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# Bike Sharing Program

### Overview

- \* A bicycle-sharing system, is a service in which bicycles 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" and return it at another dock belong to the same system.
- \* One of the such leading Bike Sharing System is Capital Bikeshare System and Client for this project.

# Why the Need

- Increase personal mobility, providing people with better access to destinations throughout the City
- Integrate bike share as an extension of public transit network
- Develop an innovative transportation system that improves livability and economic competitiveness
- Reduce the environmental impact of transportation and help achieve goal of 'Go Green'
- Develop a system that serves users in minority and low-income communities and improves their access to key destinations, such as jobs and recreation
  - safe mode of transportation that promotes active and healthy living
  - Create a system that is financially sustainable, transparently operated, and accountable to the public.

### Project Objective

The client Capital Bikeshare System needed a research analysis on their dataset to optimize their service & operations that

- Predict the Bike Rental volume/count
- Factors or features that influence Bike Rental Count

### About the Data

- \* Capital Bikeshare posts quarterly data reports of bike trip times, start and end locations, and type of user (registered or casual).
- \* Each trip is on one line of data. Dataset contains 17379 rows with 17 features.
- \* These data are readily and publicly available at https://www.kaggle.com/c/bike-sharing-demand/data

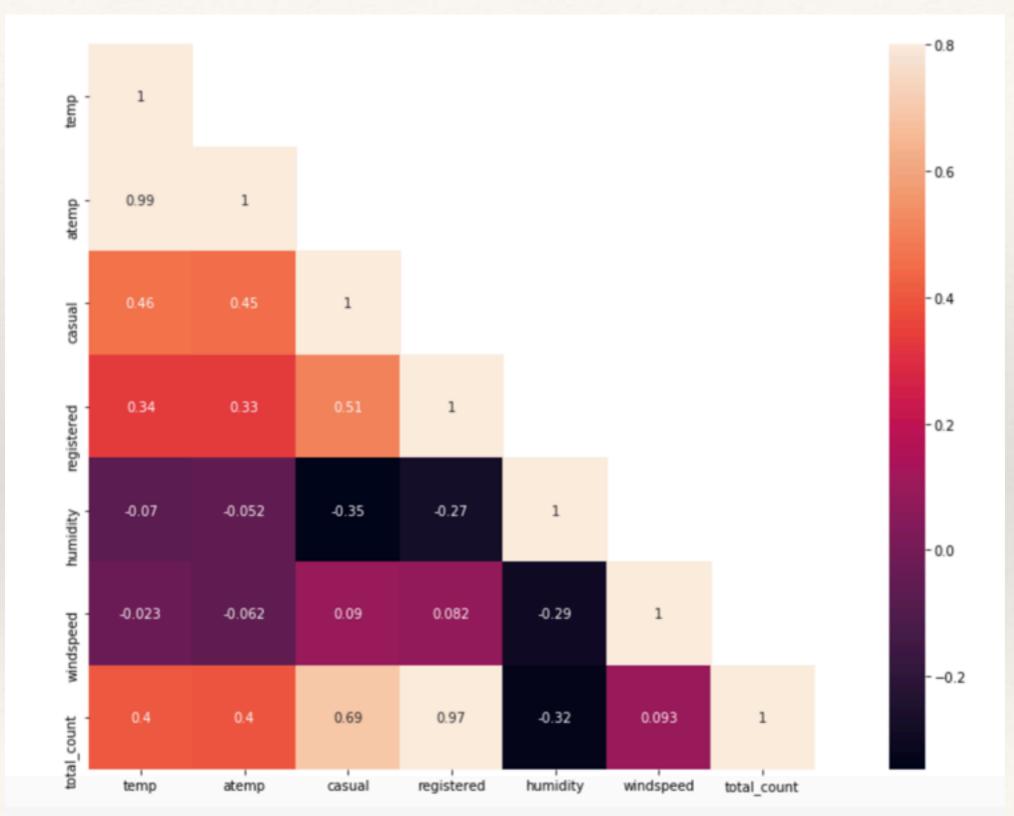
## Data Wrangling Steps

The Capital bikeshare dataset required data wrangling in terms of

- extracting dataset followed by identifying meaningful dataset
- \* renaming a few columns based on preference
- feature transforming date timestamp to day, month,
  year

