# Analysis of Hubway Bike Rental Data

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#### Introduction

#### The objective of this analysis:

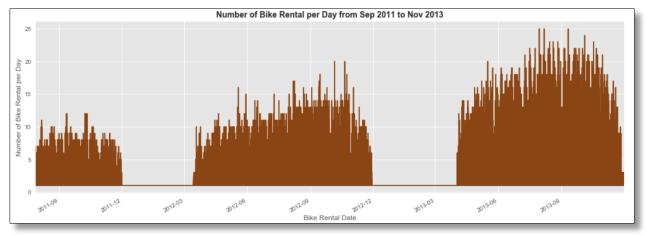
- Analyze bike rental data to know the demand for bike on weekdays
- Identify most prominent time of the day having highest demand for bike
- Create predictive model of Hubway bike demand using machine learning

#### Data sets

- Hubway Trip Data from 2011 to 2013
  - ▶ Date, time, origin and destination stations, plus the bike number
- Weather Data from 2012 to 2013
  - ▶ Daily temperature, humidity, wind speed

# **DATA EXPLORATION**

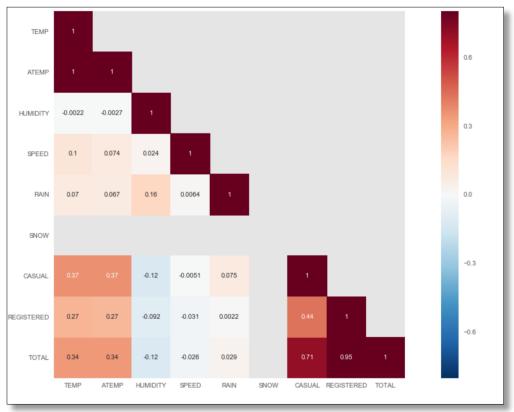
# Daily Bike Rental by Year



- ☐ The bike share program is gaining popularity over time.
- ☐ Weather is an important factor in daily demand of bike rental

# PREDICTIVE MODELS

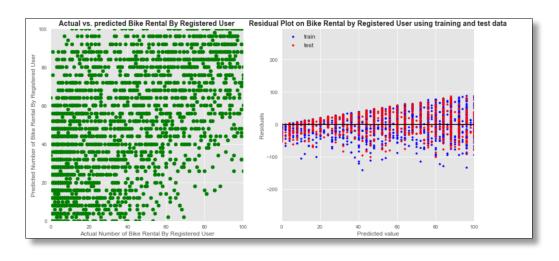
# Linear Regression



Selected 12 features	
TEMP	ATEMP
HUMIDITY	SPEED
RAIN	SNOW
CASUAL	REGISTERED

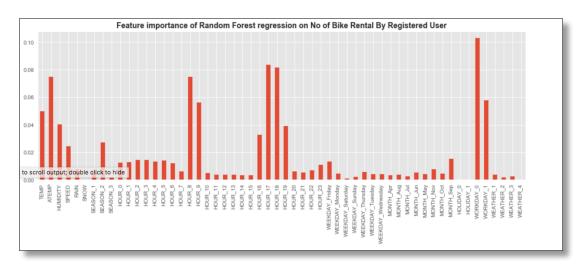
Correlation matrix shows some relationships but none strong enough to skew the model

# Linear Regression - Registered User



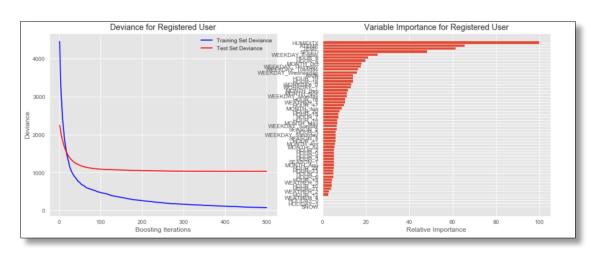
- ☐ Model explains 60% of the variance in the data a moderate value for fit
- ☐ Training data has a MSE = 3720.20 and test data has a MSE = 3776.83
- ☐ Linear regression is not a good choice for this data set

# Random Tree Regression - Registered User



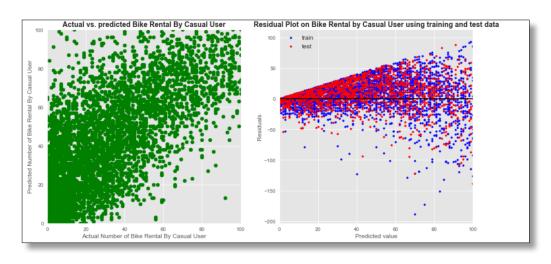
- ☐ Model explains 81% of data's variance but does not generalize well to the test set
- ☐ WORKDAY and HOUR are most important features during training

