

# DengAI Predicting Disease Spread

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## Summary



Dengue is a mosquito-borne disease. It occurs mainly in the tropical and subtropical parts of the world. Since, it is transmitted by mosquitoes, the transmission of the disease is related to the climatic conditions and environmental variables such as precipitation and temperature. The disease is prevalent in Southeast Asia and Pacific Islands and epidemics of this disease are expected based on differences in climatic conditions. Nearly half a million cases of the dengue fever every year are reported in the Latin America, as reported by DataDriven.org.

*Data Source* The dataset was collected and publically shared by “*DrivenData.org*”.The link to original dataset can be found [here](#).The environmental data has been collected by the U.S. Federal Government agencies - Centers for Disease Control (CDC) and Prevention to the National Oceanic and Atmospheric Administration (NOAA).

*Goal* The goal of this project is to build supervised learning model to predict the number of dengue fever cases each week in the cities of *San Juan, Puerto Rico and Iquitos, Peru* based on the features provided in the test data set, from 2008 (week 18) till 2013 (week 13) for San Juan , and from 2010 (week 26) till 2013

(week 26) for Iquitos. The champion model will be used for predicting the total cases per week from the features in the test set.

*Algorithms used* We used several algorithms in supervised learning models including Decision (Regression) Tree, Random Forest and Extreme Gradient Boosting, Partial Least Squares, GLMNET for building the prediction model on a training set and compare their performance. Finally the champion model was chosen for predicting on a future dataset.

## Loading Libraries

```
# install.packages("RCurl")
# install.packages("e1071")
# install.packages("caret")
# install.packages("doSNOW")
# install.packages("ipred")
# install.packages("xgboost")
# install.packages("dplyr")
# install.packages("tidyr")
# install.packages("naniar")
# install.packages("corrplot")
# install.packages("gbm")
# install.packages("mda")
# install.packages("psych")
# install.packages("kknn")
# install.packages("pls")
# install.packages("pamr")
# install.packages("mda")
# install.packages("rattle")
# install.packages("vtreat")
library(RCurl)
```

```
## Loading required package: bitops
```

```
library(e1071)
library(caret)
```

```
## Loading required package: ggplot2
```

```
## Warning: package 'ggplot2' was built under R version 3.5.3
```

```
library(doSNOW)
```

```
## Loading required package: foreach
```

```
## Warning: package 'foreach' was built under R version 3.5.3
```

```
## Loading required package: iterators
```

```
## Warning: package 'iterators' was built under R version 3.5.3
```

```
## Loading required package: snow
```

```
library(ipred)
library(xgboost)
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 3.5.3
```

```

##
## Attaching package: 'dplyr'
## The following object is masked from 'package:xgboost':
##
##     slice
## The following objects are masked from 'package:stats':
##
##     filter, lag
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
library(tidyr)

## Warning: package 'tidyr' was built under R version 3.5.3
##
## Attaching package: 'tidyr'
## The following object is masked from 'package:RCurl':
##
##     complete
library(naniar)
library(corrplot)

## corrplot 0.84 loaded
library(psych)

##
## Attaching package: 'psych'
## The following objects are masked from 'package:ggplot2':
##
##     %+%, alpha
library(grid)
library(ggplot2)
library(kknn)

##
## Attaching package: 'kknn'
## The following object is masked from 'package:caret':
##
##     contr.dummy
library(pls)

##
## Attaching package: 'pls'
## The following object is masked from 'package:corrplot':
##
##     corrplot
## The following object is masked from 'package:caret':
##

```

```
##      R2
## The following object is masked from 'package:stats':
##
##      loadings
library(pamr)

## Loading required package: cluster
## Loading required package: survival
##
## Attaching package: 'survival'
## The following object is masked from 'package:caret':
##
##      cluster
## The following object is masked from 'package:rpart':
##
##      solder
library(mda)

## Loading required package: class
## Loaded mda 0.4-10
library(rattle)

## Rattle: A free graphical interface for data science with R.
## Version 5.2.0 Copyright (c) 2006-2018 Togaware Pty Ltd.
## Type 'rattle()' to shake, rattle, and roll your data.
##
## Attaching package: 'rattle'
## The following object is masked from 'package:xgboost':
##
##      xgboost
library(vtreat)
library(glmnet)

## Loading required package: Matrix
##
## Attaching package: 'Matrix'
## The following object is masked from 'package:tidyr':
##
##      expand
## Loaded glmnet 2.0-16
```

## Data Preparation Steps

The dataset contains both train and test data. We will split train data and use one part (i.e., the major part of the split) to train the predictive model and use the other smaller part to test the performance of the predictive model/regressor. The new test dataset will be used for validation.

## Importing Datasets Into the R-Console

##Importing dengue\_features\_train and label dataset using “getURL” method from the RCurl package. This dataset contains information about the various features that can affect the incidence of the cases of dengue (mosquito-borne disease) per week.

Importing the training data features and labels and then merging them by their composite keys (i.e., a combination of ‘city’, ‘year’ and ‘week of year’)

```
trfeat <- getURL("https://s3.amazonaws.com/drivendata/data/44/public/dengue_features_train.csv")
trfeat <- read.csv(text = trfeat)
trfeat <- trfeat[, -c(4)]
trlabel <- getURL("https://s3.amazonaws.com/drivendata/data/44/public/dengue_labels_train.csv")
trlabel <- read.csv(text = trlabel)
trmerge <- merge(trfeat, trlabel, by=c("city", "year", "weekofyear"))
names(trmerge)
```

```
## [1] "city"
## [2] "year"
## [3] "weekofyear"
## [4] "ndvi_ne"
## [5] "ndvi_nw"
## [6] "ndvi_se"
## [7] "ndvi_sw"
## [8] "precipitation_amt_mm"
## [9] "reanalysis_air_temp_k"
## [10] "reanalysis_avg_temp_k"
## [11] "reanalysis_dew_point_temp_k"
## [12] "reanalysis_max_air_temp_k"
## [13] "reanalysis_min_air_temp_k"
## [14] "reanalysis_precip_amt_kg_per_m2"
## [15] "reanalysis_relative_humidity_percent"
## [16] "reanalysis_sat_precip_amt_mm"
## [17] "reanalysis_specific_humidity_g_per_kg"
## [18] "reanalysis_tdtr_k"
## [19] "station_avg_temp_c"
## [20] "station_diur_temp_rng_c"
## [21] "station_max_temp_c"
## [22] "station_min_temp_c"
## [23] "station_precip_mm"
## [24] "total_cases"
```

```
dengue_train <- trmerge[,c(-2)]
names(dengue_train)
```

```
## [1] "city"
## [2] "weekofyear"
## [3] "ndvi_ne"
## [4] "ndvi_nw"
## [5] "ndvi_se"
## [6] "ndvi_sw"
## [7] "precipitation_amt_mm"
## [8] "reanalysis_air_temp_k"
```

```
## [9] "reanalysis_avg_temp_k"
## [10] "reanalysis_dew_point_temp_k"
## [11] "reanalysis_max_air_temp_k"
## [12] "reanalysis_min_air_temp_k"
## [13] "reanalysis_precip_amt_kg_per_m2"
## [14] "reanalysis_relative_humidity_percent"
## [15] "reanalysis_sat_precip_amt_mm"
## [16] "reanalysis_specific_humidity_g_per_kg"
## [17] "reanalysis_tdtr_k"
## [18] "station_avg_temp_c"
## [19] "station_diur_temp_rng_c"
## [20] "station_max_temp_c"
## [21] "station_min_temp_c"
## [22] "station_precip_mm"
## [23] "total_cases"
```

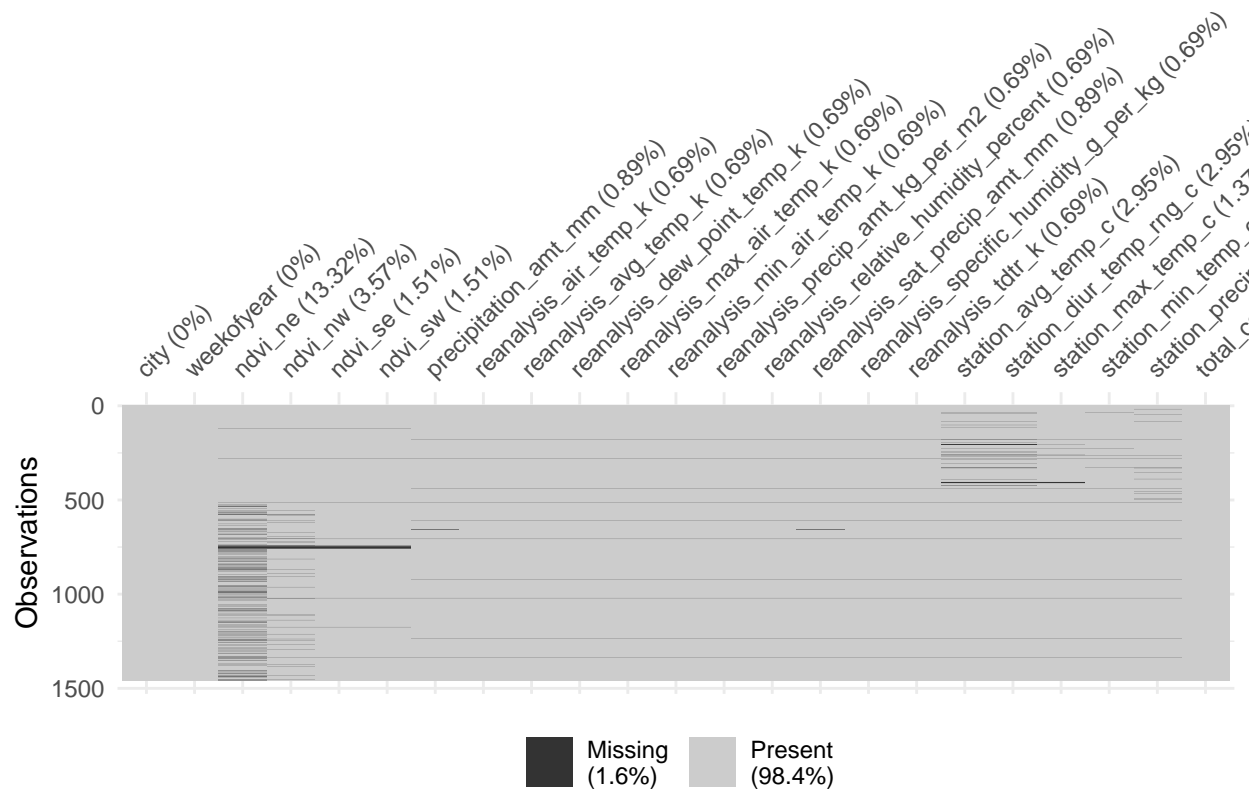
```
dim(dengue_train)
```

```
## [1] 1456    23
```

