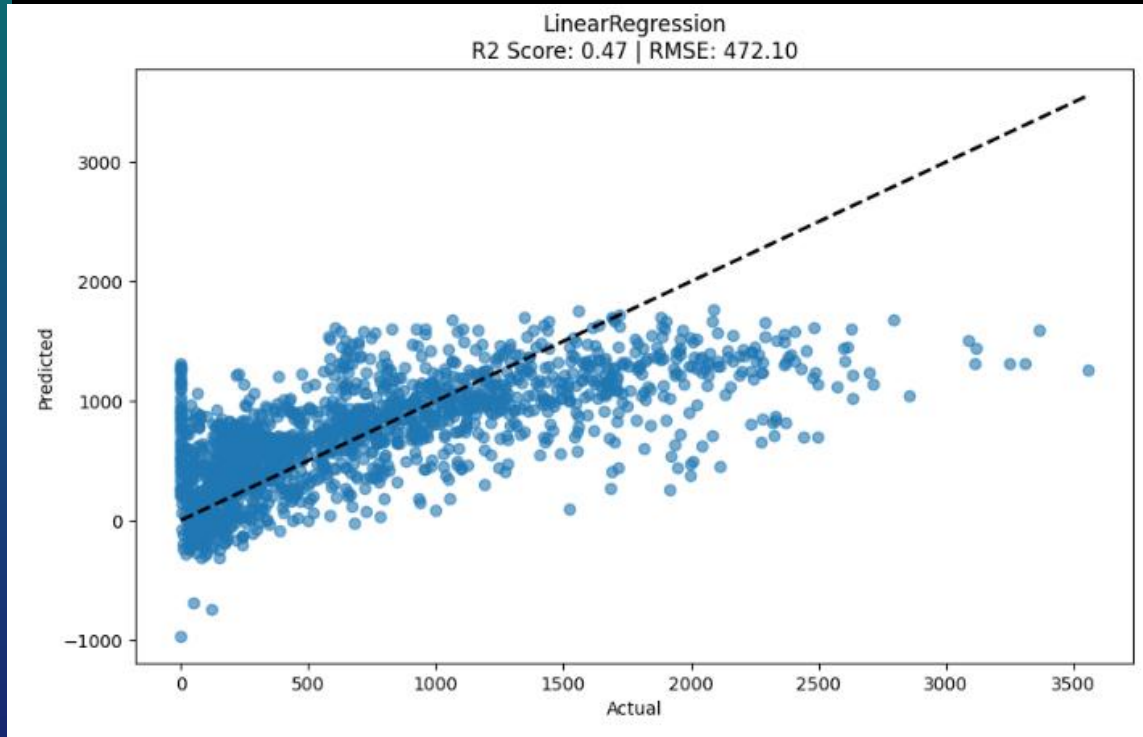
XGB Regressor Model, score 0,76

# The medium models



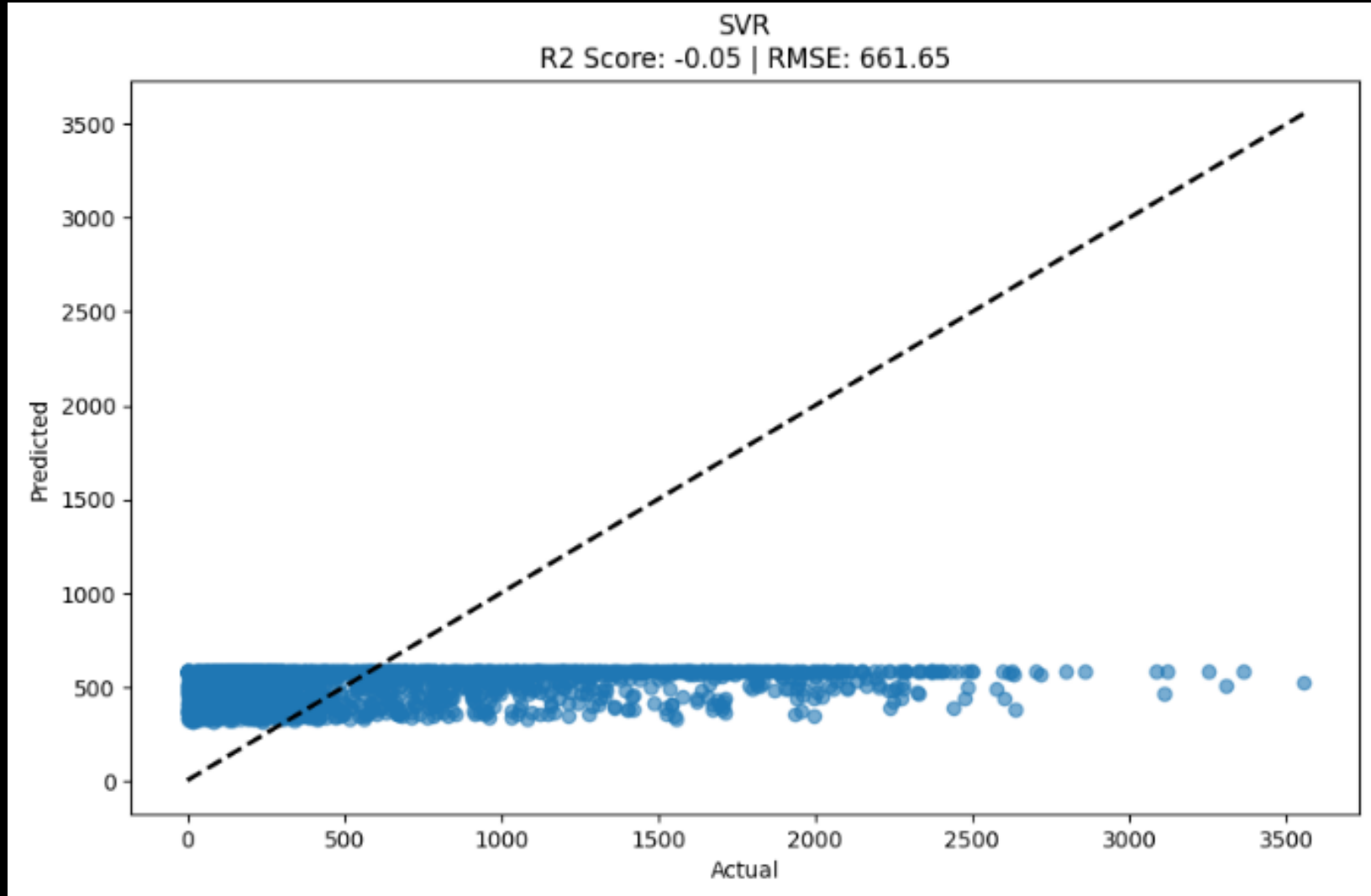
Decision Tree Regressor Model, Score 0,55

# The medium models



Linear Regression and Ridge Models, both score 0,47

# The worst Model



- SVR model, score -0,05, the only negative score for all the models we tested



# In Conclusion

- accurate Model:  $R^2$  0,78
- Predicts well using LGBM Regressor Model
- Helping the company understand and manage their rents and items

