# Final Report

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### Introduction

```
library(usethis)
library(devtools)
load_all("package/bikeSharing/")
## i Loading bikeSharing
set.seed(1)
str(london)
                   2185 obs. of 14 variables:
## 'data.frame':
                : chr "01-01" "01-01" "01-01" "01-01" ...
## $ Hour_chunks : Factor w/ 3 levels "[0,8)","[8,16)",..: 1 1 2 2 3 3 1 1 2 2 ...
                 : num 1 1 1 1 1 1 2 2 2 2 ...
## $ Is weekend : Factor w/ 2 levels "0","1": 1 2 1 2 1 2 1 2 1 2 ...
## $ Is_holiday : Factor w/ 2 levels "0","1": 2 1 2 1 2 1 2 1 2 1 ...
                 : Factor w/ 4 levels "Spring", "Summer", ..: 4 4 4 4 4 4 4 4 4 ...
## $ Season
##
   $ Min_temp
                 : num 3535351919 ...
                 : num 9 10 9 10 9 10 6 11.5 6 11.5 ...
## $ Max_temp
## $ Min_humidity: num 76 81 76 81 76 81 71 82 71 82 ...
## $ Max_humidity: num 87 93 87 93 87 93 94 93 94 ...
## $ Year
                       "Year 1" "Year 2" "Year 1" "Year 2" ...
                 : chr
## $ Wind_speed : num 2.48 3.65 4.83 4.08 6.63 ...
## $ Rain_or_snow: Factor w/ 2 levels "0","1": 1 2 2 2 2 2 1 2 1 2 ...
## $ Bike_count : int 2715 2962 4460 2450 2622 1009 438 475 7756 4263 ...
str(dc)
## 'data.frame':
                   2187 obs. of 14 variables:
                        "01-01" "01-01" "01-01" "01-01" ...
   $ Date
                 : chr
## $ Hour_chunks : Factor w/ 3 levels "[0,8)","[8,16)",..: 1 1 2 2 3 3 1 1 2 2 ...
## $ Day
                 : num 1 1 1 1 1 1 2 2 2 2 ...
## $ Is_weekend : Factor w/ 2 levels "0","1": 2 2 2 2 2 2 1 2 1 2 ...
## $ Is_holiday : Factor w/ 2 levels "0","1": 1 1 1 1 1 1 2 1 2 1 ...
## $ Season
                 : Factor w/ 4 levels "Spring", "Summer", ..: 1 1 1 1 1 1 1 1 1 1 ...
                 : num 1.4 4.22 1.4 4.22 1.4 4.22 2.34 2.34 2.34 2.34 ...
## $ Min_temp
## $ Max_temp
                 : num 13.6 14.6 13.6 14.6 13.6 ...
## $ Min humidity: num 72 48 72 48 72 48 32 39 32 39 ...
## $ Max_humidity: num 94 93 94 93 94 93 45 100 45 100 ...
                        "Year 1" "Year 2" "Year 1" "Year 2" ...
## $ Year
                 : chr
## $ Wind_speed : num 0.208 1.458 3.958 3.75 4.791 ...
## $ Rain_or_snow: Factor w/ 2 levels "0","1": 1 1 1 1 2 2 1 2 1 2 ...
```

```
## $ Bike_count : int 108 290 508 1218 369 786 96 55 1102 452 ...
str(seoul)
## 'data.frame': 1059 obs. of 13 variables:
                : chr "01-01" "01-01" "01-01" "01-02" ...
## $ Date
## $ Hour_chunks : Factor w/ 3 levels "[0,8)","[8,16)",..: 1 2 3 1 2 3 1 2 3 1 ...
## $ Day
                 : num 1 1 1 2 2 2 3 3 3 4 ...
## $ Is_weekend : Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 1 1 1 ...
## $ Is_holiday : Factor w/ 2 levels "0","1": 2 2 2 1 1 1 1 1 1 1 ...
## $ Season
                 : Factor w/ 4 levels "Spring", "Summer", ...: 4 4 4 4 4 4 4 4 4 4 ...
## $ Min temp : num -5 -5 -5 -3.8 -3.8 -7 -7 -7 -8.6 ...
## $ Max_temp
                 : num 3.7 3.7 3.7 1.7 1.7 1.7 -0.4 -0.4 -0.4 -0.8 ...
## $ Min humidity: int 20 20 20 20 20 29 29 29 31 ...
## $ Max_humidity: int 56 56 56 71 71 71 54 54 54 57 ...
## $ Wind_speed : num 0.9 1.85 1.61 0.65 2.26 ...
## $ Rain_or_snow: Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 1 1 1 ...
## $ Bike_count : int 1002 1633 1655 938 2610 2898 1022 2624 2866 1015 ...
london train <- london[london$Year == "Year 1",]</pre>
london_test <- london[london$Year == "Year 2",]</pre>
```

