

## main.R Markup

This program Performs a Prediction of MKT\_TOTAL\_VAL

First Header	Second Header
File Name	main.R
Author	K. Rajesh Jagannath
Uses	functions_MKT_VAL.R
Uses	Assessors_yearly_data_ETL_MKT_VAL.R
Uses	Buffer_parcel ETL_MKT_VAL.R
Uses	streetcarbuffer_parcelsgis_shape_files/StreetCar
Final Springboard Project submitted	4/30/2017

```
# Libraries used
```

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 3.2.5
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## 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 3.2.5
```

```
library(ggplot2)
```

```
library(readr)
```

```
library(stringr)
```

```
library(scales)
```

```
##
```

```
## Attaching package: 'scales'
```

```
## The following objects are masked from 'package:readr':
```

```
##
```

```
## col_factor, col_numeric
```

```
library(mixtools)
```

```
## mixtools package, version 1.0.4, Released 2016-01-11
```

```
## This package is based upon work supported by the National Science Foundation under Grant No. SES-051
```

```
library(readxl)
```

```
library(lubridate)
```

```
## Warning: package 'lubridate' was built under R version 3.2.5
```

```
##
```

```
## Attaching package: 'lubridate'
```

```

## The following object is masked from 'package:base':
##
##      date
library(reshape2)

##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
##      smiths
library(forecast)

## Warning: package 'forecast' was built under R version 3.2.5
## Loading required package: zoo
## Warning: package 'zoo' was built under R version 3.2.5
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric
## Loading required package: timeDate
## This is forecast 7.2
library(TSclust)

## Loading required package: wmtsa
## Warning: package 'wmtsa' was built under R version 3.2.5
## Loading required package: pdc
## Loading required package: cluster
## Warning: package 'cluster' was built under R version 3.2.5
library(sqldf)

## Loading required package: gsubfn
## Loading required package: proto
## Loading required package: RSQLite
## Loading required package: DBI
## Warning: package 'DBI' was built under R version 3.2.5
library(ClustOfVar)
library(digest)
library(zoo)
library(maptools)

## Loading required package: sp
## Warning: package 'sp' was built under R version 3.2.5

```

```

## Checking rgeos availability: FALSE
##      Note: when rgeos is not available, polygon geometry      computations in maptools depend on gpcl
##      which has a restricted licence. It is disabled by default;
##      to enable gpclib, type gpclibPermit()

library(rgdal)

## Warning: package 'rgdal' was built under R version 3.2.5
## rgdal: version: 1.1-10, (SVN revision 622)
##  Geospatial Data Abstraction Library extensions to R successfully loaded
##  Loaded GDAL runtime: GDAL 1.11.4, released 2016/01/25
##  Path to GDAL shared files: /Library/Frameworks/R.framework/Versions/3.2/Resources/library/rgdal/gdal
##  Loaded PROJ.4 runtime: Rel. 4.9.1, 04 March 2015, [PJ_VERSION: 491]
##  Path to PROJ.4 shared files: /Library/Frameworks/R.framework/Versions/3.2/Resources/library/rgdal/proj
##  Linking to sp version: 1.2-3

# set the seed for repeatability

set.seed(100)

options(scipen=12)

# Set variables

#
# Analysis is done on non_buffer_zone as well as CORE, CENTER and EDGE buffer zones
# set skip_non_buffer = 1 to skip directly to CORE, CENTER and EDGE analysis
# used for bring up and debug only
skip_non_buffer = 0
#
# Choose a smaller sample for non buffer zone : skip_1500 = 1 for dev
# Choose about 1500 samples don't skip : for analysis
# used for debug and bring up only
skip_1500 = 0

print(" Setting current working directory ")

## [1] " Setting current working directory "

# set the working dir
setwd("/Users/rajesh/Desktop/Coursera/SpringBoardGithub/Streetcar0719")

print(" sourcing functions ")

## [1] " sourcing functions "

setwd("/Users/rajesh/Desktop/Coursera/SpringBoardGithub/Streetcar0719/taxinfoTimeSeries")
source ('/Users/rajesh/Desktop/Coursera/SpringBoardGithub/Streetcar0719/taxinfoTimeSeries/functions_MKT.R')

setwd("/Users/rajesh/Desktop/Coursera/SpringBoardGithub/Streetcar0719")

print("Reading in 8 years of parcel data 1/4 Million observations X 8 years")

```

```

## [1] "Reading in 8 years of parcel data 1/4 Million observations X 8 years"
print(" Performing Extraction, Transformation and Loading ... ")

## [1] " Performing Extraction, Transformation and Loading ... "
# Here we read in 8 years data for 250,000 - 275,000 parcels for all of Cincinnati
# Extraction Transformation and Cleaning is done here
source('taxinfoTimeSeries/Assessors_yearly_data_ETL_MKT_VAL.R')

## [1] "Processing the year : 2007"
## [1] "df_taxinfo_2007"
## [1] "Processing the year : 2008"
## [1] "df_taxinfo_2008"
## [1] "Processing the year : 2009"
## [1] "df_taxinfo_2009"
## [1] "Processing the year : 2010"
## [1] "df_taxinfo_2010"
## [1] "Processing the year : 2011"
## [1] "df_taxinfo_2011"
## [1] "Processing the year : 2012"
## [1] "df_taxinfo_2012"
## [1] "Processing the year : 2013"
## [1] "df_taxinfo_2013"
## [1] "Processing the year : 2014"
## [1] "df_taxinfo_2014"
## [1] "Processing the year : 2015"
## [1] "df_taxinfo_2015"
## [1] "processing df_taxinfo_2007$MKT_LAND_VAL"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$MKT_IMPR_VAL"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$MKT_TOTAL_VAL"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$ANNUAL_TAXES"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$TAXES_PAID"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$DELQ_TAXES"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$ACRES"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$SALE_AMOUNT"

## Warning: NAs introduced by coercion

## [1] "processing df_taxinfo_2007$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2007$FORECL_FLAG"

```

```

## [1] "processing df_taxinfo_2008$MKT_LAND_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2008$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2008$FORECL_FLAG"
## [1] "processing df_taxinfo_2009$MKT_LAND_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2009$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2009$FORECL_FLAG"
## [1] "processing df_taxinfo_2010$MKT_LAND_VAL"

```

```

## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2010$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2010$FORECL_FLAG"
## [1] "processing df_taxinfo_2011$MKT_LAND_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2011$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2011$FORECL_FLAG"
## [1] "processing df_taxinfo_2012$MKT_LAND_VAL"
## Warning: NAs introduced by coercion

```

```

## [1] "processing df_taxinfo_2012$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2012$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2012$FORECL_FLAG"
## [1] "processing df_taxinfo_2013$MKT_LAND_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2013$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2013$FORECL_FLAG"
## [1] "processing df_taxinfo_2014$MKT_LAND_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$MKT_IMPR_VAL"

```

```

## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2014$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2014$FORECL_FLAG"
## [1] "processing df_taxinfo_2015$MKT_LAND_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$MKT_IMPR_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$MKT_TOTAL_VAL"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$ANNUAL_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$TAXES_PAID"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$DELQ_TAXES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$ACRES"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$SALE_AMOUNT"
## Warning: NAs introduced by coercion
## [1] "processing df_taxinfo_2015$NEW_CONS_FLAG"
## [1] "processing df_taxinfo_2015$FORECL_FLAG"
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion

```



```

## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: NAs introduced by coercion
## Warning: Setting row names on a tibble is deprecated.
print( "Reading Buffer zone Parcel information ")

## [1] "Reading Buffer zone Parcel information "
print(" Performing Extraction, Transformation and Loading ... ")

## [1] " Performing Extraction, Transformation and Loading ... "
# here we read in and clean up the Parcels of the Buffer
# extraction , Transormaton and Cleaning is done in this file
source('taxinfoTimeSeries/Buffer_parcel_ETL_MKT_VAL.R')

## [1] "Processing the zone : CENTER"
## [1] "df_buffer_CENTER"
## [1] "Processing the zone : EDGE"
## [1] "df_buffer_EDGE"
## [1] "Processing the zone : CORE"
## [1] "df_buffer_CORE"

setwd("/Users/rajesh/Desktop/Coursera/SpringBoardGithub/Streetcar0719")

if(skip_non_buffer == 0) {

  print(" NON BUFFER ZONE : Hierarchical Clustering ")
  # Begin Hierarchical Clustering of parcel ids

  # For illustartive purposes we choose first 10 columns
  Clustvar <-hclustvar(taxinfo.matrix_no_na[, 1:10])
  plot(Clustvar)

  print(" NON BUFFER ZONE : 100 random samples Hierarchical Clustering ")

  #then random 100 columns
  n <- ncol(taxinfo.matrix_no_na)
  taxinfo.matrix_shuffled <- taxinfo.matrix_no_na[,sample(n) ]
  #0.01248 is 2500 values

```

```

train_indices <- 1:round(0.00042226 * n)
train <- taxinfo.matrix_shuffled[,train_indices ]

Clustvar100 <-hclustvar(train)
plot(Clustvar100)
rect.hclust(Clustvar100,k=12, border="red")
part100 <- cutreevar(Clustvar100,12)

plot(Clustvar100$height)

print(" NON BUFFER ZONE : 100 random samples  Stability graph ")

#stab100 <- stability(Clustvar100,B = 60, graph = TRUE)

if(skip_1500 == 0) {
  #Now let's get some random PARCEL_IDs
  # About 1500 should do

  n <- ncol(taxinfo.matrix_no_na)
  taxinfo.matrix_shuffled <- taxinfo.matrix_no_na[,sample(n) ]
  #0.0051699 of 241748 observations is 1249
  train_indices <- 1:round(0.0051699 * n)
  train <- taxinfo.matrix_shuffled[,train_indices ]

  Clustvar1500 <-hclustvar(train)
  plot(Clustvar1500)
  rect.hclust(Clustvar1500,k=12, border="red")

  #####
  # B is the Boot strapping sample and we determine the stability
  # Evaluates the stability of partitions obtained from a hierarchy of p variables. This hierarchy is
  # with hclustvar and the stability of the partitions of 2 to p-1 clusters is evaluated with a
  # bootstrap approach. The bootstrap approach is the following: hclustvar is applied to B bootstrap
  # samples of the n rows. The partitions of 2 to p-1 clusters obtained from the B bootstrap hierarch
  # are compared with the partitions from the initial hierarchy . The mean of the corrected Rand
  # indices is plotted according to the number of clusters. This graphical representation helps in th
  # determination of a suitable numbers of clusters.

  #stab1500 <- stability(Clustvar1500,B = 5, graph = TRUE)

  # Could not get a graph : Make a rough partition of 12 clusters

  part1500 <- cutreevar(Clustvar1500,12)
  plot(Clustvar1500$height)

  a <- part1500$var

```

```

} else {
  part100 <- cutreevar(Clustvar100,12)
  plot(Clustvar100$height)

  a <- part100$var
}
#Remove xs from Parcel ids

train_cl1 <- as.data.frame(row.names(a$cluster1), stringsAsFactors = FALSE)
train_cl2 <- as.data.frame(row.names(a$cluster2), stringsAsFactors = FALSE)
train_cl3 <- as.data.frame(row.names(a$cluster3), stringsAsFactors = FALSE)
train_cl4 <- as.data.frame(row.names(a$cluster4), stringsAsFactors = FALSE)
train_cl5 <- as.data.frame(row.names(a$cluster5), stringsAsFactors = FALSE)
train_cl6 <- as.data.frame(row.names(a$cluster6), stringsAsFactors = FALSE)
train_cl7 <- as.data.frame(row.names(a$cluster7), stringsAsFactors = FALSE)
train_cl8 <- as.data.frame(row.names(a$cluster8), stringsAsFactors = FALSE)
train_cl9 <- as.data.frame(row.names(a$cluster9), stringsAsFactors = FALSE)
train_cl10 <- as.data.frame(row.names(a$cluster10), stringsAsFactors = FALSE)
train_cl11 <- as.data.frame(row.names(a$cluster11), stringsAsFactors = FALSE)
train_cl12 <- as.data.frame(row.names(a$cluster12), stringsAsFactors = FALSE)

# Remove X's from the PARCEL IDs
train_cl1 <- apply(train_cl1,2, function(y) as.character(gsub("X", "", y)))
train_cl2 <- apply(train_cl2,2, function(y) as.character(gsub("X", "", y)))
train_cl3 <- apply(train_cl3,2, function(y) as.character(gsub("X", "", y)))
train_cl4 <- apply(train_cl4,2, function(y) as.character(gsub("X", "", y)))
train_cl5 <- apply(train_cl5,2, function(y) as.character(gsub("X", "", y)))
train_cl6 <- apply(train_cl6,2, function(y) as.character(gsub("X", "", y)))
train_cl7 <- apply(train_cl7,2, function(y) as.character(gsub("X", "", y)))
train_cl8 <- apply(train_cl8,2, function(y) as.character(gsub("X", "", y)))
train_cl9 <- apply(train_cl9,2, function(y) as.character(gsub("X", "", y)))
train_cl10 <- apply(train_cl10,2, function(y) as.character(gsub("X", "", y)))
train_cl11 <- apply(train_cl11,2, function(y) as.character(gsub("X", "", y)))
train_cl12 <- apply(train_cl12,2, function(y) as.character(gsub("X", "", y)))

# Use Holt-Winters for Simple Exponential Smoothing
for ( clst in c(seq(1:12))) {
  forecast_clusters(clst,a, taxinfo.matrix_uniq)
}

# we create 5 models of regressions
# 1st order - a + T ( linear)
# 2nd order a + bT + cT^2 ( quadratic)

```

```

# 3rd order  $a = bT + cT^2 + dT^3$  ( cubic)
# 4th order and
# 5th order as well

print("NON BUFFER ZONE : Plotting Actuals vs. Fitted for models in clusters")
print("NON BUFFER ZONE : Evaluating 12 Clusters with 1st, 2nd, 3rd and 4th and 5th order polynomial r")
print("NON BUFFER ZONE : Years 2007-2014 serve as training data")
print("NON BUFFER ZONE : Years 2015-2018 serve as test data for validation")

for( poly_order in c(seq(1:5))) {

  for ( clst in c(seq(1:12))) {

    df_mape_train_clusters <- plot_act_pred_trg(clst,a, taxinfo.matrix[c(1:6),], poly_order)
    df_mape_test_clusters <- plot_act_pred_test(clst,a, taxinfo.matrix[c(1:6),], taxinfo.matrix[c(7:12),])

    if (clst == 1) {
      # df_mape_train_clst_poly <- df_mape_train_clusters["mape"]
      df_actpred_train_clst_poly <- df_mape_train_clusters["act_pred"]
      df_mape_test_clst_poly <- df_mape_test_clusters["mape"]

    }
    else {

      #df_mape_train_clst_poly <- bind_rows(df_mape_train_clst_poly, df_mape_train_clusters["mape"])
      df_actpred_train_clst_poly <- bind_rows(df_actpred_train_clst_poly, df_mape_train_clusters["act_pred"])

      df_mape_test_clst_poly <- bind_rows(df_mape_test_clst_poly, df_mape_test_clusters["mape"])
    }
  }

  if (poly_order == 1) {
    #df_mape_train_all <- df_mape_train_clst_poly
    df_actpred_train_all <- df_actpred_train_clst_poly

    df_mape_test_all <- df_mape_test_clst_poly
  }
  else
  {
    #df_mape_train_all <- bind_rows(df_mape_train_all, df_mape_train_clst_poly)
    df_actpred_train_all <- bind_rows(df_actpred_train_all, df_actpred_train_clst_poly)

    df_mape_test_all <- bind_rows(df_mape_test_all, df_mape_test_clst_poly)
  }
}

```

```

}

print("NON BUFFER ZONE : Computing MAPE - Mean Absolute Percentage error")

# Chart the actuals and predicted for the different clusters

print("NON BUFFER ZONE: Plotting the Actual vs. Fitted for Training set ")

for( poly_order in c(seq(1:5))) {

  plot_actpred_chart(df_actpred_train_all, " Non BUFFER ZONE training set ", poly_order)

} # poly_order

# Evaluate the Mean Absolute Percentage error(MAPE)
# Training data
# Years 2007-2012
#plot_mape_chart(df_mape_train_all, "non Buffer zone : training data")

# Evaluate the MAPE for
# Testing data
# years 2013-2014

plot_mape_chart(df_mape_test_all, "non Buffer zone : test data")

print("NON BUFFER ZONE : Plotting MKT_TOT_VAL source and predicted data")
print("NON BUFFER ZONE : Creating df_final_mkt_val data frame for later Visualization")

for( poly_order in c(seq(1:5))) {

  # facet the clusters on a single plot
  # visualize fitted and predicted

  df_final_mkt_val_non_bfr_zone <- non_linear_regression_model(a, taxinfo.matrix, poly_order, 12, "non
  if ( poly_order == 1) {
    df_final_mkt_val <- df_final_mkt_val_non_bfr_zone
  }
  else{ df_final_mkt_val <- bind_rows(df_final_mkt_val, df_final_mkt_val_non_bfr_zone )}