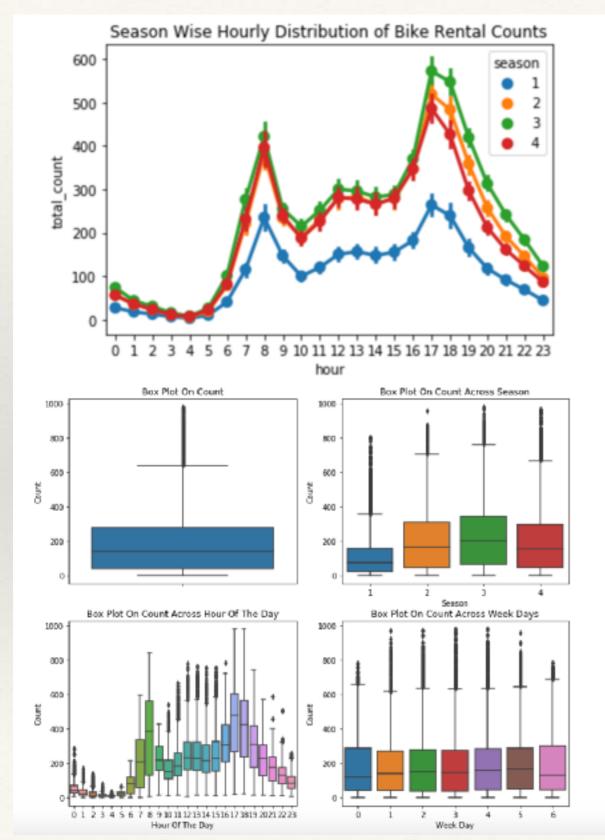
### Exploratory Data Analysis

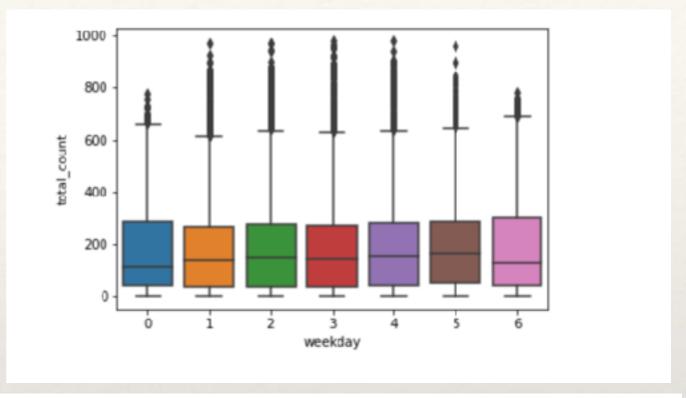
What impacts the Bike Rental Count

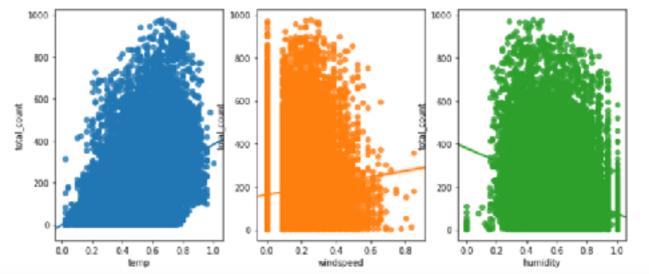


### Exploratory Data Analysis

When impacts the Bike Rental Count

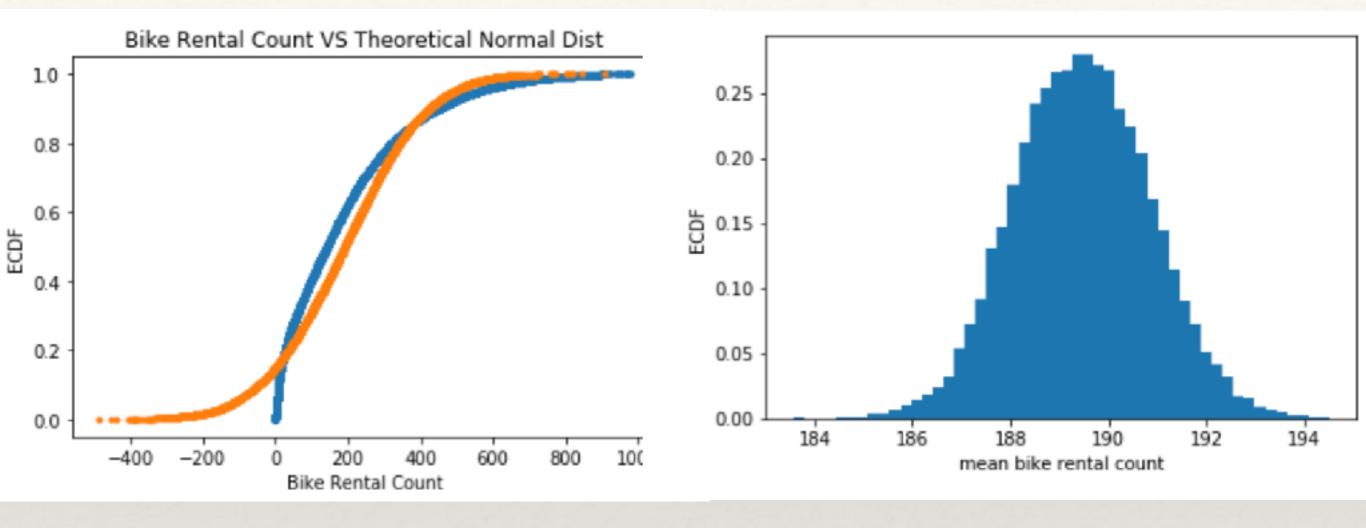