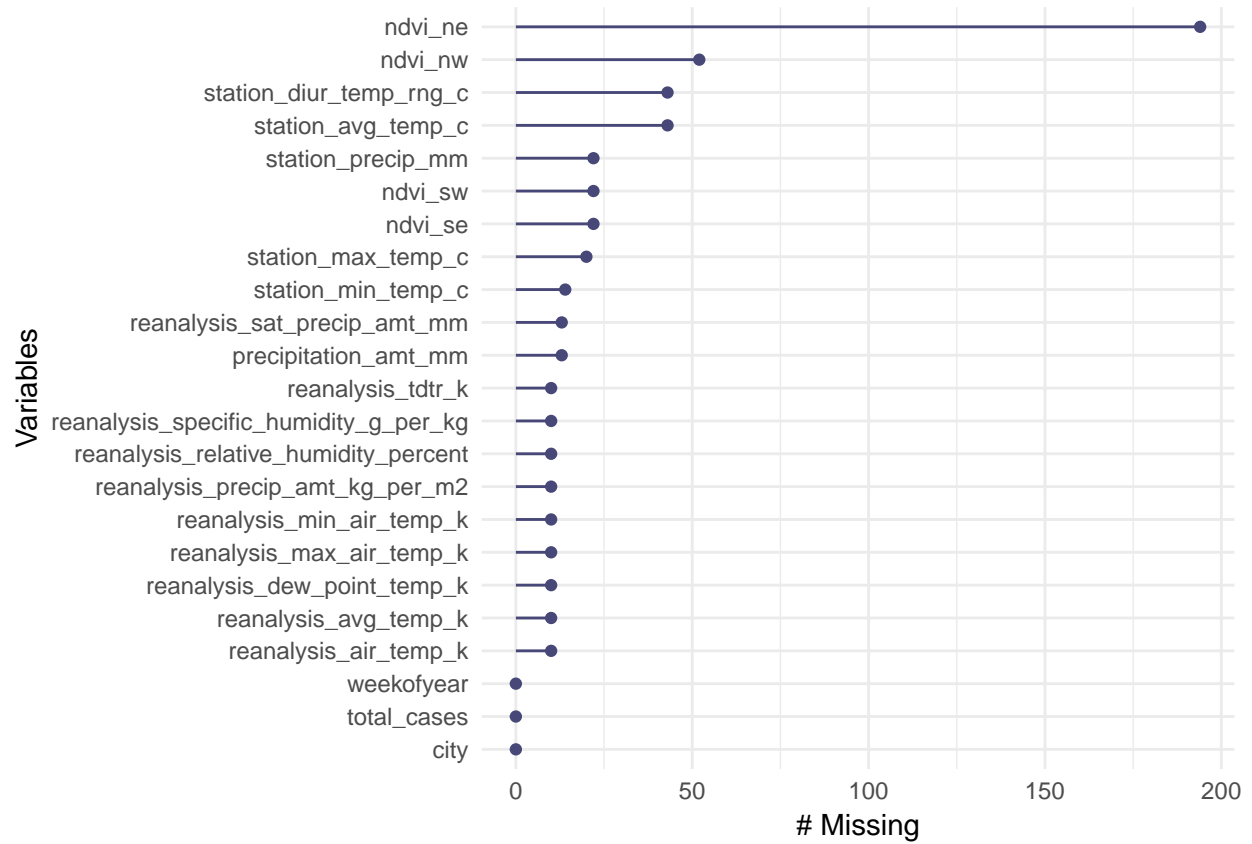
## Checking and visualizing missing values in the merged training data

```
# Visualizing missing values for the training data
```

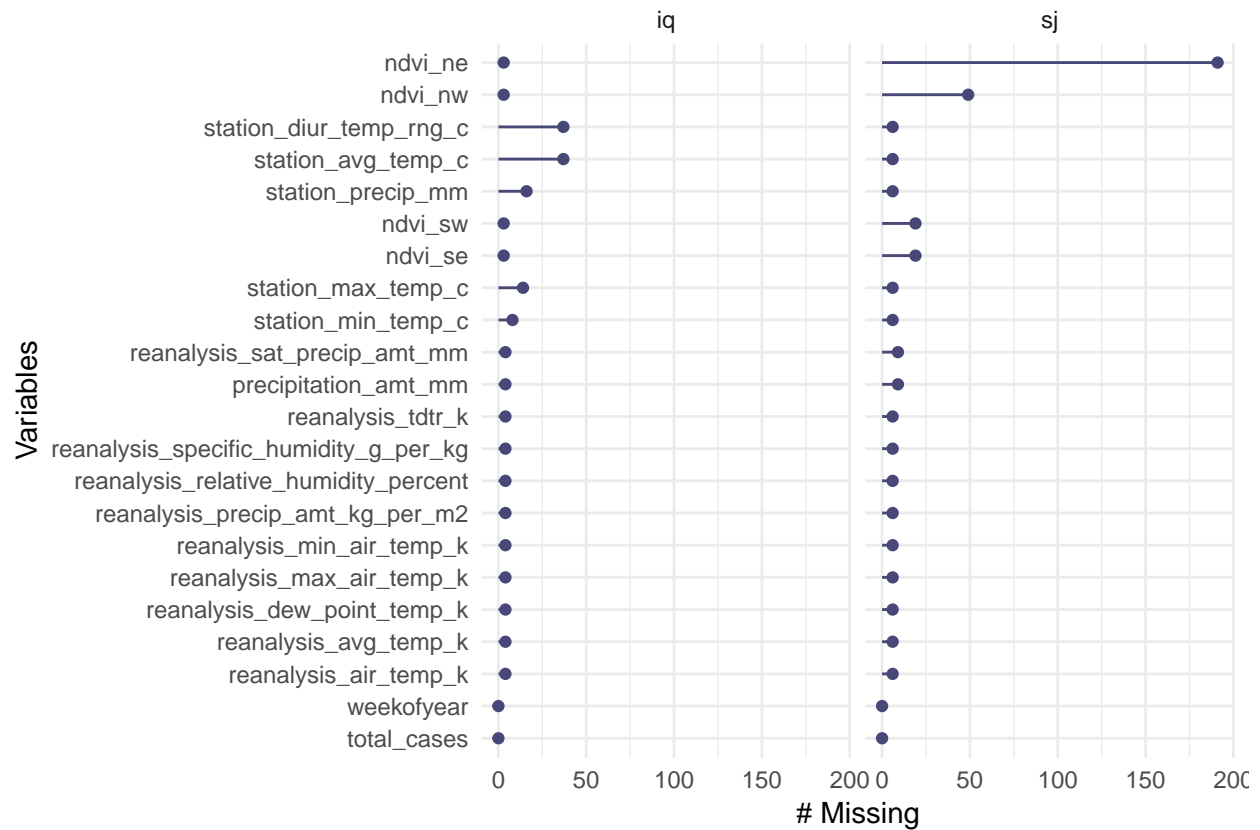
```
vis_miss(dengue_train)
```



```
gg_miss_var(dengue_train) + theme_minimal()
```



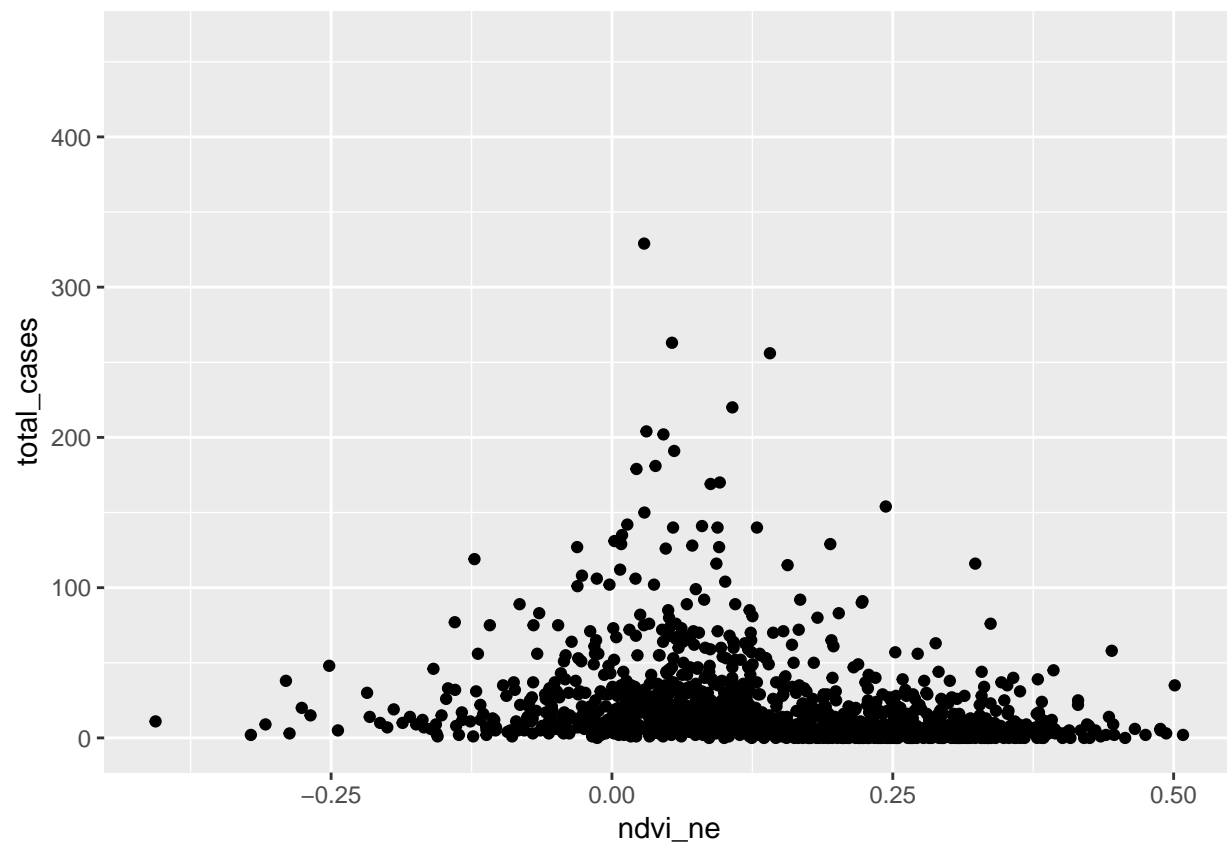
```
gg_miss_var(dengue_train, facet = city) + theme_minimal()
```



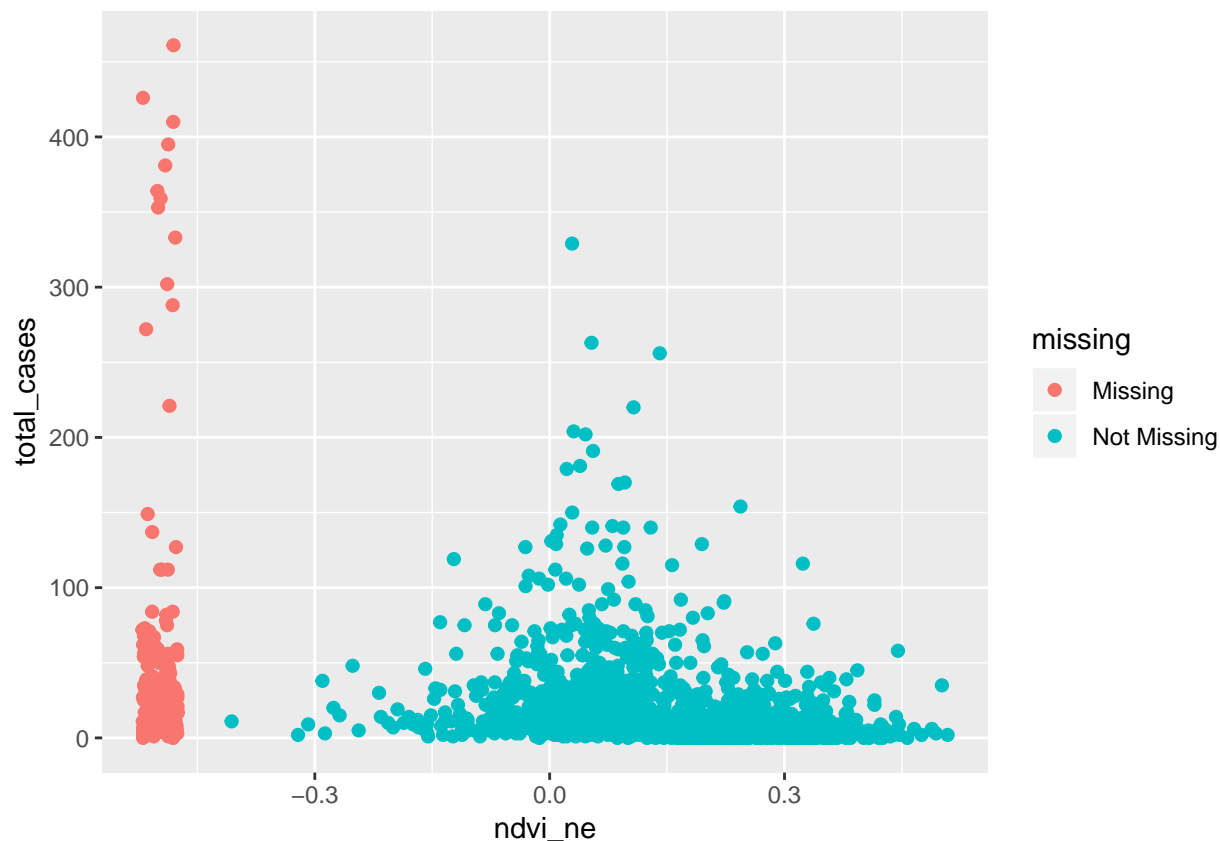
```
ggplot(dengue_train, aes(x=ndvi_ne, y = total_cases)) + geom_point()
```

```
## Warning: Removed 194 rows containing missing values (geom_point).
```





```
ggplot(dengue_train, aes(x=ndvi_ne, y = total_cases)) + geom_miss_point()
```



Conclusion: Most of the missing values can be classified as 'Missing Not At Random'.