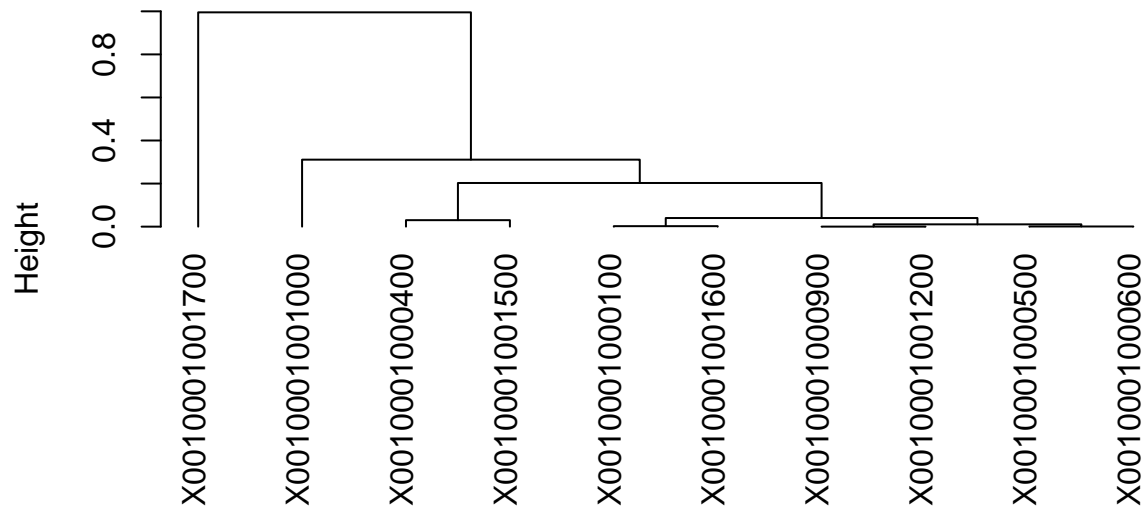
}

} # skip_non_buffer_zone

## [1] " NON BUFFER ZONE : Hierarchical Clustering "

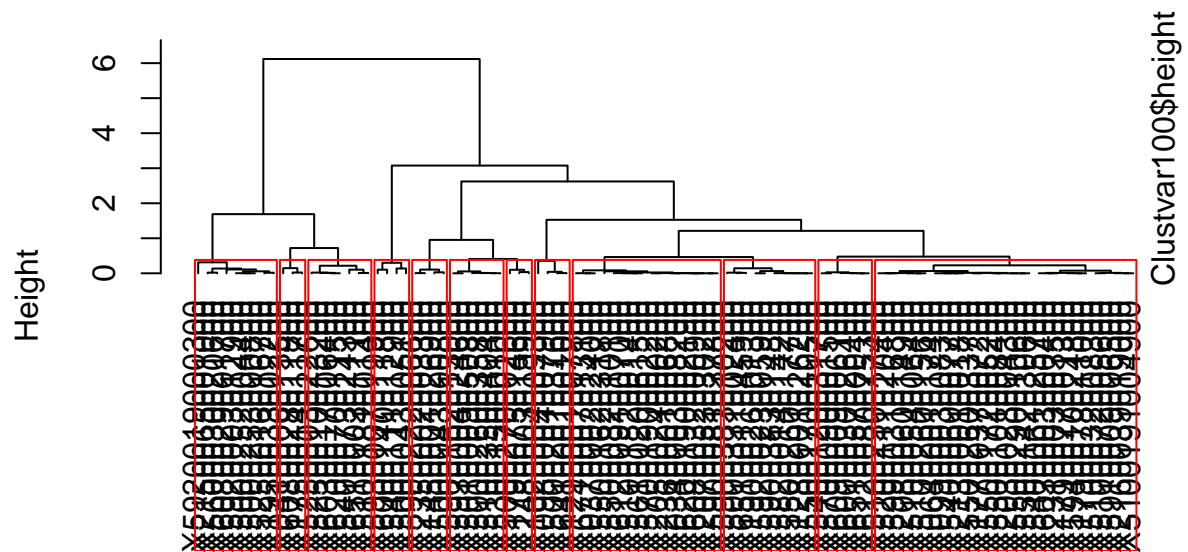
```

### Cluster Dendrogram

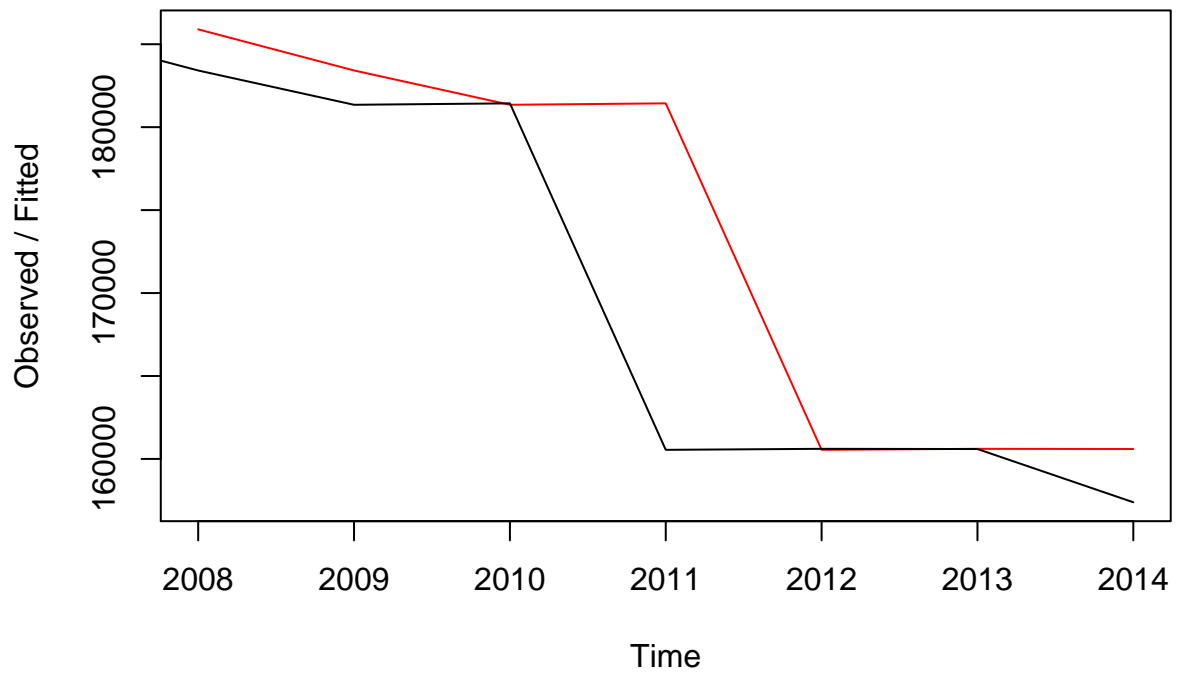
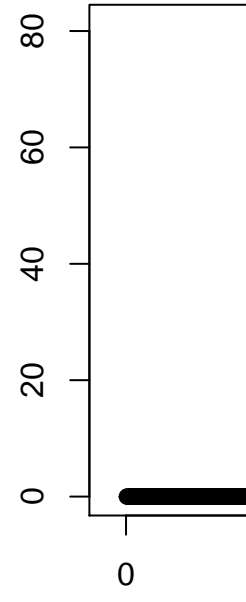
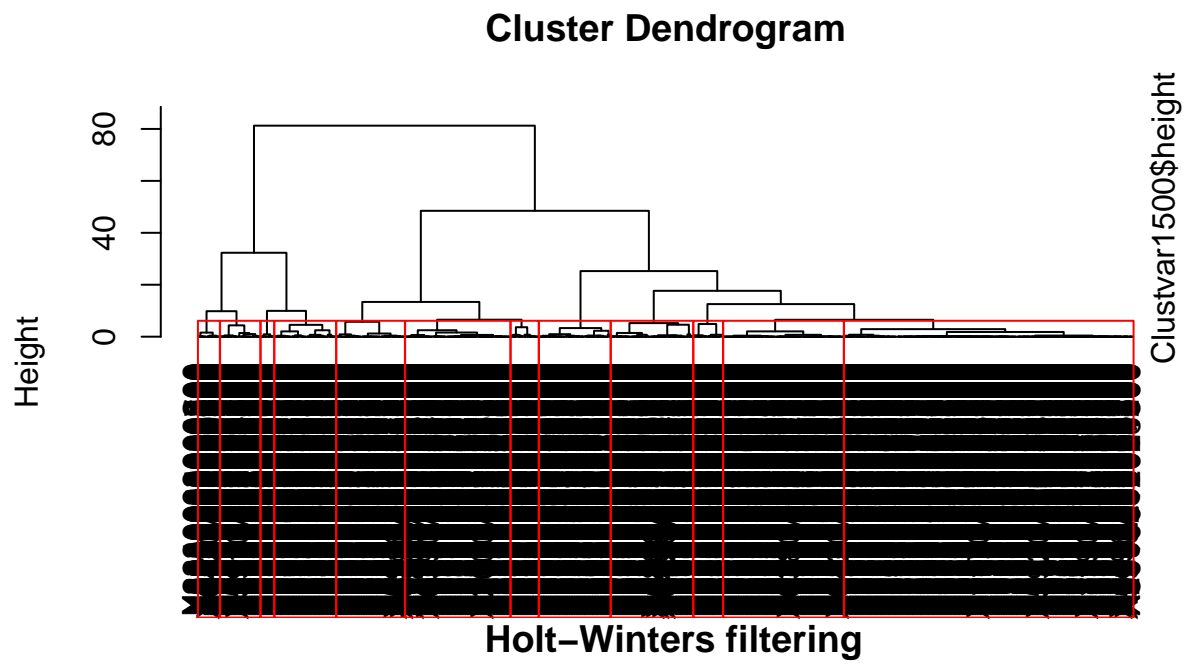


## [1] " NON BUFFER ZONE : 100 random samples Hierarchical Clustering "

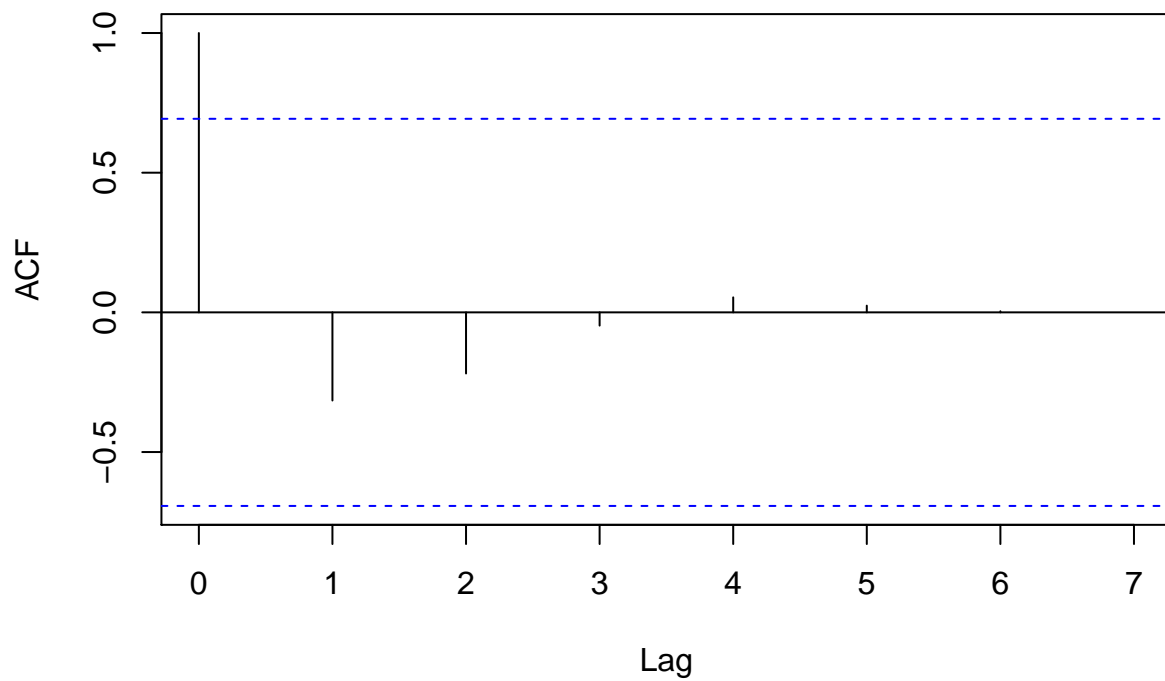
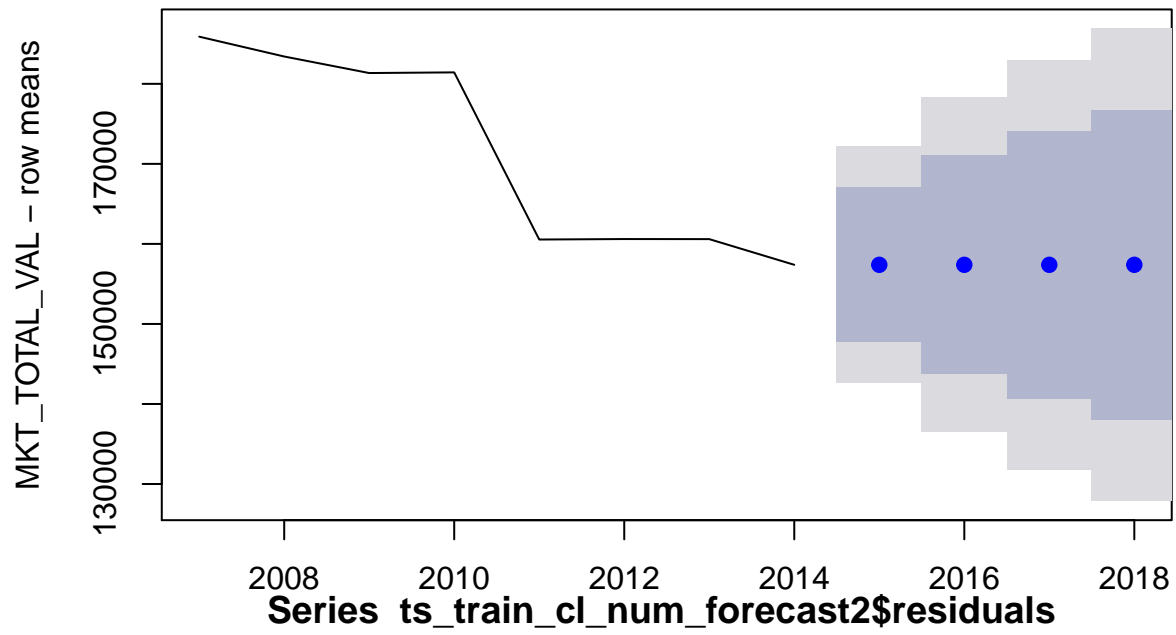
### Cluster Dendrogram



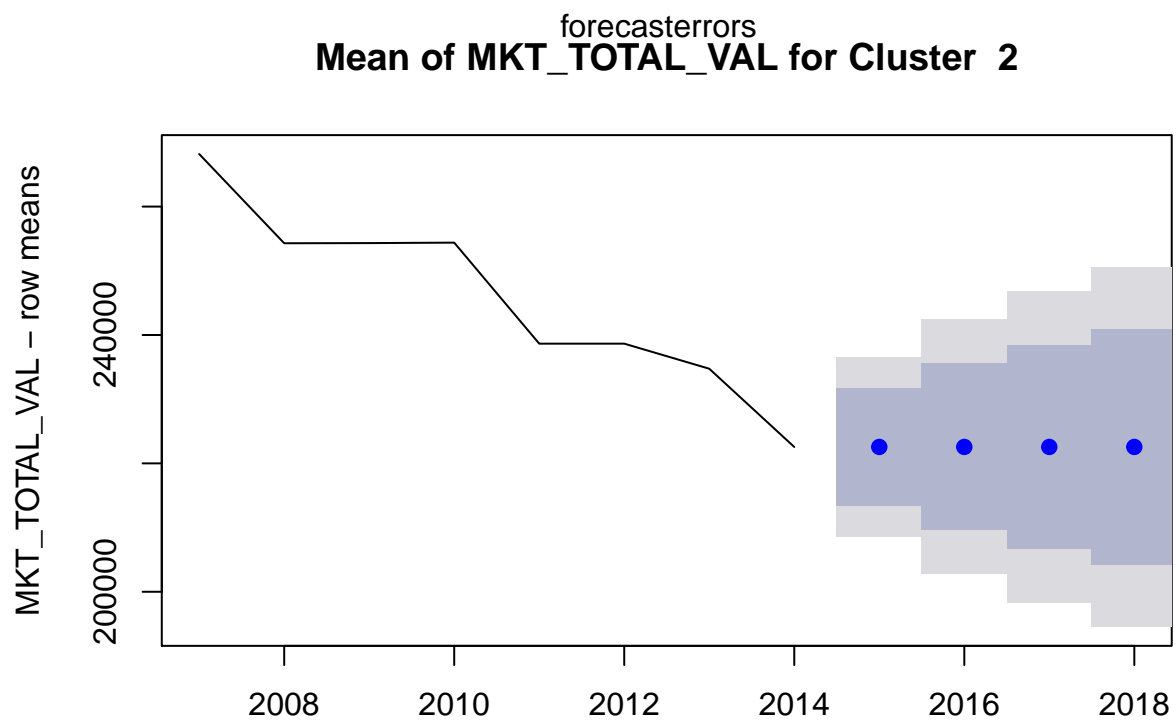
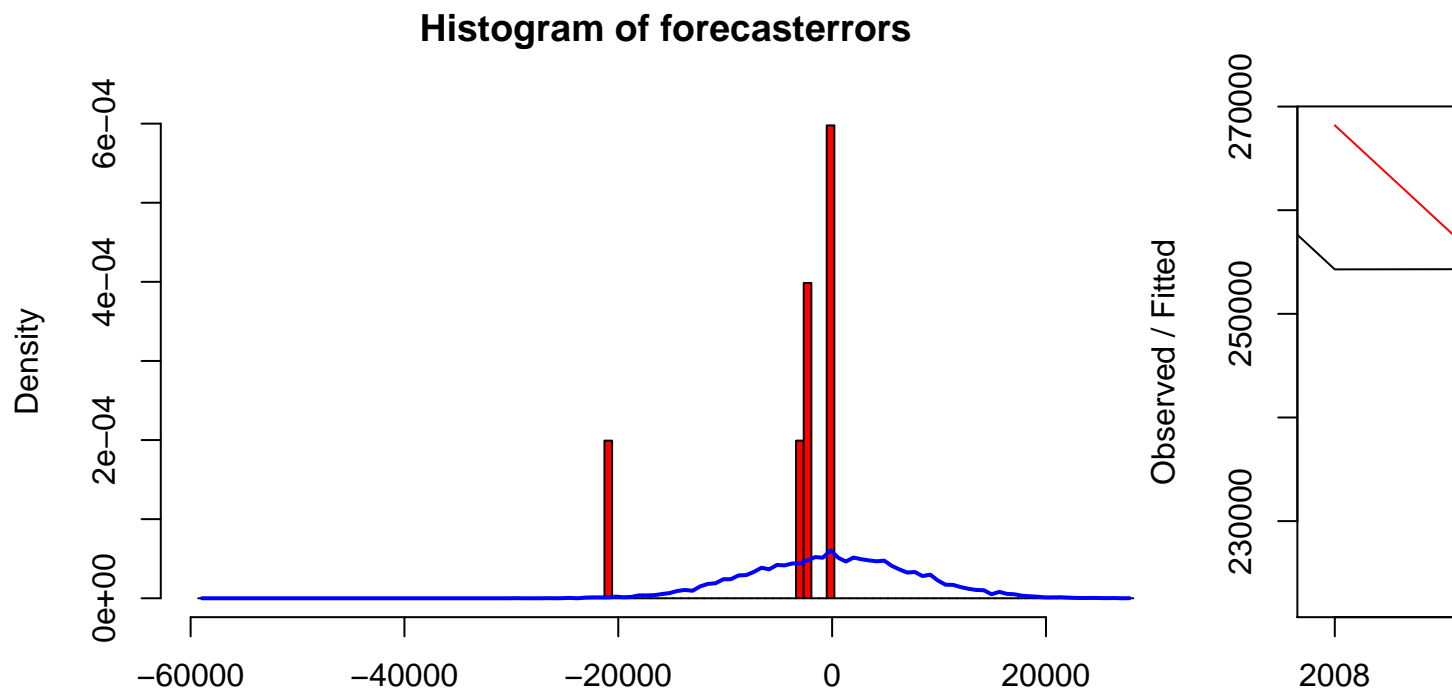
## [1] " NON BUFFER ZONE : 100 random samples Stability graph "



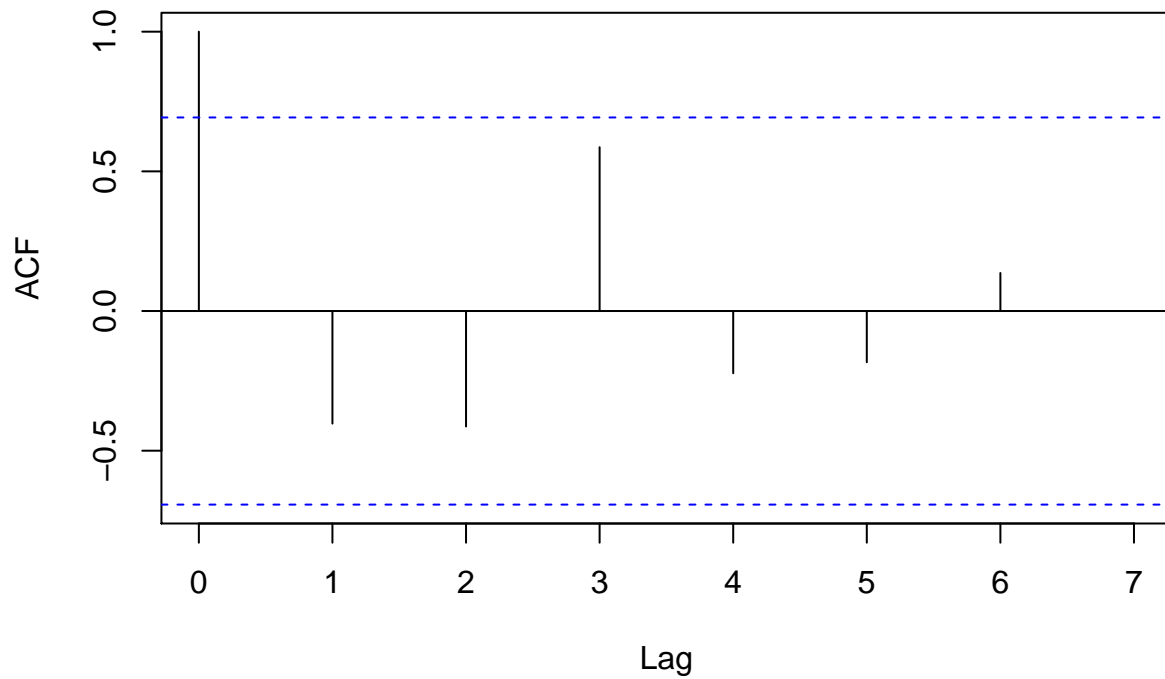
Mean of MKT\_TOTAL\_VAL for Cluster 1



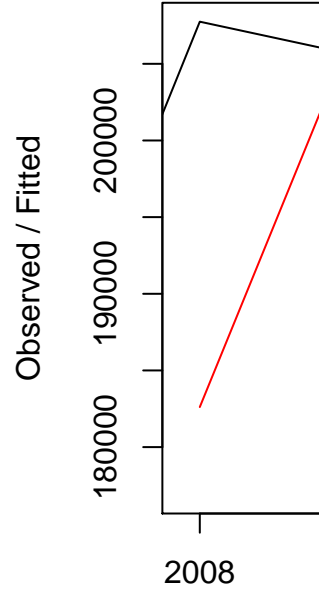
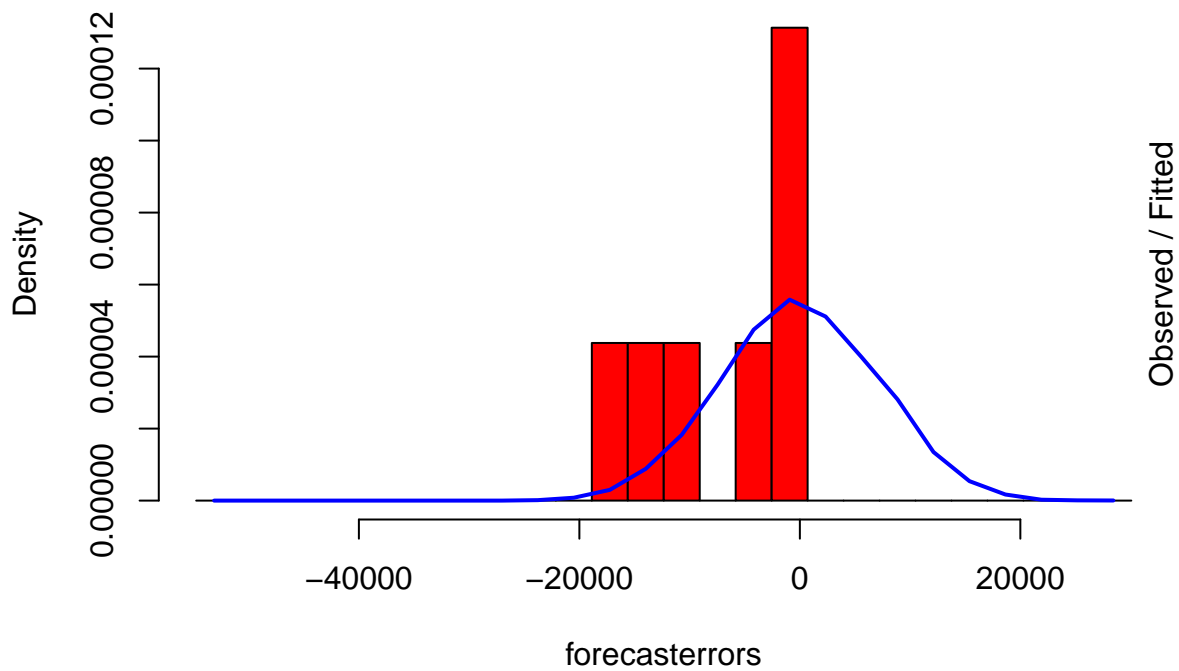


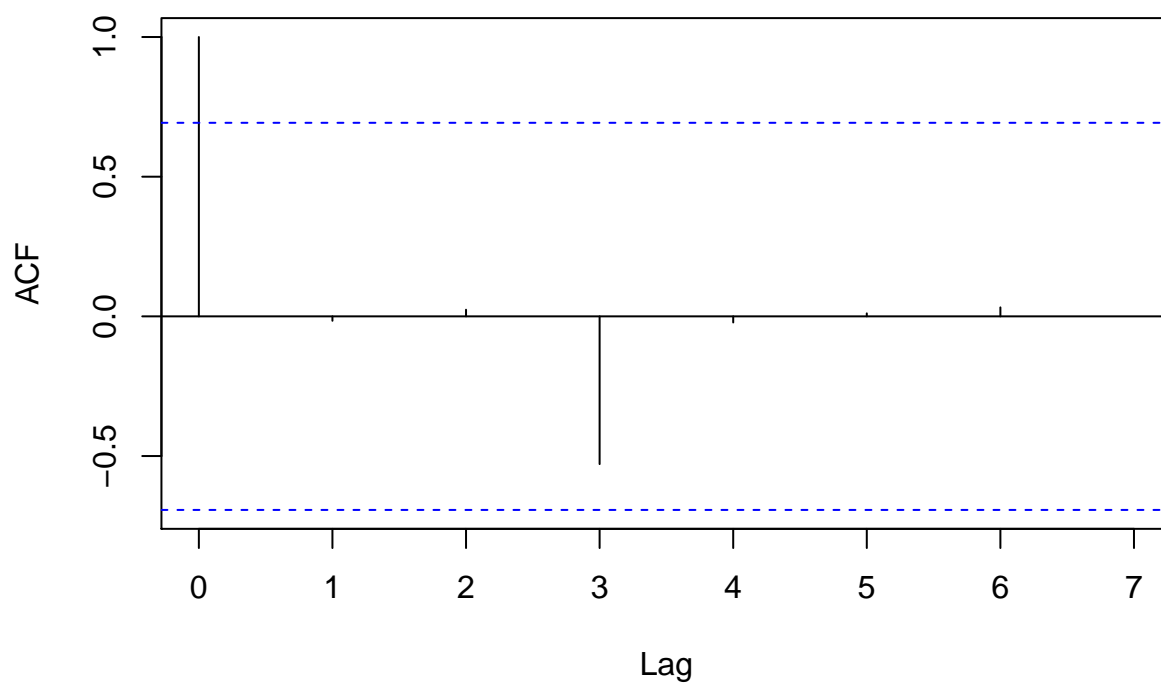
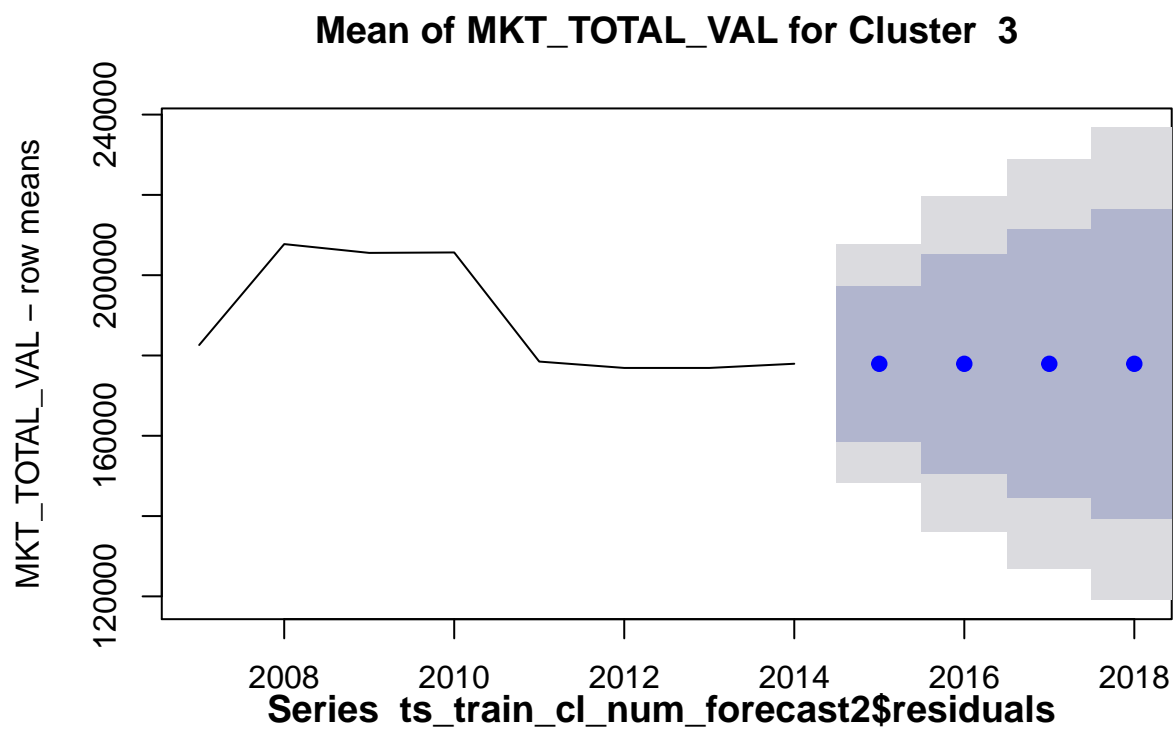


Series ts\_train\_cl\_num\_forecast2\$residuals

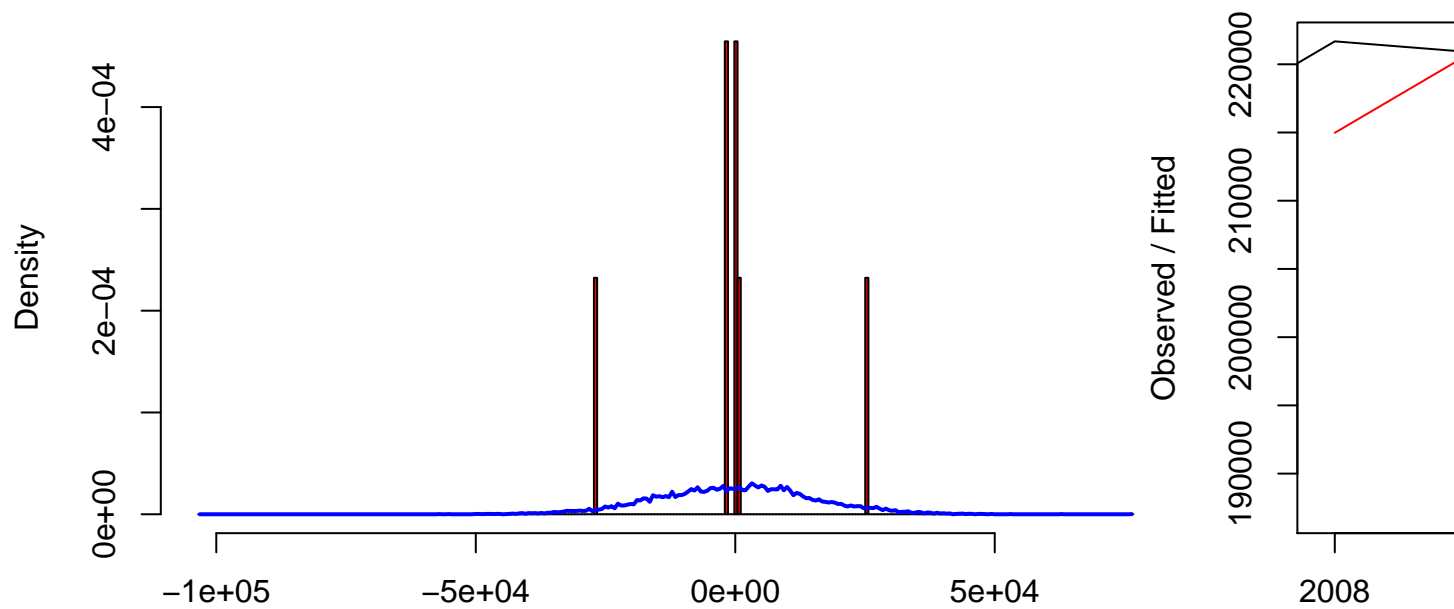


Histogram of forecasterrors

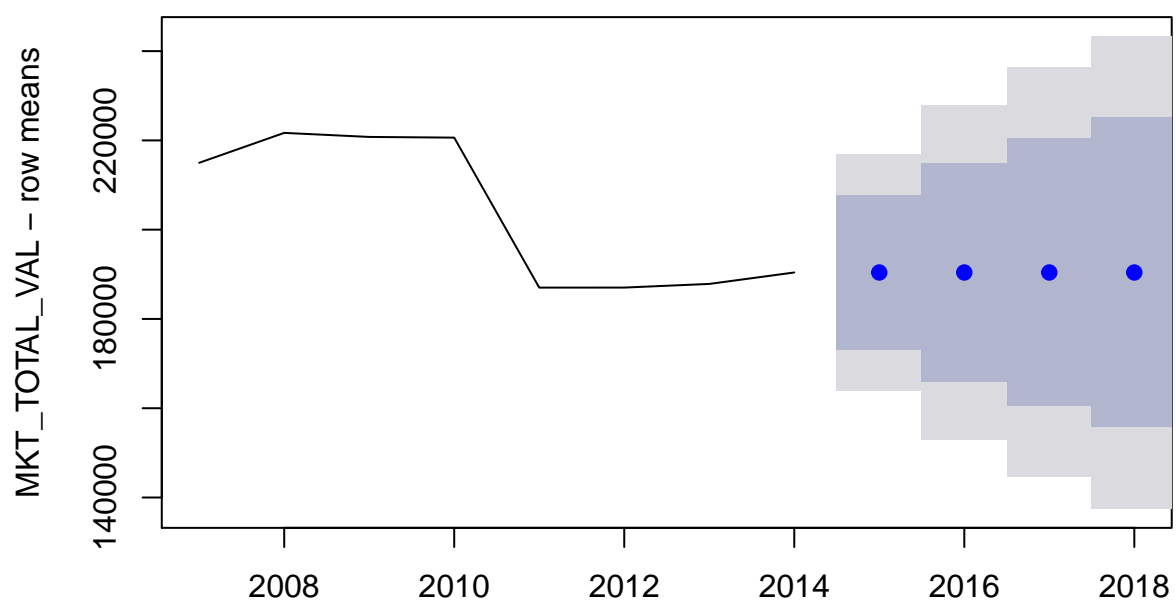




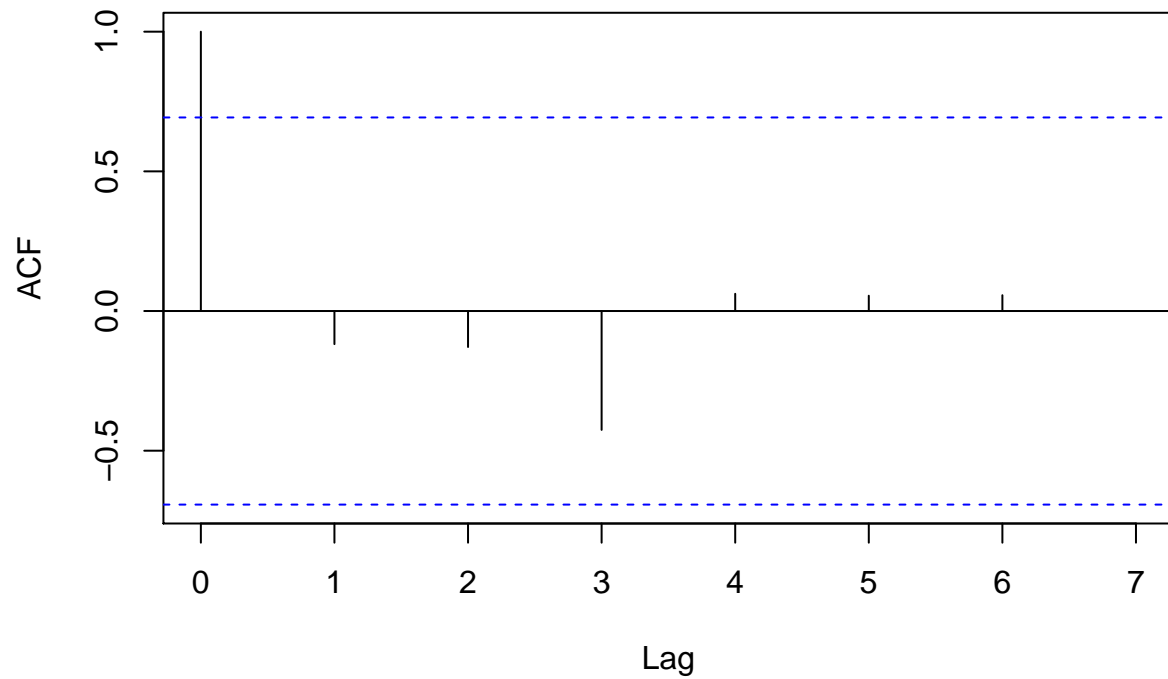
Histogram of forecasterrors



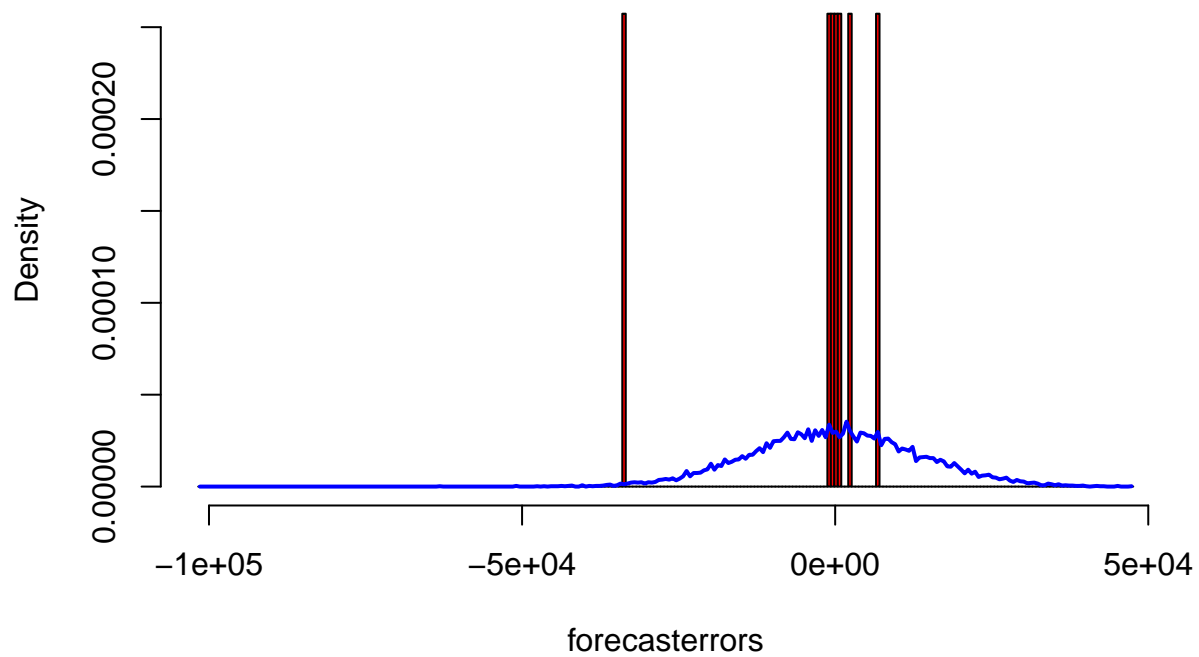
Mean of MKT\_TOTAL\_VAL for Cluster 4



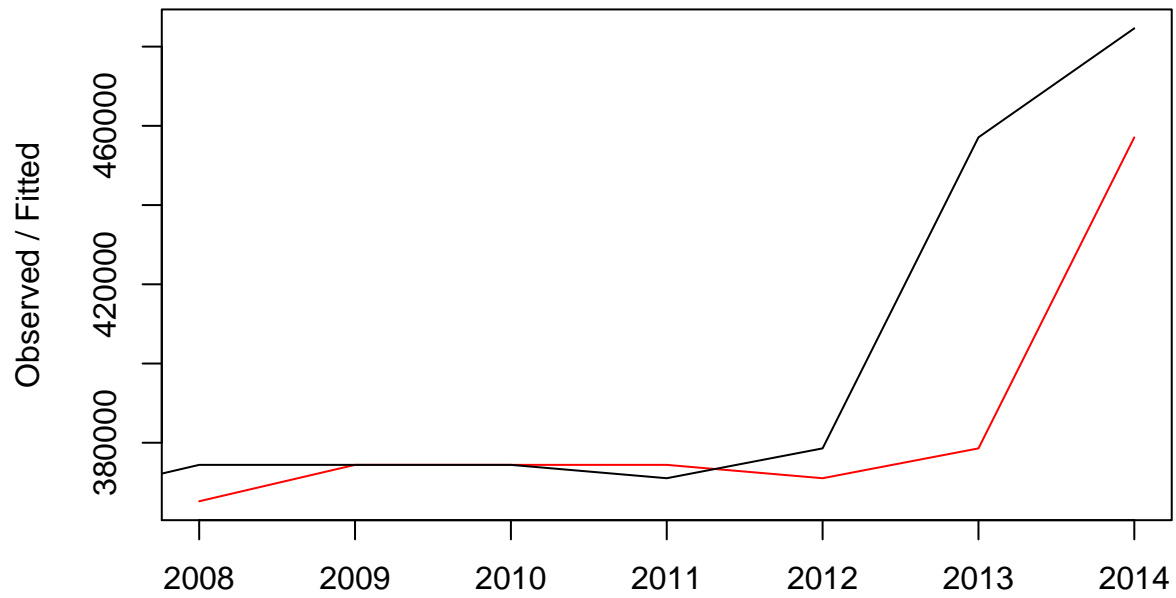
Series ts\_train\_cl\_num\_forecast2\$residuals



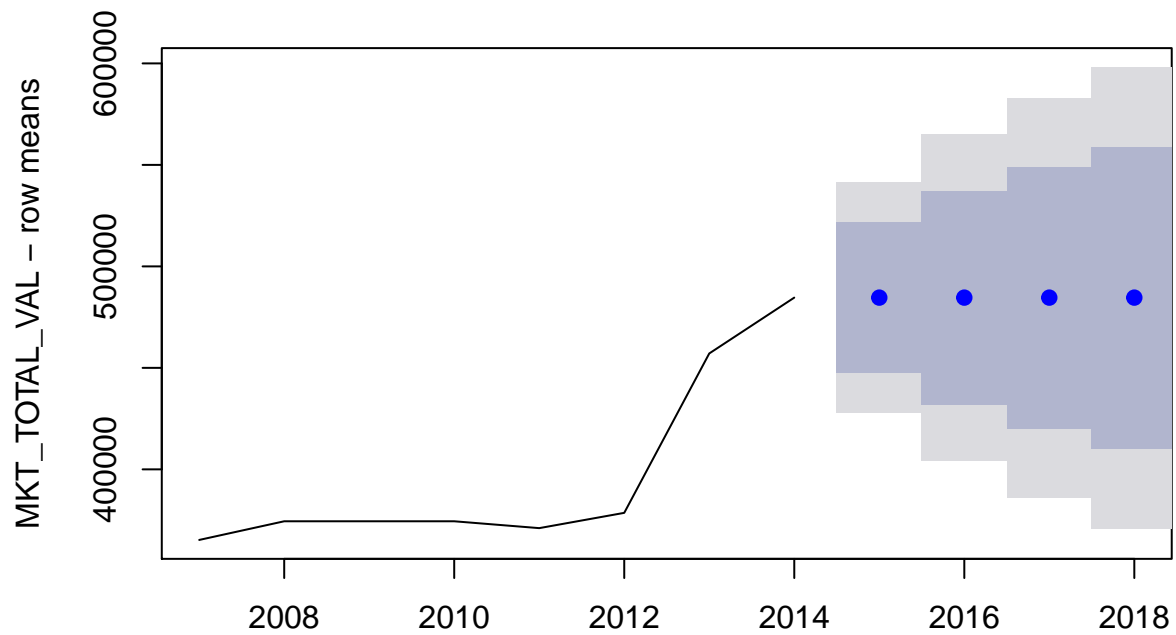
Histogram of forecasterrors



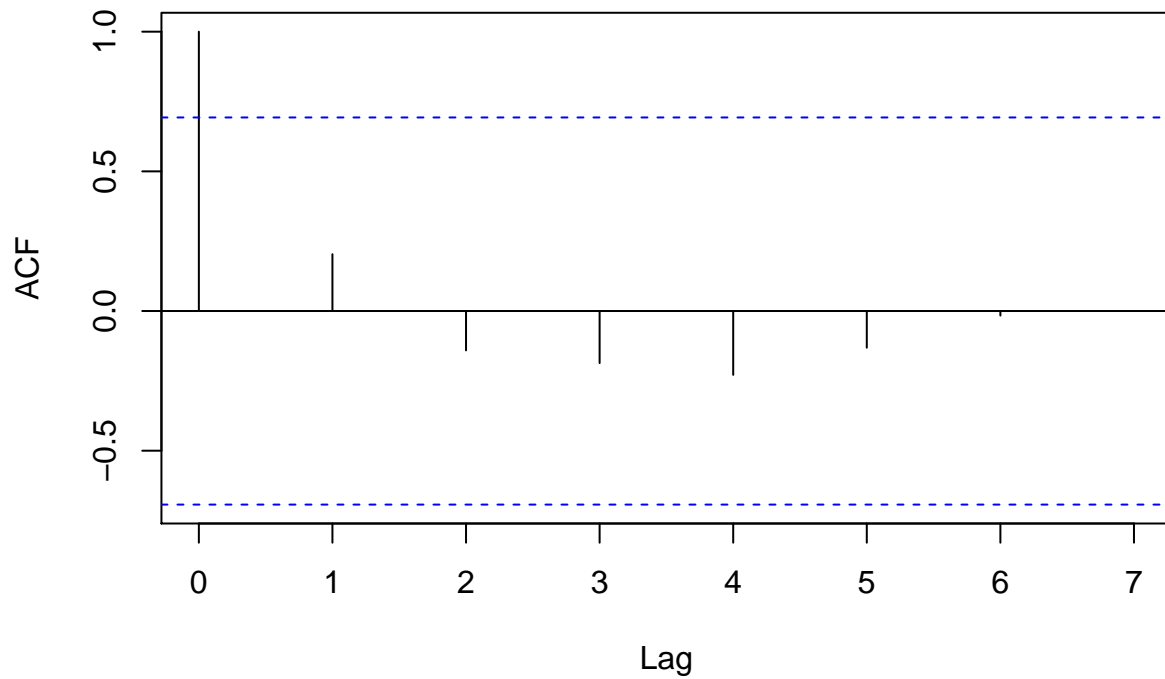
### Holt-Winters filtering



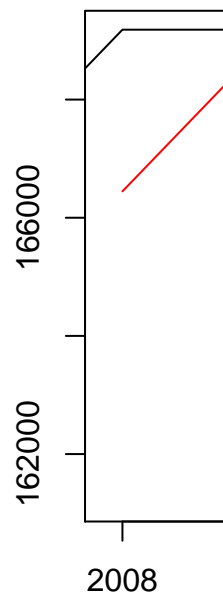
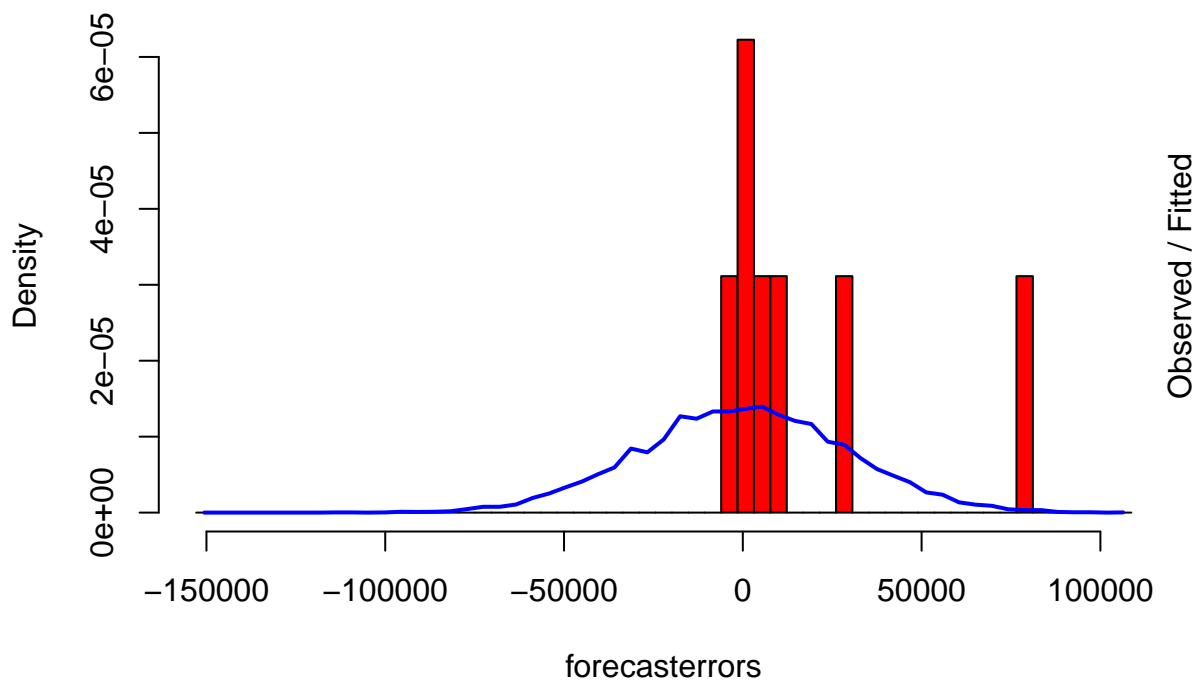
### Mean of MKT\_TOTAL\_VAL for Cluster 5



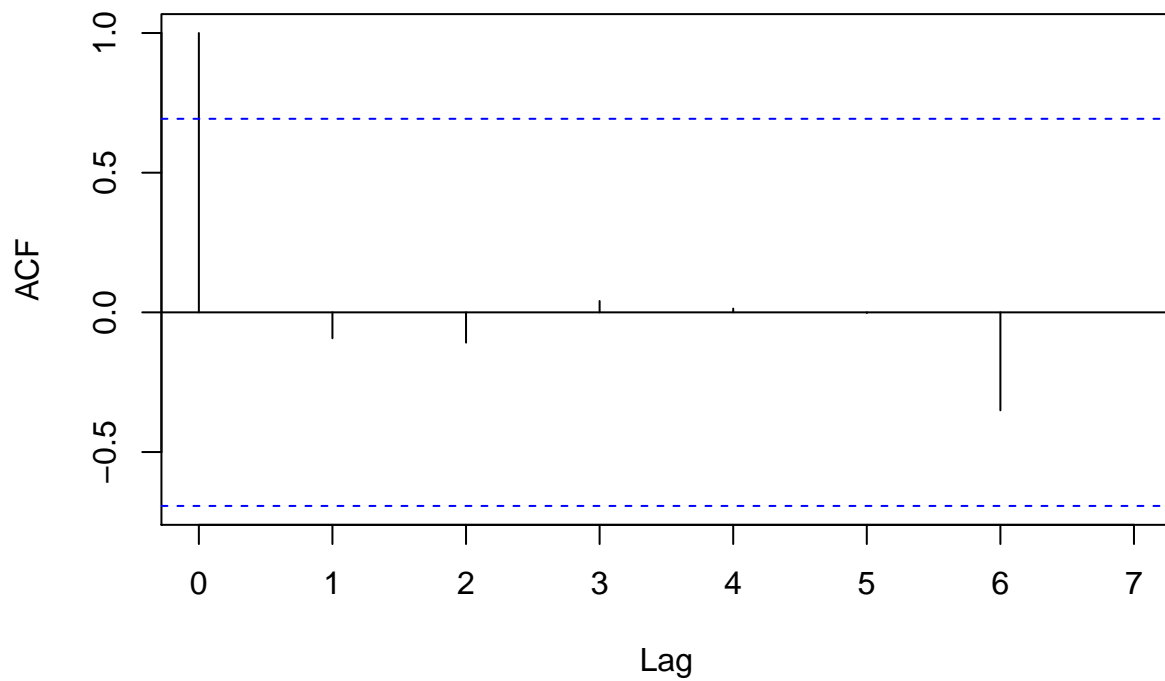
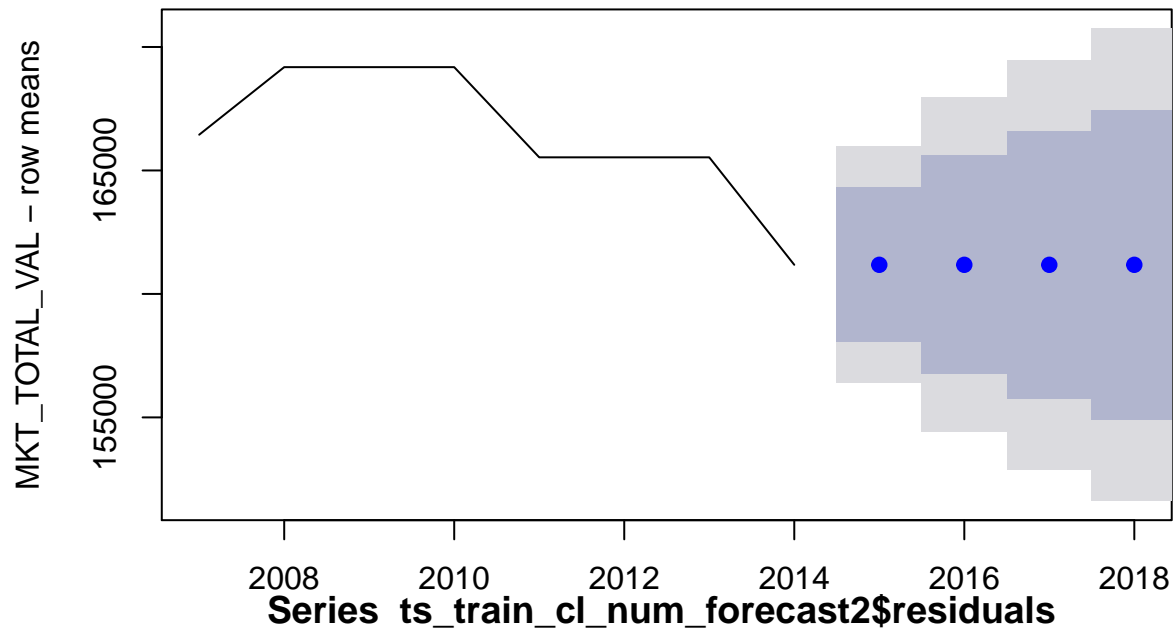