# Gradient Boosting Regression - Registered User



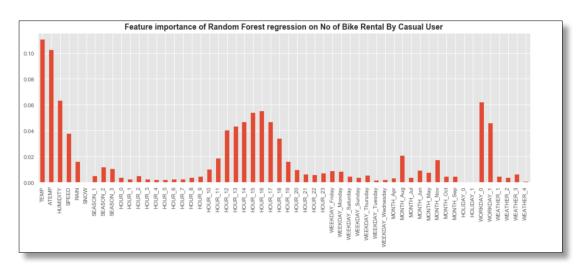
- ☐ Model explains 85% of data's variance but does not generalize well to the test set
- ☐ Temperature and Humidity are most important features during training

# Linear Regression - Casual User



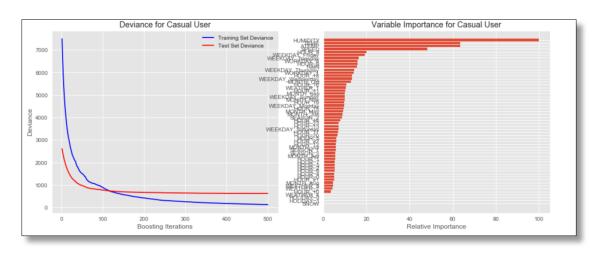
- ☐ Model explains 58% of the variance in the data a moderate value for fit
- ☐ Training data has a MSE = 897.87 and test data has a MSE = 830.70
- ☐ Linear regression is not a good choice for this data set

## Random Tree Regression - Casual User



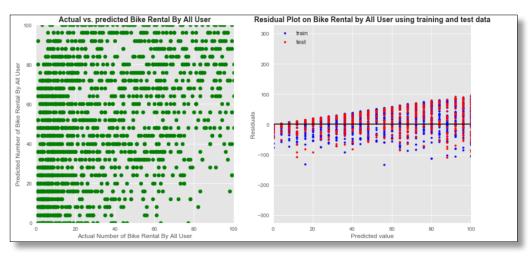
- ☐ Model explains 76% of data's variance but does not generalize well to the test set
- ☐ Temperature and Humidity are most important features during training

### Gradient Boosting Regression - Casual User



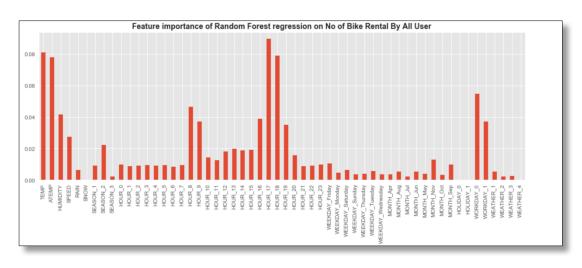
- ☐ Model explains 84% of data's variance but does not generalize well to the test set
- ☐ Temperature and Humidity are most important features during training

## Linear Regression - All User



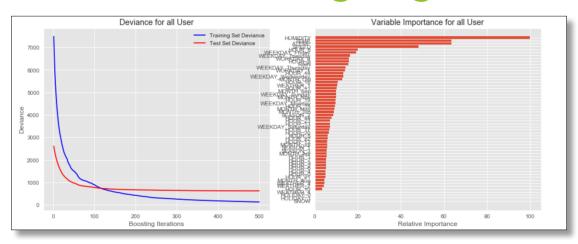
- ☐ Model explains 63% of the variance in the data a moderate value for fit
- ☐ Training data has a MSE = 5756.49 and test data has a MSE = 5714.19
- ☐ Linear regression is not a good choice for this data set

### Random Tree Regression - All User



- ☐ Model explains 81% of data's variance but does not generalize well to the test set
- ☐ Temperature and Humidity are most important features during training

# Gradient Boosting Regression - All User



- ☐ Model explains 86% of data's variance but does not generalize well to the test set
- ☐ Temperature and Humidity are most important features during training

#### Recommendations and future work

#### For Registered User:

Conduct a study to determine the root cause if there is any relationship between a stating station and ending station and then predict the demand for such combination

#### For Casual User:

Investigate if the number of casual user can be converted in to registered user so that the demand can be prediction with more accuracy.

Future opportunities to continue this work include further training of the regression model on similar city with more accurate dataset to determine if the demand for bike rental holds good.