## Imputation of missing values

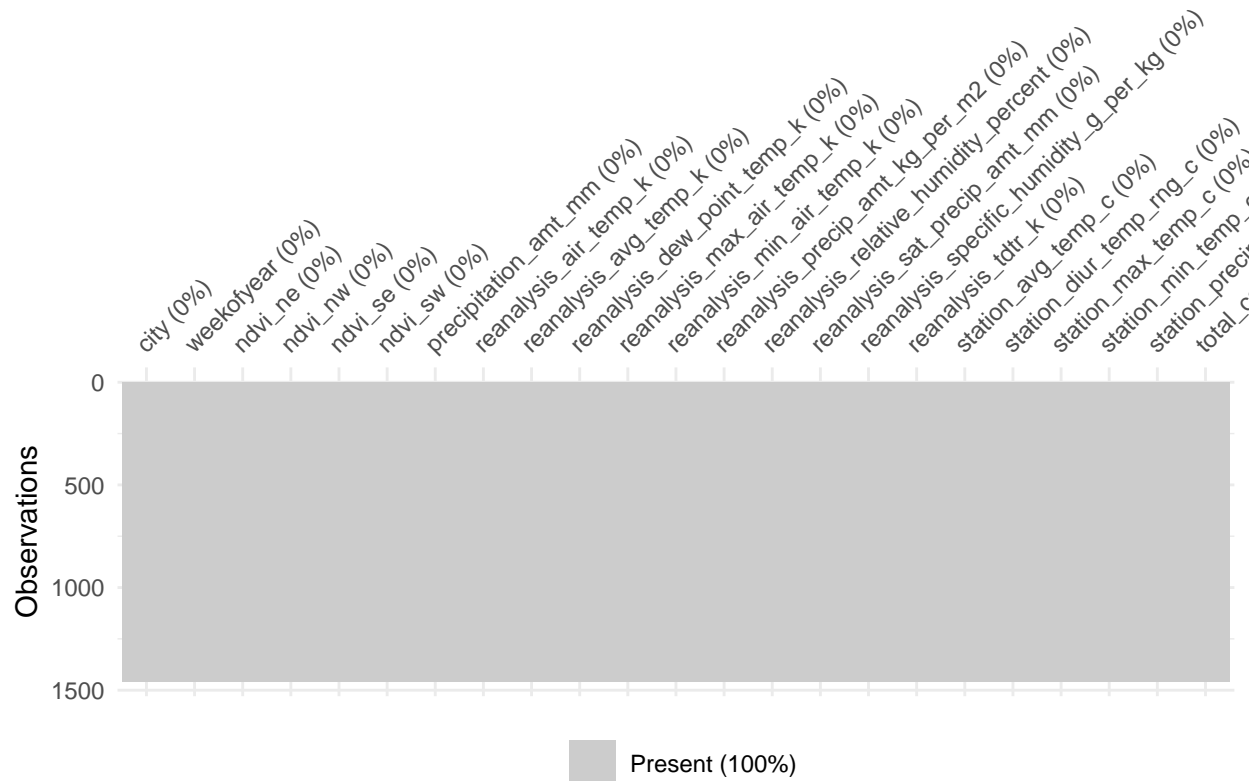
```
pre.process <- preProcess(dengue_train, method = "bagImpute")
imputed.data <- predict(pre.process, dengue_train)
dengue_train$ndvi_ne <- imputed.data[,3]
dengue_train$ndvi_nw <- imputed.data[,4]
dengue_train$ndvi_se <- imputed.data[,5]
dengue_train$ndvi_sw <- imputed.data[,6]
dengue_train$precipitation_amt_mm <- imputed.data[,7]
dengue_train$reanalysis_air_temp_k <- imputed.data[, 8]
dengue_train$reanalysis_avg_temp_k <- imputed.data[,9]
dengue_train$reanalysis_dew_point_temp_k <- imputed.data[,10]
dengue_train$reanalysis_max_air_temp_k <- imputed.data[,11]
dengue_train$reanalysis_min_air_temp_k <- imputed.data[,12]
dengue_train$reanalysis_precip_amt_kg_per_m2 <- imputed.data[,13]
dengue_train$reanalysis_relative_humidity_percent <- imputed.data[,14]
dengue_train$reanalysis_sat_precip_amt_mm <- imputed.data[,15]
dengue_train$reanalysis_specific_humidity_g_per_kg <- imputed.data[,16]
dengue_train$reanalysis_tdtr_k <- imputed.data[,17]
dengue_train$station_avg_temp_c <- imputed.data[,18]
dengue_train$station_diur_temp_rng_c <- imputed.data[,19]
dengue_train$station_max_temp_c <- imputed.data[,20]
dengue_train$station_min_temp_c <- imputed.data[,21]
```

```
dengue_train$station_precip_mm <- imputed.data[,22]
```

```
anyNA(dengue_train)
```

```
## [1] FALSE
```

```
vis_miss(dengue_train)
```



Conclusion: All of the missing values were imputed.

## Randomize the training data

```
random_index <- sample(1:nrow(dengue_train), nrow(dengue_train))
random_train <- dengue_train[random_index, ]
names(random_train)
```

```
## [1] "city"
## [2] "weekofyear"
## [3] "ndvi_ne"
## [4] "ndvi_nw"
## [5] "ndvi_se"
## [6] "ndvi_sw"
## [7] "precipitation_amt_mm"
## [8] "reanalysis_air_temp_k"
## [9] "reanalysis_avg_temp_k"
```

```
## [10] "reanalysis_dew_point_temp_k"
## [11] "reanalysis_max_air_temp_k"
## [12] "reanalysis_min_air_temp_k"
## [13] "reanalysis_precip_amt_kg_per_m2"
## [14] "reanalysis_relative_humidity_percent"
## [15] "reanalysis_sat_precip_amt_mm"
## [16] "reanalysis_specific_humidity_g_per_kg"
## [17] "reanalysis_tdtr_k"
## [18] "station_avg_temp_c"
## [19] "station_diur_temp_rng_c"
## [20] "station_max_temp_c"
## [21] "station_min_temp_c"
## [22] "station_precip_mm"
## [23] "total_cases"
```

```
dim(random_train)
```

```
## [1] 1456 23
```

```
anyNA(random_train)
```

```
## [1] FALSE
```

## Defining the tuning grid

```
grid <- expand.grid(eta = c(0.05, 0.5),
                    nrounds = c(70, 90),
                    max_depth = 1:6,
                    min_child_weight = c(1.0, 4),
                    colsample_bytree = c(0.5, 1),
                    gamma = c(0, 0.1),
                    subsample = c(0.8, 1))
```

## Defining trainControl for the ML Algorithms

```
train.control <- trainControl(method = "repeatedcv",
                              number = 10,
                              repeats = 5,
                              search = "grid")
```

#Applying ML Algorithms For Training the Prediction Model

## K-Nearest Neighbor (knn) Algorithm to Train The Prediction Model

```
set.seed(45220)
model_kknn <- caret::train(total_cases ~ .,
                           data = random_train[,c(2,3:6,12,16,20,23)],
                           type="prob",
                           method = "knn",
                           tuneLength = 10,
                           preProcess = NULL,
```

```

                                trControl = train.control)
model_kknn

## k-Nearest Neighbors
##
## 1456 samples
##    8 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1310, 1309, 1310, 1311, 1311, 1311, ...
## Resampling results across tuning parameters:
##
##   kmax  RMSE      Rsquared  MAE
##    5    32.08063  0.4449575  17.70653
##    7    31.72765  0.4490605  17.61308
##    9    31.84266  0.4471819  17.61459
##   11    31.95553  0.4452667  17.63847
##   13    31.97997  0.4447213  17.64092
##   15    32.01345  0.4444422  17.64742
##   17    32.02947  0.4439074  17.64962
##   19    32.04736  0.4435655  17.65310
##   21    32.07460  0.4430591  17.66140
##   23    32.07171  0.4433770  17.65868
##
## Tuning parameter 'distance' was held constant at a value of 2
##
## Tuning parameter 'kernel' was held constant at a value of optimal
## RMSE was used to select the optimal model using the smallest value.
## The final values used for the model were kmax = 7, distance = 2 and
## kernel = optimal.

```

**GLMNET Algorithm to Train The Prediction Model:** generalized linear model via penalized maximum likelihood; the regularization path is computed for elasticnet penalty at a grid of values for the regularization parameter lambda

```

set.seed(45220)
model_glmnet <- caret::train(total_cases ~ .,
                             data = random_train[,c(2,3:6,12,16,20,23)],
                             method = "glmnet",
                             preProcess = NULL,
                             trControl = train.control)
model_glmnet

## glmnet
##
## 1456 samples
##    8 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1310, 1309, 1310, 1311, 1311, 1311, ...

```

```
## Resampling results across tuning parameters:
##
##   alpha  lambda      RMSE      Rsquared    MAE
##   0.10   0.02843102  39.26132  0.1500108  20.93842
##   0.10   0.28431018  39.25446  0.1501951  20.90350
##   0.10   2.84310180  39.20494  0.1517795  20.59967
##   0.55   0.02843102  39.26191  0.1500124  20.94252
##   0.55   0.28431018  39.23127  0.1510429  20.81516
##   0.55   2.84310180  39.22430  0.1519902  20.40782
##   1.00   0.02843102  39.25970  0.1500870  20.93313
##   1.00   0.28431018  39.21615  0.1515980  20.74542
##   1.00   2.84310180  39.31499  0.1506882  20.32348
##
## RMSE was used to select the optimal model using the smallest value.
## The final values used for the model were alpha = 0.1 and lambda = 2.843102.
```

## Random Forest Algorithm to Train The Prediction Model

```
x <- random_train[,2:22]

metric <- "MAE"
mtry <- sqrt(ncol(x))
model_rf <- caret::train(total_cases ~ .,
                          data = random_train[,c(2,3:6,12,16,20,23)],
                          method = "rf",
                          preProcess = NULL,
                          metric = metric,
                          tuneGrid = expand.grid(.mtry = mtry),
                          trControl = train.control)

model_rf

## Random Forest
##
## 1456 samples
##    8 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1310, 1310, 1309, 1311, 1311, 1310, ...
## Resampling results:
##
##   RMSE      Rsquared    MAE
##  30.52333  0.4839504  16.83161
##
## Tuning parameter 'mtry' was held constant at a value of 4.582576
```

## Regression Tree Algorithm to Train The Prediction Model

```
set.seed(123)
model_rpart <- caret::train(total_cases ~ ., data = random_train,
                             method = "rpart",
```