Series ts\_train\_cl\_num\_forecast2\$residuals



Histogram of forecasterrors

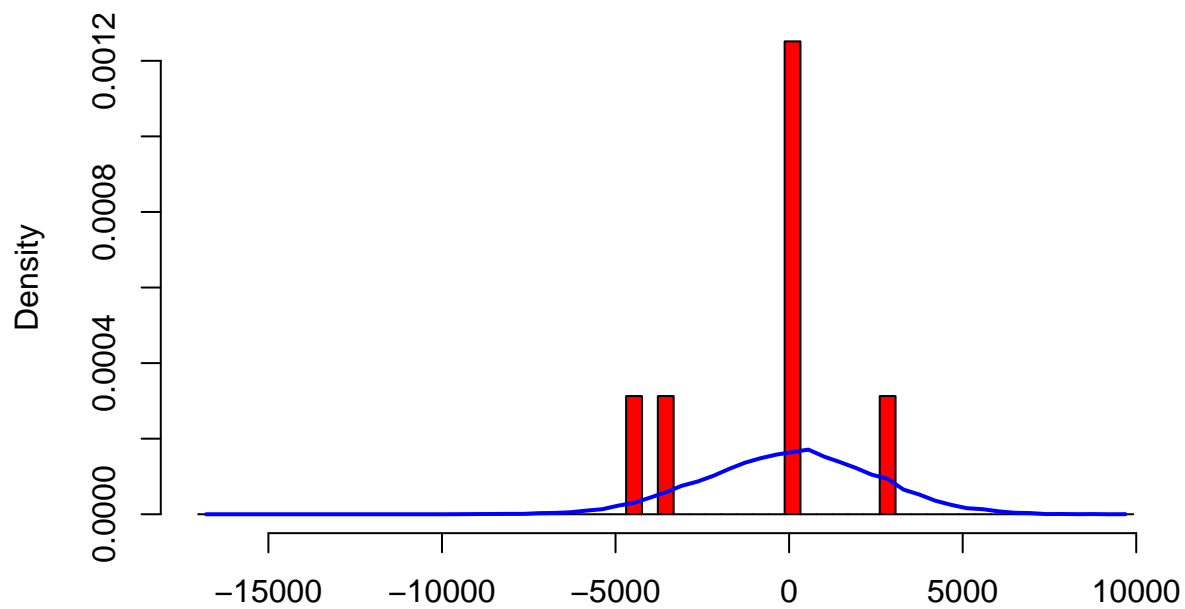


Mean of MKT\_TOTAL\_VAL for Cluster 6

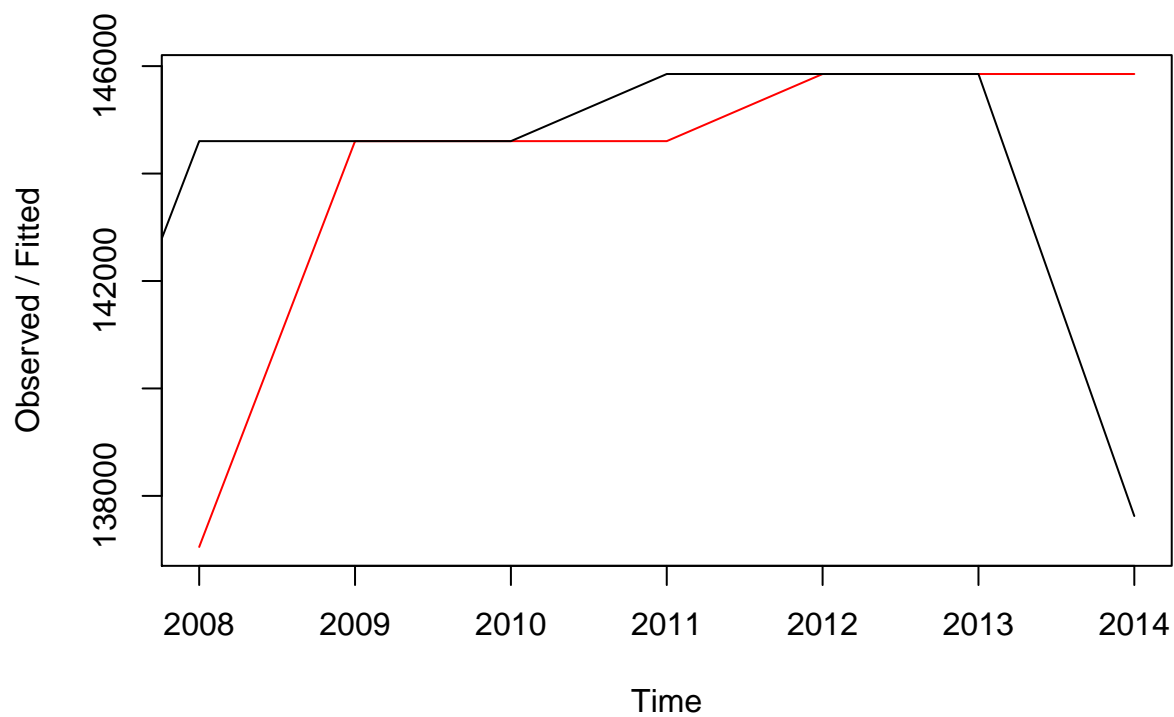




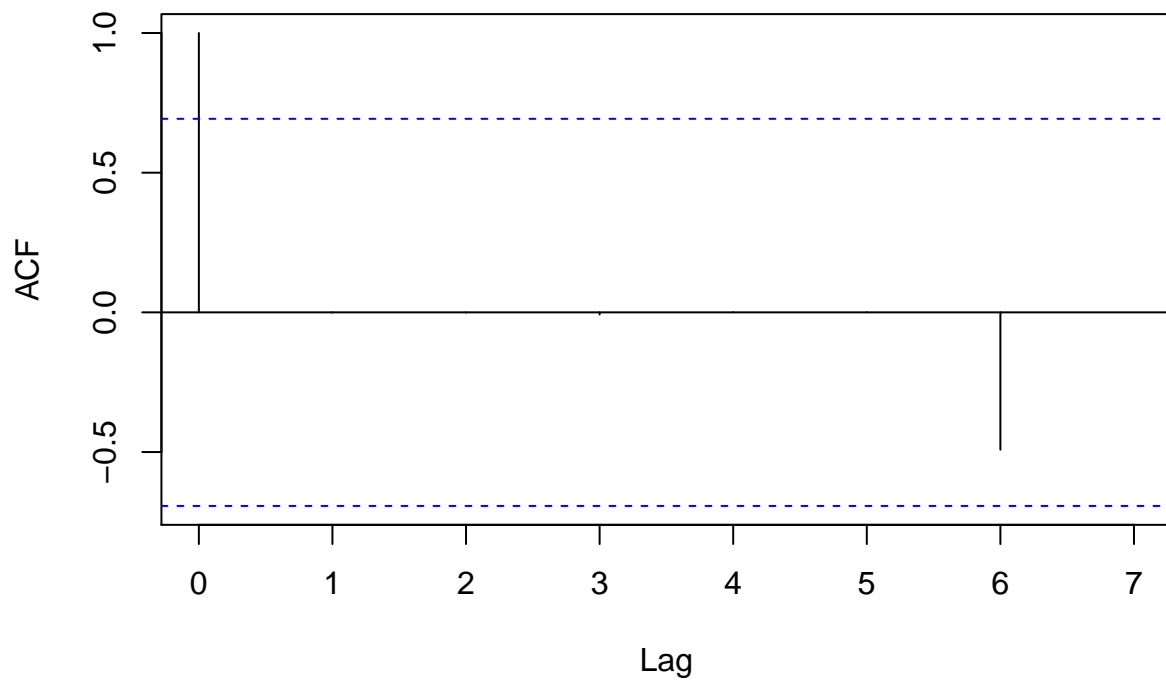
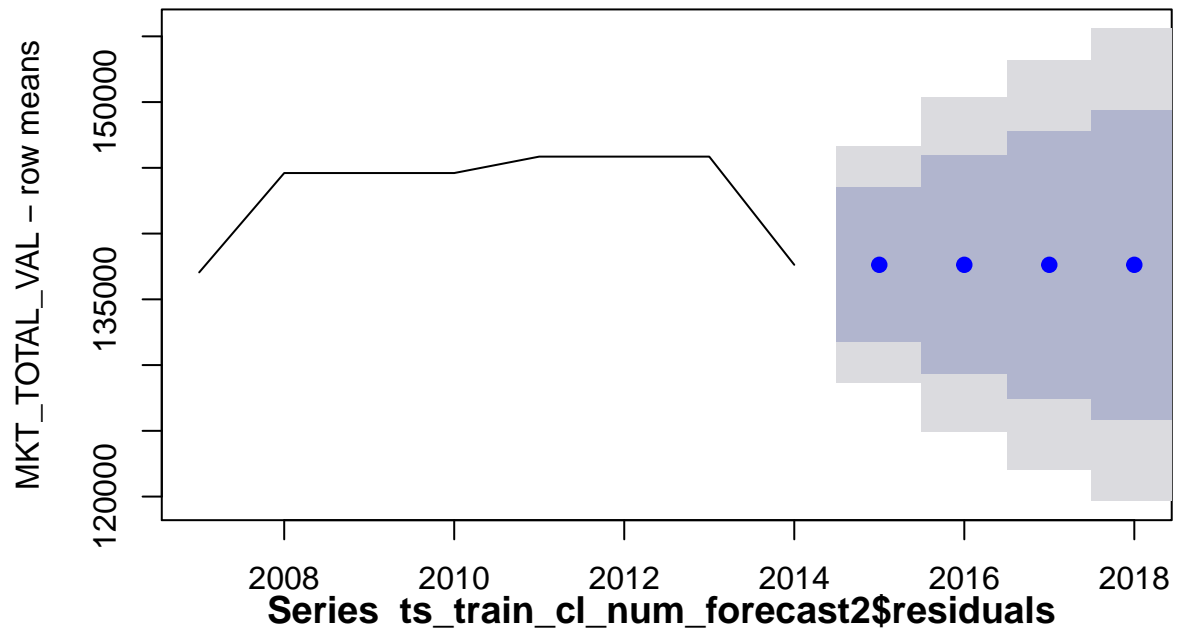
**Histogram of forecasterrors**



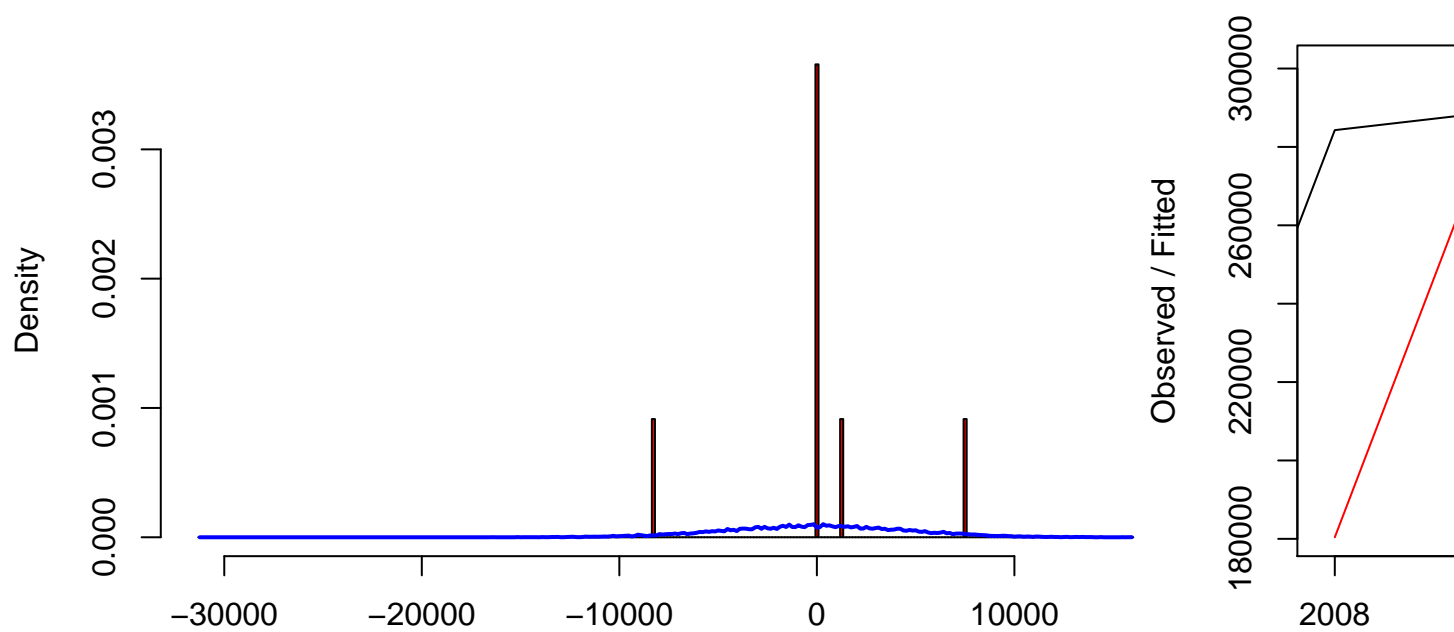
forecasterrors  
**Holt-Winters filtering**



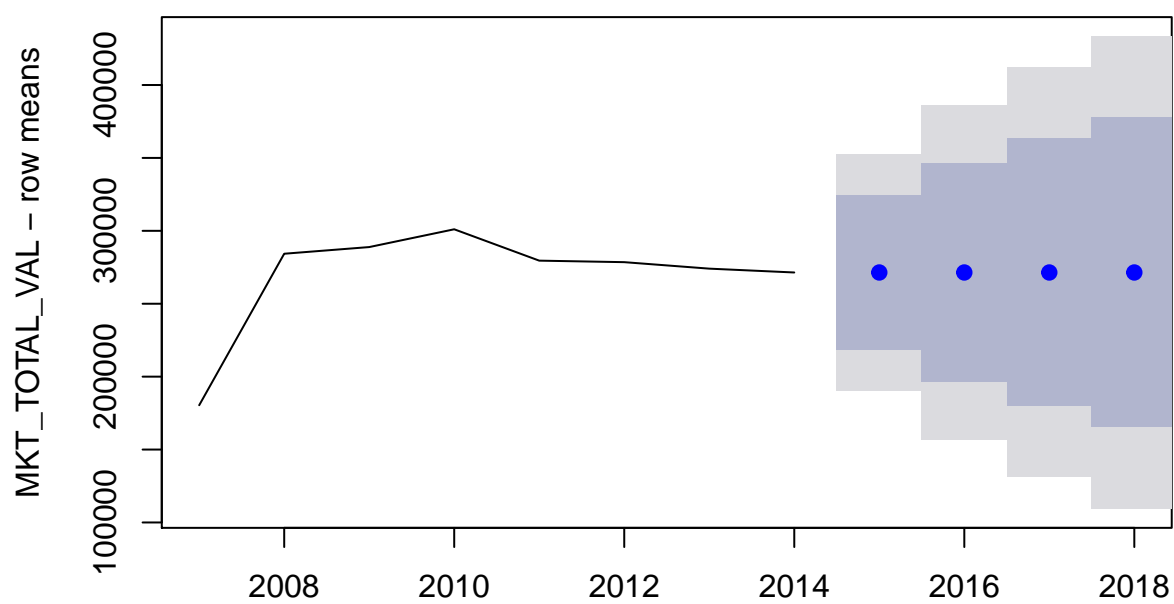
Mean of MKT\_TOTAL\_VAL for Cluster 7



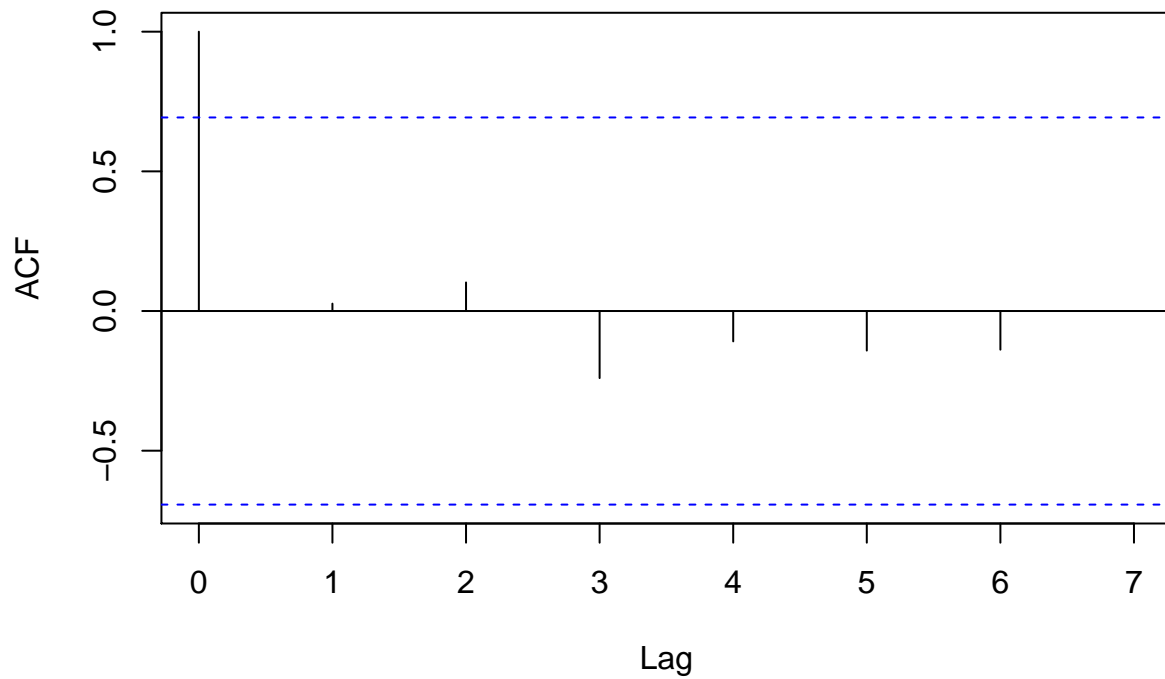
**Histogram of forecasterrors**



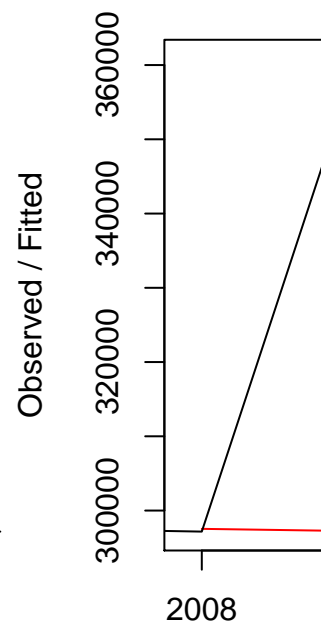
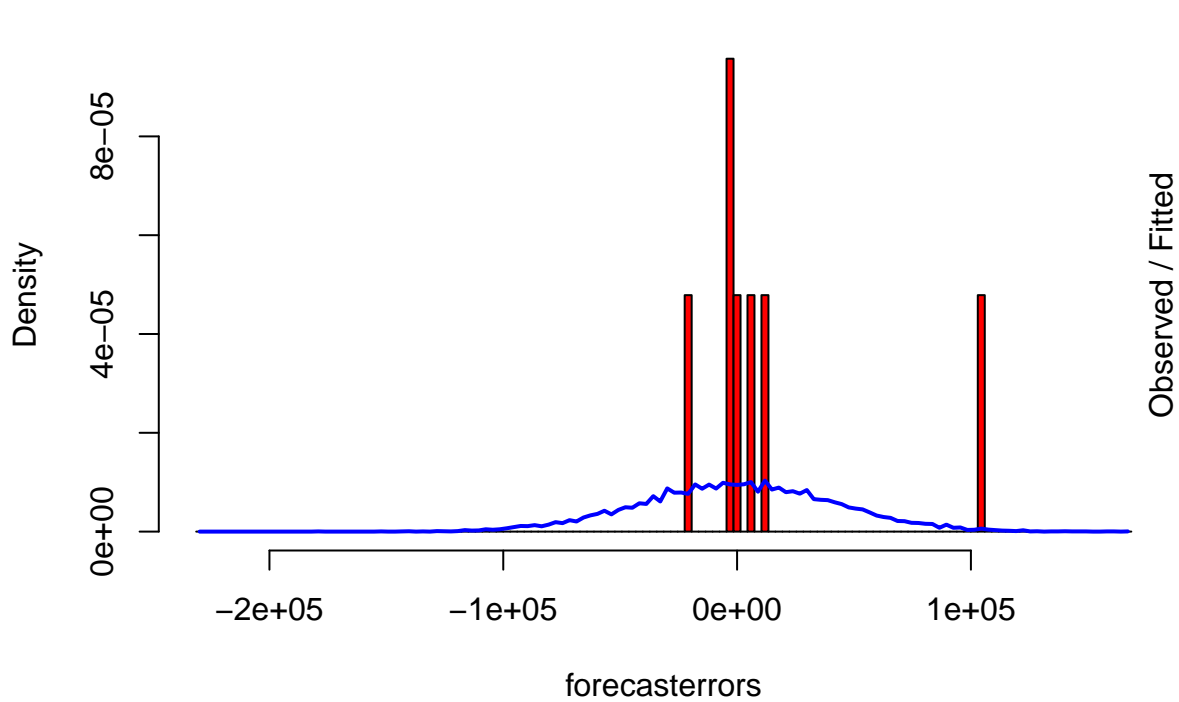
**Mean of MKT\_TOTAL\_VAL for Cluster 8**



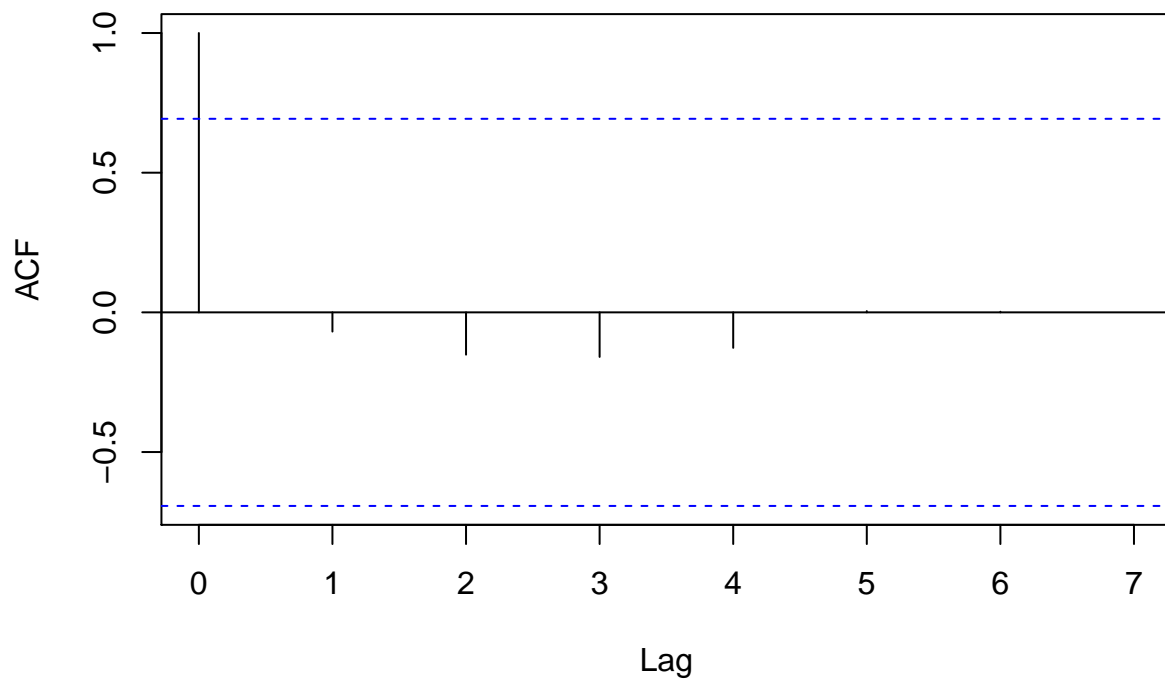
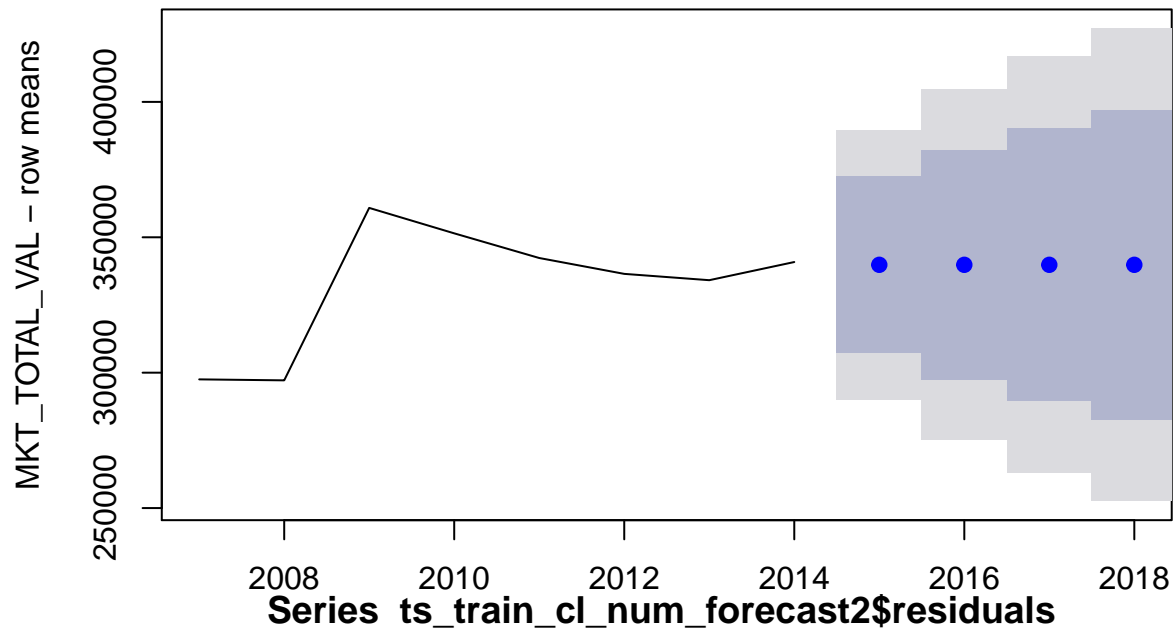
Series ts\_train\_cl\_num\_forecast2\$residuals



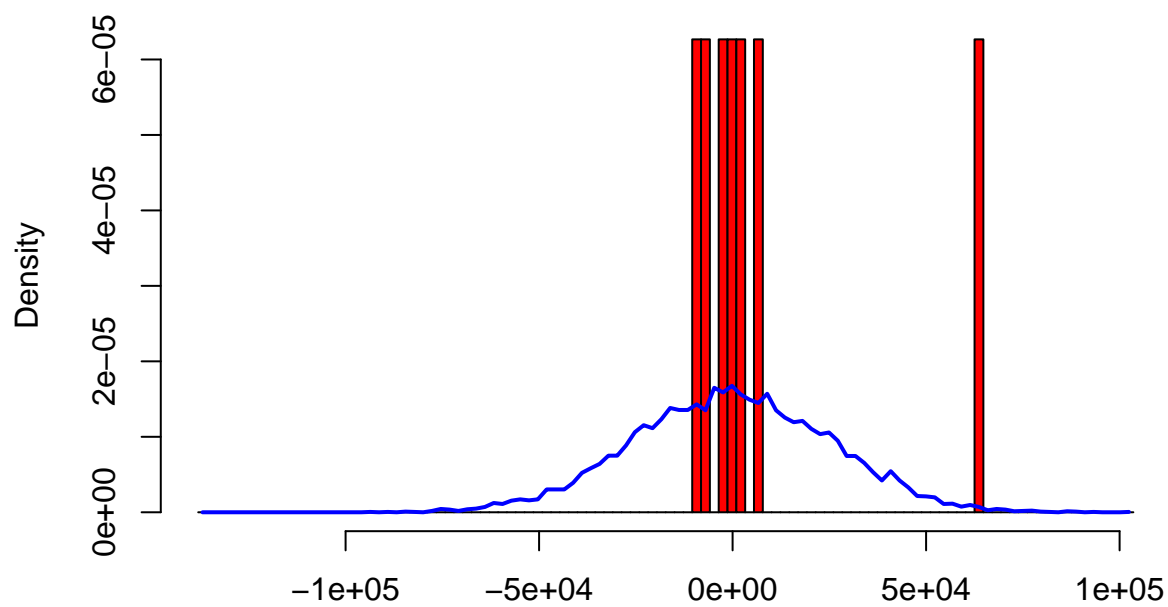
Histogram of forecasterrors



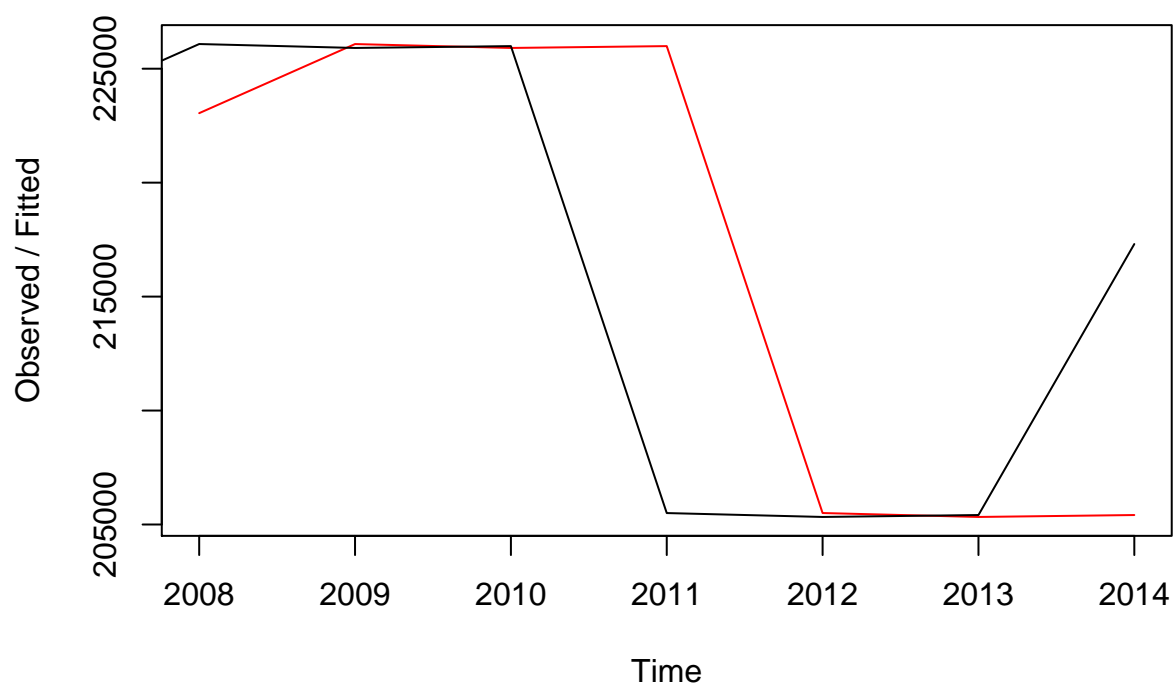
Mean of MKT\_TOTAL\_VAL for Cluster 9



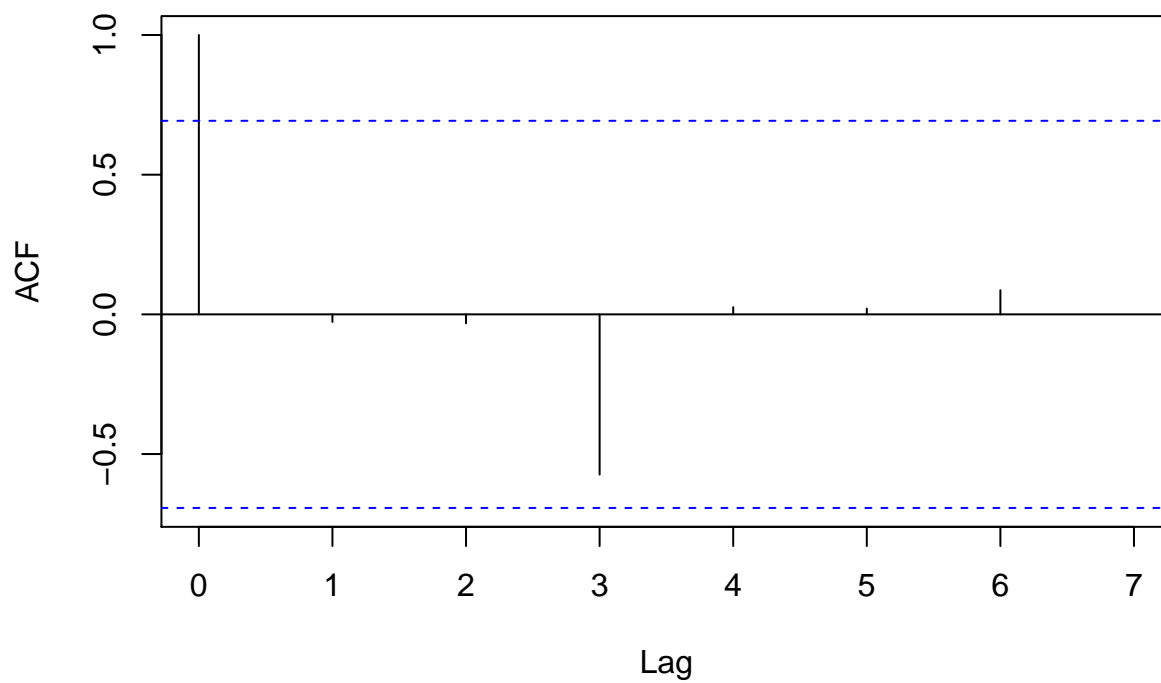
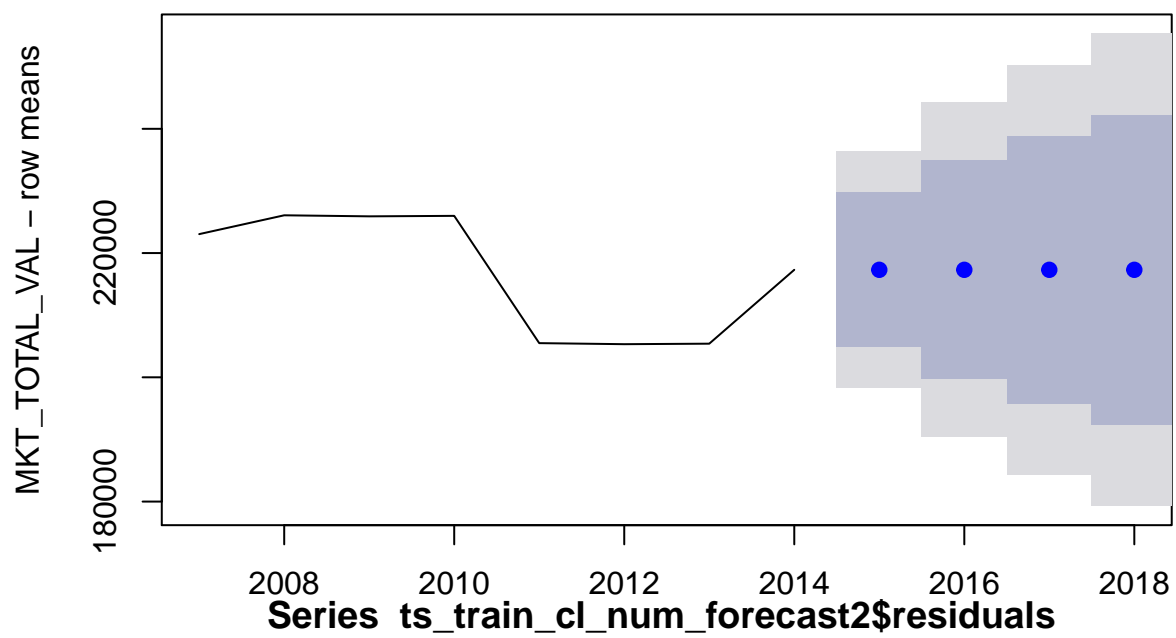
**Histogram of forecasterrors**



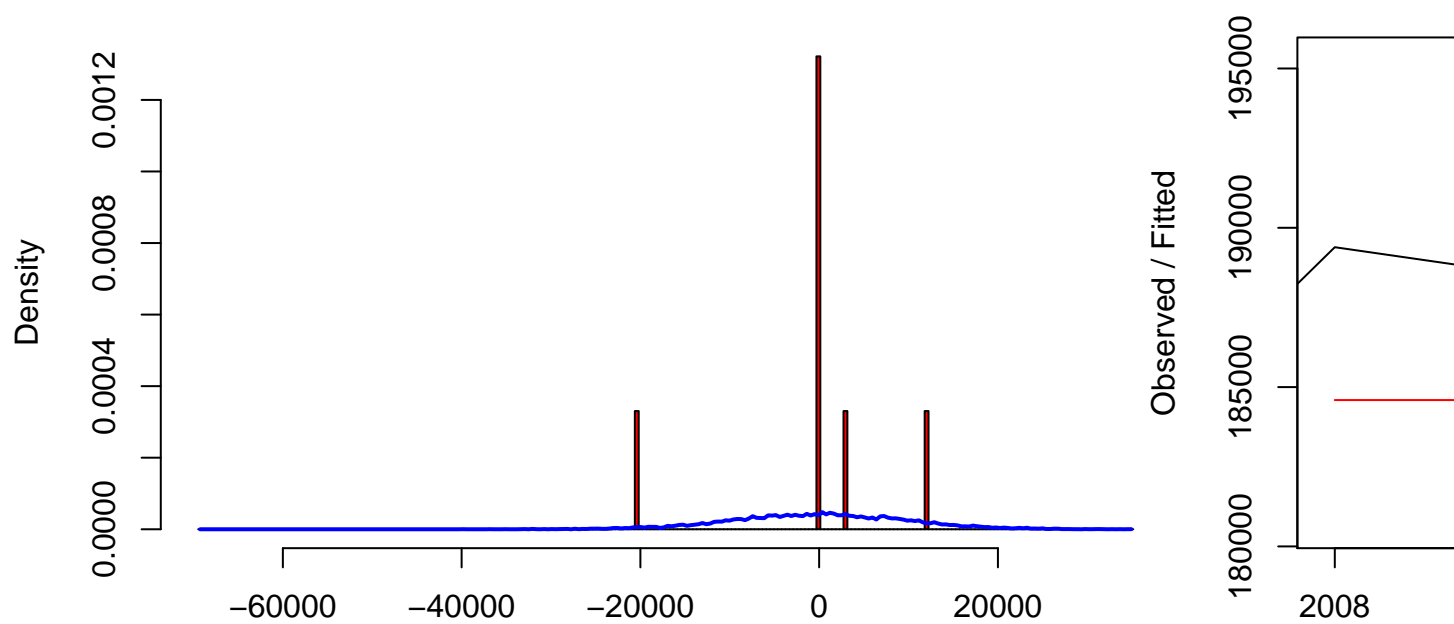
**Holt-Winters filtering**



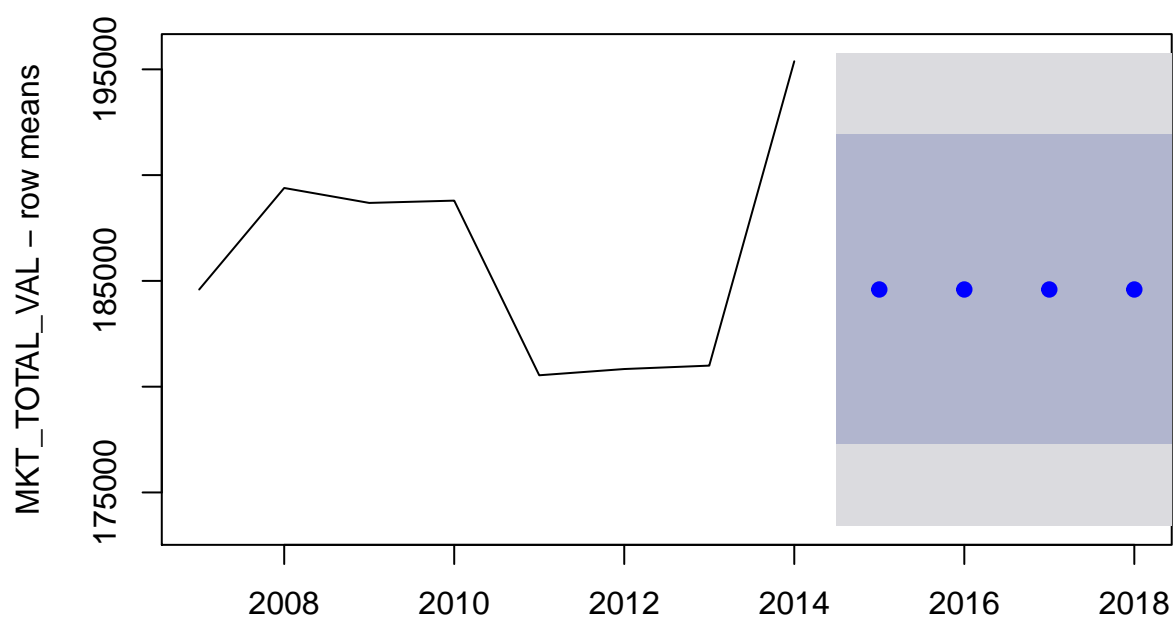
Mean of MKT\_TOTAL\_VAL for Cluster 10



# Histogram of forecasterrors

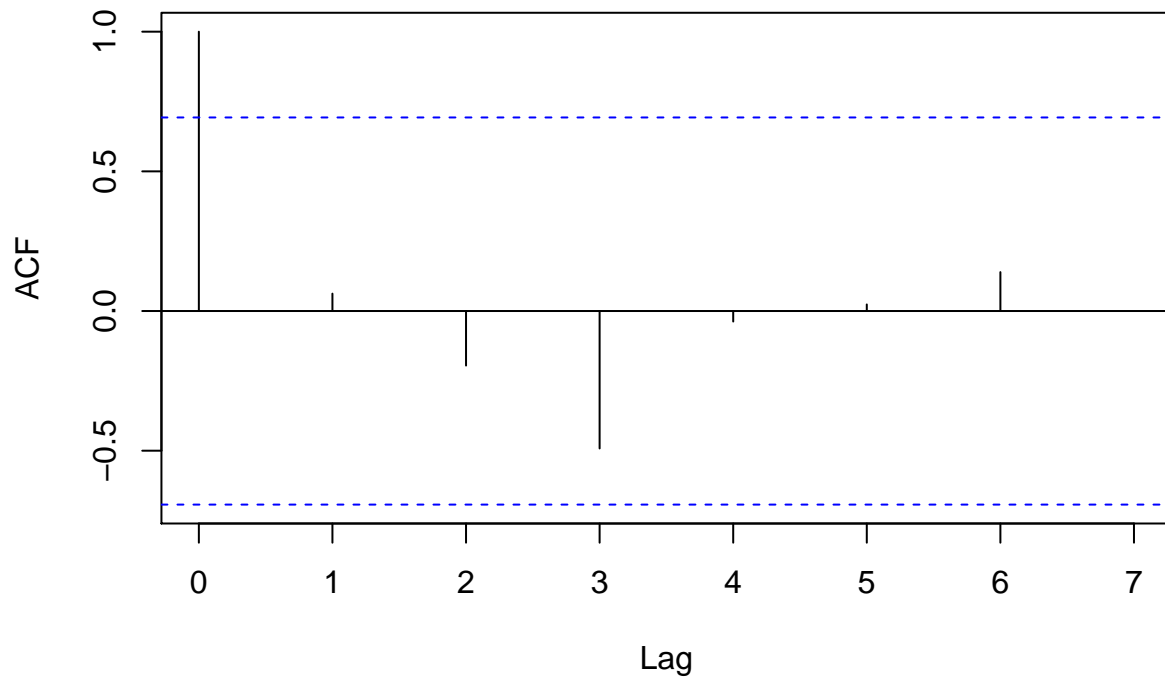


## Mean of MKT\_TOTAL\_VAL for Cluster 11

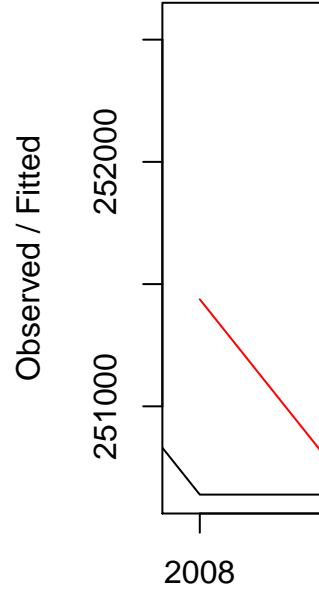
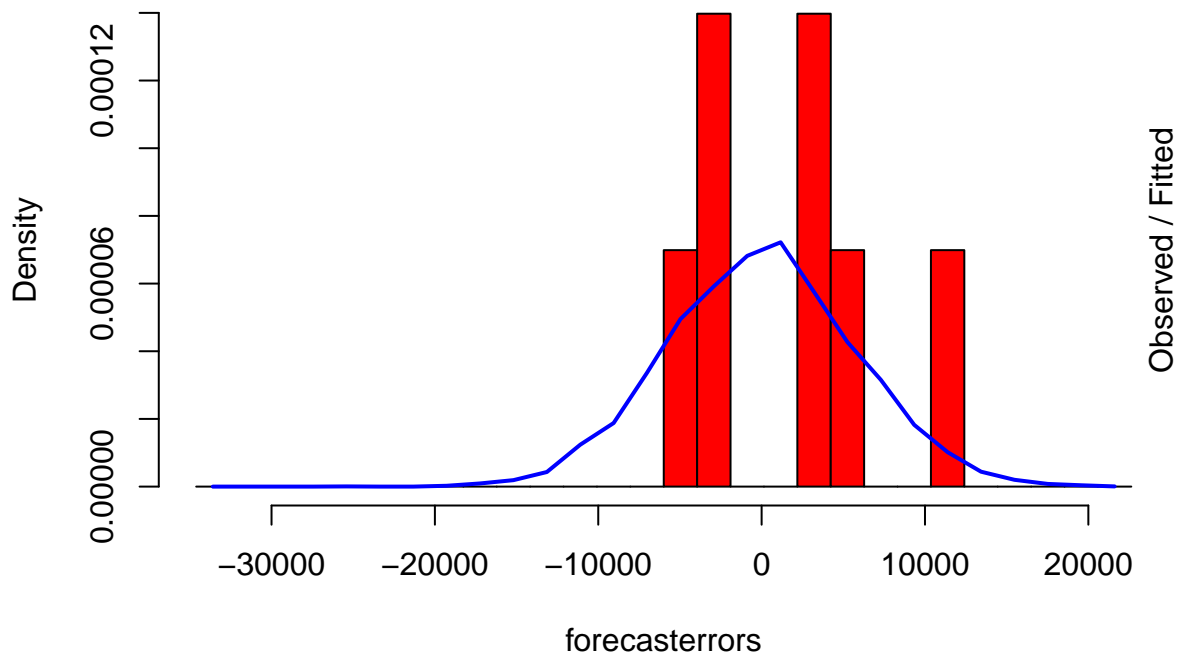




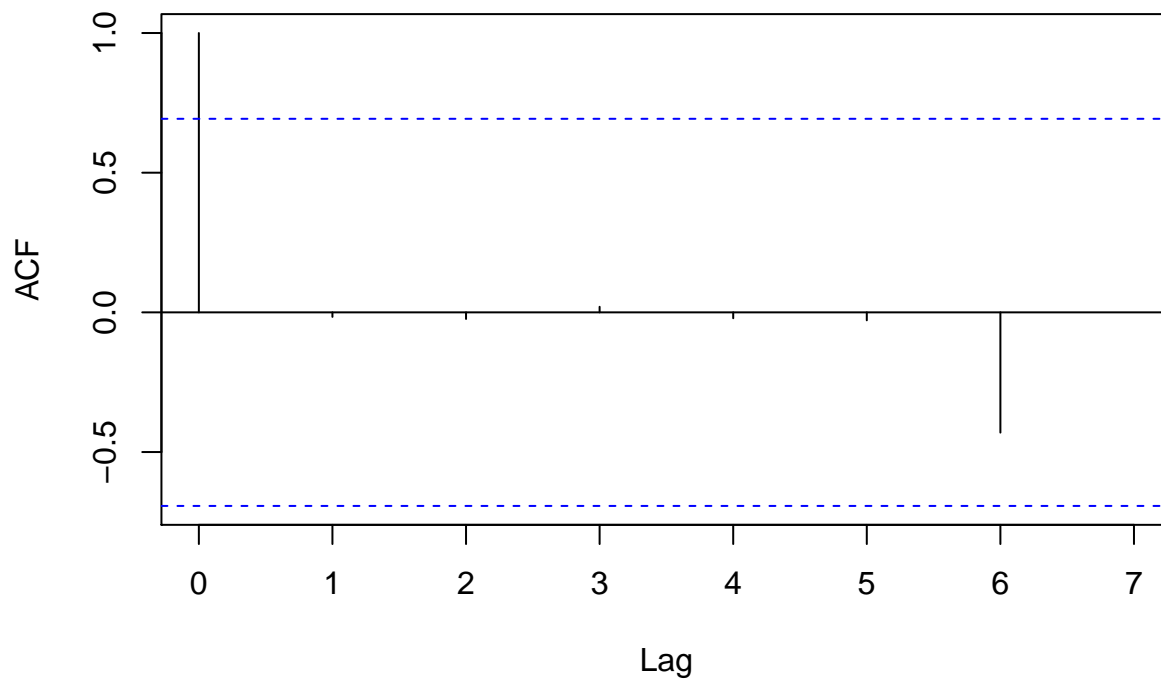
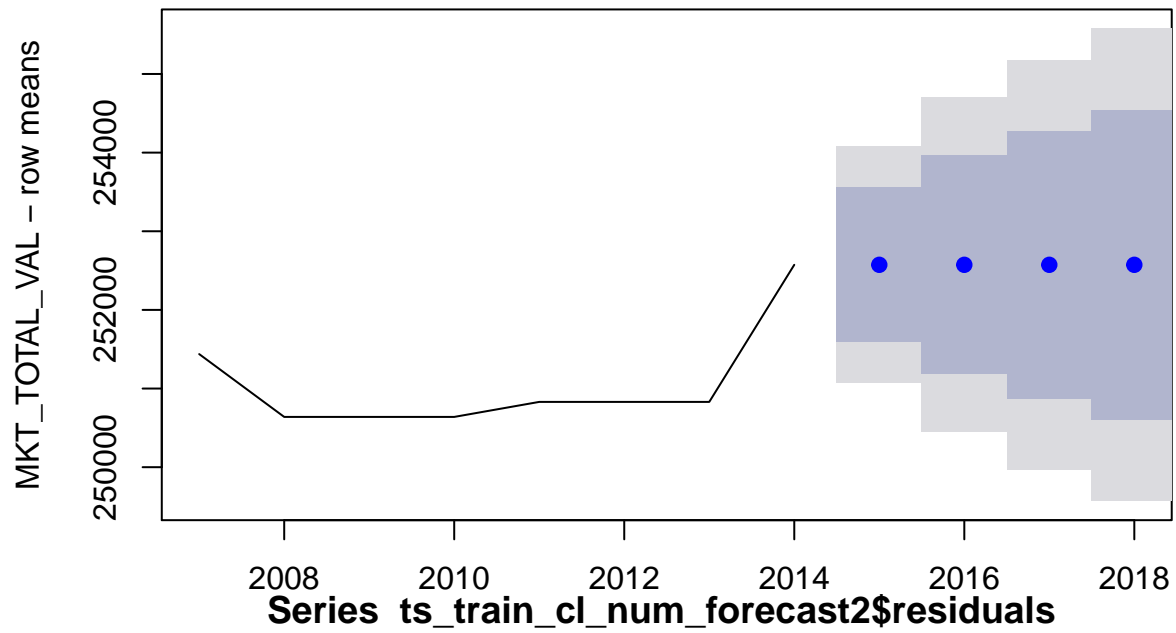
Series ts\_train\_cl\_num\_forecast2\$residuals



Histogram of forecasterrors



## Mean of MKT\_TOTAL\_VAL for Cluster 12