#### EDA/Inferential Statistics

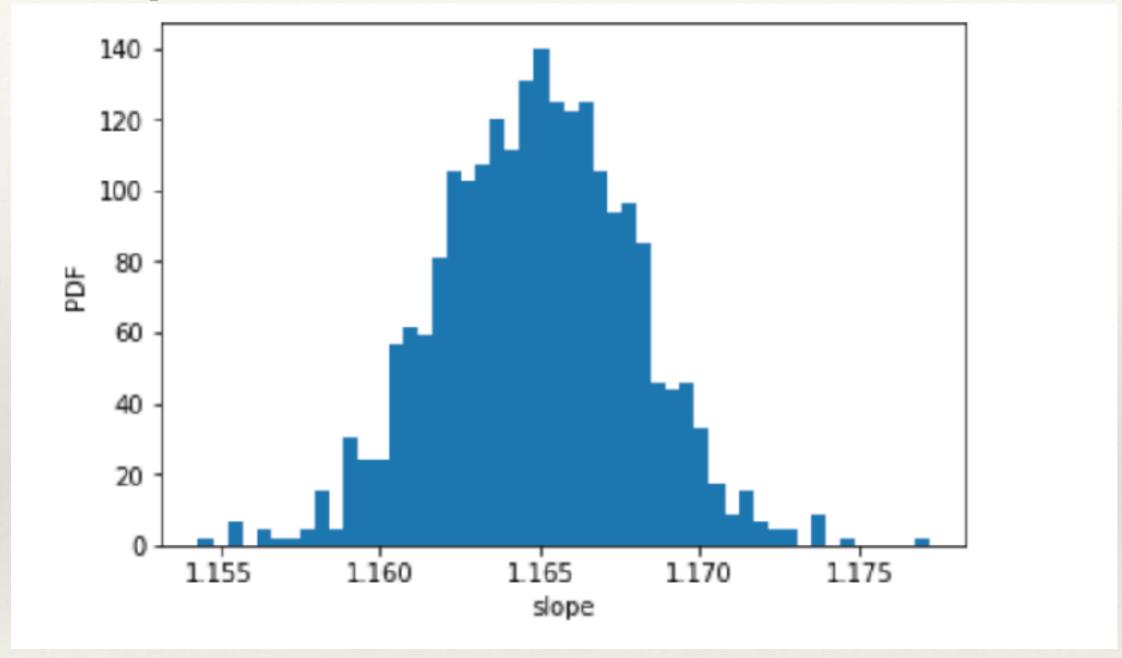
Bike Rental Count Distribution



This is bootstrap estimate of the probability distribution function of the mean Bike Rental Count at the Capital Bikeshare System. It assumes 95% Confidence Interval.