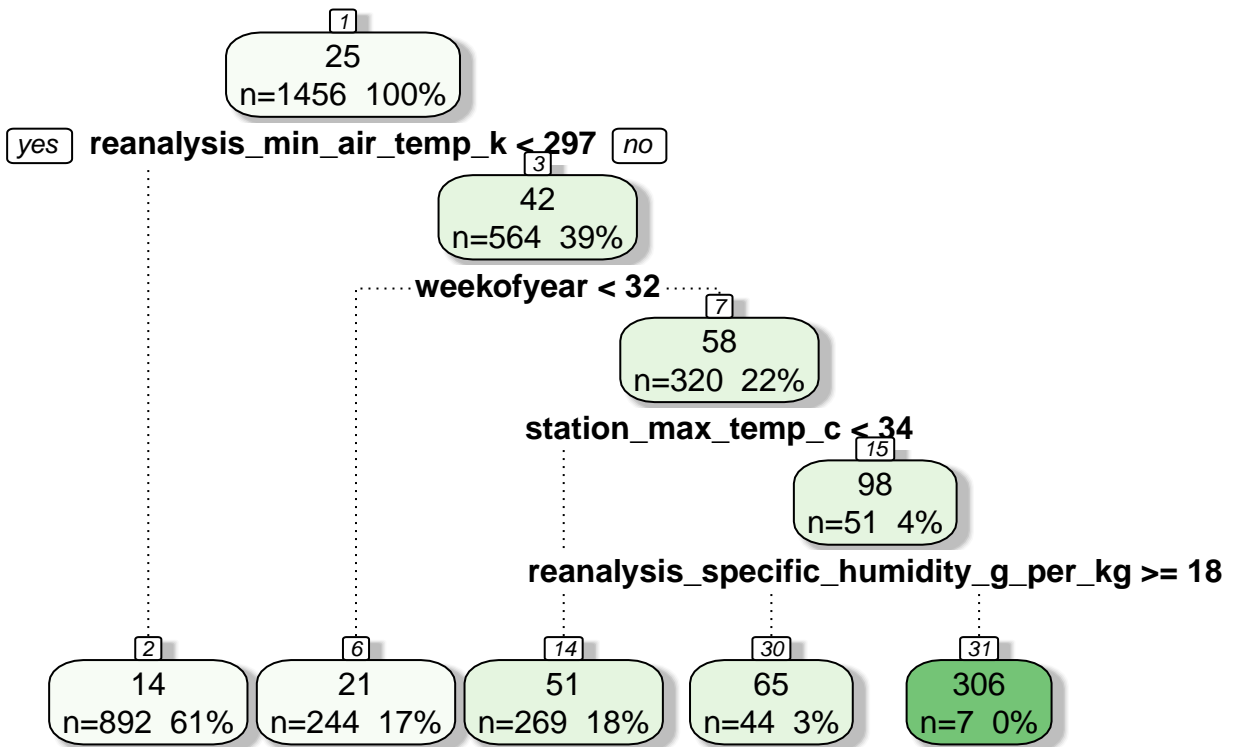
```

preProcess = NULL,
trControl = train.control)

## Warning in nominalTrainWorkflow(x = x, y = y, wts = weights, info =
## trainInfo, : There were missing values in resampled performance measures.
model_rpart

## CART
##
## 1456 samples
## 22 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1311, 1310, 1312, 1310, 1310, 1309, ...
## Resampling results across tuning parameters:
##
##   cp          RMSE      Rsquared    MAE
## 0.02426438 38.40981 0.22829163 19.70701
## 0.07763399 40.63197 0.13067041 20.48623
## 0.10154375 41.88394 0.07935057 22.35448
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was cp = 0.02426438.
fancyRpartPlot(model_rpart$finalModel)

```



Rattle 2020-Apr-07 10:24:31 Jesus

## Partial Least Squares (PLS) to Train The Prediction Model

```
set.seed(27)
model_pls <- caret::train(total_cases ~ .,
                          data = random_train[,c(2,3:6,12,16,20,23)],
                          method = "pls",
                          preProcess = NULL,
                          trControl = train.control)

model_pls

## Partial Least Squares
##
## 1456 samples
##    8 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1312, 1310, 1312, 1310, 1308, 1309, ...
## Resampling results across tuning parameters:
##
##  ncomp  RMSE      Rsquared    MAE
##   1      41.20706  0.05916277  22.39654
##   2      39.43403  0.14559185  20.74194
##   3      39.43646  0.14565291  20.85513
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was ncomp = 2.
```

## Extreme Gradient Boosting

```
cl <- makeCluster(3, type = "SOCK")
registerDoSNOW(cl)

model_xgb <- caret::train(total_cases ~ .,
                          data = random_train[,c(2,3:6,12,16,20,23)],
                          method = "xgbTree",
                          tuneGrid = grid,
                          trControl = train.control)

model_xgb

## eXtreme Gradient Boosting
##
## 1456 samples
##    8 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1312, 1310, 1311, 1310, 1311, ...
## Resampling results across tuning parameters:
##
```



##	eta	max_depth	gamma	colsample_bytree	min_child_weight	subsample
##	0.05	1	0.0	0.5	1	0.8
##	0.05	1	0.0	0.5	1	0.8
##	0.05	1	0.0	0.5	1	1.0
##	0.05	1	0.0	0.5	1	1.0
##	0.05	1	0.0	0.5	4	0.8
##	0.05	1	0.0	0.5	4	0.8
##	0.05	1	0.0	0.5	4	1.0
##	0.05	1	0.0	0.5	4	1.0
##	0.05	1	0.0	1.0	1	0.8
##	0.05	1	0.0	1.0	1	0.8
##	0.05	1	0.0	1.0	1	1.0
##	0.05	1	0.0	1.0	1	1.0
##	0.05	1	0.0	1.0	4	0.8
##	0.05	1	0.0	1.0	4	0.8
##	0.05	1	0.0	1.0	4	1.0
##	0.05	1	0.0	1.0	4	1.0
##	0.05	1	0.1	0.5	1	0.8
##	0.05	1	0.1	0.5	1	0.8
##	0.05	1	0.1	0.5	1	1.0
##	0.05	1	0.1	0.5	1	1.0
##	0.05	1	0.1	0.5	4	0.8
##	0.05	1	0.1	0.5	4	0.8
##	0.05	1	0.1	0.5	4	1.0
##	0.05	1	0.1	0.5	4	1.0
##	0.05	1	0.1	1.0	1	0.8
##	0.05	1	0.1	1.0	1	0.8
##	0.05	1	0.1	1.0	1	1.0
##	0.05	1	0.1	1.0	1	1.0
##	0.05	1	0.1	1.0	4	0.8
##	0.05	1	0.1	1.0	4	0.8
##	0.05	1	0.1	1.0	4	1.0
##	0.05	1	0.1	1.0	4	1.0
##	0.05	2	0.0	0.5	1	0.8
##	0.05	2	0.0	0.5	1	0.8
##	0.05	2	0.0	0.5	1	1.0
##	0.05	2	0.0	0.5	1	1.0
##	0.05	2	0.0	0.5	4	0.8
##	0.05	2	0.0	0.5	4	0.8
##	0.05	2	0.0	0.5	4	1.0
##	0.05	2	0.0	0.5	4	1.0
##	0.05	2	0.0	1.0	1	0.8
##	0.05	2	0.0	1.0	1	0.8
##	0.05	2	0.0	1.0	1	1.0
##	0.05	2	0.0	1.0	1	1.0
##	0.05	2	0.0	1.0	4	0.8
##	0.05	2	0.0	1.0	4	0.8
##	0.05	2	0.0	1.0	4	1.0
##	0.05	2	0.0	1.0	4	1.0
##	0.05	2	0.1	0.5	1	0.8
##	0.05	2	0.1	0.5	1	0.8
##	0.05	2	0.1	0.5	1	1.0
##	0.05	2	0.1	0.5	1	1.0
##	0.05	2	0.1	0.5	4	0.8

##	0.05	2	0.1	0.5	4	0.8
##	0.05	2	0.1	0.5	4	1.0
##	0.05	2	0.1	0.5	4	1.0
##	0.05	2	0.1	1.0	1	0.8
##	0.05	2	0.1	1.0	1	0.8
##	0.05	2	0.1	1.0	1	1.0
##	0.05	2	0.1	1.0	1	1.0
##	0.05	2	0.1	1.0	4	0.8
##	0.05	2	0.1	1.0	4	0.8
##	0.05	2	0.1	1.0	4	1.0
##	0.05	2	0.1	1.0	4	1.0
##	0.05	3	0.0	0.5	1	0.8
##	0.05	3	0.0	0.5	1	0.8
##	0.05	3	0.0	0.5	1	1.0
##	0.05	3	0.0	0.5	1	1.0
##	0.05	3	0.0	0.5	4	0.8
##	0.05	3	0.0	0.5	4	0.8
##	0.05	3	0.0	0.5	4	1.0
##	0.05	3	0.0	0.5	4	1.0
##	0.05	3	0.0	1.0	1	0.8
##	0.05	3	0.0	1.0	1	0.8
##	0.05	3	0.0	1.0	1	1.0
##	0.05	3	0.0	1.0	1	1.0
##	0.05	3	0.0	1.0	4	0.8
##	0.05	3	0.0	1.0	4	0.8
##	0.05	3	0.0	1.0	4	1.0
##	0.05	3	0.0	1.0	4	1.0
##	0.05	3	0.1	0.5	1	0.8
##	0.05	3	0.1	0.5	1	0.8
##	0.05	3	0.1	0.5	1	1.0
##	0.05	3	0.1	0.5	1	1.0
##	0.05	3	0.1	0.5	4	0.8
##	0.05	3	0.1	0.5	4	0.8
##	0.05	3	0.1	0.5	4	1.0
##	0.05	3	0.1	0.5	4	1.0
##	0.05	3	0.1	1.0	1	0.8
##	0.05	3	0.1	1.0	1	0.8
##	0.05	3	0.1	1.0	1	1.0
##	0.05	3	0.1	1.0	1	1.0
##	0.05	3	0.1	1.0	4	0.8
##	0.05	3	0.1	1.0	4	0.8
##	0.05	3	0.1	1.0	4	1.0
##	0.05	3	0.1	1.0	4	1.0
##	0.05	4	0.0	0.5	1	0.8
##	0.05	4	0.0	0.5	1	0.8
##	0.05	4	0.0	0.5	1	1.0
##	0.05	4	0.0	0.5	1	1.0
##	0.05	4	0.0	0.5	4	0.8
##	0.05	4	0.0	0.5	4	0.8
##	0.05	4	0.0	0.5	4	1.0
##	0.05	4	0.0	0.5	4	1.0
##	0.05	4	0.0	1.0	1	0.8
##	0.05	4	0.0	1.0	1	0.8
##	0.05	4	0.0	1.0	1	1.0

##	0.05	4	0.0	1.0	1	1.0
##	0.05	4	0.0	1.0	4	0.8
##	0.05	4	0.0	1.0	4	0.8
##	0.05	4	0.0	1.0	4	1.0
##	0.05	4	0.0	1.0	4	1.0
##	0.05	4	0.1	0.5	1	0.8
##	0.05	4	0.1	0.5	1	0.8
##	0.05	4	0.1	0.5	1	1.0
##	0.05	4	0.1	0.5	1	1.0
##	0.05	4	0.1	0.5	4	0.8
##	0.05	4	0.1	0.5	4	0.8
##	0.05	4	0.1	0.5	4	1.0
##	0.05	4	0.1	0.5	4	1.0
##	0.05	4	0.1	1.0	1	0.8
##	0.05	4	0.1	1.0	1	0.8
##	0.05	4	0.1	1.0	1	1.0
##	0.05	4	0.1	1.0	1	1.0
##	0.05	4	0.1	1.0	4	0.8
##	0.05	4	0.1	1.0	4	0.8
##	0.05	4	0.1	1.0	4	1.0
##	0.05	4	0.1	1.0	4	1.0
##	0.05	5	0.0	0.5	1	0.8
##	0.05	5	0.0	0.5	1	0.8
##	0.05	5	0.0	0.5	1	1.0
##	0.05	5	0.0	0.5	1	1.0
##	0.05	5	0.0	0.5	4	0.8
##	0.05	5	0.0	0.5	4	0.8
##	0.05	5	0.0	0.5	4	1.0
##	0.05	5	0.0	0.5	4	1.0
##	0.05	5	0.0	1.0	1	0.8
##	0.05	5	0.0	1.0	1	0.8
##	0.05	5	0.0	1.0	1	1.0
##	0.05	5	0.0	1.0	1	1.0
##	0.05	5	0.0	1.0	4	0.8
##	0.05	5	0.0	1.0	4	0.8
##	0.05	5	0.0	1.0	4	1.0
##	0.05	5	0.0	1.0	4	1.0
##	0.05	5	0.1	0.5	1	0.8
##	0.05	5	0.1	0.5	1	0.8
##	0.05	5	0.1	0.5	1	1.0
##	0.05	5	0.1	0.5	1	1.0
##	0.05	5	0.1	0.5	4	0.8
##	0.05	5	0.1	0.5	4	0.8
##	0.05	5	0.1	0.5	4	1.0
##	0.05	5	0.1	0.5	4	1.0
##	0.05	5	0.1	1.0	1	0.8
##	0.05	5	0.1	1.0	1	0.8
##	0.05	5	0.1	1.0	1	1.0
##	0.05	5	0.1	1.0	1	1.0
##	0.05	5	0.1	1.0	4	0.8
##	0.05	5	0.1	1.0	4	0.8
##	0.05	5	0.1	1.0	4	1.0
##	0.05	5	0.1	1.0	4	1.0
##	0.05	6	0.0	0.5	1	0.8