```
## [1] "NON BUFFER ZONE : Plotting Actuals vs. Fitted for models in clusters"
## [1] "NON BUFFER ZONE : Evaluating 12 Clusters with 1st, 2nd, 3rd and 4th and 5th order polynomial regression"
## [1] "NON BUFFER ZONE : Years 2007-2014 serve as training data"
## [1] "NON BUFFER ZONE : Years 2015-2018 serve as test data for validation"

## Warning in window.default(x, ...): 'end' value not changed

## Warning in window.default(x, ...): 'end' value not changed
```

[illegible]

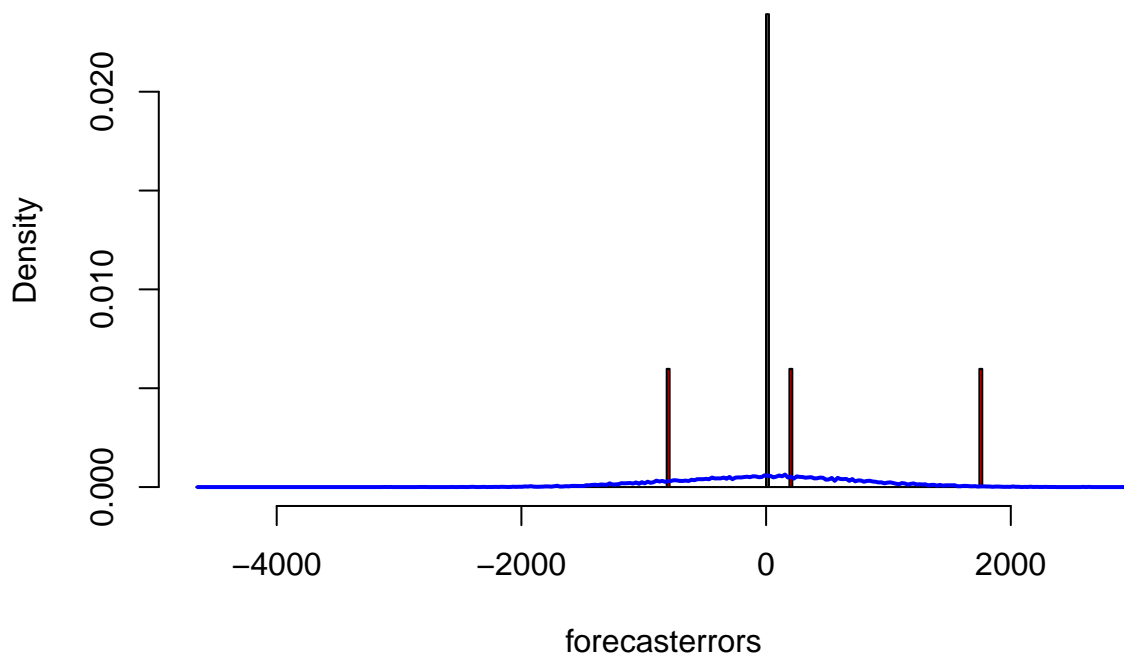
[illegible]

[illegible]

[illegible]

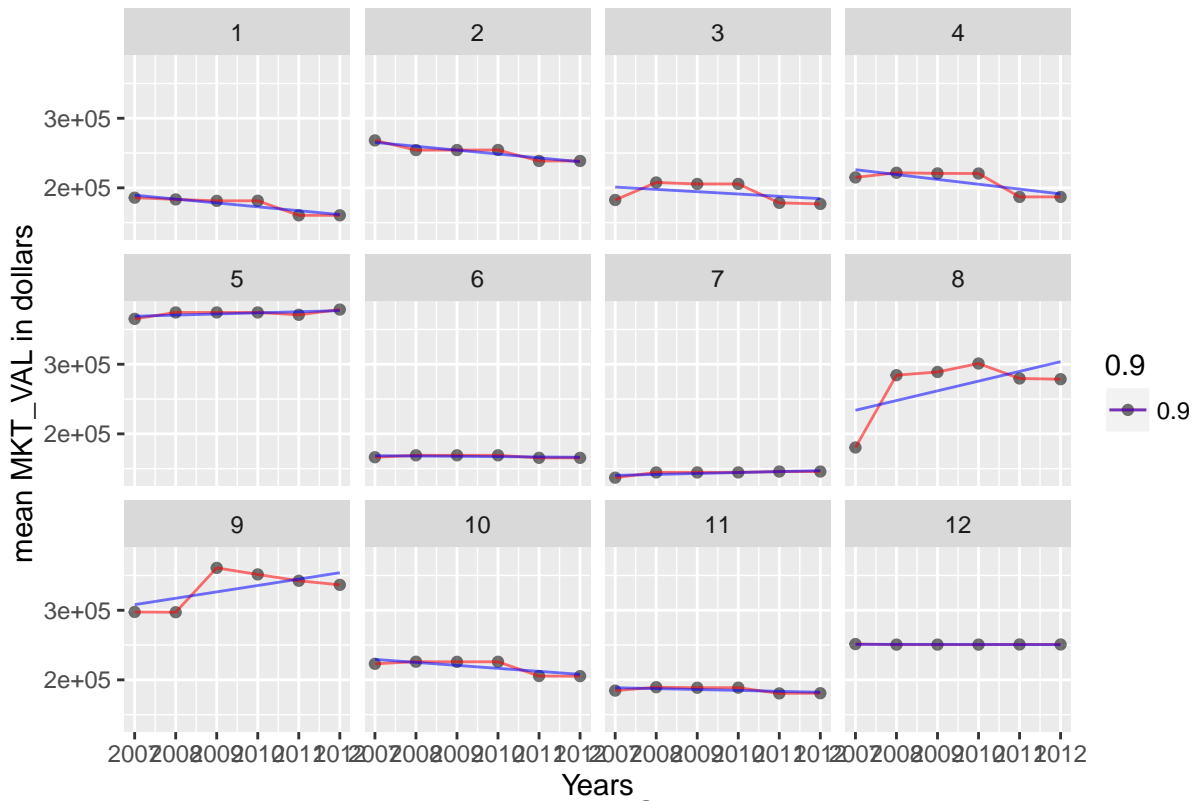
```
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
```

### Histogram of forecasterrors

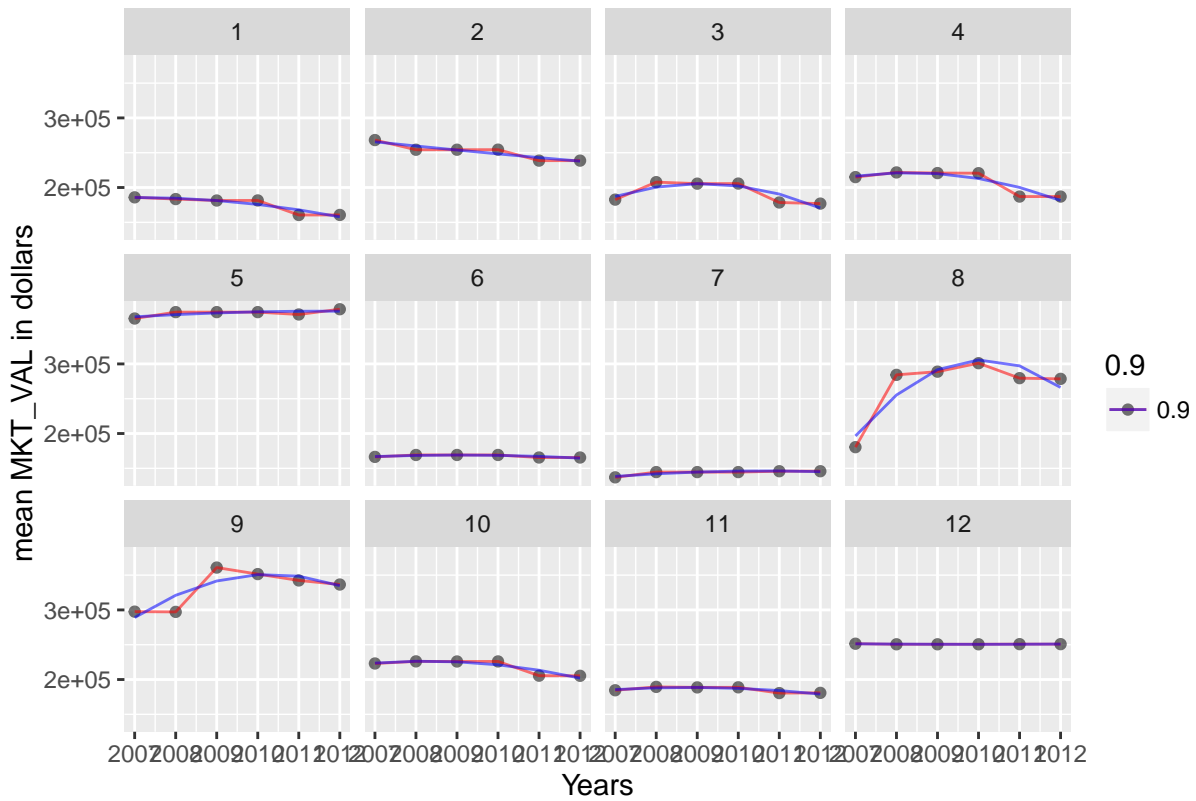