### EDA/Inferential Statistics

Who impacts Bike Rental Count Distribution

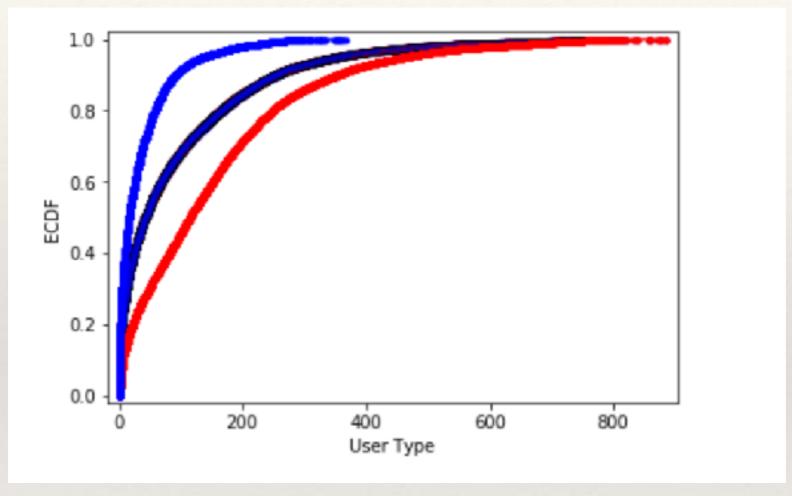


Extending Confidence Interval Concept to Pairs Bootstrap between Bike Rental Count and Registered User Type

### Exploratory Data Analysis

How many impacts the Bike Rental Count

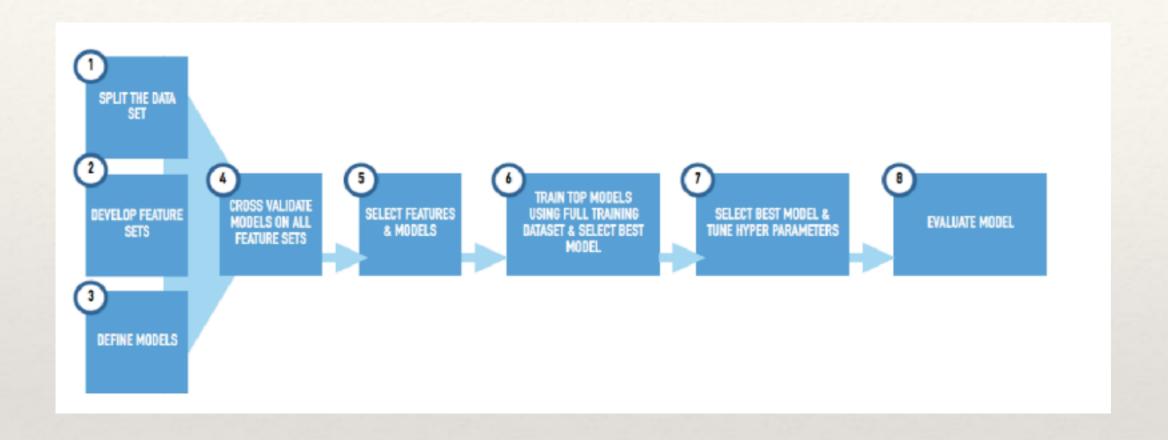
Null Hypothesis- There is no significant difference between registered and casual user type mean on Bike Rental Count.



Registered & Casual User Type Means are not identically distributed and do not influence data in similar way. So Null Hypothesis is rejected.

### Supervised Regression

Steps in Machine Learning Modeling



## Supervised Regression

Machine Learning Algorithms Analyzed

Regressor	Hyperparameters
Linear Regression	R2, CV
Ridge Regression	R2, CV
Lasso Regression	Min Samples Split = 10, Max Leaf Nodes = 500, criterion = MSE, Max
Decision Tree Regression	R2, CV, Pipeline

# Supervised Regression

Model Evaluation

Regressor	R^2/MSE	Grid Search Cross Validation R squared/Elastic Net using Pipeline
Linear Regression	0.387/ -4525.62	0.389, 1.0
Ridge Regression	0.390	0.389, 1.0
Lasso Regression	0.378	0.389, 1.0
Decision Tree Regression	0.886/-97.24	0.858

#### Recommendations & Future Work

With this predictive model, Client may benefit in better prediction of Bike Rental Count:

- \* Demand for bike share program is maximum between 7 AM to 9 AM and 4 PM to 6 PM
- Weather conditions have direct correlation with Bike Rental
  Count
- \* Registered and Casual users are in equation to total bike rental count
- \* Bike Rental Count distribution by Registered and Casual Users is good predictive model

### Future Work

There is lot of potential to enhance the model by:

- \* Collection of more features in the dataset like Gender and Age to help customer know if bike rental preference is by any age or gender group
- \* Model improvement using other Regression models like Random Forest, Support Vector

#### References

- \* https://www.washingtonpost.com/local/trafficandcommuting/capital-bikeshare-gears-up-for-another-expansion/2017/10/02/bcf81b4a-a2fe-11e7-ade1-76d061d56efa\_story.html?utm\_term=.7690885fb8e3
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### Thank You