##	0.05	6	0.0	0.5	1	0.8
##	0.05	6	0.0	0.5	1	1.0
##	0.05	6	0.0	0.5	1	1.0
##	0.05	6	0.0	0.5	4	0.8
##	0.05	6	0.0	0.5	4	0.8
##	0.05	6	0.0	0.5	4	1.0
##	0.05	6	0.0	0.5	4	1.0
##	0.05	6	0.0	1.0	1	0.8
##	0.05	6	0.0	1.0	1	0.8
##	0.05	6	0.0	1.0	1	1.0
##	0.05	6	0.0	1.0	1	1.0
##	0.05	6	0.0	1.0	4	0.8
##	0.05	6	0.0	1.0	4	0.8
##	0.05	6	0.0	1.0	4	1.0
##	0.05	6	0.0	1.0	4	1.0
##	0.05	6	0.1	0.5	1	0.8
##	0.05	6	0.1	0.5	1	0.8
##	0.05	6	0.1	0.5	1	1.0
##	0.05	6	0.1	0.5	1	1.0
##	0.05	6	0.1	0.5	4	0.8
##	0.05	6	0.1	0.5	4	0.8
##	0.05	6	0.1	0.5	4	1.0
##	0.05	6	0.1	0.5	4	1.0
##	0.05	6	0.1	1.0	1	0.8
##	0.05	6	0.1	1.0	1	0.8
##	0.05	6	0.1	1.0	1	1.0
##	0.05	6	0.1	1.0	1	1.0
##	0.05	6	0.1	1.0	4	0.8
##	0.05	6	0.1	1.0	4	0.8
##	0.05	6	0.1	1.0	4	1.0
##	0.05	6	0.1	1.0	4	1.0
##	0.50	1	0.0	0.5	1	0.8
##	0.50	1	0.0	0.5	1	0.8
##	0.50	1	0.0	0.5	1	1.0
##	0.50	1	0.0	0.5	1	1.0
##	0.50	1	0.0	0.5	4	0.8
##	0.50	1	0.0	0.5	4	0.8
##	0.50	1	0.0	0.5	4	1.0
##	0.50	1	0.0	0.5	4	1.0
##	0.50	1	0.0	1.0	1	0.8
##	0.50	1	0.0	1.0	1	0.8
##	0.50	1	0.0	1.0	1	1.0
##	0.50	1	0.0	1.0	1	1.0
##	0.50	1	0.0	1.0	4	0.8
##	0.50	1	0.0	1.0	4	0.8
##	0.50	1	0.0	1.0	4	1.0
##	0.50	1	0.0	1.0	4	1.0
##	0.50	1	0.1	0.5	1	0.8
##	0.50	1	0.1	0.5	1	0.8
##	0.50	1	0.1	0.5	1	1.0
##	0.50	1	0.1	0.5	1	1.0
##	0.50	1	0.1	0.5	4	0.8
##	0.50	1	0.1	0.5	4	0.8
##	0.50	1	0.1	0.5	4	1.0

##	0.50	1	0.1	0.5	4	1.0
##	0.50	1	0.1	1.0	1	0.8
##	0.50	1	0.1	1.0	1	0.8
##	0.50	1	0.1	1.0	1	1.0
##	0.50	1	0.1	1.0	1	1.0
##	0.50	1	0.1	1.0	4	0.8
##	0.50	1	0.1	1.0	4	0.8
##	0.50	1	0.1	1.0	4	1.0
##	0.50	1	0.1	1.0	4	1.0
##	0.50	2	0.0	0.5	1	0.8
##	0.50	2	0.0	0.5	1	0.8
##	0.50	2	0.0	0.5	1	1.0
##	0.50	2	0.0	0.5	1	1.0
##	0.50	2	0.0	0.5	4	0.8
##	0.50	2	0.0	0.5	4	0.8
##	0.50	2	0.0	0.5	4	1.0
##	0.50	2	0.0	0.5	4	1.0
##	0.50	2	0.0	1.0	1	0.8
##	0.50	2	0.0	1.0	1	0.8
##	0.50	2	0.0	1.0	1	1.0
##	0.50	2	0.0	1.0	1	1.0
##	0.50	2	0.0	1.0	4	0.8
##	0.50	2	0.0	1.0	4	0.8
##	0.50	2	0.0	1.0	4	1.0
##	0.50	2	0.0	1.0	4	1.0
##	0.50	2	0.1	0.5	1	0.8
##	0.50	2	0.1	0.5	1	0.8
##	0.50	2	0.1	0.5	1	1.0
##	0.50	2	0.1	0.5	1	1.0
##	0.50	2	0.1	0.5	4	0.8
##	0.50	2	0.1	0.5	4	0.8
##	0.50	2	0.1	0.5	4	1.0
##	0.50	2	0.1	0.5	4	1.0
##	0.50	2	0.1	1.0	1	0.8
##	0.50	2	0.1	1.0	1	0.8
##	0.50	2	0.1	1.0	1	1.0
##	0.50	2	0.1	1.0	1	1.0
##	0.50	2	0.1	1.0	4	0.8
##	0.50	2	0.1	1.0	4	0.8
##	0.50	2	0.1	1.0	4	1.0
##	0.50	2	0.1	1.0	4	1.0
##	0.50	3	0.0	0.5	1	0.8
##	0.50	3	0.0	0.5	1	0.8
##	0.50	3	0.0	0.5	1	1.0
##	0.50	3	0.0	0.5	1	1.0
##	0.50	3	0.0	0.5	4	0.8
##	0.50	3	0.0	0.5	4	0.8
##	0.50	3	0.0	0.5	4	1.0
##	0.50	3	0.0	0.5	4	1.0
##	0.50	3	0.0	1.0	1	0.8
##	0.50	3	0.0	1.0	1	0.8
##	0.50	3	0.0	1.0	1	1.0
##	0.50	3	0.0	1.0	1	1.0
##	0.50	3	0.0	1.0	4	0.8

##	0.50	3	0.0	1.0	4	0.8
##	0.50	3	0.0	1.0	4	1.0
##	0.50	3	0.0	1.0	4	1.0
##	0.50	3	0.1	0.5	1	0.8
##	0.50	3	0.1	0.5	1	0.8
##	0.50	3	0.1	0.5	1	1.0
##	0.50	3	0.1	0.5	1	1.0
##	0.50	3	0.1	0.5	4	0.8
##	0.50	3	0.1	0.5	4	0.8
##	0.50	3	0.1	0.5	4	1.0
##	0.50	3	0.1	0.5	4	1.0
##	0.50	3	0.1	1.0	1	0.8
##	0.50	3	0.1	1.0	1	0.8
##	0.50	3	0.1	1.0	1	1.0
##	0.50	3	0.1	1.0	1	1.0
##	0.50	3	0.1	1.0	4	0.8
##	0.50	3	0.1	1.0	4	0.8
##	0.50	3	0.1	1.0	4	1.0
##	0.50	3	0.1	1.0	4	1.0
##	0.50	4	0.0	0.5	1	0.8
##	0.50	4	0.0	0.5	1	0.8
##	0.50	4	0.0	0.5	1	1.0
##	0.50	4	0.0	0.5	1	1.0
##	0.50	4	0.0	0.5	4	0.8
##	0.50	4	0.0	0.5	4	0.8
##	0.50	4	0.0	0.5	4	1.0
##	0.50	4	0.0	0.5	4	1.0
##	0.50	4	0.0	1.0	1	0.8
##	0.50	4	0.0	1.0	1	0.8
##	0.50	4	0.0	1.0	1	1.0
##	0.50	4	0.0	1.0	1	1.0
##	0.50	4	0.0	1.0	4	0.8
##	0.50	4	0.0	1.0	4	0.8
##	0.50	4	0.0	1.0	4	1.0
##	0.50	4	0.0	1.0	4	1.0
##	0.50	4	0.1	0.5	1	0.8
##	0.50	4	0.1	0.5	1	0.8
##	0.50	4	0.1	0.5	1	1.0
##	0.50	4	0.1	0.5	1	1.0
##	0.50	4	0.1	0.5	4	0.8
##	0.50	4	0.1	0.5	4	0.8
##	0.50	4	0.1	0.5	4	1.0
##	0.50	4	0.1	0.5	4	1.0
##	0.50	4	0.1	1.0	1	0.8
##	0.50	4	0.1	1.0	1	0.8
##	0.50	4	0.1	1.0	1	1.0
##	0.50	4	0.1	1.0	1	1.0
##	0.50	4	0.1	1.0	4	0.8
##	0.50	4	0.1	1.0	4	0.8
##	0.50	4	0.1	1.0	4	1.0
##	0.50	4	0.1	1.0	4	1.0
##	0.50	4	0.1	1.0	4	1.0
##	0.50	5	0.0	0.5	1	0.8
##	0.50	5	0.0	0.5	1	0.8
##	0.50	5	0.0	0.5	1	1.0

##	0.50	5	0.0	0.5	1	1.0
##	0.50	5	0.0	0.5	4	0.8
##	0.50	5	0.0	0.5	4	0.8
##	0.50	5	0.0	0.5	4	1.0
##	0.50	5	0.0	0.5	4	1.0
##	0.50	5	0.0	1.0	1	0.8
##	0.50	5	0.0	1.0	1	0.8
##	0.50	5	0.0	1.0	1	1.0
##	0.50	5	0.0	1.0	1	1.0
##	0.50	5	0.0	1.0	4	0.8
##	0.50	5	0.0	1.0	4	0.8
##	0.50	5	0.0	1.0	4	1.0
##	0.50	5	0.0	1.0	4	1.0
##	0.50	5	0.1	0.5	1	0.8
##	0.50	5	0.1	0.5	1	0.8
##	0.50	5	0.1	0.5	1	1.0
##	0.50	5	0.1	0.5	1	1.0
##	0.50	5	0.1	0.5	4	0.8
##	0.50	5	0.1	0.5	4	0.8
##	0.50	5	0.1	0.5	4	1.0
##	0.50	5	0.1	0.5	4	1.0
##	0.50	5	0.1	1.0	1	0.8
##	0.50	5	0.1	1.0	1	0.8
##	0.50	5	0.1	1.0	1	1.0
##	0.50	5	0.1	1.0	1	1.0
##	0.50	5	0.1	1.0	4	0.8
##	0.50	5	0.1	1.0	4	0.8
##	0.50	5	0.1	1.0	4	1.0
##	0.50	5	0.1	1.0	4	1.0
##	0.50	6	0.0	0.5	1	0.8
##	0.50	6	0.0	0.5	1	0.8
##	0.50	6	0.0	0.5	1	1.0
##	0.50	6	0.0	0.5	1	1.0
##	0.50	6	0.0	0.5	4	0.8
##	0.50	6	0.0	0.5	4	0.8
##	0.50	6	0.0	0.5	4	1.0
##	0.50	6	0.0	0.5	4	1.0
##	0.50	6	0.0	1.0	1	0.8
##	0.50	6	0.0	1.0	1	0.8
##	0.50	6	0.0	1.0	1	1.0
##	0.50	6	0.0	1.0	1	1.0
##	0.50	6	0.0	1.0	4	0.8
##	0.50	6	0.0	1.0	4	0.8
##	0.50	6	0.0	1.0	4	1.0
##	0.50	6	0.0	1.0	4	1.0
##	0.50	6	0.1	0.5	1	0.8
##	0.50	6	0.1	0.5	1	0.8
##	0.50	6	0.1	0.5	1	1.0
##	0.50	6	0.1	0.5	1	1.0
##	0.50	6	0.1	0.5	4	0.8
##	0.50	6	0.1	0.5	4	0.8
##	0.50	6	0.1	0.5	4	1.0
##	0.50	6	0.1	0.5	4	1.0
##	0.50	6	0.1	1.0	1	0.8