```
## [1] "NON BUFFER ZONE : Computing MAPE - Mean Absolute Percentage error"
## [1] "NON BUFFER ZONE: Plotting the Actual vs. Fitted for Training set "
```

Actual vs. Fitted – Non BUFFER ZONE training set poly\_order: 1

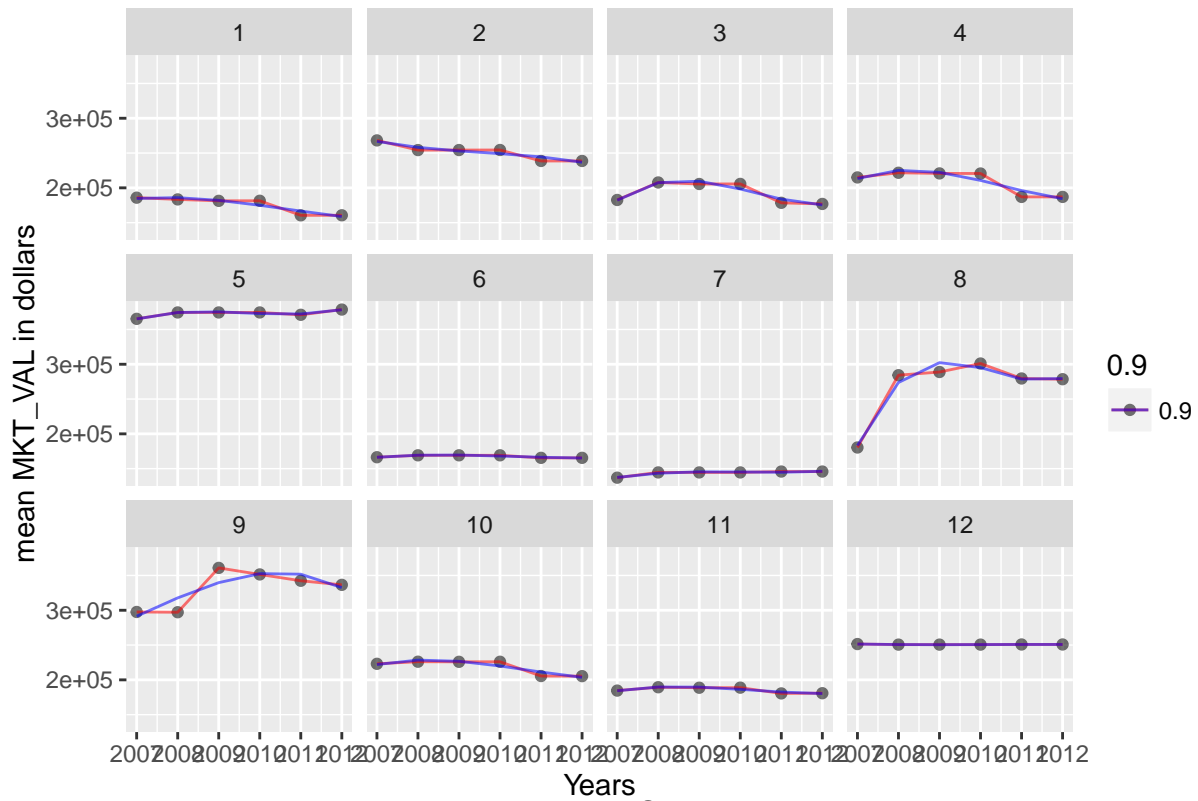


Actual vs. Fitted – Non BUFFER ZONE training set poly\_order: 2

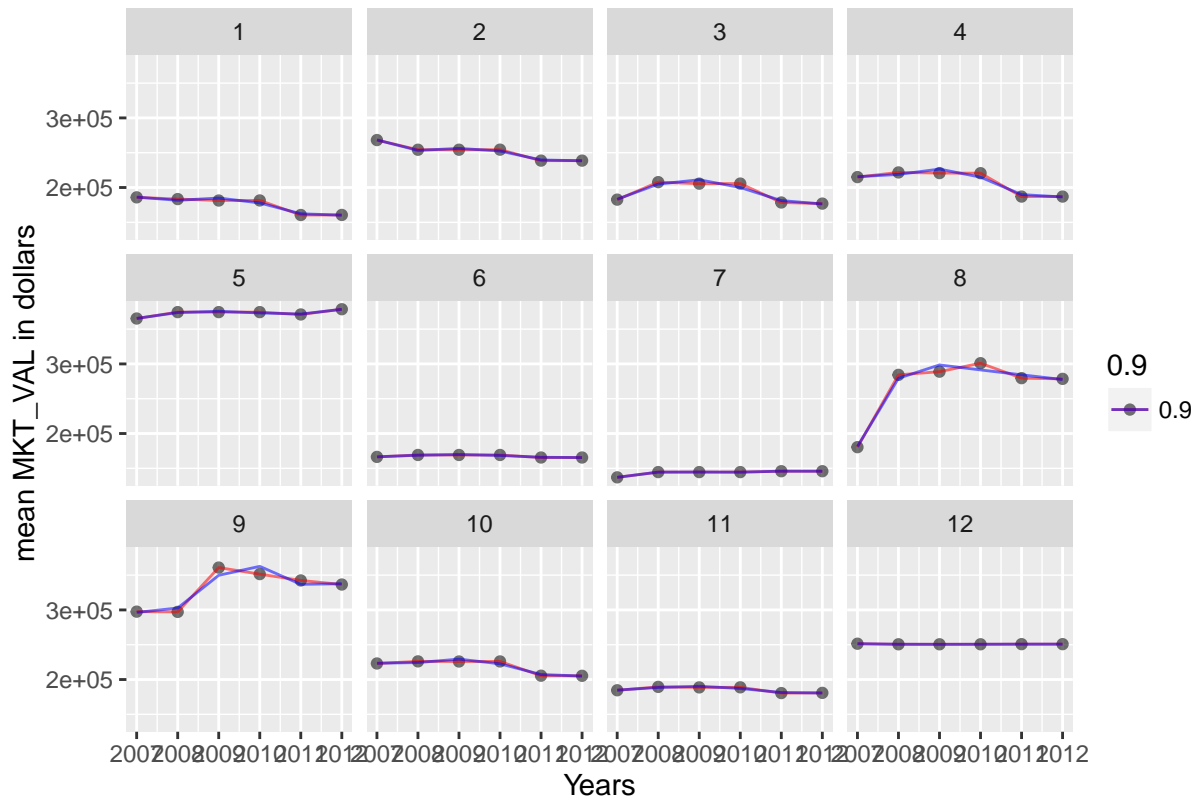




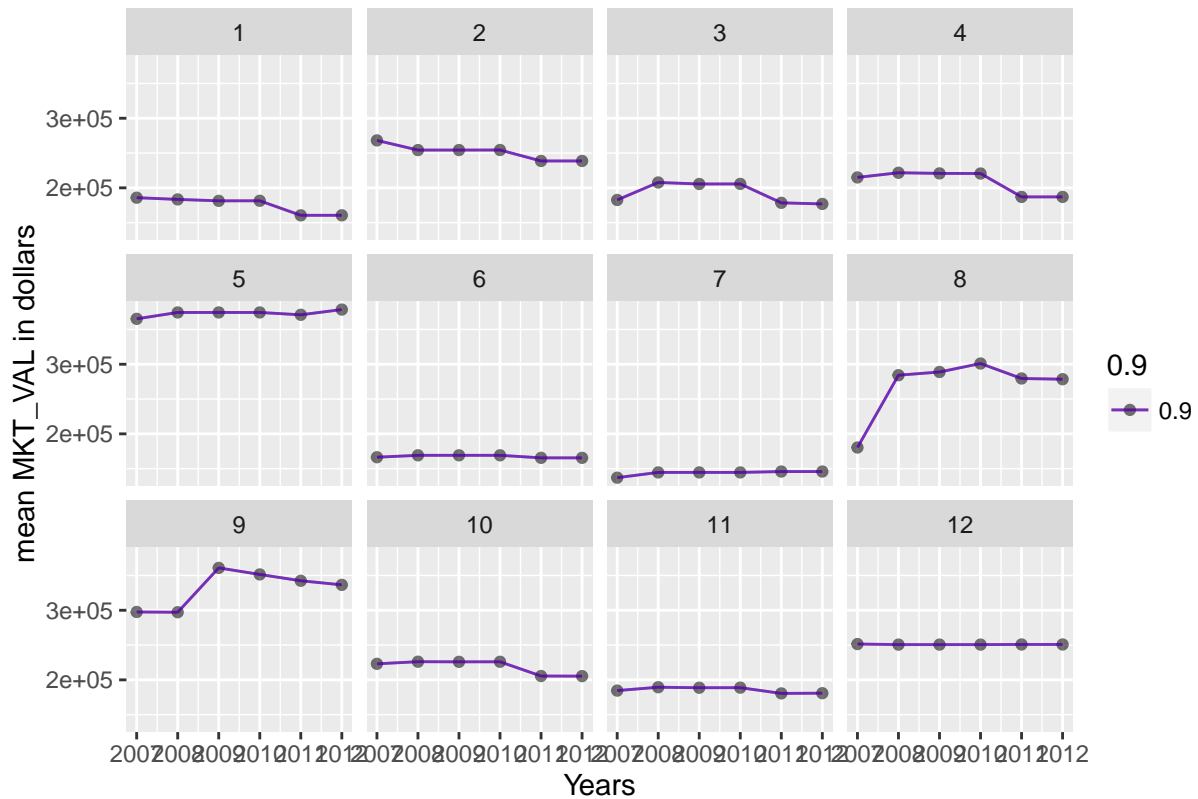
Actual vs. Fitted – Non BUFFER ZONE training set poly\_order: 3



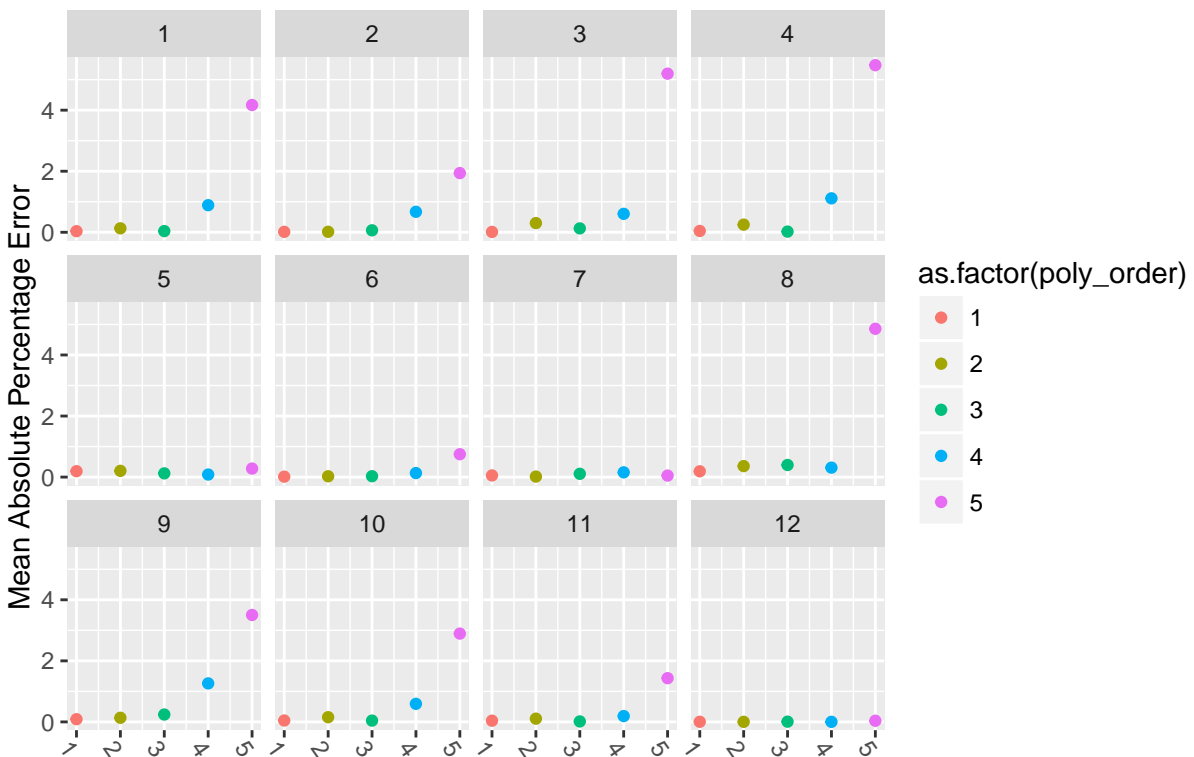
Actual vs. Fitted – Non BUFFER ZONE training set poly\_order: 4



Actual vs. Fitted – Non BUFFER ZONE training set poly\_order: 5



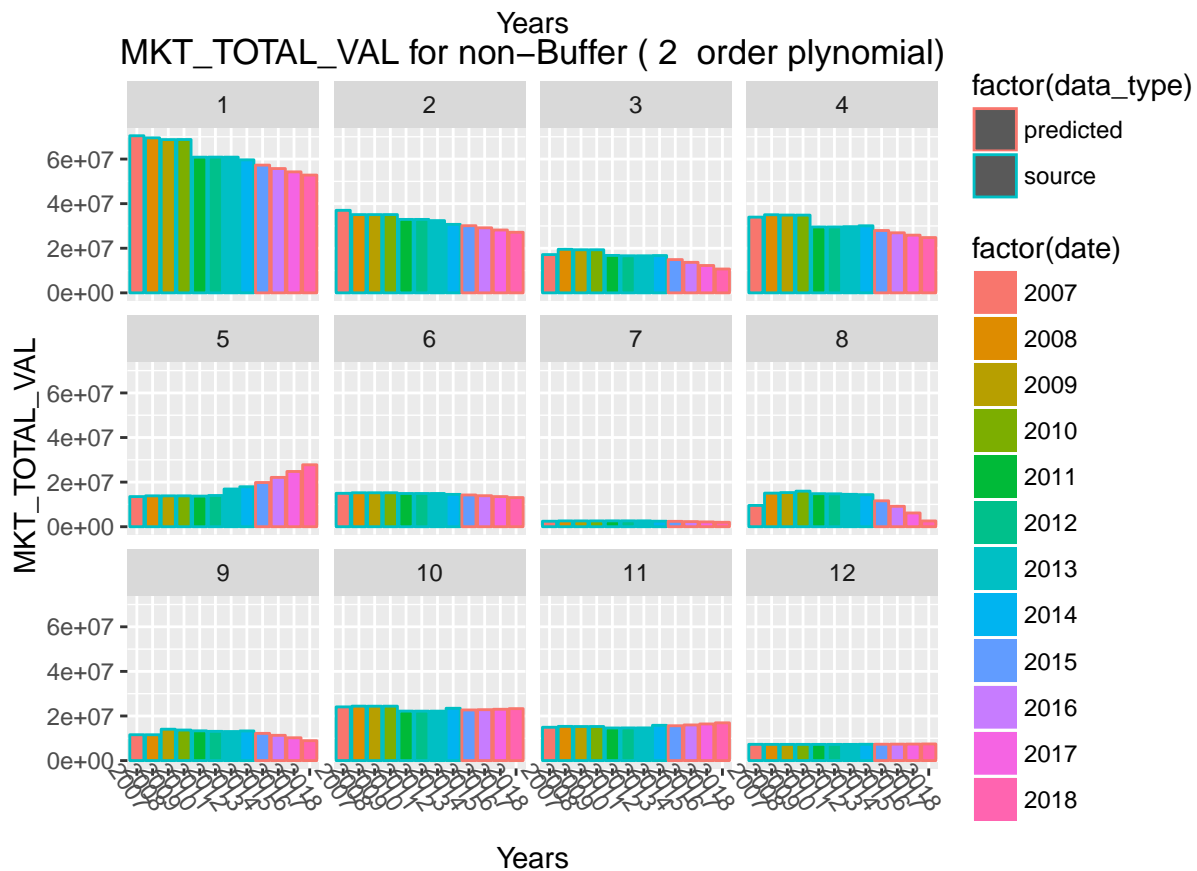
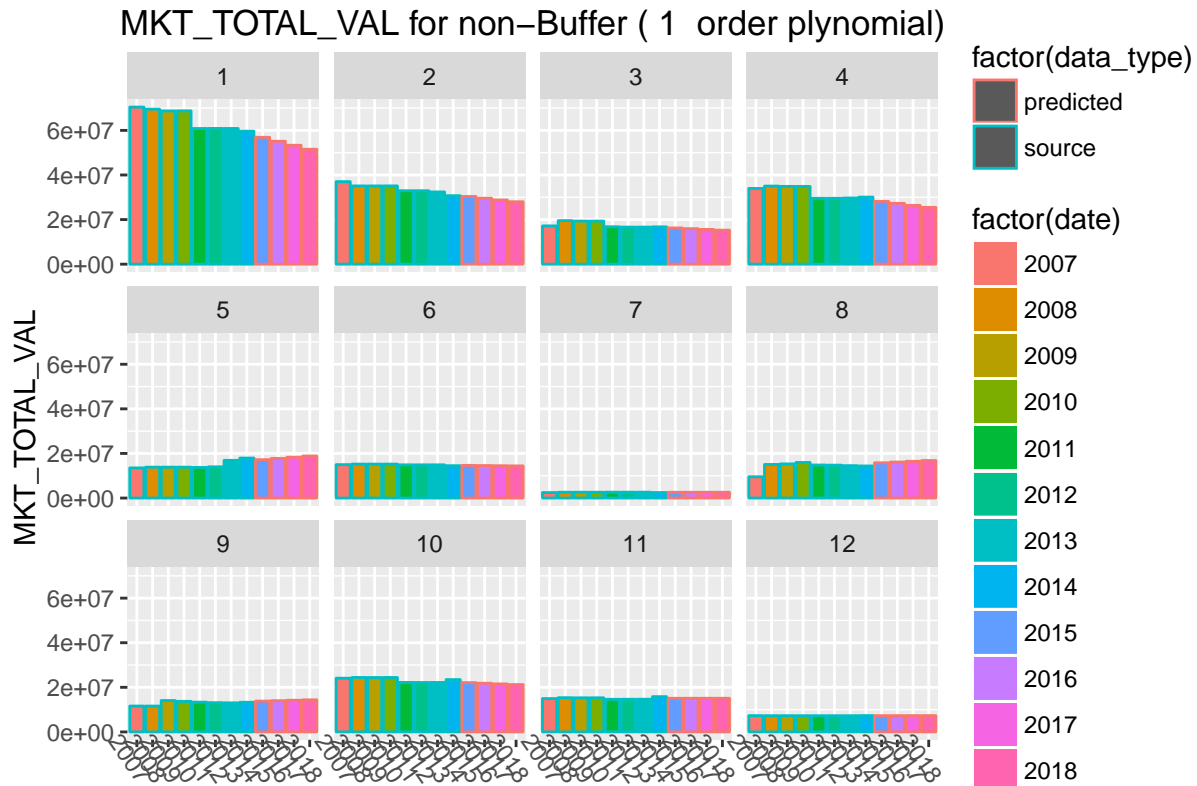
MAPE – non Buffer zone : test data

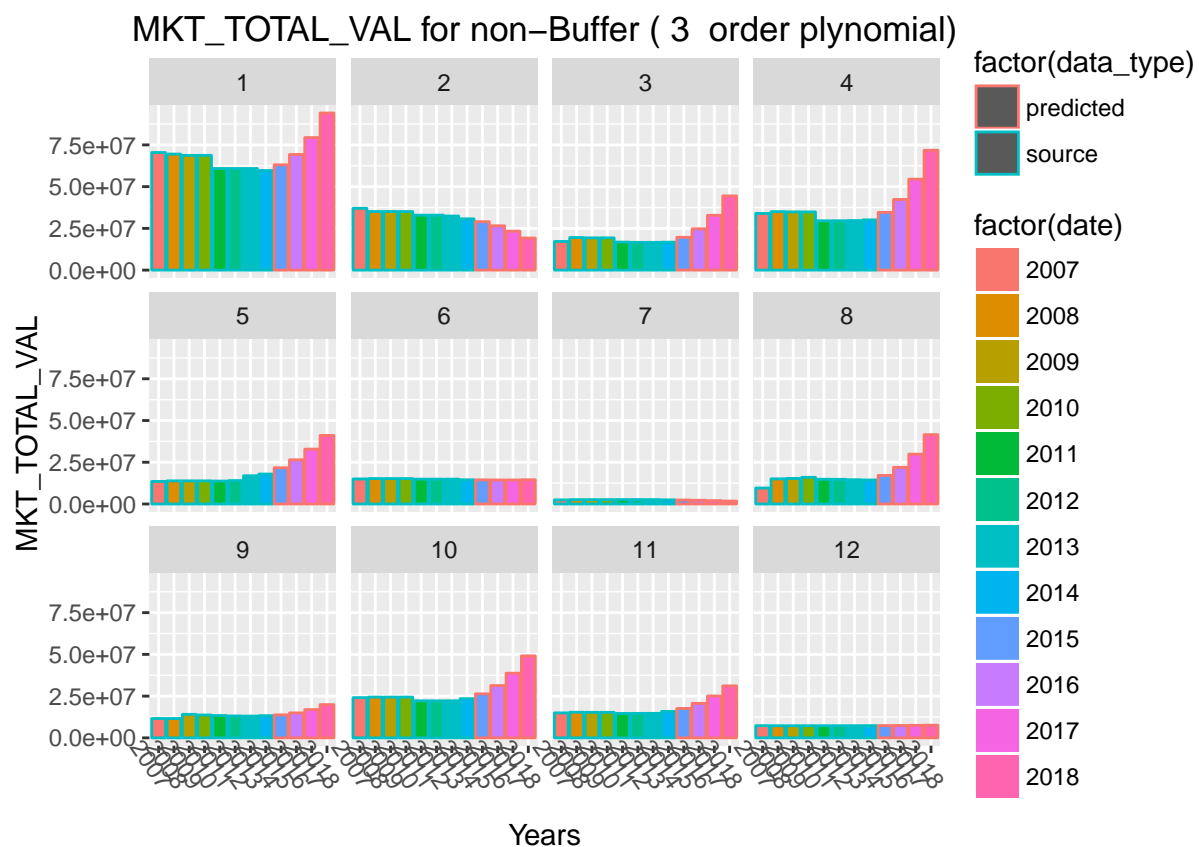


Polynomial Order – ( Linear, Quadratic, Cubic, 4th and 5th order polynomial)

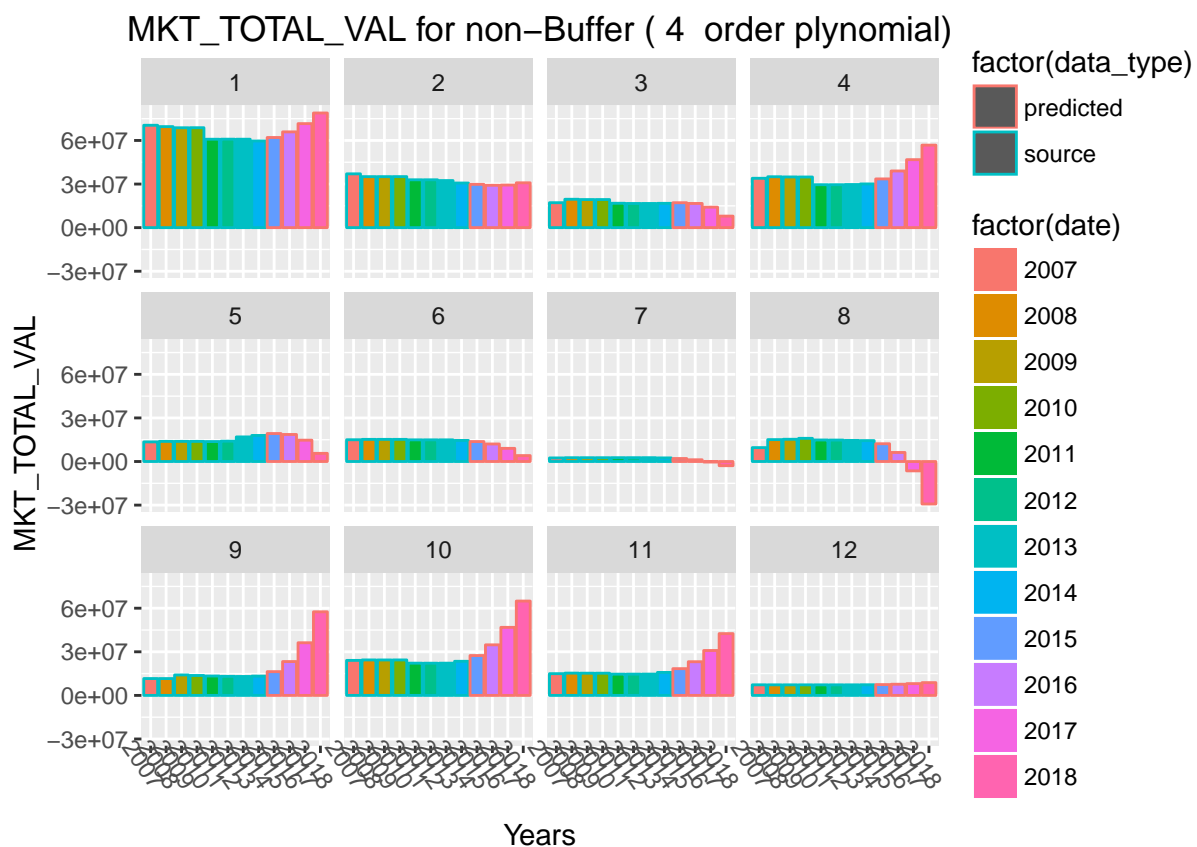
```
## [1] "NON BUFFER ZONE : Plotting MKT_TOT_VAL source and predicted data"
```

```
## [1] "NON BUFFER ZONE : Creating df_final_mkt_val data frame for later Visualization"
```

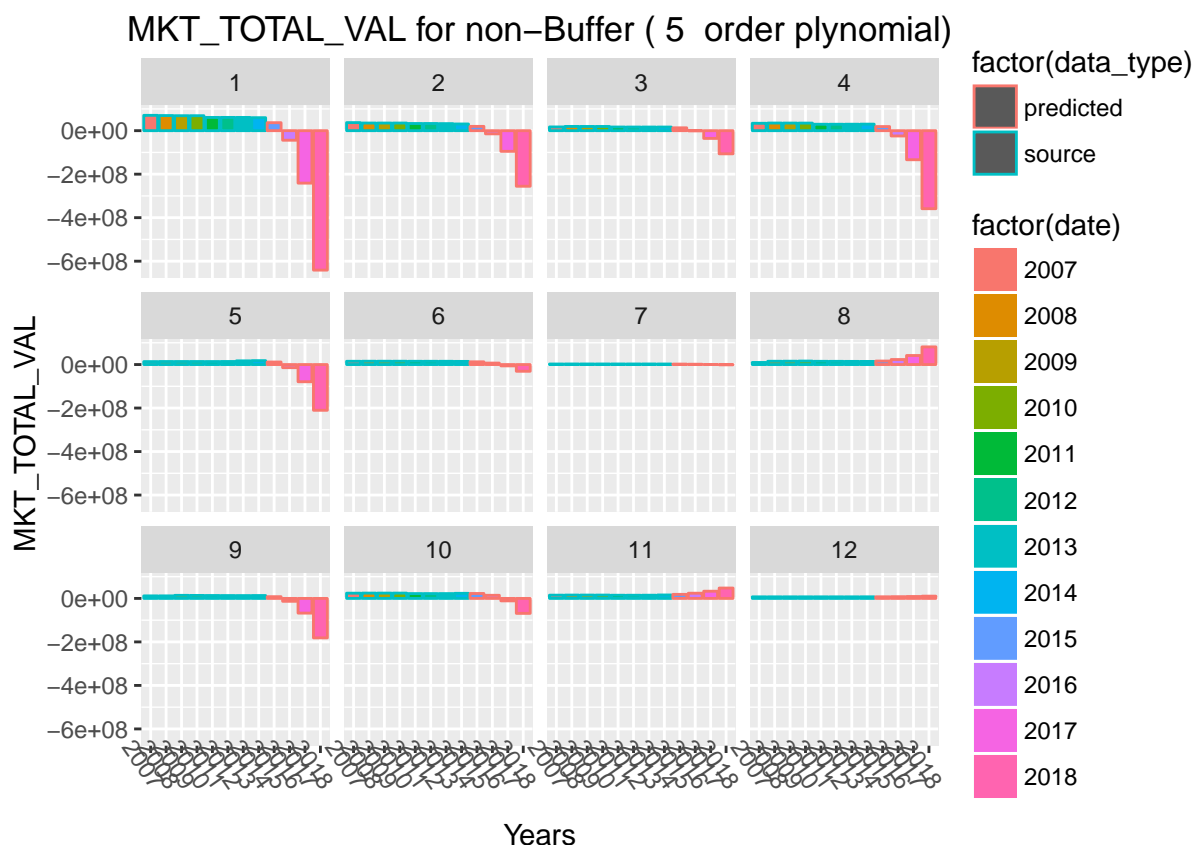




## Warning: Stacking not well defined when ymin != 0



## Warning: Stacking not well defined when ymin != 0



```
print("BUFFER ZONE: Training and test data set ")
```

```
## [1] "BUFFER ZONE: Training and test data set "
```

```
#####
# EDGE
#####
```

```
print("EDGE BUFFER ZONE : Training Data - Dendrogram")
```

```
## [1] "EDGE BUFFER ZONE : Training Data - Dendrogram"
```

```
#then random 100 columns
```

```
n <- ncol(taxinfo_EDGE.matrix_no_na)
```

```
taxinfo_EDGE.matrix_shuffled <- taxinfo_EDGE.matrix_no_na[,sample(n) ]
```

```
#0.7 is 813 values
```

```
train_indices <- 1:round(0.7 * n)
```

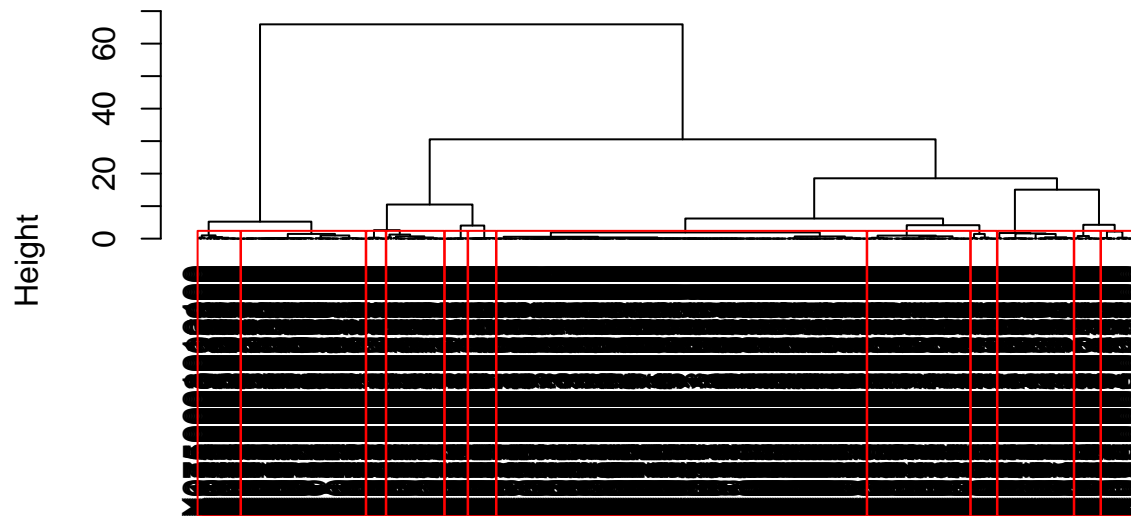
```
train <- taxinfo_EDGE.matrix_shuffled[,train_indices ]
```

```
ClustvarEDGE100 <-hclustvar(train)
```

```
plot(ClustvarEDGE100)
```

```
rect.hclust(ClustvarEDGE100,k=12, border="red")
```

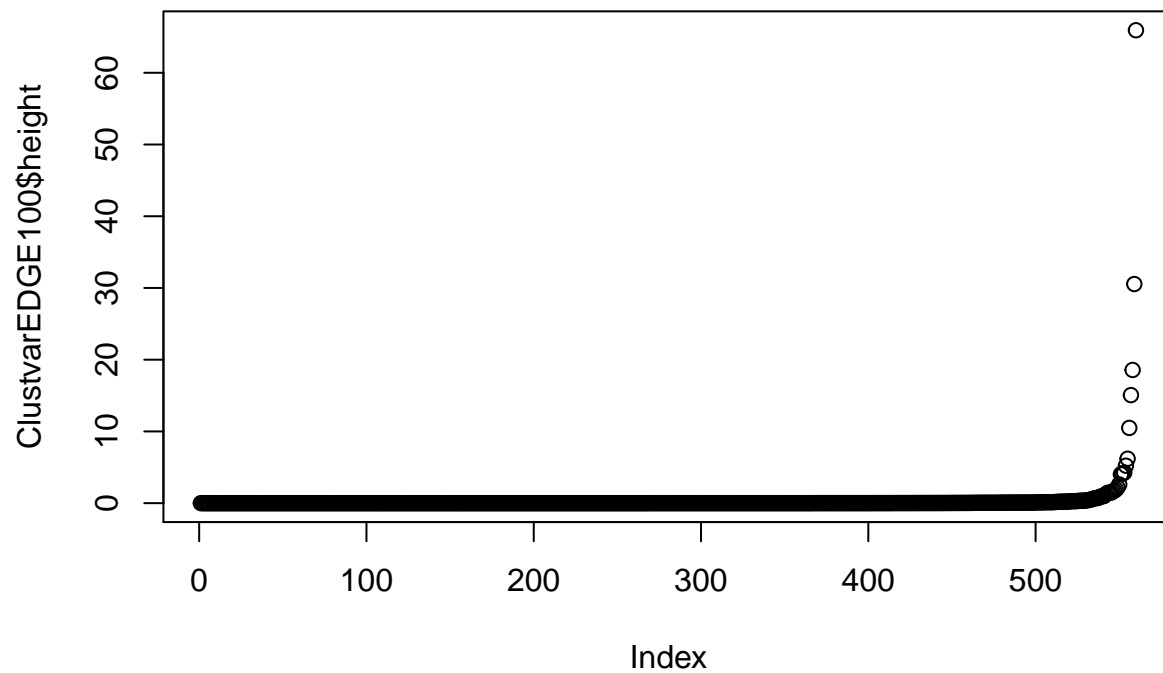
## Cluster Dendrogram



```
partEDGE100 <- cutreevar(ClustvarEDGE100,12)
```

```
a_EDGE <- partEDGE100$var
```

```
plot(ClustvarEDGE100$height)
```



```
#####
# CENTER
#####
print("CENTER BUFFER ZONE : Training Data - Dendrogram")
```

```
## [1] "CENTER BUFFER ZONE : Training Data - Dendrogram"
```

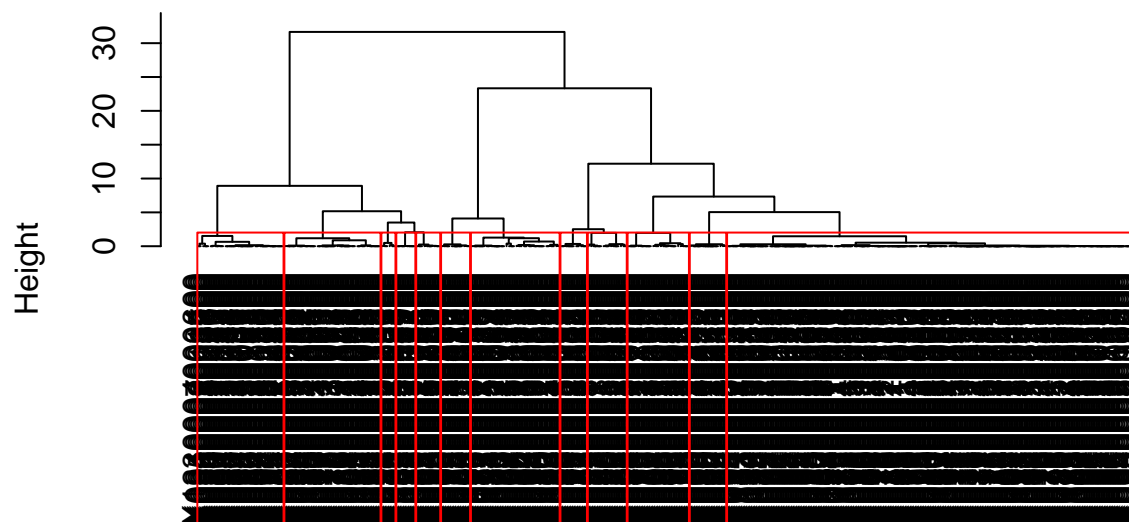
```

#then random 100 columns
n <- ncol(taxinfo_CENTER.matrix_no_na)
taxinfo_CENTER.matrix_shuffled <- taxinfo_CENTER.matrix_no_na[,sample(n) ]
#0.7 is 561 values
train_indices <- 1:round(0.7 * n)
train <- taxinfo_CENTER.matrix_shuffled[,train_indices ]

ClustvarCENTER100 <-hclustvar(train)
plot(ClustvarCENTER100)
rect.hclust(ClustvarCENTER100,k=12, border="red")

```

## Cluster Dendrogram

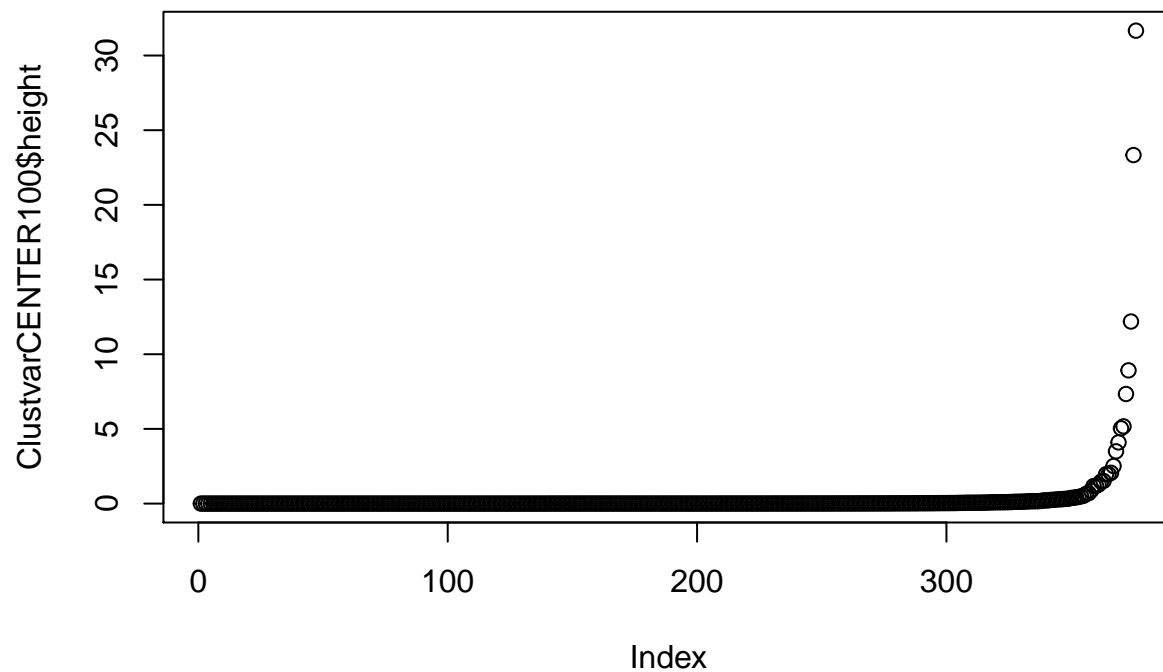


