**Capstone Project Submission**

**Instructions:**

i) Please fill in all the required information.

ii) Avoid grammatical errors.

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| **Team Member’s Name, Email, and Contribution:** |
| **Name**: Snehal D. Ramteke  **Email**: [snehalramteke27@gmail.com](mailto:snehalramteke27@gmail.com)  **Contribution**:   * Whole Project |
| **Please paste the GitHub Repo link.** |
| **GitHub Link**:-  <https://github.com/snehalramteke27/Soeul-Bike-Sharing-Demand-Prediction> |
| **Please write a summary of your Capstone project and its components. Describe the problem statement, your approaches, and your conclusions. (200-400 words)** |

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| The contents of the data came from a city called Seoul. A bike-sharing system is a service in which bikes are made available for shared use to individuals on a short-term basis for a price or free. Many bike share systems allow people to borrow a bike from a "dock" which is usually computer-controlled wherein the user enters the payment information, and the system unlocks it. This bike can then be returned to another dock belonging to the same system. The data had variables such as date, hour, temperature, humidity, wind speed, visibility, dew point temperature, solar radiation, rainfall, snowfall, seasons, holiday, functioning days, and rented bike count.  The problem statement was to build a machine learning model that could predict the rented bikes count required for an hour, given other variables. The first step in the exercise involved exploratory data analysis where we tried to dig for insights from the data in hand. It included univariate and multivariate analysis in which we identified certain trends, relationships, and correlations and found out the features that had some impact on our dependent variable. The second step was to clean the data and perform modifications. We checked for missing values and outliers and removed irrelevant features. We also encoded the categorical variables. The third step was to try various machine-learning algorithms on our split and standardized data. We tried different algorithms namely; Linear regression, Random forest, and XGBoost. We did hyperparameter tuning and evaluated the performance of each model using various metrics. The best performance was given by the Gradient boosting and Random forest model where the R2\_score for training and test set was 0.95 and 0.92 respectively.  The most important features that had a major impact on the model predictions were; hour, temperature, wind speed, solar radiation, month, and seasons. Demand for bikes got higher when the temperature and hour values were more. Demand was high for low values of wind speed and solar radiation. Demand was high during spring and summer and very low during winter.  The model performed well in this case but as the data is time-dependent, values of temperature, wind speed, solar radiation, etc. will not always be consistent. Therefore, there will be scenarios where the model might not perform well. As Machine learning is an exponentially evolving field, we will have to be prepared for all contingencies and also keep checking our model from time to time |

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| **Conclusion:**   * ‘Hour’ of the day holds the most important feature. * Bike rental count is mostly correlated with the time of the day as speak at 10 am morning and 8 pm evening. * We observed that the bike rental count is high during working days than on nonworking days. * We see that people generally prefer to bike at moderate to high temperatures, and when a little windy. * It is observed that the highest number of bike rentals counts in the Autumn & Summer seasons & the lowest in the winter season. * We observed 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. |