##	0.50	6	0.1	1.0	1	0.8
##	0.50	6	0.1	1.0	1	1.0
##	0.50	6	0.1	1.0	1	1.0
##	0.50	6	0.1	1.0	4	0.8
##	0.50	6	0.1	1.0	4	0.8
##	0.50	6	0.1	1.0	4	1.0
##	0.50	6	0.1	1.0	4	1.0
##	nrounds	RMSE	Rsquared	MAE		
##	70	38.93380	0.1811865	19.65144		
##	90	38.75062	0.1884744	19.71423		
##	70	39.00722	0.1778792	19.68096		
##	90	38.86030	0.1834527	19.79050		
##	70	38.93968	0.1804497	19.64199		
##	90	38.73566	0.1890442	19.70852		
##	70	39.00119	0.1781800	19.68745		
##	90	38.85714	0.1837445	19.79708		
##	70	38.91931	0.1815206	19.62425		
##	90	38.71840	0.1901745	19.68537		
##	70	39.00397	0.1784542	19.68371		
##	90	38.86922	0.1833651	19.80421		
##	70	38.92391	0.1811875	19.64981		
##	90	38.73350	0.1890720	19.71377		
##	70	39.00397	0.1784542	19.68371		
##	90	38.86922	0.1833651	19.80421		
##	70	38.92781	0.1815651	19.65428		
##	90	38.73998	0.1893634	19.70127		
##	70	39.00390	0.1780704	19.68443		
##	90	38.85916	0.1837981	19.79145		
##	70	38.94398	0.1806536	19.64717		
##	90	38.74769	0.1886212	19.71482		
##	70	39.00367	0.1780555	19.68086		
##	90	38.85777	0.1836602	19.78702		
##	70	38.90915	0.1817254	19.63188		
##	90	38.70962	0.1902086	19.68265		
##	70	39.00397	0.1784542	19.68371		
##	90	38.86922	0.1833651	19.80421		
##	70	38.92868	0.1808875	19.61471		
##	90	38.74498	0.1887385	19.68799		
##	70	39.00397	0.1784542	19.68371		
##	90	38.86922	0.1833651	19.80421		
##	70	35.84972	0.3235034	17.96664		
##	90	34.54454	0.3705581	17.88460		
##	70	35.56104	0.3361276	17.96228		
##	90	34.03300	0.3879934	17.86556		
##	70	35.79299	0.3264867	17.95570		
##	90	34.41189	0.3772544	17.87272		
##	70	35.50927	0.3383005	17.96661		
##	90	33.82476	0.3941956	17.81327		
##	70	35.65256	0.3285653	17.95976		
##	90	34.42660	0.3714123	17.91153		
##	70	35.48176	0.3379096	17.97487		
##	90	34.04353	0.3854159	17.86416		
##	70	35.60412	0.3332009	17.87980		
##	90	34.33363	0.3775531	17.85228		



##	70	35.44613	0.3398192	17.95465
##	90	33.95001	0.3891928	17.83760
##	70	35.82011	0.3242790	17.92563
##	90	34.53060	0.3702405	17.87523
##	70	35.70765	0.3304450	17.98320
##	90	33.91057	0.3924805	17.82293
##	70	35.78347	0.3269908	17.94749
##	90	34.38717	0.3778985	17.81983
##	70	35.56691	0.3369841	17.99790
##	90	33.89499	0.3936127	17.87117
##	70	35.47190	0.3371538	17.85537
##	90	34.27929	0.3786043	17.81701
##	70	35.48176	0.3379096	17.97487
##	90	34.04353	0.3854159	17.86416
##	70	35.50708	0.3371729	17.86153
##	90	34.36183	0.3765412	17.83044
##	70	35.44613	0.3398192	17.95465
##	90	33.95001	0.3891928	17.83760
##	70	32.41080	0.4398004	17.17372
##	90	31.54636	0.4582108	17.13542
##	70	32.36896	0.4427821	17.13233
##	90	31.59248	0.4585362	17.13127
##	70	32.17453	0.4466124	17.08531
##	90	31.38715	0.4626057	17.08961
##	70	32.29724	0.4427478	17.12071
##	90	31.50800	0.4594526	17.11193
##	70	32.40625	0.4386393	17.01008
##	90	31.50130	0.4603295	16.95997
##	70	32.66315	0.4271446	17.14496
##	90	31.78915	0.4500060	17.09879
##	70	32.23860	0.4435239	17.00490
##	90	31.38326	0.4634397	16.99090
##	70	32.61626	0.4289618	17.17503
##	90	31.75532	0.4501522	17.14432
##	70	32.26378	0.4459287	17.08666
##	90	31.44187	0.4636288	17.07080
##	70	32.32827	0.4422645	17.13189
##	90	31.49046	0.4602367	17.09333
##	70	31.99076	0.4574056	17.04988
##	90	31.28202	0.4695641	17.08201
##	70	32.27846	0.4409741	17.11747
##	90	31.40688	0.4593071	17.06677
##	70	32.39534	0.4367507	17.04653
##	90	31.54369	0.4574322	17.01059
##	70	32.66315	0.4271446	17.14496
##	90	31.78915	0.4500060	17.09879
##	70	32.34883	0.4381627	17.03074
##	90	31.43616	0.4600881	16.98290
##	70	32.61626	0.4289618	17.17503
##	90	31.75532	0.4501522	17.14432
##	70	30.72661	0.4830078	16.72953
##	90	30.17882	0.4900561	16.74778
##	70	30.72395	0.4800030	16.68122
##	90	30.23124	0.4889268	16.70962

##	70	30.59136	0.4815418	16.68365
##	90	30.11921	0.4882307	16.70035
##	70	30.74853	0.4757912	16.69117
##	90	30.19466	0.4855680	16.71121
##	70	30.41586	0.4853927	16.44326
##	90	29.91159	0.4948687	16.47957
##	70	30.00040	0.4904151	16.37641
##	90	29.53965	0.5012919	16.42901
##	70	30.06253	0.4970370	16.38223
##	90	29.52312	0.5065568	16.41140
##	70	29.87647	0.4932038	16.40944
##	90	29.44787	0.5024174	16.46693
##	70	30.73638	0.4792939	16.72448
##	90	30.19514	0.4887178	16.70334
##	70	30.95246	0.4729681	16.76501
##	90	30.45559	0.4816629	16.78971
##	70	30.61908	0.4845244	16.66176
##	90	30.07100	0.4927244	16.66709
##	70	30.56755	0.4824449	16.68460
##	90	30.09377	0.4900970	16.72591
##	70	30.25158	0.4921180	16.36409
##	90	29.72739	0.5013216	16.40979
##	70	30.00040	0.4904151	16.37641
##	90	29.53965	0.5012905	16.42835
##	70	30.14351	0.4934133	16.39469
##	90	29.58956	0.5040379	16.40149
##	70	29.87647	0.4932038	16.40944
##	90	29.44787	0.5024174	16.46693
##	70	29.50097	0.5082819	16.28136
##	90	29.18651	0.5099628	16.37007
##	70	29.83411	0.4982878	16.39700
##	90	29.59209	0.5005620	16.52407
##	70	29.71906	0.5002562	16.38785
##	90	29.38988	0.5033017	16.46059
##	70	29.91350	0.4928540	16.49860
##	90	29.60380	0.4967459	16.56496
##	70	29.60024	0.5021617	16.17512
##	90	29.25608	0.5075869	16.26582
##	70	29.57206	0.4987872	16.31189
##	90	29.29491	0.5053331	16.42451
##	70	29.55402	0.5042854	16.19260
##	90	29.25599	0.5076697	16.30515
##	70	29.40492	0.5009533	16.32415
##	90	29.13154	0.5066029	16.40671
##	70	29.56013	0.5113736	16.30482
##	90	29.23734	0.5132021	16.37972
##	70	29.83760	0.4958452	16.40741
##	90	29.59310	0.4980113	16.52002
##	70	29.63124	0.5039334	16.33061
##	90	29.26756	0.5068694	16.37611
##	70	29.96231	0.4903694	16.50984
##	90	29.62852	0.4949486	16.58379
##	70	29.56852	0.5049576	16.15761
##	90	29.18899	0.5112107	16.25278

##	70	29.57213	0.4987838	16.31189
##	90	29.29553	0.5053027	16.42526
##	70	29.39959	0.5077494	16.11165
##	90	29.07132	0.5128366	16.20769
##	70	29.40492	0.5009533	16.32415
##	90	29.13154	0.5066029	16.40671
##	70	29.26479	0.5107976	16.23977
##	90	29.02907	0.5118618	16.35765
##	70	29.38017	0.5048296	16.27899
##	90	29.18387	0.5053661	16.40903
##	70	29.38038	0.5051692	16.31876
##	90	29.11215	0.5073081	16.41483
##	70	29.25268	0.5085446	16.36441
##	90	29.07403	0.5084985	16.48799
##	70	29.07501	0.5130483	15.94604
##	90	28.81851	0.5163021	16.05299
##	70	29.88874	0.4860657	16.27820
##	90	29.72878	0.4902940	16.43458
##	70	28.85962	0.5172747	15.98583
##	90	28.71478	0.5179434	16.12768
##	70	29.46742	0.4969129	16.23030
##	90	29.31771	0.4993093	16.37965
##	70	29.27494	0.5114814	16.28642
##	90	29.04310	0.5121681	16.41456
##	70	29.45984	0.5041248	16.26302
##	90	29.27336	0.5043921	16.41240
##	70	29.22788	0.5113015	16.28249
##	90	28.96773	0.5129881	16.37434
##	70	29.07563	0.5114832	16.28838
##	90	28.91115	0.5108919	16.40839
##	70	29.28461	0.5076716	16.00617
##	90	29.05540	0.5111310	16.13832
##	70	29.88383	0.4861456	16.28020
##	90	29.69717	0.4909166	16.43553
##	70	29.05052	0.5119041	16.01769
##	90	28.83842	0.5144064	16.15154
##	70	29.46739	0.4969126	16.23033
##	90	29.31742	0.4992894	16.37951
##	70	35.98060	0.3015864	20.13145
##	90	35.43582	0.3210418	20.11454
##	70	35.65601	0.3155451	19.76704
##	90	35.06844	0.3370865	19.68742
##	70	35.99480	0.3004137	20.09721
##	90	35.48595	0.3197571	20.07503
##	70	35.67726	0.3149085	19.69424
##	90	35.07329	0.3366886	19.65049
##	70	35.47556	0.3203613	19.77163
##	90	34.97233	0.3389534	19.83753
##	70	35.04275	0.3442549	19.29808
##	90	34.36665	0.3668574	19.19739
##	70	35.50774	0.3186878	19.77951
##	90	34.90026	0.3411037	19.73372
##	70	35.04124	0.3443104	19.29092
##	90	34.37638	0.3661398	19.20255

##	70	35.93019	0.3021627	20.02438
##	90	35.49257	0.3185020	20.09216
##	70	35.65846	0.3166420	19.66582
##	90	35.03845	0.3391553	19.67980
##	70	35.88299	0.3029580	20.04783
##	90	35.37043	0.3217758	20.01334
##	70	35.62401	0.3170001	19.65158
##	90	35.00853	0.3398895	19.61109
##	70	35.51376	0.3205009	19.86105
##	90	34.91719	0.3420988	19.85246
##	70	35.04275	0.3442549	19.29808
##	90	34.36665	0.3668574	19.19739
##	70	35.59372	0.3154529	19.85094
##	90	34.99581	0.3373584	19.85453
##	70	35.04124	0.3443104	19.29092
##	90	34.37638	0.3661398	19.20255
##	70	32.34011	0.4183742	19.14101
##	90	32.37309	0.4177610	19.30704
##	70	31.82108	0.4352718	18.60892
##	90	31.93202	0.4333120	18.81224
##	70	32.41569	0.4203684	19.24165
##	90	32.55981	0.4171633	19.50413
##	70	31.78268	0.4333030	19.00007
##	90	31.97464	0.4291061	19.23146
##	70	32.33535	0.4195302	19.19326
##	90	32.50347	0.4170781	19.45657
##	70	31.55345	0.4371784	18.60847
##	90	31.63112	0.4365241	18.76661
##	70	32.01270	0.4305010	19.02135
##	90	32.21501	0.4264963	19.35959
##	70	31.85675	0.4318786	18.80310
##	90	31.88410	0.4320160	19.04337
##	70	32.01470	0.4304105	19.06212
##	90	32.22609	0.4275873	19.31758
##	70	31.70401	0.4359145	18.82903
##	90	31.86282	0.4338760	19.04207
##	70	32.25203	0.4238945	19.30807
##	90	32.33842	0.4233952	19.54238
##	70	32.24746	0.4229531	18.89919
##	90	32.50817	0.4166323	19.27577
##	70	32.61527	0.4156039	19.15539
##	90	32.83104	0.4131757	19.49814
##	70	31.55345	0.4371784	18.60847
##	90	31.63112	0.4365241	18.76661
##	70	32.30332	0.4241118	19.15850
##	90	32.46076	0.4215501	19.38086
##	70	31.85675	0.4318786	18.80310
##	90	31.88410	0.4320160	19.04337
##	70	32.96675	0.4177799	19.85073
##	90	33.28630	0.4114856	20.18791
##	70	32.35368	0.4222981	19.26682
##	90	32.49920	0.4210122	19.48663
##	70	33.29089	0.4005706	19.95211
##	90	33.47952	0.3971331	20.26343