```

partCENTER100 <- cutreevar(ClustvarCENTER100,12)
a_CENTER <- partCENTER100$var
plot(ClustvarCENTER100$height)

```





```
#stab100 <- stability(Clustvar100,B = 60, graph = TRUE)

#####
# CORE
#####

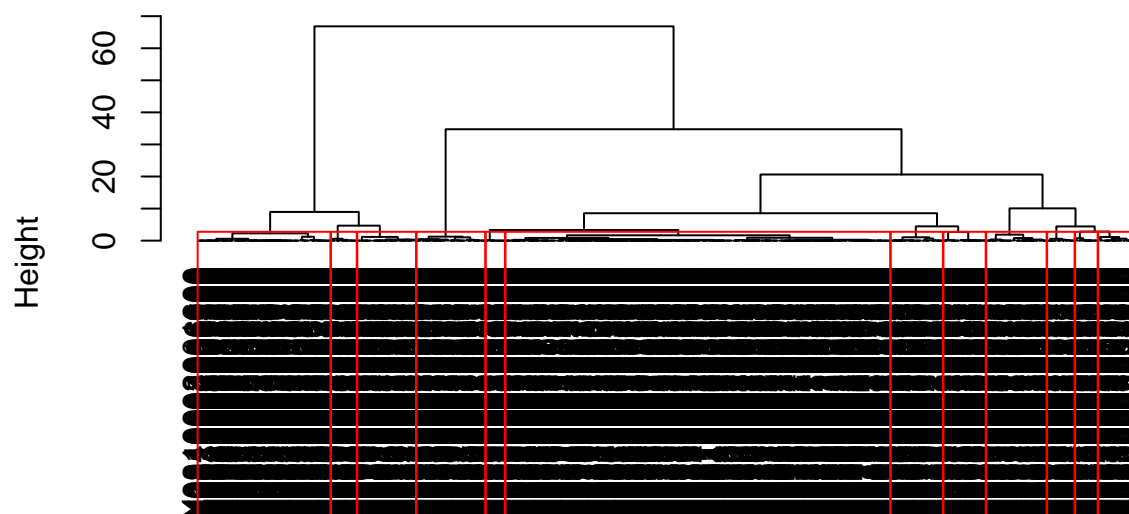
print("CORE BUFFER ZONE : Training Data - Dendogram")

## [1] "CORE BUFFER ZONE : Training Data - Dendogram"

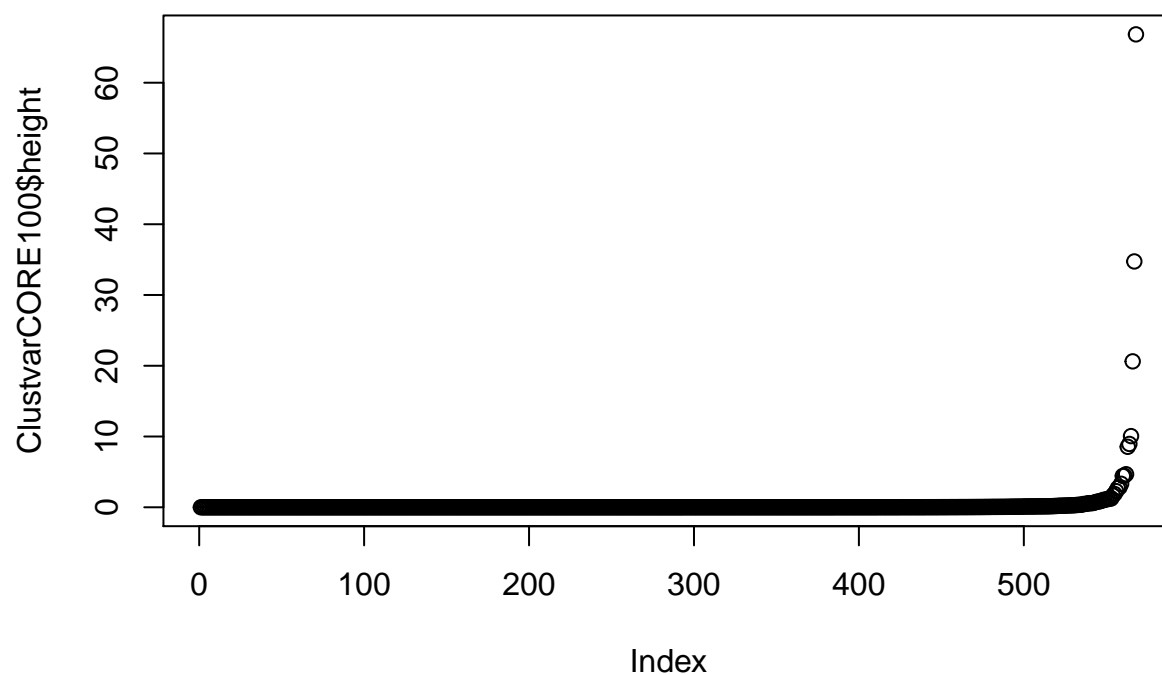
#then random 100 columns
n <- ncol(taxinfo_CORE.matrix_no_na)
taxinfo_CORE.matrix_shuffled <- taxinfo_CORE.matrix_no_na[,sample(n) ]
#0.7 is 561 values
train_indices <- 1:round(0.7 * n)
train <- taxinfo_CORE.matrix_shuffled[,train_indices ]

ClustvarCORE100 <-hclustvar(train)
plot(ClustvarCORE100)
rect.hclust(ClustvarCORE100,k=12, border="red")
```

## Cluster Dendrogram



```
partCORE100 <- cutreevar(ClustvarCORE100,12)
a_CORE <-partCORE100$var
plot(ClustvarCORE100$height)
```



```
#stab100 <- stability(Clustvar100,B = 60, graph = TRUE)
print("BUFFER_ZONE: About to plot actuals and fitted")

## [1] "BUFFER_ZONE: About to plot actuals and fitted"
for( poly_order in c(seq(1:5))){
```

```

for ( clst in c(seq(1:12))) {

  df_mape_train_CORE_parcel    <- plot_act_pred_trg(clst,a_CORE,  taxinfo_CORE.matrix[c(1:6),], poly_
  df_mape_train_CENTER_parcel  <- plot_act_pred_trg(clst,a_CENTER, taxinfo_CENTER.matrix[c(1:6),], poly_
  df_mape_train_EDGE_parcel    <- plot_act_pred_trg(clst,a_EDGE,  taxinfo_EDGE.matrix[c(1:6),], poly_

  df_mape_test_CORE_parcel    <- plot_act_pred_test(clst,a_CORE,  taxinfo_CORE.matrix[c(1:6),], taxinfo_
  df_mape_test_CENTER_parcel  <- plot_act_pred_test(clst,a_CENTER, taxinfo_CENTER.matrix[c(1:6),], taxinfo_
  df_mape_test_EDGE_parcel    <- plot_act_pred_test(clst,a_EDGE,  taxinfo_EDGE.matrix[c(1:6),], taxinfo_

  if (clst == 1) {

    df_mape_train_CORE_clst_poly <- df_mape_train_CORE_parcel["mape"]
    df_mape_train_CENTER_clst_poly <- df_mape_train_CENTER_parcel["mape"]
    df_mape_train_EDGE_clst_poly <- df_mape_train_EDGE_parcel["mape"]

    df_actpred_train_CORE_clst_poly <- df_mape_train_CORE_parcel["act_pred"]
    df_actpred_train_CENTER_clst_poly <- df_mape_train_CENTER_parcel["act_pred"]
    df_actpred_train_EDGE_clst_poly <- df_mape_train_EDGE_parcel["act_pred"]

    df_mape_test_CORE_clst_poly <- df_mape_test_CORE_parcel["mape"]
    df_mape_test_CENTER_clst_poly <- df_mape_test_CENTER_parcel["mape"]
    df_mape_test_EDGE_clst_poly <- df_mape_test_EDGE_parcel["mape"]

  }
  else {

    #df_mape_train_CORE_clst_poly <- bind_rows(df_mape_train_CORE_clst_poly,      df_mape_train_CORE_p
    #df_mape_train_CENTER_clst_poly <- bind_rows(df_mape_train_CENTER_clst_poly,      df_mape_train_CE
    #df_mape_train_EDGE_clst_poly <- bind_rows(df_mape_train_EDGE_clst_poly,      df_mape_train_EDGE_p

    df_actpred_train_CORE_clst_poly <- bind_rows(df_actpred_train_CORE_clst_poly,      df_mape_train_C
    df_actpred_train_CENTER_clst_poly <- bind_rows(df_actpred_train_CENTER_clst_poly, df_mape_train_C
    df_actpred_train_EDGE_clst_poly <- bind_rows(df_actpred_train_EDGE_clst_poly,      df_mape_train_E

```

```

df_mape_test_CORE_clst_poly <- bind_rows(df_mape_test_CORE_clst_poly,      df_mape_test_CORE_parce
df_mape_test_CENTER_clst_poly <- bind_rows(df_mape_test_CENTER_clst_poly,      df_mape_test_CENTER
df_mape_test_EDGE_clst_poly <- bind_rows(df_mape_test_EDGE_clst_poly,      df_mape_test_EDGE_parce

}
}

if (poly_order == 1) {

  #df_mape_train_CORE_all <- df_mape_train_CORE_clst_poly
  #df_mape_train_CENTER_all <- df_mape_train_CENTER_clst_poly
  #df_mape_train_EDGE_all <- df_mape_train_EDGE_clst_poly

  df_actpred_train_CORE_all <- df_actpred_train_CORE_clst_poly
  df_actpred_train_CENTER_all <- df_actpred_train_CENTER_clst_poly
  df_actpred_train_EDGE_all <- df_actpred_train_EDGE_clst_poly

  df_mape_test_CORE_all <- df_mape_test_CORE_clst_poly
  df_mape_test_CENTER_all <- df_mape_test_CENTER_clst_poly
  df_mape_test_EDGE_all <- df_mape_test_EDGE_clst_poly

}
else
{

  # df_mape_train_CORE_all <- bind_rows(df_mape_train_CORE_all, df_mape_train_CORE_clst_poly)
  # df_mape_train_CENTER_all <- bind_rows(df_mape_train_CENTER_all, df_mape_train_CENTER_clst_poly)
  # df_mape_train_EDGE_all <- bind_rows(df_mape_train_EDGE_all, df_mape_train_EDGE_clst_poly)

  df_actpred_train_CORE_all <- bind_rows(df_actpred_train_CORE_all, df_actpred_train_CORE_clst_poly)
  df_actpred_train_CENTER_all <- bind_rows(df_actpred_train_CENTER_all, df_actpred_train_CENTER_clst_poly)
  df_actpred_train_EDGE_all <- bind_rows(df_actpred_train_EDGE_all, df_actpred_train_EDGE_clst_poly)

  df_mape_test_CORE_all <- bind_rows(df_mape_test_CORE_all, df_mape_test_CORE_clst_poly)
  df_mape_test_CENTER_all <- bind_rows(df_mape_test_CENTER_all, df_mape_test_CENTER_clst_poly)
  df_mape_test_EDGE_all <- bind_rows(df_mape_test_EDGE_all, df_mape_test_EDGE_clst_poly)

```



[illegible]

[illegible]

[illegible]



[illegible]

[illegible]

[illegible]



[illegible]

[illegible]

[illegible]

[illegible]



[illegible]

```

## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
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## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
## Warning in window.default(x, ...): 'end' value not changed
# print("BUFFER ZONE: Plotting the Actual vs. Fitted for Training set ")
#
#
# for( poly_order in c(seq(1:5))){
#
#   plot_actpred_chart(df_actpred_train_CORE_all, " CORE BUFFER ZONE training set ",poly_order)
#   plot_actpred_chart(df_actpred_train_CENTER_all, " CENTER BUFFER ZONE training set", poly_order)
#   plot_actpred_chart(df_actpred_train_EDGE_all, " EDGE BUFFER ZONE training set ", poly_order)
#
# } # poly_order

print("BUFFER ZONE: Plotting the mape charts ")

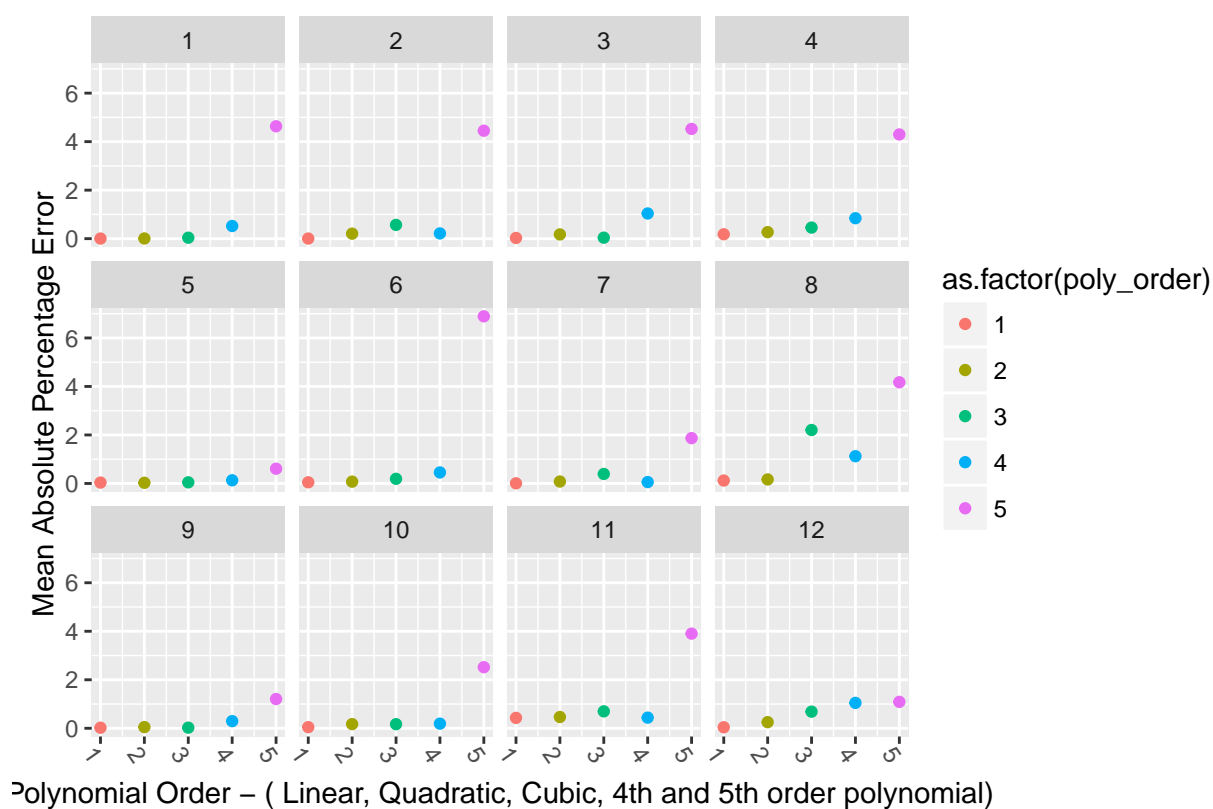
## [1] "BUFFER ZONE: Plotting the mape charts "
# Plot the mape chart for the buffer zones

# plot_mape_chart(df_mape_train_CORE_all, " CORE BUFFER ZONE training set data")
# plot_mape_chart(df_mape_train_CENTER_all, " CENTER BUFFER ZONE training set data")
# plot_mape_chart(df_mape_train_EDGE_all, " EDGE BUFFER ZONE training set data")
#

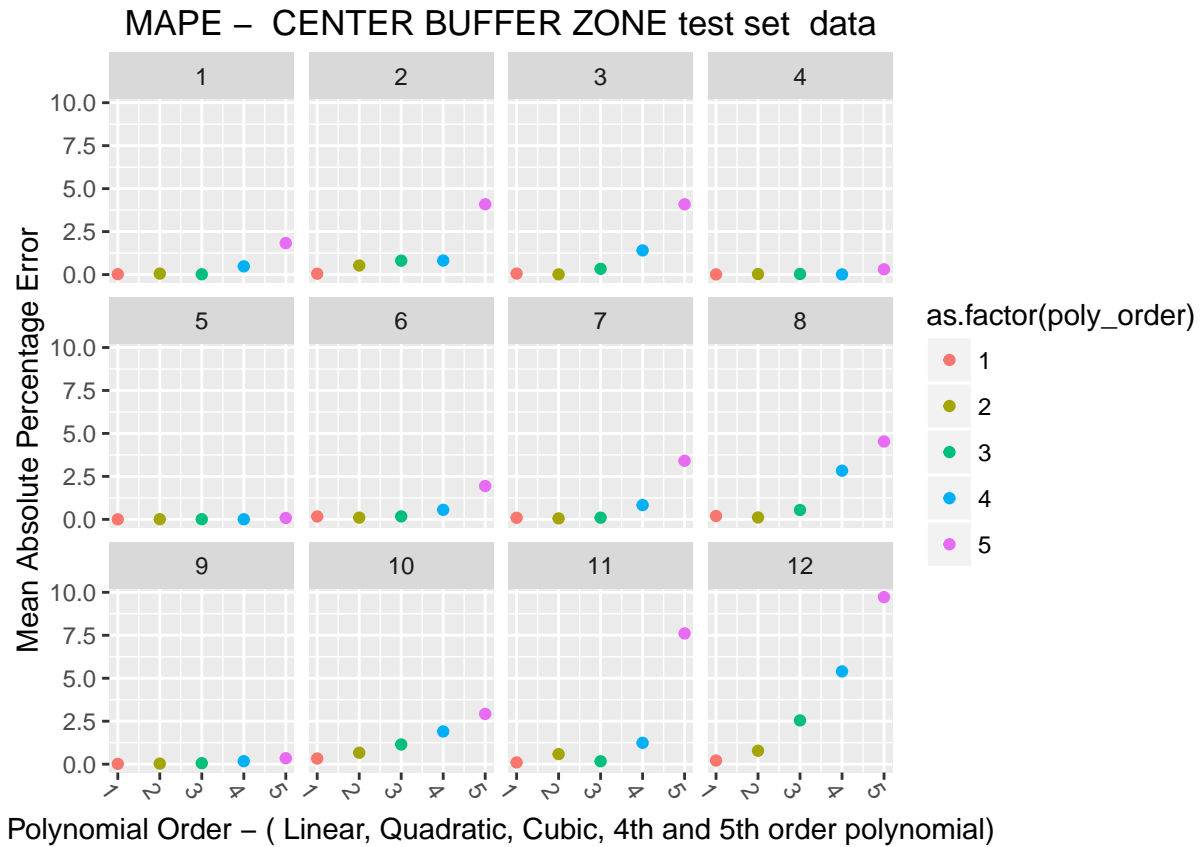
plot_mape_chart(df_mape_test_CORE_all, " CORE BUFFER ZONE test set data")

```

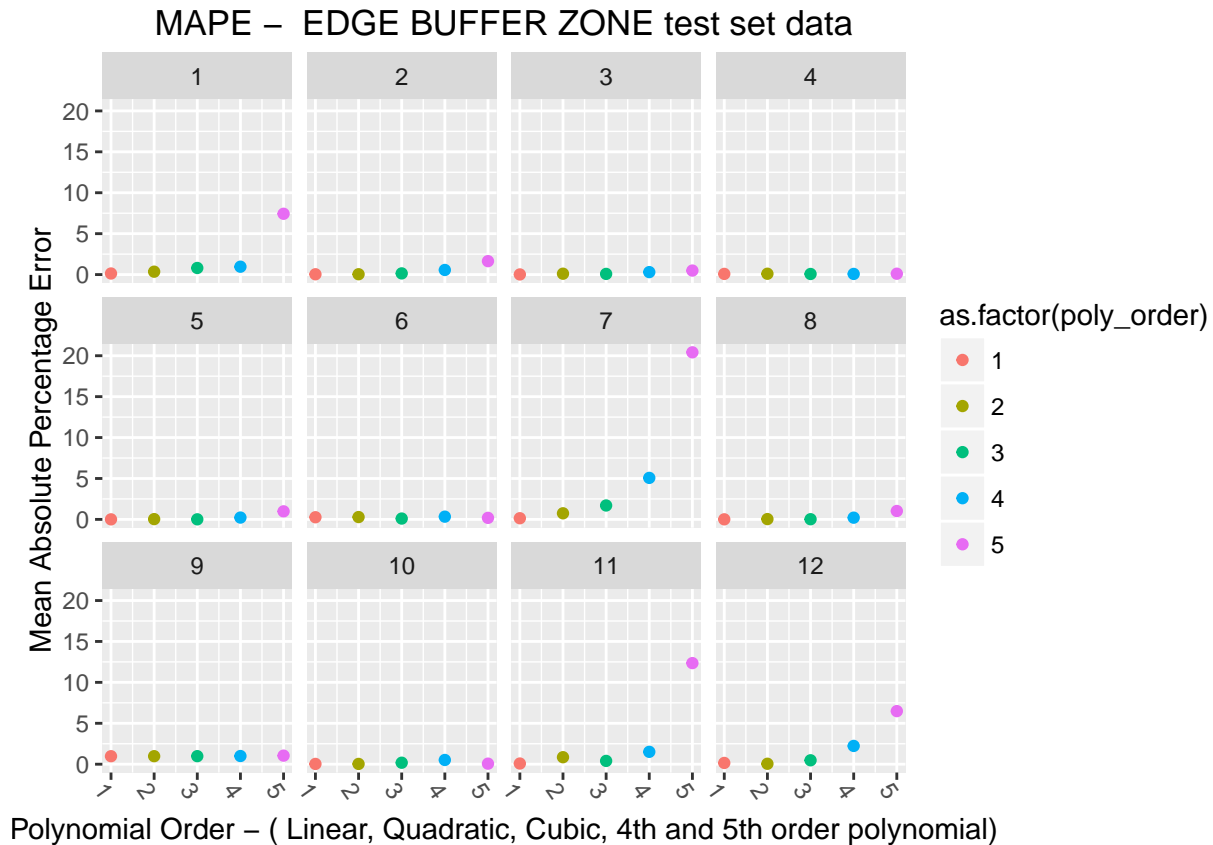
MAPE – CORE BUFFER ZONE test set data



```
plot_mape_chart(df_mape_test_CENTER_all, " CENTER BUFFER ZONE test set data")
```



```
plot_mape_chart(df_mape_test_EDGE_all, " EDGE BUFFER ZONE test set data")
```



```
# Write out to csv files
write.csv(df_mape_test_CORE_all, file = "test_set_mape_CORE.csv" )
write.csv(df_mape_test_CENTER_all, file = "test_set_mape_CENTER.csv" )
write.csv(df_mape_test_EDGE_all, file = "test_set_mape_EDGE.csv" )

print( "BUFFER_ZONE : Faceting over clusters for each zone")

## [1] "BUFFER_ZONE : Faceting over clusters for each zone"

# facet the clusters on a single plot
# visualize fitted and predicted

#####
# EDGE
#####

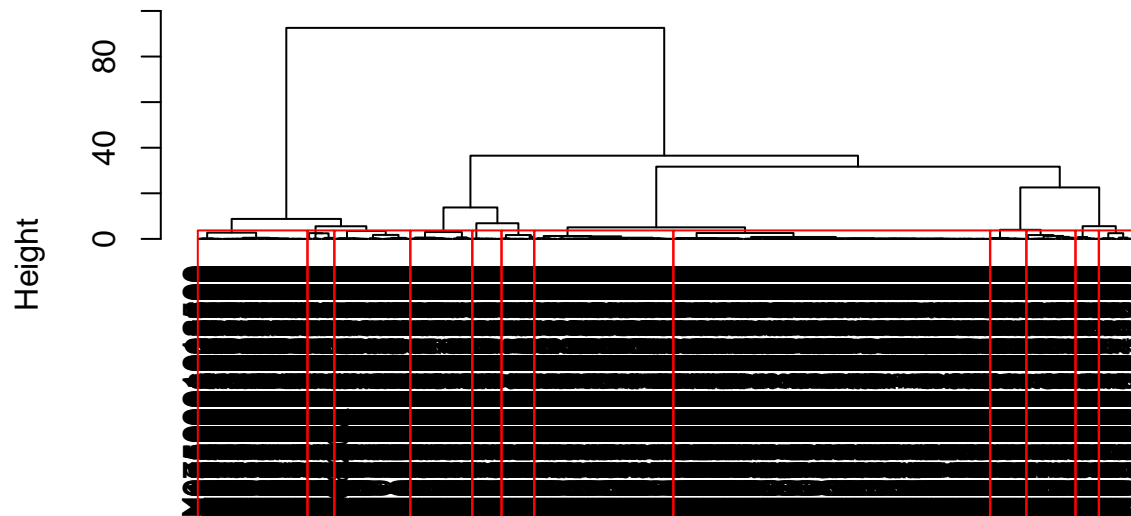
# here all of the Buffer zone parcel ids are partitioned into clusters

# Find the number of observations
n <- ncol(taxinfo_EDGE.matrix_no_na)
# Shuffle it
ntaxinfo_EDGE.matrix_shuffled <- taxinfo_EDGE.matrix_no_na[,sample(n) ]
train_indices <- 1:round(1.0 * n)
train <- taxinfo_EDGE.matrix_shuffled[,train_indices ]

# perform Clustering of Vars / dimension reduction
ClustvarEDGE100 <-hclustvar(train)
```

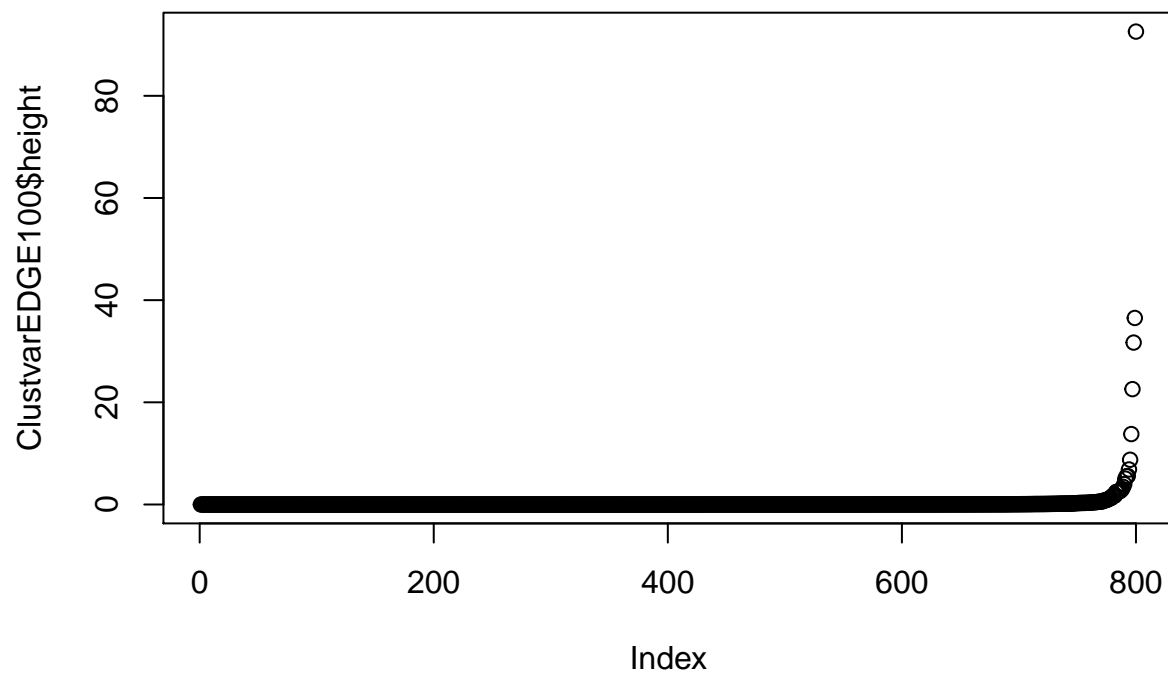
```
# plot the dendrogram
plot(ClustvarEDGE100)
rect.hclust(ClustvarEDGE100,k=12, border="red")
```

## Cluster Dendrogram



```
partEDGE100 <- cutreevar(ClustvarEDGE100,12)

a_EDGE <- partEDGE100$var
# Height of ClustVarEDGE is plotted
plot(ClustvarEDGE100$height)
```



```
#####
# CENTER
#####
```

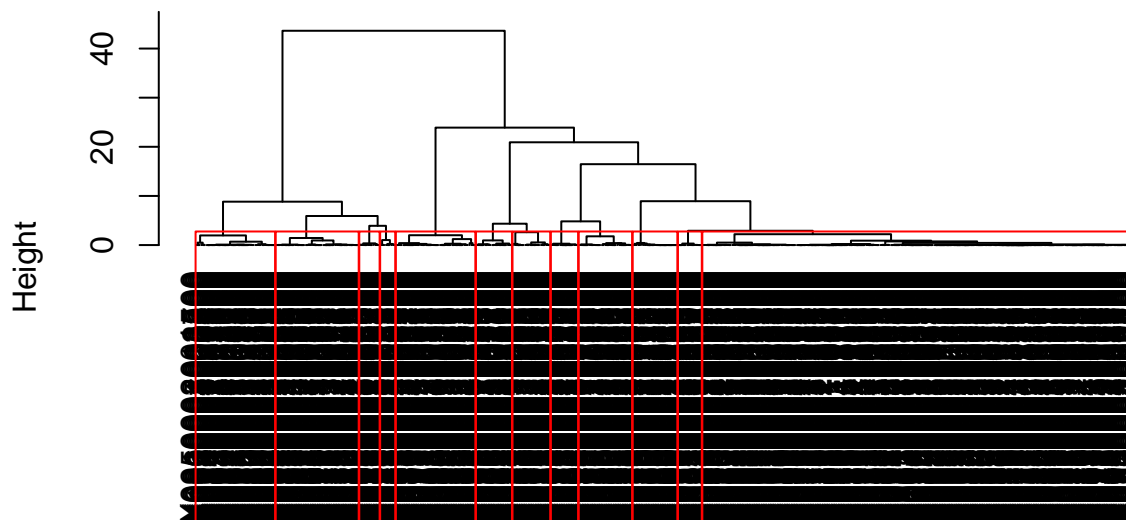
```

# Find the number of observations and shuffle it
n <- ncol(taxinfo_CENTER.matrix_no_na)
taxinfo_CENTER.matrix_shuffled <- taxinfo_CENTER.matrix_no_na[,sample(n) ]
#Take all of the observations
train_indices <- 1:round(1.0 * n)
train <- taxinfo_CENTER.matrix_shuffled[,train_indices ]

# create a dendrogram
ClustvarCENTER100 <-hclustvar(train)
plot(ClustvarCENTER100)
rect.hclust(ClustvarCENTER100,k=12, border="red")

```

## Cluster Dendrogram



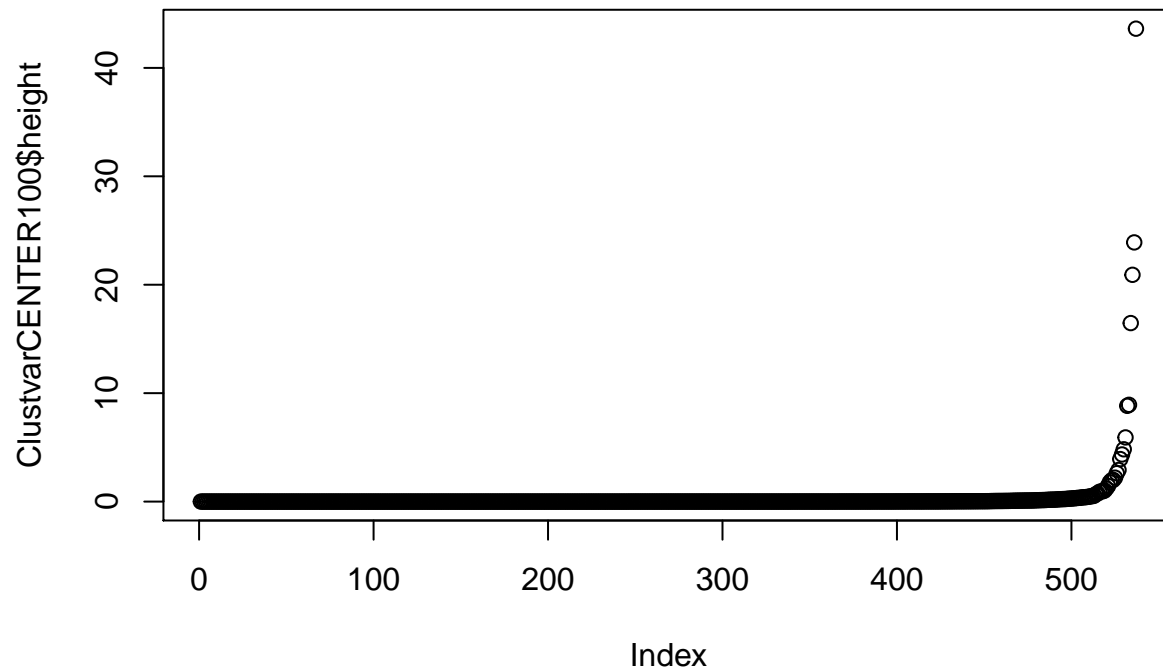
```

partCENTER100 <- cutreevar(ClustvarCENTER100,12)

a_CENTER <- partCENTER100$var

# plot the height
plot(ClustvarCENTER100$height)

```



```
#####
# CORE
#####

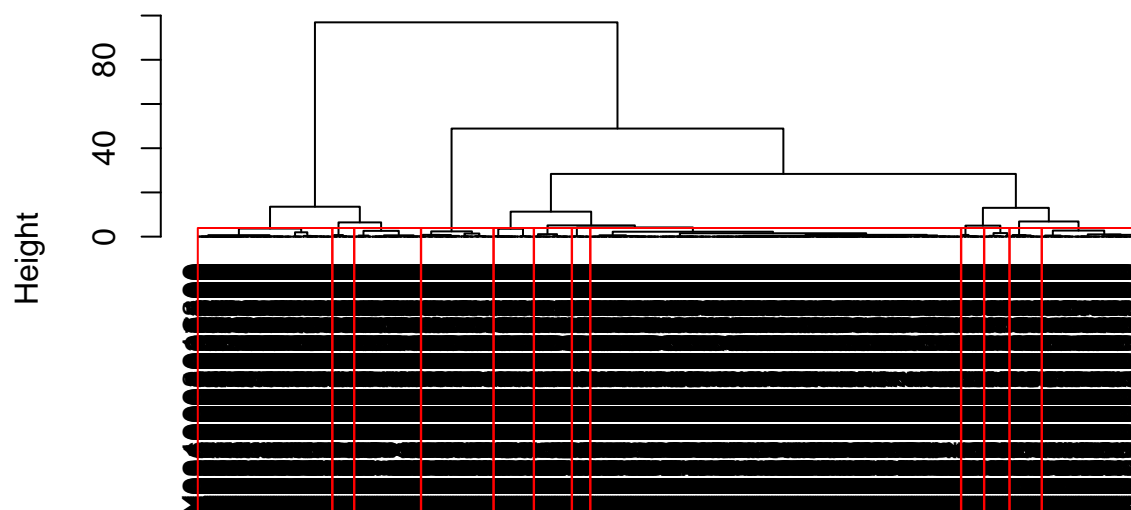
#Take all of the observations and shuffle them
n <- ncol(taxinfo_CORE.matrix_no_na)
taxinfo_CORE.matrix_shuffled <- taxinfo_CORE.matrix_no_na[,sample(n) ]

train_indices <- 1:round(1.0 * n)
train <- taxinfo_CORE.matrix_shuffled[,train_indices ]