#### Methods

Negative Binomial Generalized Linear Mixed Model

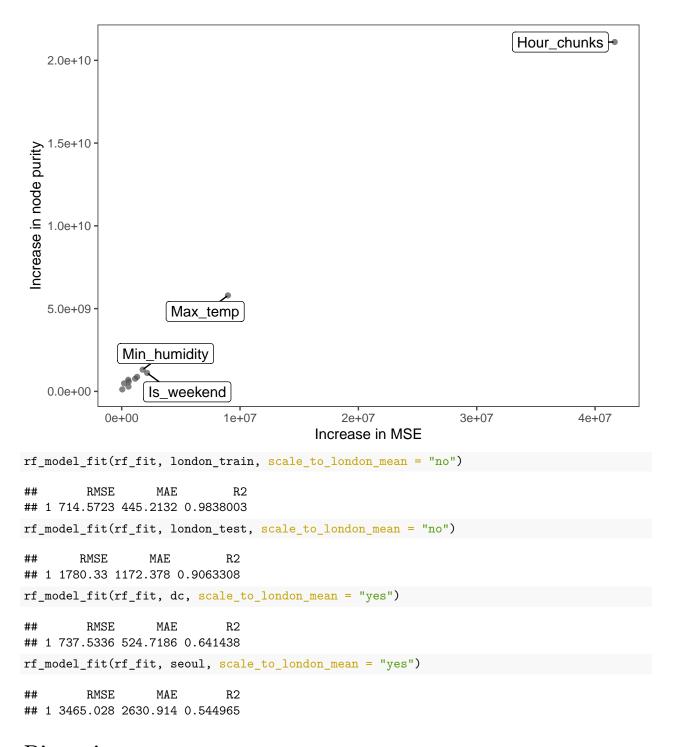
Random Forest

#### Results

Negative Binomial Generalized Linear Mixed Model

```
## List of 7
## $ beta : num [1:14] 8.353 1.534 1.415 -0.337 -0.393 ...
## $ s2gamma : num 0.0296
## $ theta : num 18.4
## $ eps : num 5.15e-05
## $ qfunction: num -9520
## $ day_ranef: num [1:365] 0.0653 -0.398 -0.5165 -0.2612 -0.0374 ...
## $ iter : num 23
glmm_model_fit(glmm_fit, london_train, scale_to_london_mean = "no")
```

```
##
         RMSE
                   MAE
                              R2
## 1 1886.267 1291.106 0.8831618
glmm_model_fit(glmm_fit, london_test, scale_to_london_mean = "no")
         RMSE
                   MAE
##
                              R.2
## 1 2491.293 1647.064 0.8142036
glmm_model_fit(glmm_fit, dc, scale_to_london_mean = "yes")
##
        RMSE
                 MAE
## 1 845.741 605.217 0.521788
glmm_model_fit(glmm_fit, seoul, scale_to_london_mean = "yes")
         RMSE
                   MAE
                              R2
## 1 3519.413 2719.021 0.4999935
Random Forest
rf_fit <- train_random_forest(data = london_train)</pre>
## Loading required package: ggplot2
## Loading required package: lattice
rf_fit
## Random Forest
##
## 1095 samples
##
     11 predictor
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 876, 878, 875, 875, 876
## Resampling results across tuning parameters:
##
##
     mtry RMSE
                     Rsquared
                                MAE
##
      2
           2262.068 0.8841247 1721.956
##
      6
           1797.963 0.8980543 1174.833
           1789.815 0.8964189 1167.026
##
     11
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was mtry = 11.
plot_rf_importance(london_train)
## Warning: ggrepel: 1 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



## Discussion