##	70	32.15653	0.4287342	19.38228
##	90	32.39967	0.4243681	19.70438
##	70	32.68858	0.4219741	19.43899
##	90	32.89982	0.4184413	19.76925
##	70	31.48422	0.4467396	18.70313
##	90	31.66028	0.4439748	18.92441
##	70	32.55196	0.4267835	19.49737
##	90	32.74326	0.4250784	19.82013
##	70	32.09785	0.4313415	19.12810
##	90	32.30688	0.4286341	19.43530
##	70	32.99750	0.4087248	19.60822
##	90	33.07423	0.4096525	19.82749
##	70	32.51953	0.4125217	19.22148
##	90	32.56703	0.4127274	19.44190
##	70	32.59423	0.4225366	19.78123
##	90	32.81494	0.4191601	20.13688
##	70	32.23464	0.4273166	19.20146
##	90	32.44600	0.4229090	19.47061
##	70	32.74845	0.4233864	19.44138
##	90	32.94088	0.4203954	19.73249
##	70	31.48422	0.4467395	18.70314
##	90	31.66028	0.4439747	18.92442
##	70	32.34641	0.4360200	19.52222
##	90	32.70338	0.4293619	19.89032
##	70	32.09785	0.4313415	19.12810
##	90	32.30688	0.4286341	19.43530
##	70	33.69960	0.4022402	20.11107
##	90	33.87817	0.3985232	20.34104
##	70	32.44009	0.4201828	19.14083
##	90	32.61572	0.4161615	19.38832
##	70	33.79062	0.4015656	20.77950
##	90	33.93451	0.4000871	20.97369
##	70	32.53174	0.4209192	19.41026
##	90	32.73461	0.4184506	19.64508
##	70	32.99633	0.4241486	19.68822
##	90	33.09239	0.4226283	19.90933
##	70	31.43337	0.4456181	18.74513
##	90	31.49702	0.4444286	18.87851
##	70	33.31093	0.4105693	20.13628
##	90	33.67669	0.4038387	20.45039
##	70	31.51826	0.4434107	18.86351
##	90	31.61327	0.4426255	19.10036
##	70	34.04311	0.3938530	20.39246
##	90	34.07959	0.3948376	20.55346
##	70	32.11633	0.4282034	19.11877
##	90	32.21371	0.4268985	19.28714
##	70	33.58896	0.4059543	20.39236
##	90	33.76634	0.4031223	20.62969
##	70	33.00855	0.4032486	19.86942
##	90	33.17189	0.4011758	20.08813
##	70	32.80520	0.4170382	19.64462
##	90	32.89685	0.4160966	19.79776
##	70	31.44111	0.4455216	18.75737
##	90	31.52020	0.4441008	18.89260

##	70	33.20114	0.4077427	20.02558
##	90	33.36473	0.4049219	20.25950
##	70	31.51826	0.4434107	18.86351
##	90	31.61468	0.4425241	19.10614
##	70	34.22135	0.3848097	20.51856
##	90	34.27109	0.3841094	20.61150
##	70	32.64281	0.4178022	19.36089
##	90	32.68431	0.4176388	19.42442
##	70	34.11638	0.3973618	20.72923
##	90	34.23368	0.3949044	20.86619
##	70	33.19060	0.4035324	19.72386
##	90	33.29929	0.4015923	19.88236
##	70	33.52457	0.4115801	19.84656
##	90	33.63427	0.4095034	19.96908
##	70	32.24681	0.4308471	18.65851
##	90	32.25383	0.4312087	18.70005
##	70	33.32195	0.4209169	19.87993
##	90	33.45185	0.4186099	20.04155
##	70	32.67800	0.4251655	19.35587
##	90	32.77037	0.4235386	19.48998
##	70	33.85171	0.3941979	20.00044
##	90	33.91974	0.3931417	20.12051
##	70	33.23194	0.4060054	19.42678
##	90	33.29580	0.4051601	19.52709
##	70	33.91710	0.3898881	20.66078
##	90	34.02794	0.3881594	20.83186
##	70	32.83932	0.4134830	19.56506
##	90	32.92899	0.4111087	19.71729
##	70	33.26958	0.4144730	19.61832
##	90	33.30261	0.4139914	19.69236
##	70	32.37112	0.4271569	18.75771
##	90	32.42407	0.4263558	18.83244
##	70	34.28660	0.3972902	20.39188
##	90	34.31807	0.3975530	20.49382
##	70	32.67408	0.4251443	19.35504
##	90	32.78314	0.4233457	19.50483
##	70	33.78737	0.3905845	19.75666
##	90	33.81378	0.3902440	19.79304
##	70	33.00400	0.4030792	19.13792
##	90	33.03172	0.4027419	19.17733
##	70	34.75917	0.3760577	21.29134
##	90	34.79627	0.3758256	21.35274
##	70	33.38699	0.4020020	19.94065
##	90	33.41120	0.4019735	20.02273
##	70	32.67476	0.4219345	19.13517
##	90	32.70360	0.4213693	19.15593
##	70	32.06768	0.4290151	18.37188
##	90	32.06913	0.4290303	18.39056
##	70	33.84588	0.3978720	19.84670
##	90	33.86619	0.3977249	19.89071
##	70	32.74757	0.4169792	19.12366
##	90	32.78798	0.4161733	19.18363
##	70	33.28438	0.3948133	19.77152
##	90	33.32312	0.3940366	19.82632

```
## 70      32.63134  0.4161330  19.02502
## 90      32.63669  0.4160452  19.04708
## 70      34.51287  0.3810938  21.00366
## 90      34.52759  0.3810824  21.06806
## 70      32.18748  0.4241761  19.24108
## 90      32.23077  0.4234210  19.31860
## 70      33.39993  0.4022029  19.31937
## 90      33.40500  0.4024852  19.34629
## 70      31.99655  0.4302456  18.34360
## 90      32.02546  0.4295806  18.37356
## 70      33.48658  0.4101938  19.84974
## 90      33.51046  0.4099644  19.90047
## 70      32.78186  0.4161460  19.14843
## 90      32.83146  0.4149973  19.20984
##
## RMSE was used to select the optimal model using the smallest value.
## The final values used for the model were nrounds = 90, max_depth = 6,
## eta = 0.05, gamma = 0, colsample_bytree = 1, min_child_weight = 4
## and subsample = 0.8.
```

## Comparing Prediction Models

```
models <- list( xgb = model_xgb,
               rf = model_rf,
               glmnet = model_glmnet,
               kknn = model_kknn,
               pls = model_pls,
               tree = model_rpart
)
resample_results <- resamples(models)
resample_results

##
## Call:
## resamples.default(x = models)
##
## Models: xgb, rf, glmnet, kknn, pls, tree
## Number of resamples: 50
## Performance metrics: MAE, RMSE, Rsquared
## Time estimates for: everything, final model fit
summary(resample_results)
```

```
##
## Call:
## summary.resamples(object = resample_results)
##
## Models: xgb, rf, glmnet, kknn, pls, tree
## Number of resamples: 50
##
## MAE
##           Min. 1st Qu.  Median    Mean 3rd Qu.    Max. NA's
## xgb    12.45974 15.07062 15.79873 16.12768 17.35844 21.05400     0
## rf     12.90208 15.15153 16.89774 16.83161 18.13347 22.23094     0
```

```
## glmnet 15.54462 18.44451 21.09668 20.59967 22.18272 26.84386 0
## kkn 14.36115 15.85349 17.71759 17.61308 18.88676 21.77379 0
## pls 15.68592 18.86746 20.49156 20.74194 22.59253 27.48153 0
## tree 16.59666 18.13871 19.76543 19.70701 20.98316 25.09659 0
##
## RMSE
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max. NA's
## xgb    20.36910 24.77598 28.11116 28.71478 31.22733 45.04036 0
## rf     20.45757 24.54976 30.26079 30.52333 35.55357 43.44913 0
## glmnet 19.79341 29.83750 41.17531 39.20494 46.35801 62.35546 0
## kkn    23.16103 27.69886 31.41306 31.72765 35.80104 45.65694 0
## pls    19.66215 31.15205 39.53229 39.43403 45.76994 64.13915 0
## tree   24.80703 32.61077 38.19259 38.40981 43.57984 56.62911 0
##
## Rsquared
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max. NA's
## xgb    0.12043064 0.3544449 0.5336537 0.5179434 0.6995061 0.8208837 0
## rf     0.16720157 0.3758397 0.5001517 0.4839504 0.6040610 0.7114627 0
## glmnet 0.09559345 0.1254328 0.1376330 0.1517795 0.1649333 0.2775887 0
## kkn    0.18126067 0.3035724 0.4302695 0.4490605 0.5936667 0.7679223 0
## pls    0.08464352 0.1169886 0.1334244 0.1455918 0.1669473 0.2601101 0
## tree   0.05603703 0.1175996 0.1792506 0.2282916 0.3035135 0.5714219 0
```

Conclusion: The prediction model based on **extreme gradient boosting** algorithm is the champion model.

## Prediction of Total Cases in Future Using XGB model

Importing the test data features on which the predictive model will be applied to predict total number of cases per week at a future date)

```
testset <- getURL("https://s3.amazonaws.com/drivendata/data/44/public/dengue_features_test.csv")
testset <- read.csv(text=testset)
names(testset)
```

```
## [1] "city"
## [2] "year"
## [3] "weekofyear"
## [4] "week_start_date"
## [5] "ndvi_ne"
## [6] "ndvi_nw"
## [7] "ndvi_se"
## [8] "ndvi_sw"
## [9] "precipitation_amt_mm"
## [10] "reanalysis_air_temp_k"
## [11] "reanalysis_avg_temp_k"
## [12] "reanalysis_dew_point_temp_k"
## [13] "reanalysis_max_air_temp_k"
## [14] "reanalysis_min_air_temp_k"
## [15] "reanalysis_precip_amt_kg_per_m2"
## [16] "reanalysis_relative_humidity_percent"
## [17] "reanalysis_sat_precip_amt_mm"
## [18] "reanalysis_specific_humidity_g_per_kg"
## [19] "reanalysis_tdtr_k"
```



```

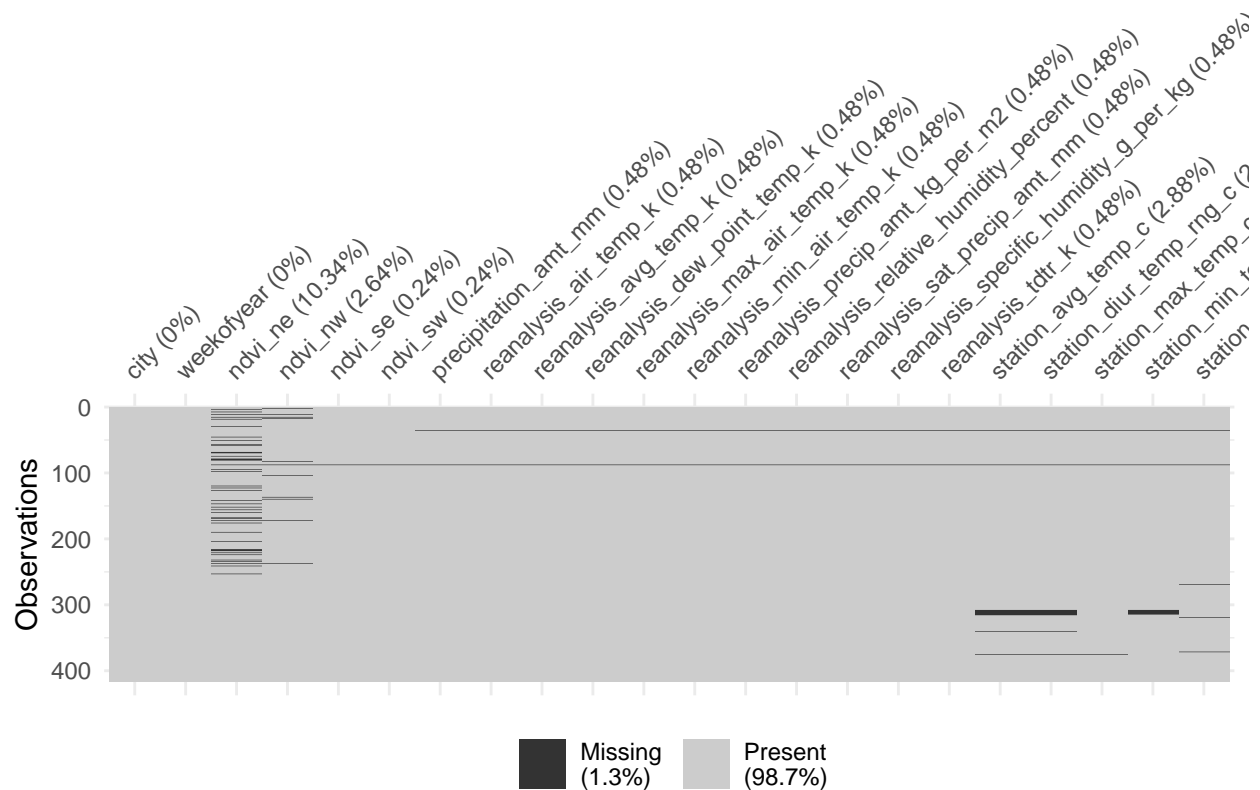
## [20] "station_avg_temp_c"
## [21] "station_diur_temp_rng_c"
## [22] "station_max_temp_c"
## [23] "station_min_temp_c"
## [24] "station_precip_mm"

dengue_test <- testset[, -c(2, 4)]
names(dengue_test)

## [1] "city"
## [2] "weekofyear"
## [3] "ndvi_ne"
## [4] "ndvi_nw"
## [5] "ndvi_se"
## [6] "ndvi_sw"
## [7] "precipitation_amt_mm"
## [8] "reanalysis_air_temp_k"
## [9] "reanalysis_avg_temp_k"
## [10] "reanalysis_dew_point_temp_k"
## [11] "reanalysis_max_air_temp_k"
## [12] "reanalysis_min_air_temp_k"
## [13] "reanalysis_precip_amt_kg_per_m2"
## [14] "reanalysis_relative_humidity_percent"
## [15] "reanalysis_sat_precip_amt_mm"
## [16] "reanalysis_specific_humidity_g_per_kg"
## [17] "reanalysis_tdtr_k"
## [18] "station_avg_temp_c"
## [19] "station_diur_temp_rng_c"
## [20] "station_max_temp_c"
## [21] "station_min_temp_c"
## [22] "station_precip_mm"

# Visualizing missing values for the test data
vis_miss(dengue_test)