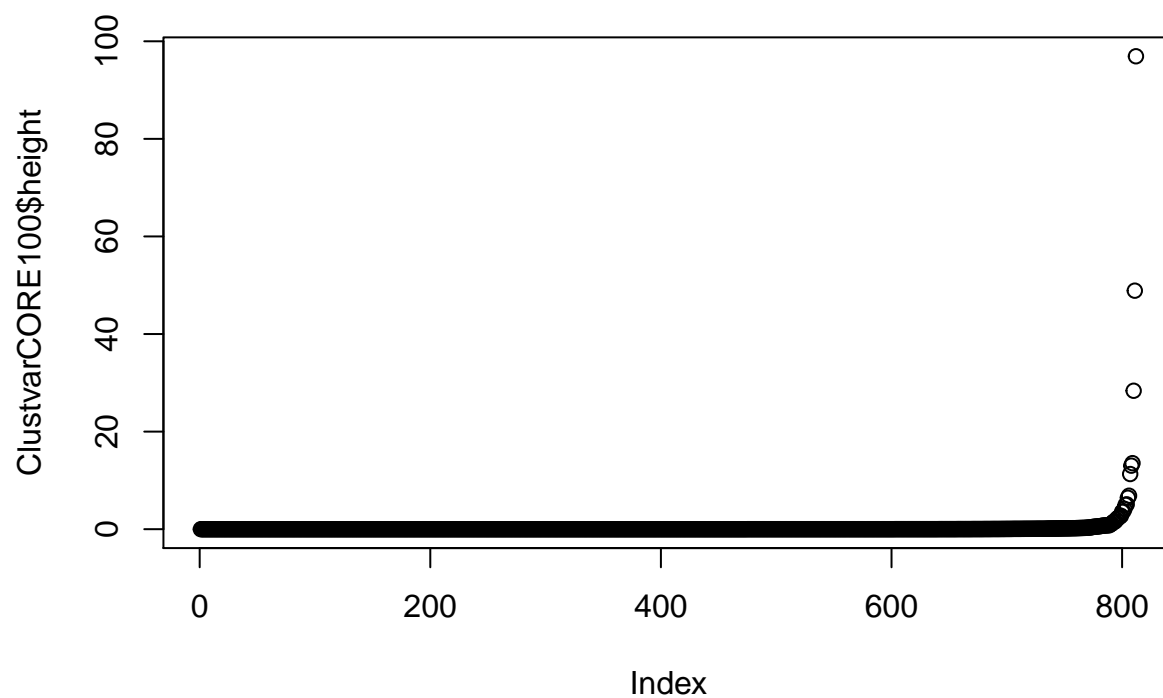
# Perform Dimension reduction
ClustvarCORE100 <-hclustvar(train)
plot(ClustvarCORE100)
partCORE100 <- cutreevar(ClustvarCORE100,12)
rect.hclust(ClustvarCORE100,k=12, border="red")
```



## Cluster Dendrogram



```
a_CORE <- partCORE100$var
plot(ClustvarCORE100$height)
```



```
#stab100 <- stability(Clustvar100,B = 60, graph = TRUE)
print("BUFFER_ZONE Completed dendograms for all the parcels")

## [1] "BUFFER_ZONE Completed dendograms for all the parcels"
print("BUFFER_ZONE - Source Data and Predicted in the same plot")

## [1] "BUFFER_ZONE - Source Data and Predicted in the same plot"
```

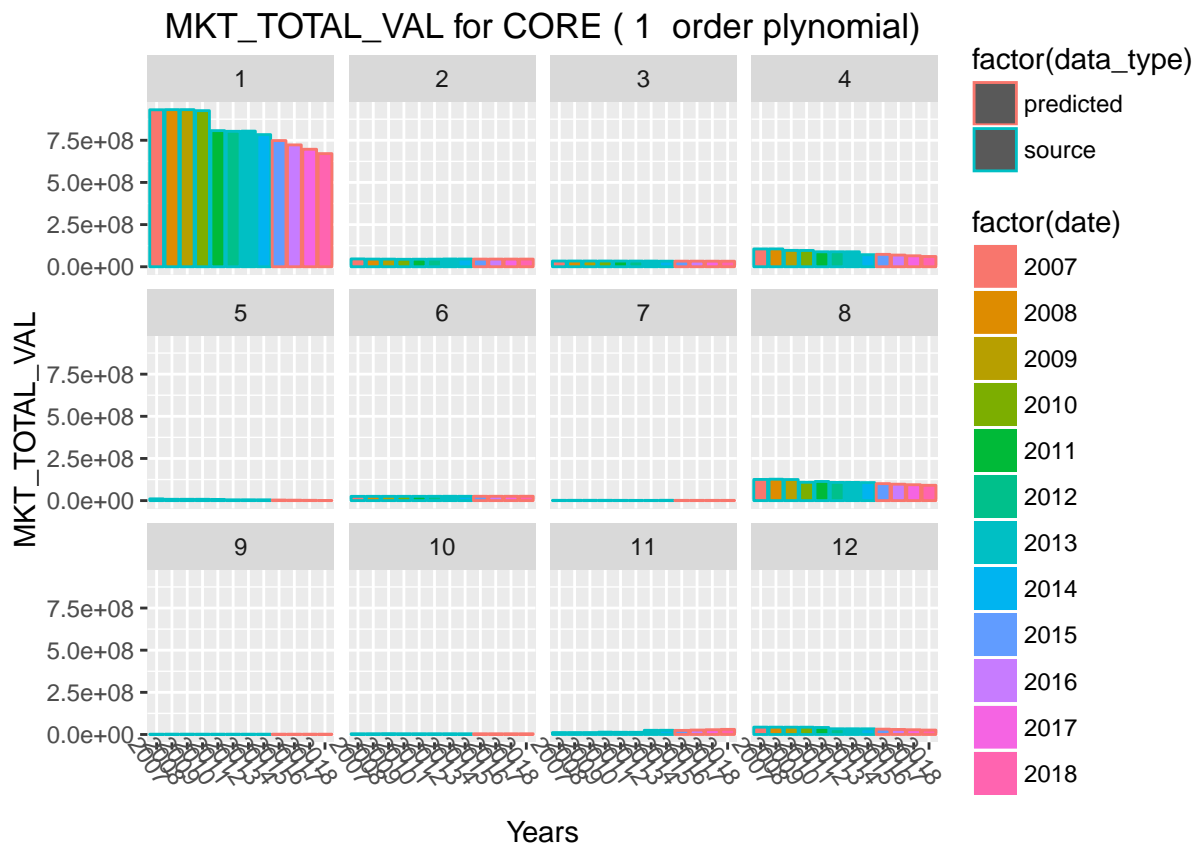
```

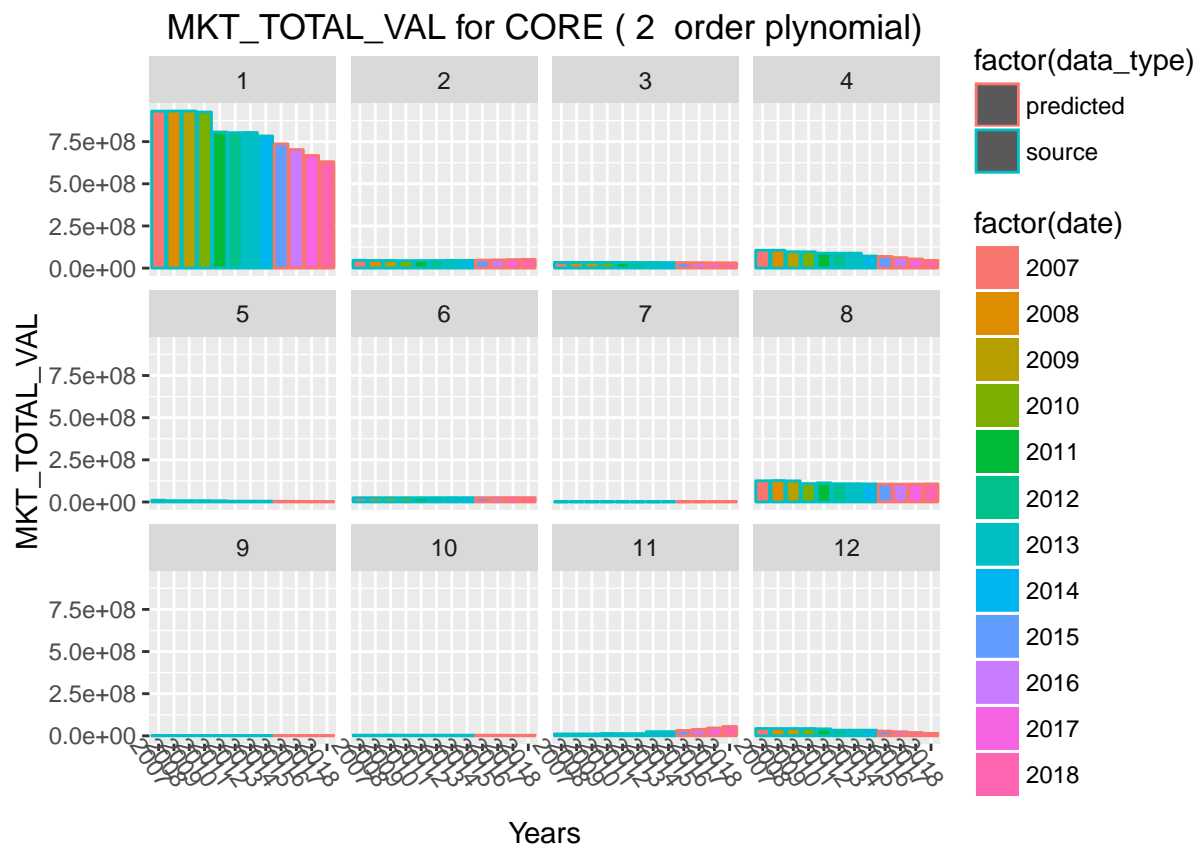
for( poly_order in c(seq(1:5))) {

  # facet the clusters on a single plot
  # visualize source and predicted

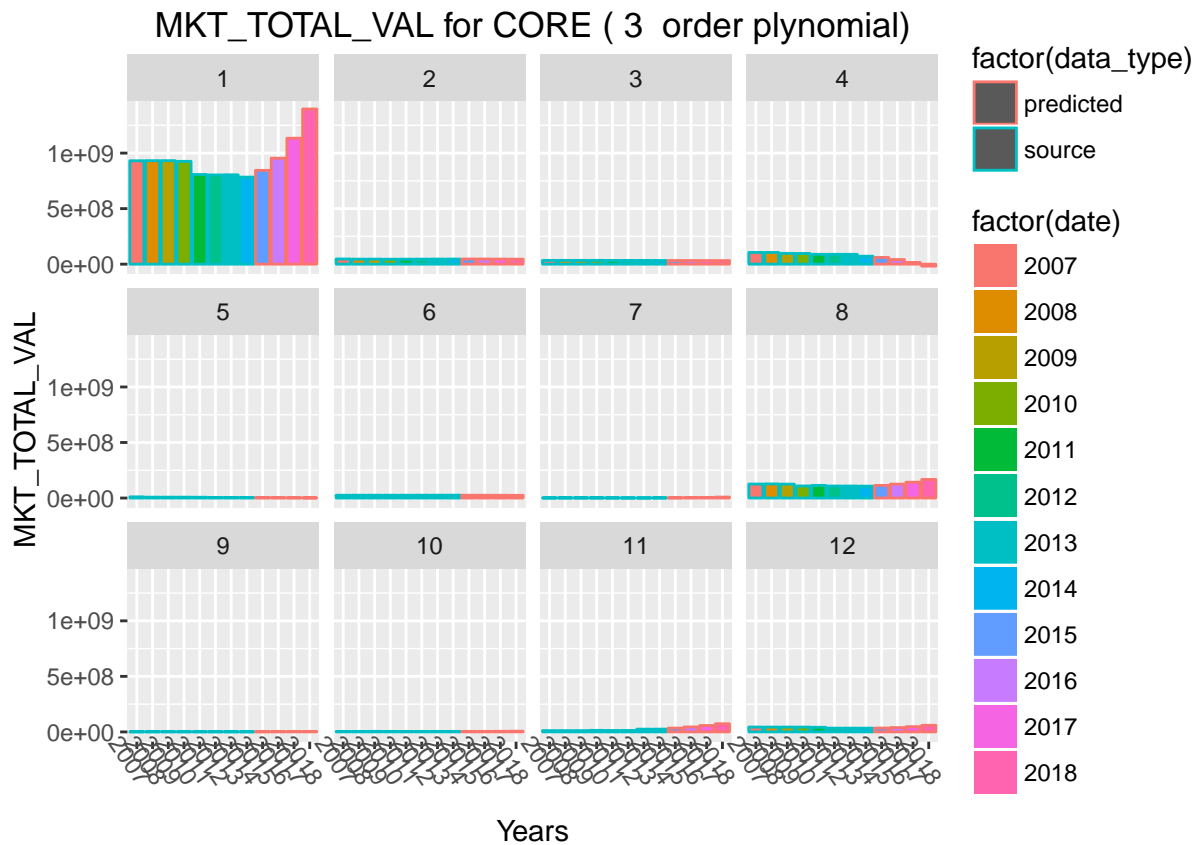
  df_final_mkt_val_core <- non_linear_regression_model(a_CORE, taxinfo_CORE.matrix, poly_order, 12, "CORE")
  if (( poly_order == 1) && (skip_non_buffer == 1)) {
    df_final_mkt_val <- df_final_mkt_val_core
  } else
  { df_final_mkt_val <- bind_rows(df_final_mkt_val, df_final_mkt_val_core )}
}

```

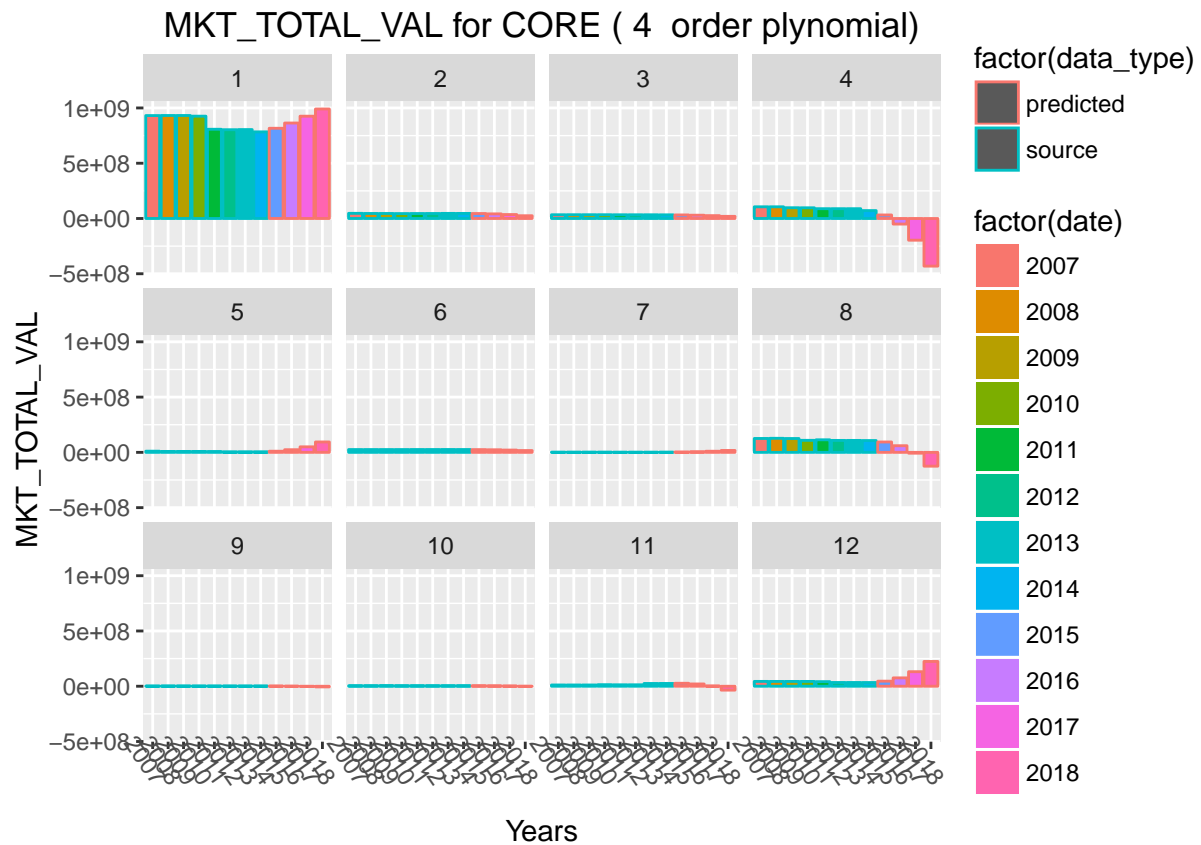




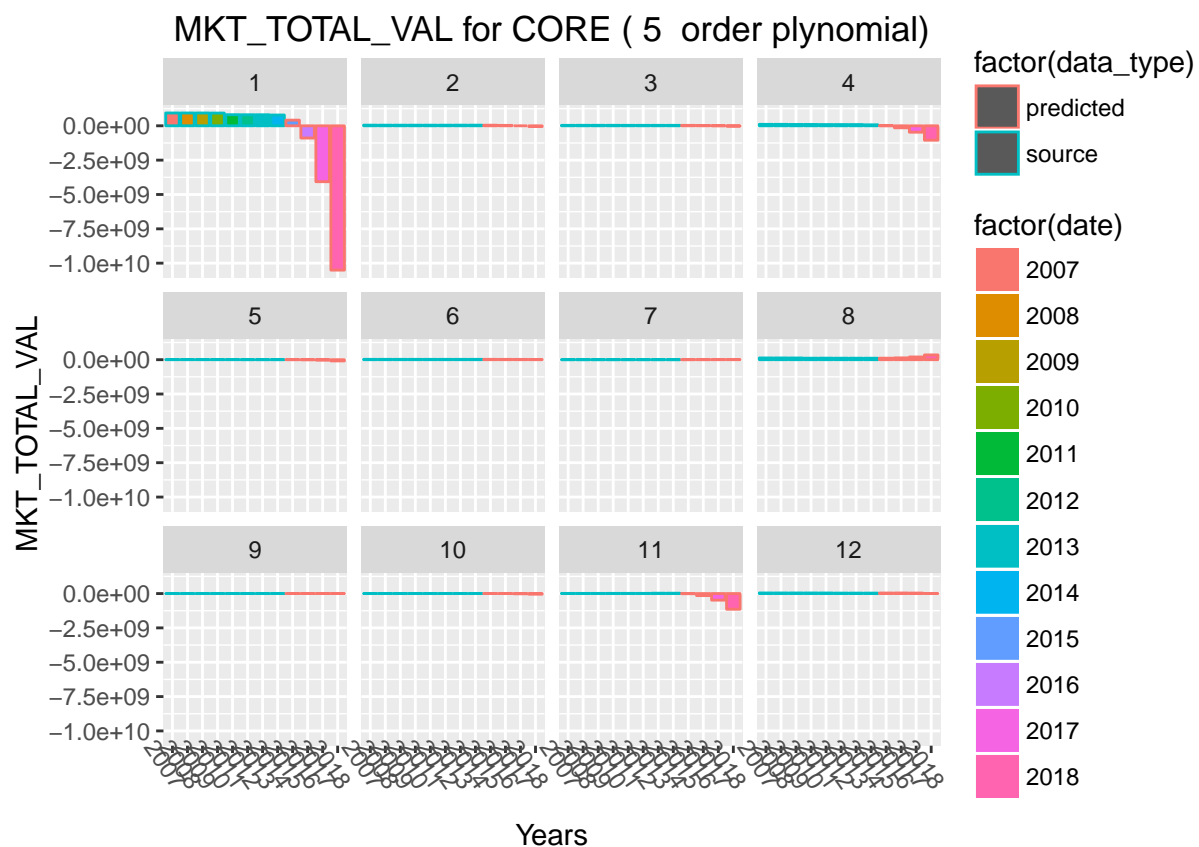
## Warning: Stacking not well defined when ymin != 0



## Warning: Stacking not well defined when ymin != 0

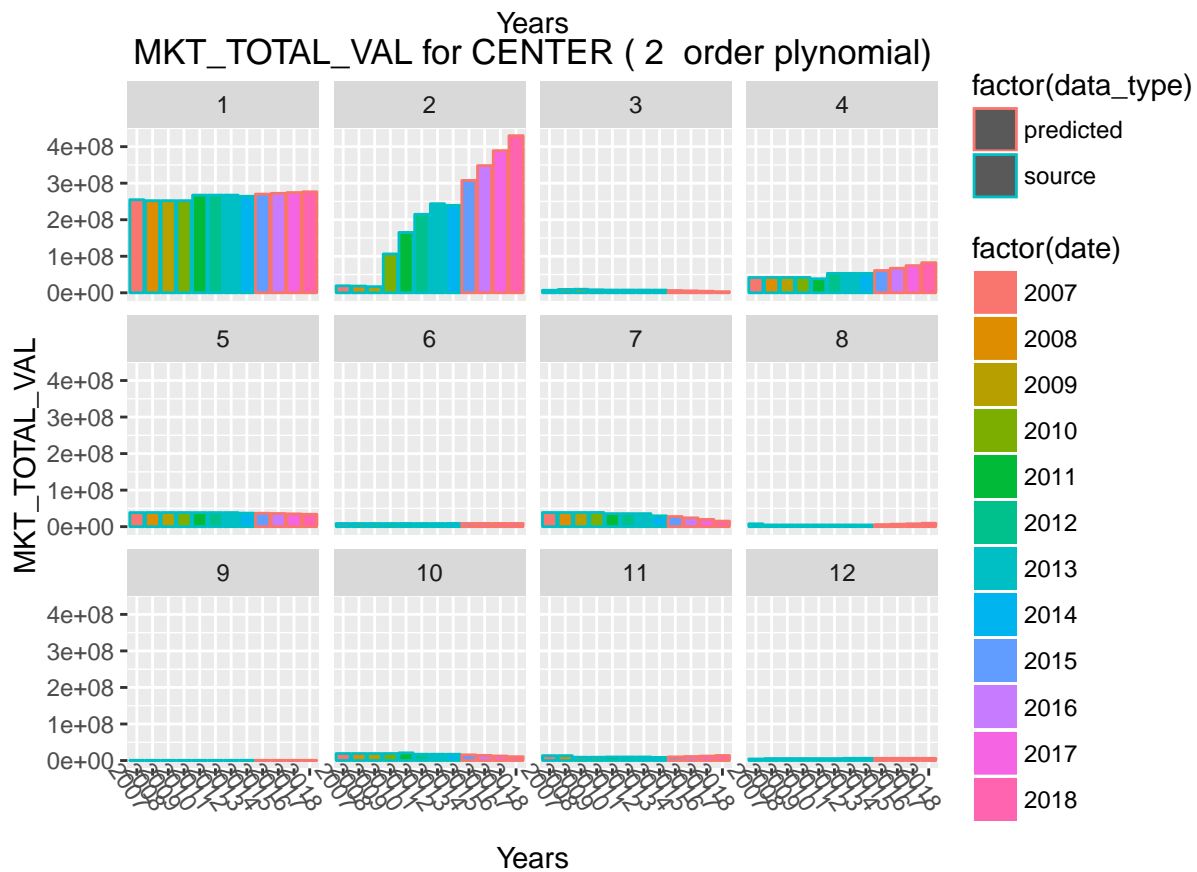
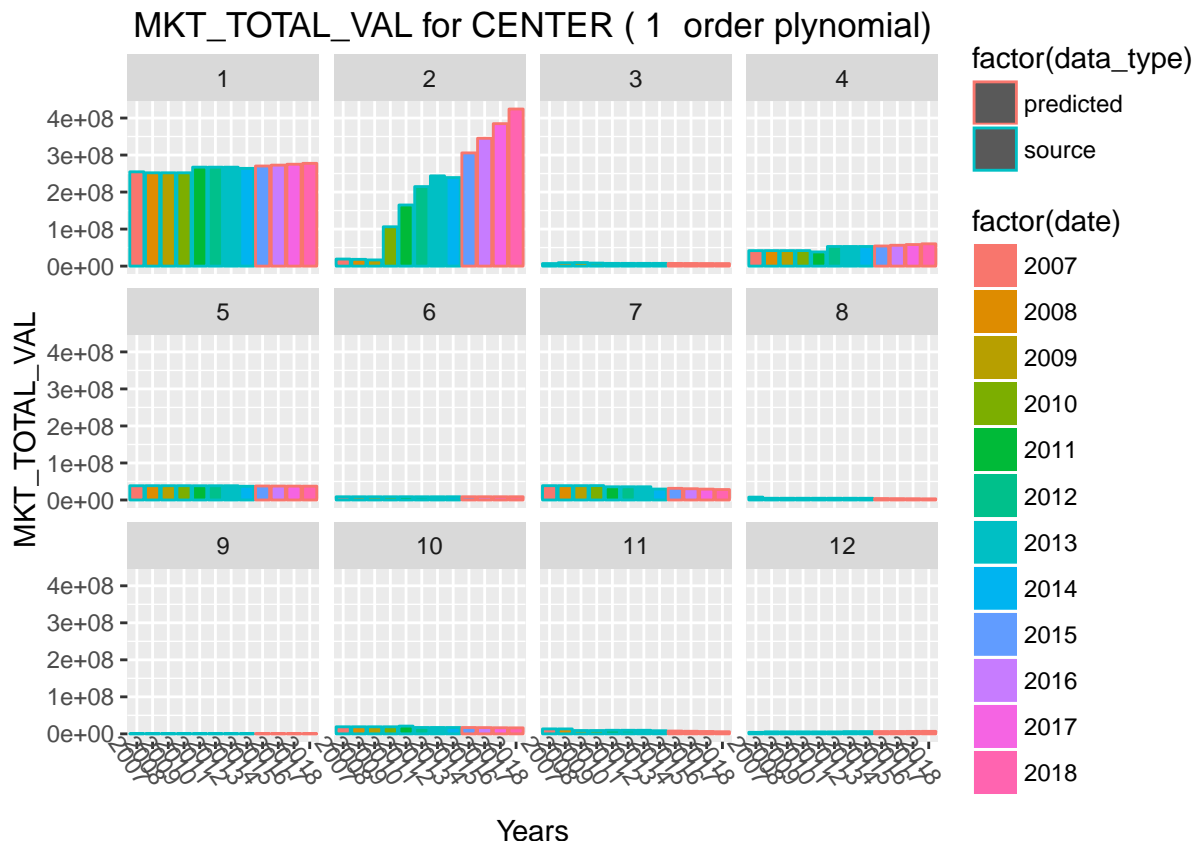


## Warning: Stacking not well defined when ymin != 0

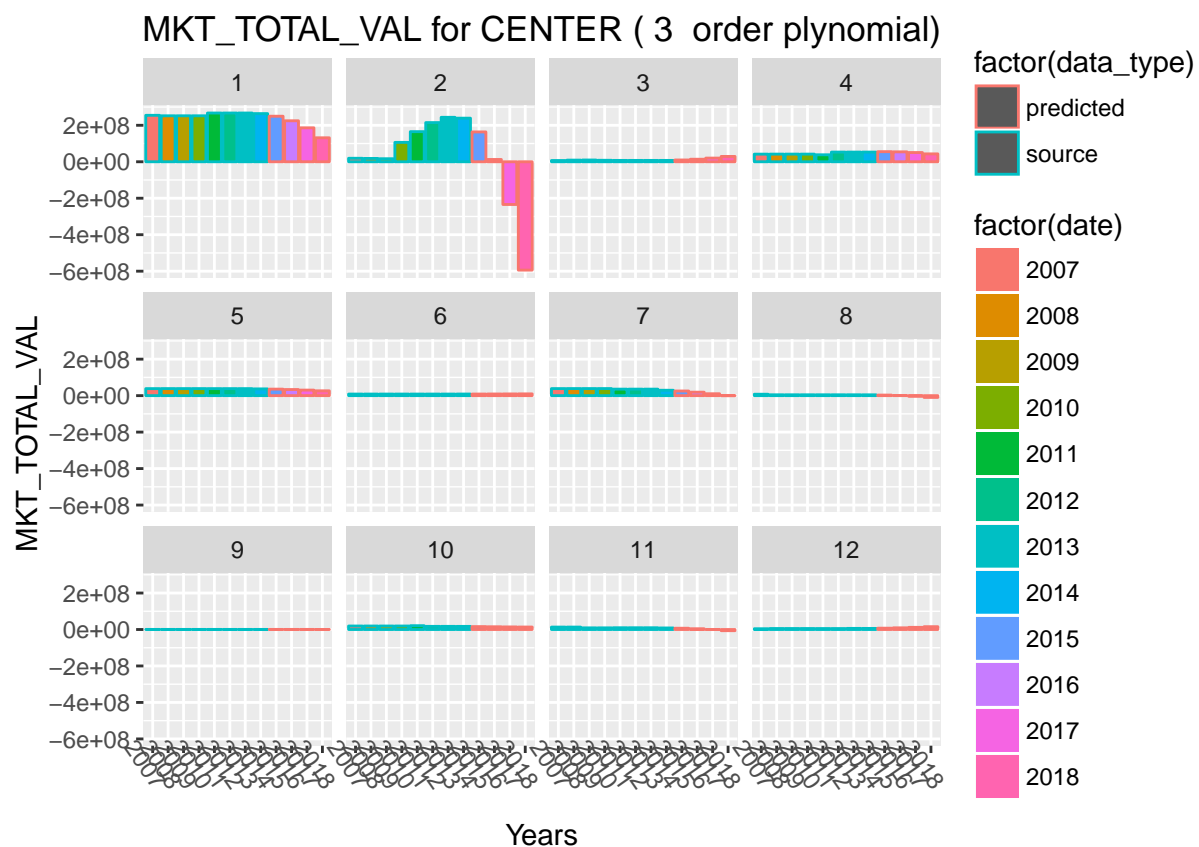