```



## Imputation of missing values in the test data

```
names(dengue_test)
```

```
## [1] "city"
## [2] "weekofyear"
## [3] "ndvi_ne"
## [4] "ndvi_nw"
## [5] "ndvi_se"
## [6] "ndvi_sw"
## [7] "precipitation_amt_mm"
## [8] "reanalysis_air_temp_k"
## [9] "reanalysis_avg_temp_k"
## [10] "reanalysis_dew_point_temp_k"
## [11] "reanalysis_max_air_temp_k"
## [12] "reanalysis_min_air_temp_k"
## [13] "reanalysis_precip_amt_kg_per_m2"
## [14] "reanalysis_relative_humidity_percent"
## [15] "reanalysis_sat_precip_amt_mm"
## [16] "reanalysis_specific_humidity_g_per_kg"
## [17] "reanalysis_tdtr_k"
## [18] "station_avg_temp_c"
## [19] "station_diur_temp_rng_c"
## [20] "station_max_temp_c"
## [21] "station_min_temp_c"
```

```
## [22] "station_precip_mm"

pre.process <- preProcess(dengue_test, method = "bagImpute")
imputed.data <- predict(pre.process, dengue_test)
dengue_test$ndvi_ne <- imputed.data[,3]
dengue_test$ndvi_nw <- imputed.data[,4]
dengue_test$ndvi_se <- imputed.data[,5]
dengue_test$ndvi_sw <- imputed.data[,6]
dengue_test$precipitation_amt_mm <- imputed.data[,7]
dengue_test$reanalysis_air_temp_k <- imputed.data[, 8]
dengue_test$reanalysis_avg_temp_k <- imputed.data[,9]
dengue_test$reanalysis_dew_point_temp_k <- imputed.data[,10]
dengue_test$reanalysis_max_air_temp_k <- imputed.data[,11]
dengue_test$reanalysis_min_air_temp_k <- imputed.data[,12]
dengue_test$reanalysis_precip_amt_kg_per_m2 <- imputed.data[,13]
dengue_test$reanalysis_relative_humidity_percent <- imputed.data[,14]
dengue_test$reanalysis_sat_precip_amt_mm <- imputed.data[,15]
dengue_test$reanalysis_specific_humidity_g_per_kg <- imputed.data[,16]
dengue_test$reanalysis_tdtr_k <- imputed.data[,17]
dengue_test$station_avg_temp_c <- imputed.data[,18]
dengue_test$station_diur_temp_rng_c <- imputed.data[,19]
dengue_test$station_max_temp_c <- imputed.data[,20]
dengue_test$station_min_temp_c <- imputed.data[,21]
dengue_test$station_precip_mm <- imputed.data[,22]

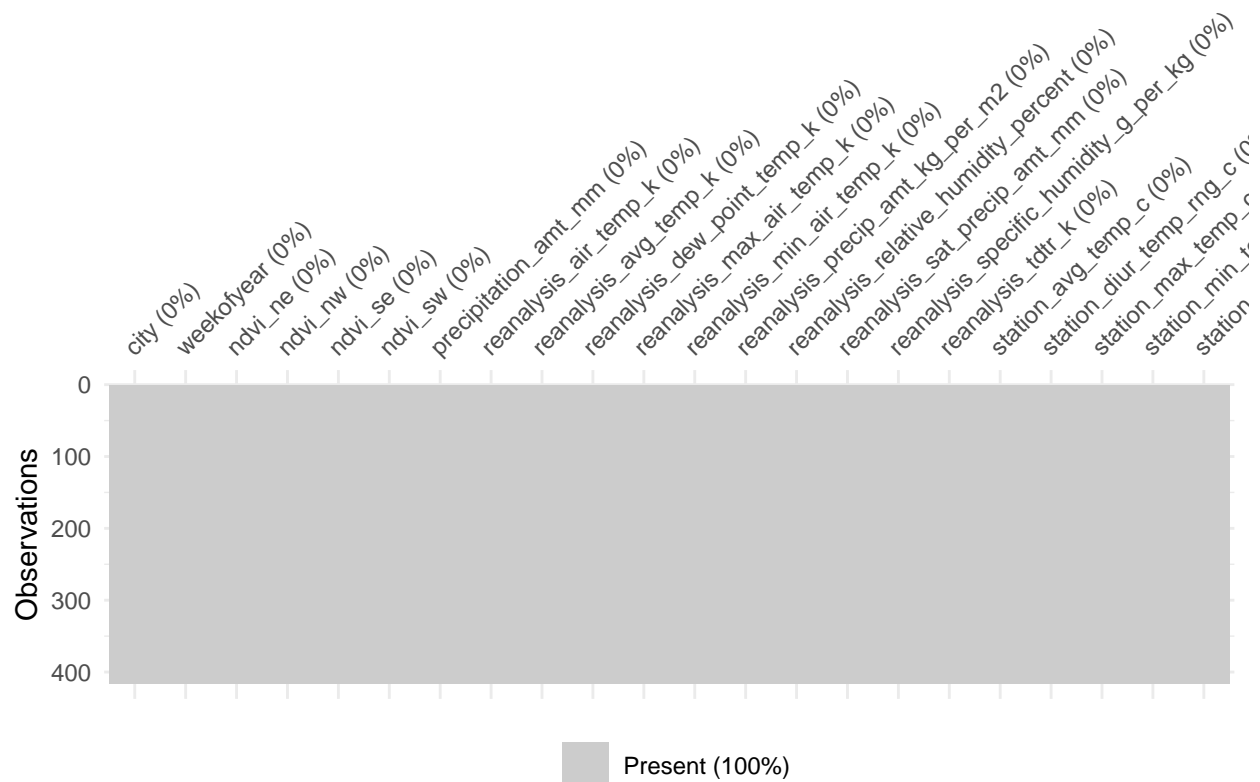
dim(dengue_test)

## [1] 416 22

anyNA(dengue_test)

## [1] FALSE

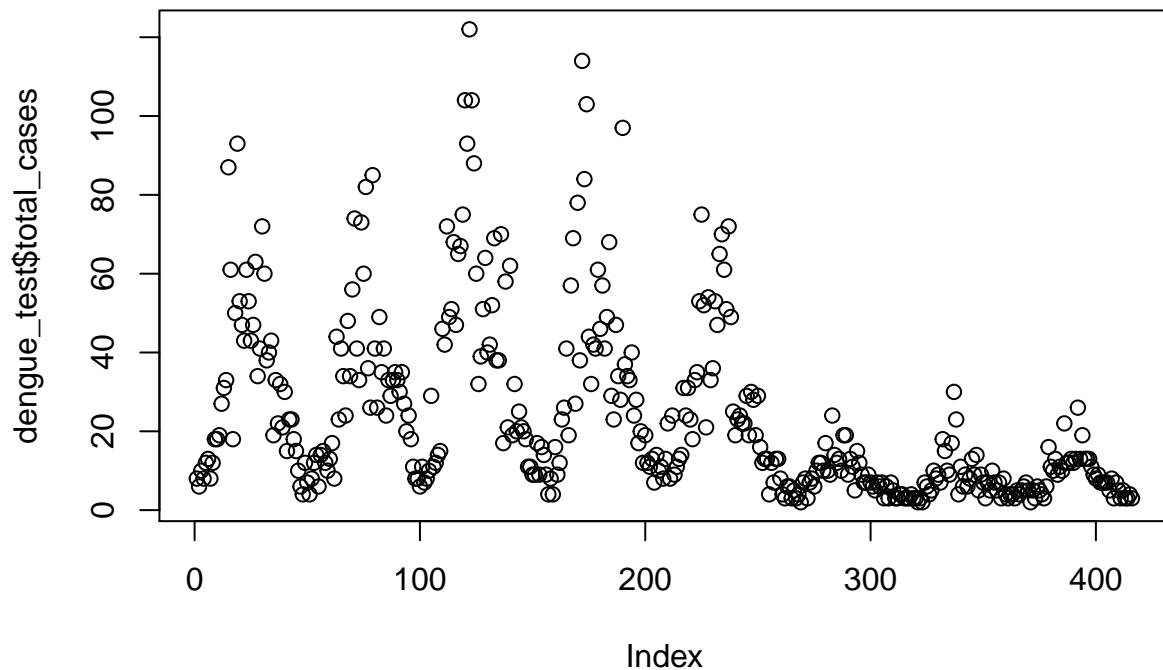
vis_miss(dengue_test)
```



## Predicting total cases on test data

```
# predict values for test data
pred <- predict(model_xgb, dengue_test)
dengue_test$total_cases <- round(pred, digits = 0)

# Visualizing the time-series total cases on the test data
plot(dengue_test$total_cases)
```



```
# Summary of the predicted total cases
```

```
summary(dengue_test$total_cases)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      2.00   8.00   14.00   23.74   33.00   122.00
```

```
#Entering the predicted 'total_cases' from the test-set into the submission form
```

```
Submitformat <- getURL("https://s3.amazonaws.com/drivendata/data/44/public/submission_format.csv")
```

```
submitformat <- read.csv(text=Submitformat)
```

```
submitformat$total_cases<- dengue_test$total_cases
```

```
# Exporting the output (total cases) to local drive as an Excel file
```

```
write.csv(submitformat, "D://STUDY//MSIS//DM//submit0407620xgb_send.csv", row.names = FALSE)
```

# Our Current Ranking in the DengAI Competition at DrivenData.org

Competition: DengAI: Predicting

drivendata.org/competitions/44/dengai-predicting-disease-spread/submissions/

DengAI: Predicting Disease Spread

HOSTED BY DRIVENDATA

GLORY!

2

MONTHS LEFT

Glory!

Submissions

BEST

CURRENT RANK

# COMPETITORS

SUBS. MADE

25.3438

983

8720

1 of 3

SUBMISSION RESTRICTIONS

Competitors are allowed 3 submissions per 1 day.

Your next submission can be on April 6, 2020 UTC.

EVALUATION METRIC

$$MAE = \frac{1}{n} \sum_{i=1}^n |f_i - y_i|$$

The metric used for this competition is mean absolute error. The absolute error is calculated for each label in the submission and then averaged across the labels. For more information on how to calculate MAE, see [wikipedia](#), [sklearn](#) in Python, or the [Metrics](#) package in R. A lower score is better. The goal is to minimize MAE.

SHARE YOUR WORK!

Show off your skills! You can add a link to your work for this competition, whether that is on your

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