```
for( poly_order in c(seq(1:5))) {
  # facet the clusters on a single plot
  # visualize source and predicted

  df_final_mkt_val_center <- non_linear_regression_model(a_CENTER, taxinfo_CENTER.matrix, poly_order, 12,
  df_final_mkt_val <- bind_rows(df_final_mkt_val, df_final_mkt_val_center )
}
```

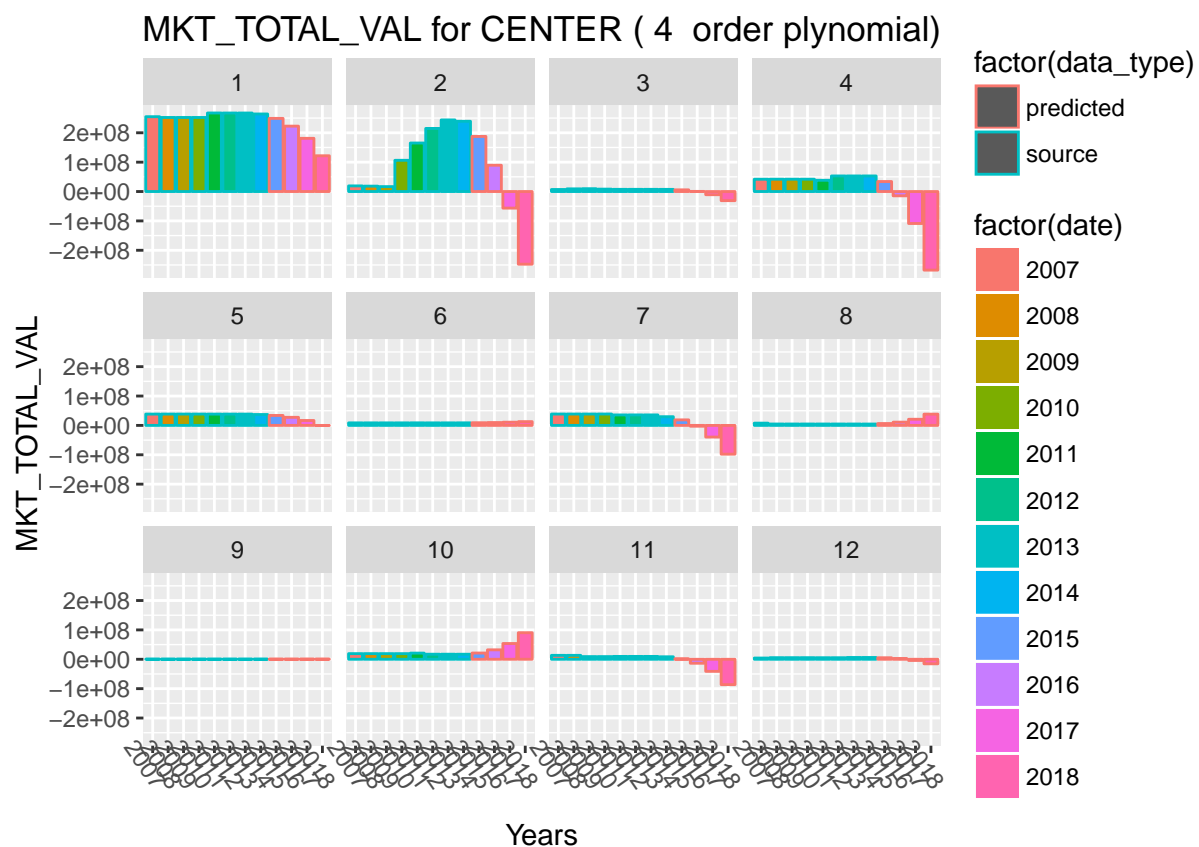


## Warning: Stacking not well defined when ymin != 0

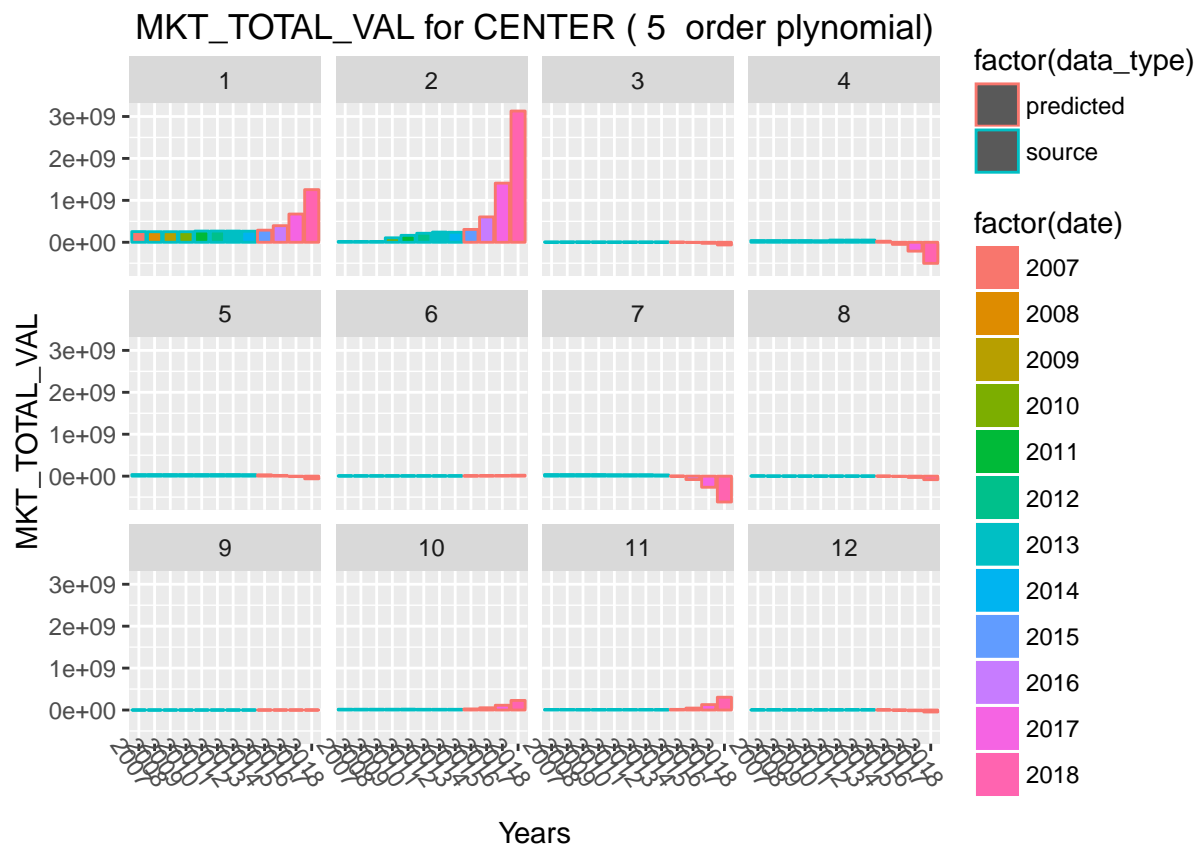


## Warning: Stacking not well defined when ymin != 0

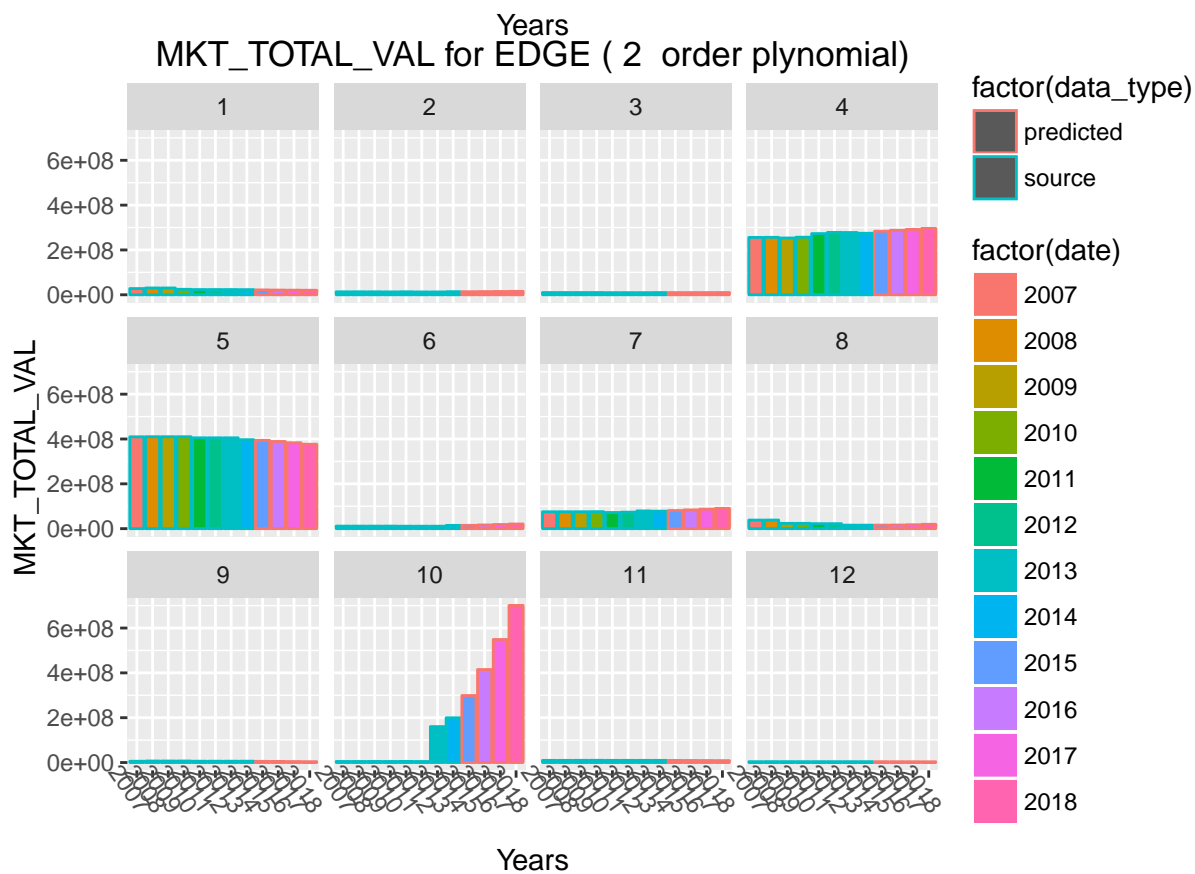
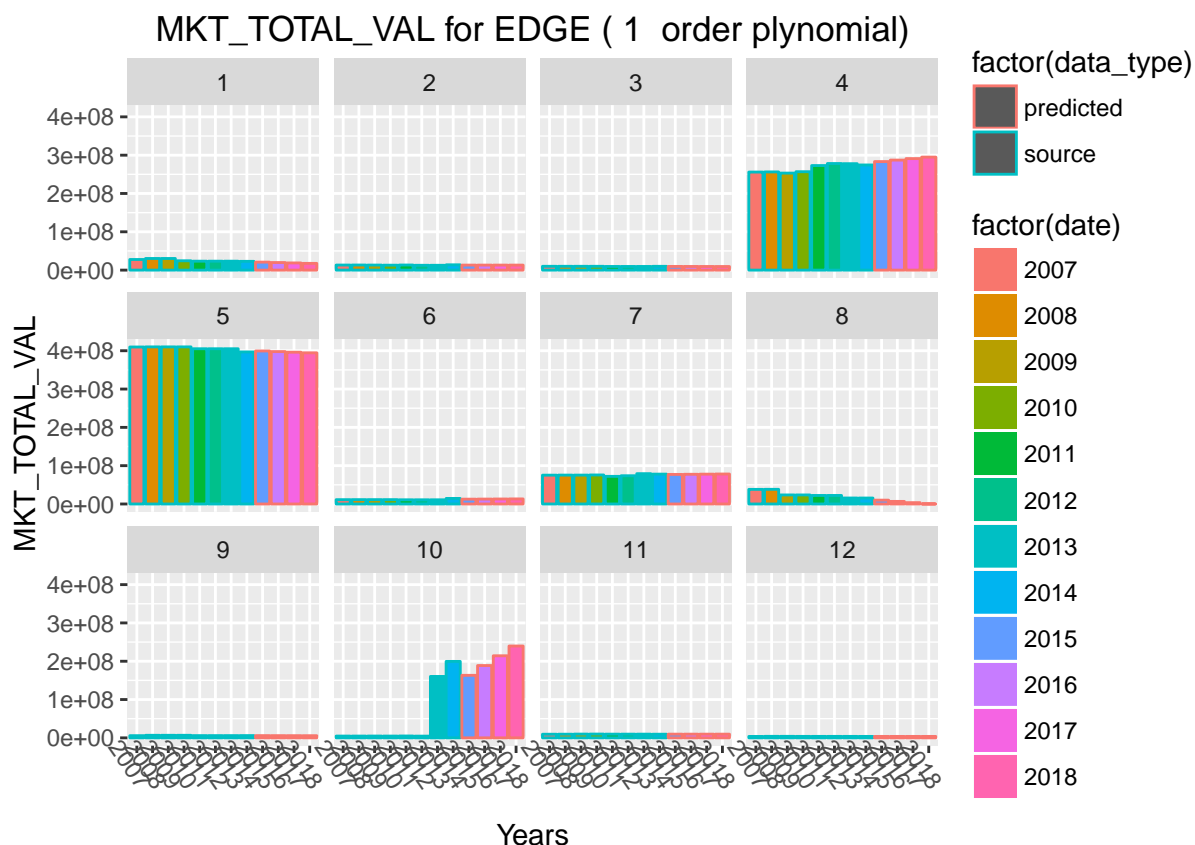




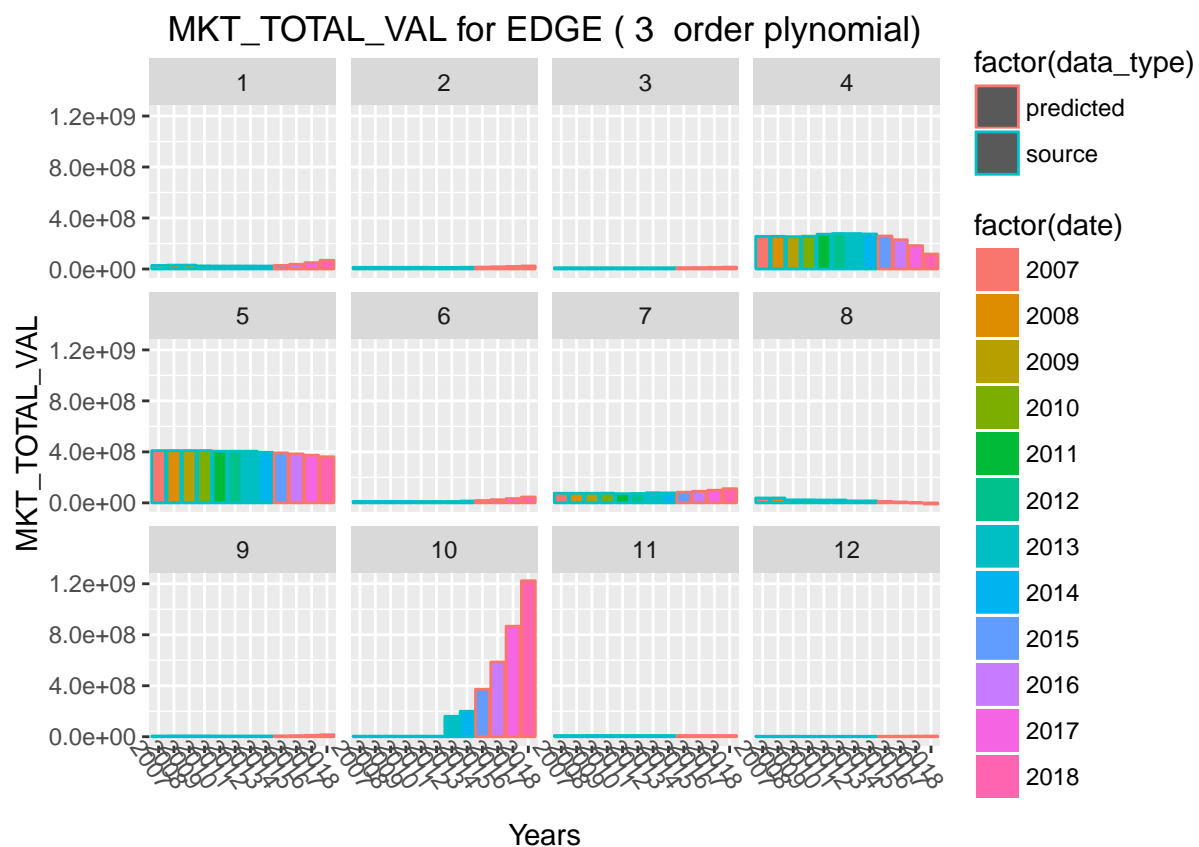
## Warning: Stacking not well defined when ymin != 0



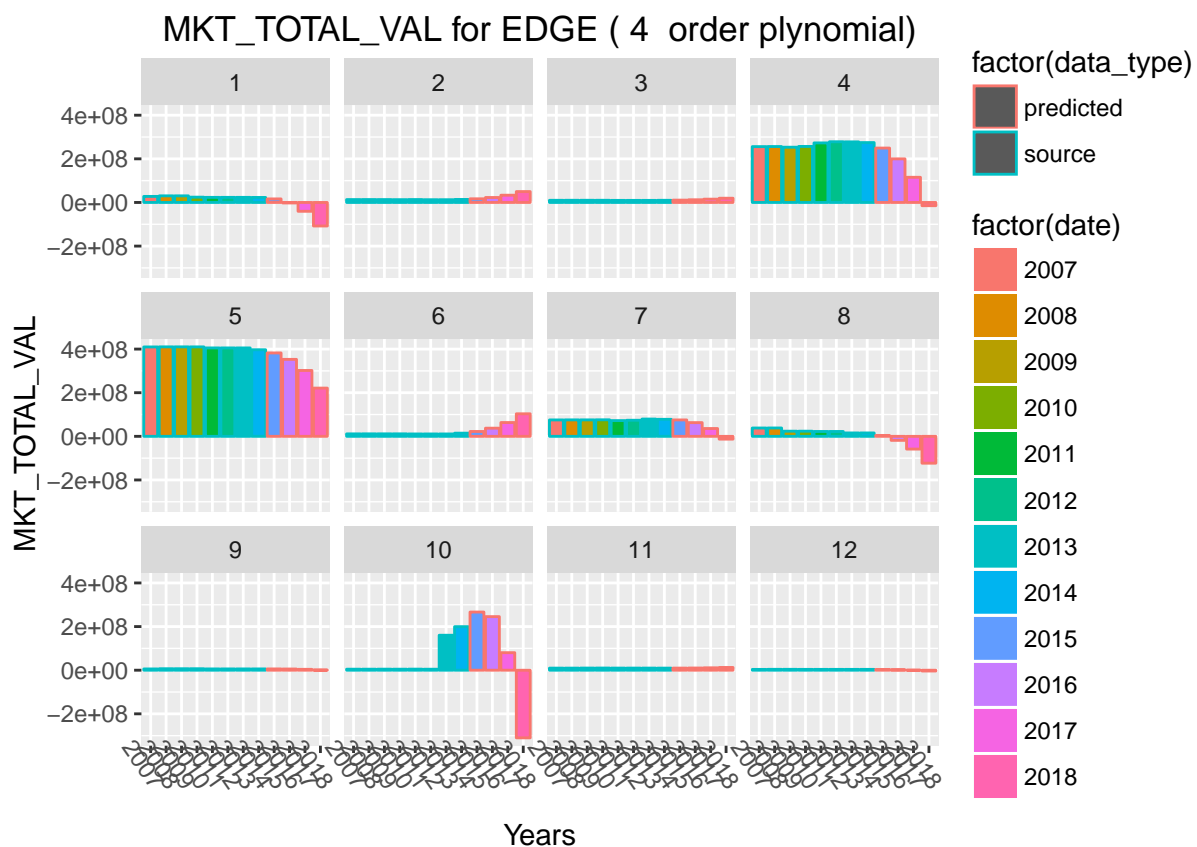
```
for( poly_order in c(seq(1:5))) {
  # facet the clusters on a single plot
  # visualize source and predicted
  df_final_mkt_val_edge <- non_linear_regresion_model(a_EDGE, taxinfo_EDGE.matrix, poly_order, 12, "EDGE")
  df_final_mkt_val <- bind_rows(df_final_mkt_val, df_final_mkt_val_edge )
}
```



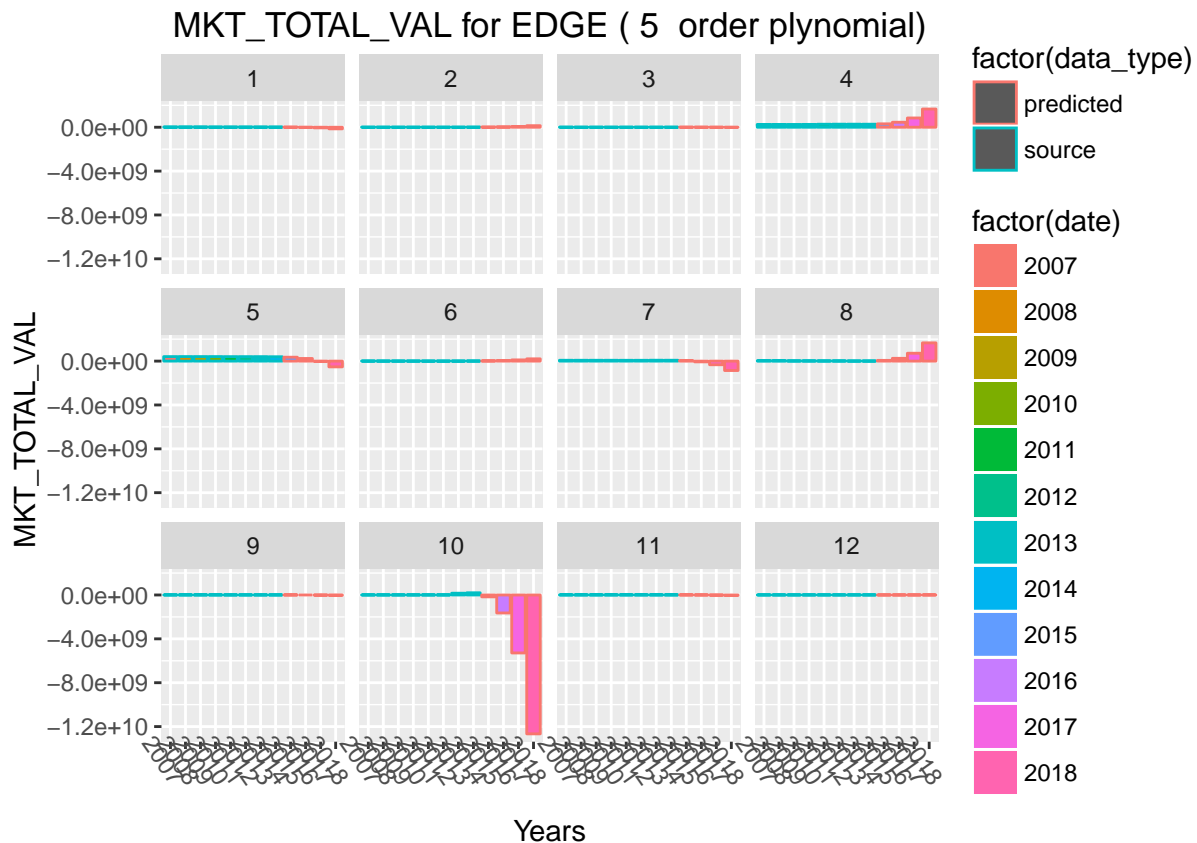
## Warning: Stacking not well defined when ymin != 0



## Warning: Stacking not well defined when ymin != 0



## Warning: Stacking not well defined when ymin != 0



```
# For the plots below we need to make the zone a factor
df_final_mkt_val$zone <- as.factor(df_final_mkt_val$zone)
df_final_mkt_val$data_type <- as.factor(df_final_mkt_val$data_type)
df_final_mkt_val$date <- as.factor(df_final_mkt_val$date)

# Also find the mkt_total/ cluster = ( mkt_val_mean/cluster * num_parcel_ids/cluster)
df_final_mkt_val <- df_final_mkt_val %>% mutate(mkt_total = mkt_val * num_parcel_ids)

# Write out to csv files
write.csv(df_final_mkt_val, file = "final_mkt_val.csv" )

print(" ZONE UNDER STUDY : Final plot ")

## [1] " ZONE UNDER STUDY : Final plot "
```

```
n3 <- NULL
# Data Layer
#title <- paste0("MKT TOTAL VALUE in different ZONES ")

title <- paste0("MKT TOTAL VALUE in different ZONES ")

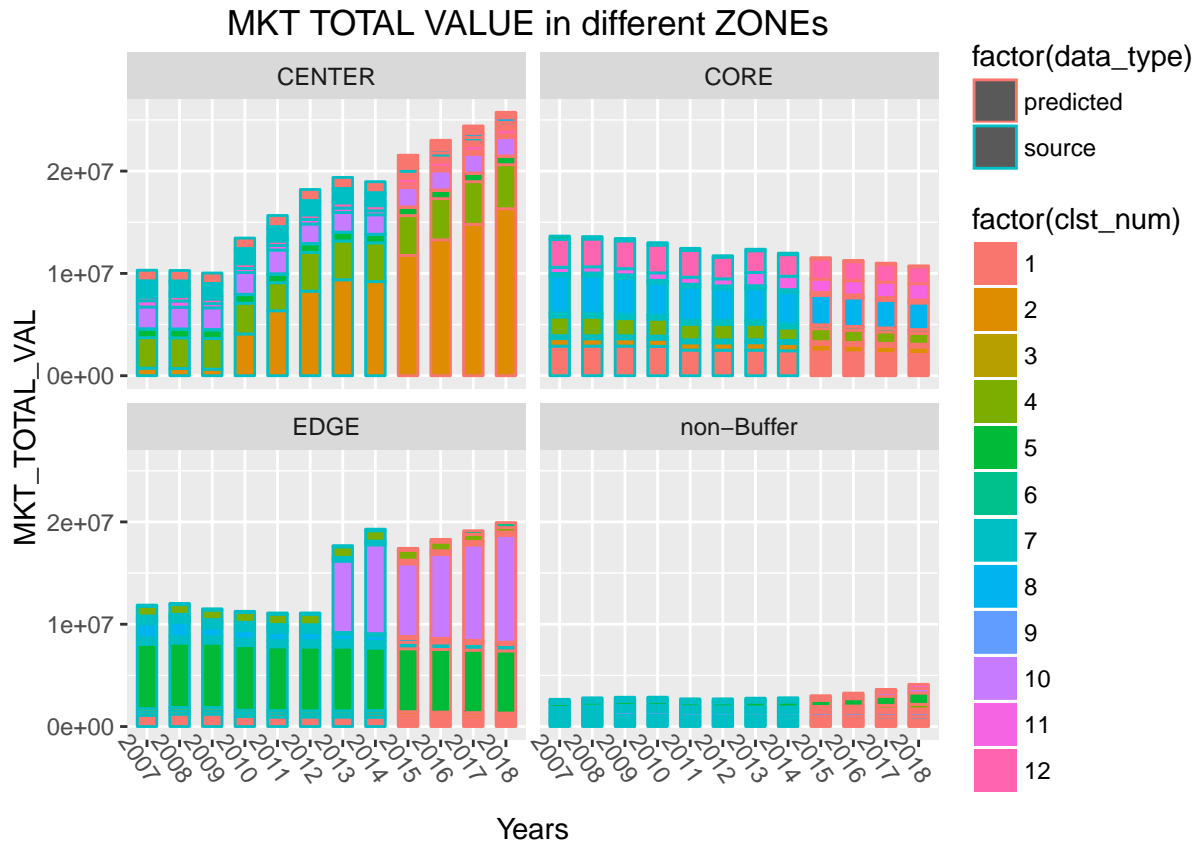
n3 <- (ggplot)
#
# # Data + Aesthetics Mapping
```

```

#n3 <- ggplot(df_final_mkt_val,aes(y = mkt_total, x = factor(date), fill = factor(clst_num), color=factor(
n3 <- ggplot(df_final_mkt_val,aes(y = mkt_val, x = factor(date), fill = factor(clst_num), color=factor(

#n3 <- ggplot(df_final_mkt_val,aes(y = mkt_val, x = date ))
#
# # Data + aes + Geometries
n3 <- n3 + geom_bar(stat="identity", width = 0.6)
#n3 <- n3 + geom_line()
#
# # Data + Aesthetic Mapping + Geom + Facets
# #n <- n + facet_grid(. ~ Department.Title )
#
# #Data + Aesthetics + Geoms + Facets + Statistics
# # ... add a linear regression model here
#n3 <- n3 + geom_smooth(method="gam" )
# #Data + Aesthetics + Geoms + Facets + Statistics + Co-ordinates
# #n <- n + scale_y_discrete(limits=c("00000","120000"), breaks=seq(00000,120000,10000))
n3 <- n3 + scale_x_discrete()
n3 <- n3 + xlab("Years")
n3 <- n3 + ylab("MKT_TOTAL_VAL ")
#
# #Data + Aesthetics + Geoms + Facets + Statistics + Co-ordinates + Theme
n3 <- n3 + theme(axis.text.x = element_text(angle=305))
#
n3 <- n3 + ggtitle(title)
n3 <- n3 +facet_wrap( ~ zone)
print(n3)

```



```
## Split it
df_fin_1 <- df_final_mkt_val %>% filter((zone %in% c("CORE","non-Buffer")))

df_fin_2 <- df_final_mkt_val %>% filter((zone %in% c("CENTER","EDGE")))

n3 <- NULL
# Data Layer

title <- paste0("MKT TOTAL VALUE in different ZONES CENTER and EDGE ")

n3 <- (ggplot)
#
# # Data + Aesthetics Mapping
#n3 <- ggplot(df_final_mkt_val, aes(y = mkt_total, x = factor(date), fill = factor(clst_num), color=factor(data_type)))
n3 <- ggplot(df_fin_2, aes(y = mkt_val, x = factor(date), fill = factor(clst_num), color=factor(data_type)))

#n3 <- ggplot(df_final_mkt_val, aes(y = mkt_val, x = date ))
#
# # Data + aes + Geometries
n3 <- n3 + geom_bar(stat="identity", width = 0.6)
#n3 <- n3 + geom_line()
#
# # Data + Aesthetic Mapping + Geom + Facets
# #n <- n + facet_grid(. ~ Department.Title )
#
```

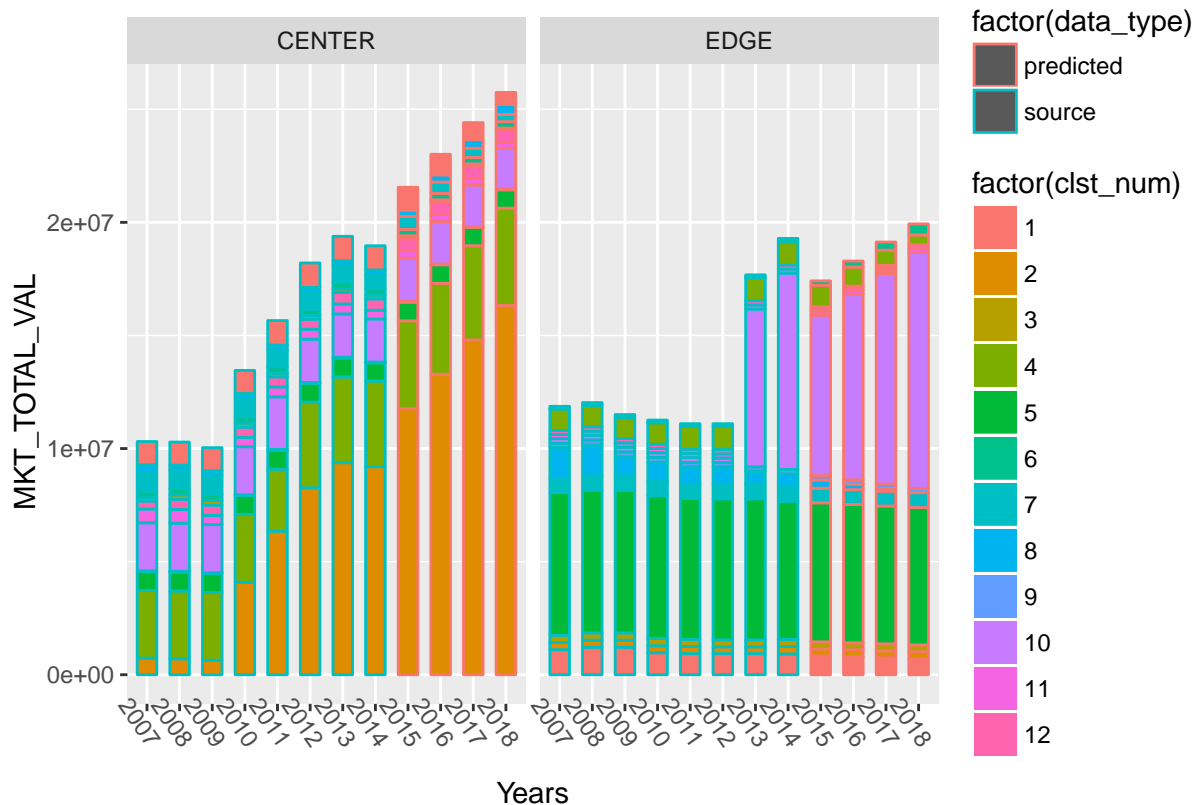


```

# #Data + Aesthetics + Geoms + Facets + Statistics
# # ... add a linear regression model here
#n3 <- n3 + geom_smooth(method="gam" )
# #Data + Aesthetics + Geoms + Facets + Statistics + Co-ordinates
# #n <- n + scale_y_discrete(limits=c("00000","120000"), breaks=seq(00000,120000,10000))
n3 <- n3 + scale_x_discrete()
n3 <- n3 + xlab("Years")
n3 <- n3 + ylab("MKT_TOTAL_VAL ")
#
# #Data + Aesthetics + Geoms + Facets + Statistics + Co-ordinates + Theme
n3 <- n3 + theme(axis.text.x = element_text(angle=305))
#
n3 <- n3 + ggtitle(title)
n3 <- n3 + facet_wrap( ~ zone)
print(n3)

```

### MKT TOTAL VALUE in different ZONES CENTER and EDGE



```

n3 <- NULL
# Data Layer
#title <- paste0("MKT TOTAL VALUE in different ZONES ")

title <- paste0("MKT TOTAL VALUE in different ZONES CORE and non-Buffer ")

n3 <- (ggplot)
#
# # Data + Aesthetics Mapping
#n3 <- ggplot(df_final_mkt_val, aes(y = mkt_total, x = factor(date), fill = factor(clst_num), color=fact

```

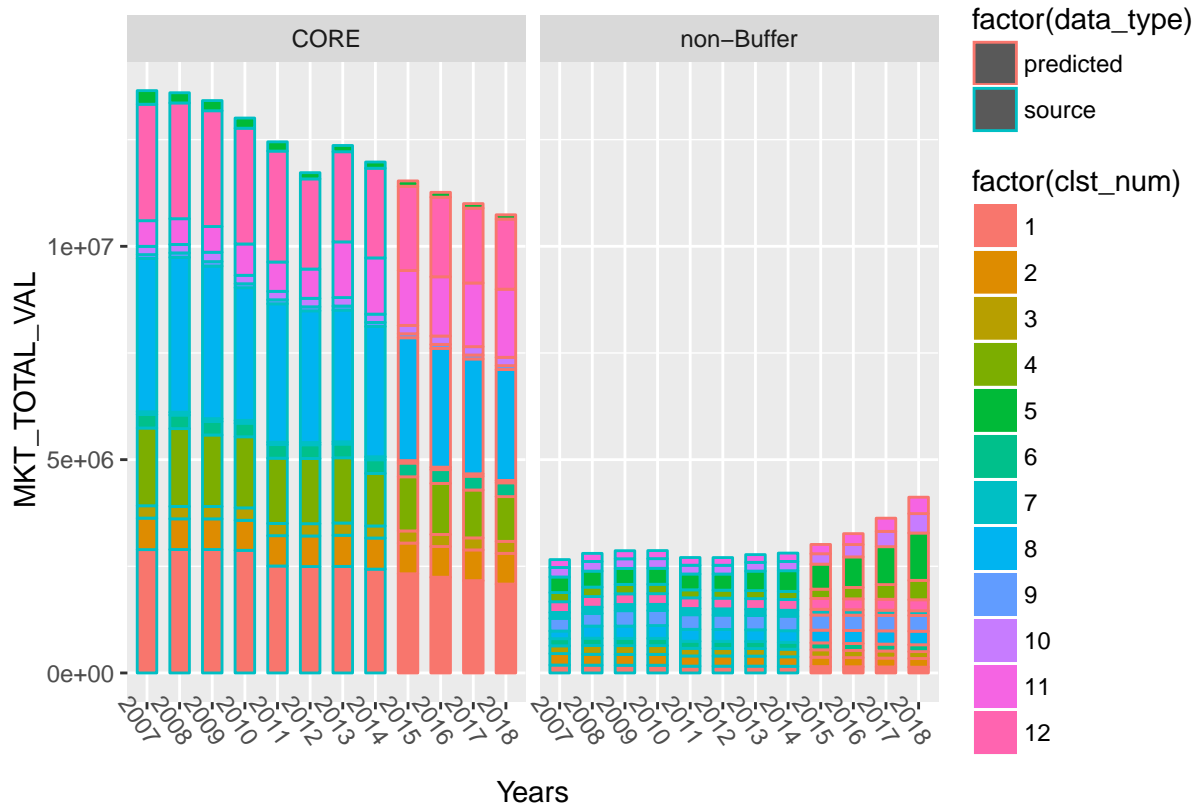
```

n3 <- ggplot(df_fin_1,aes(y = mkt_val, x = factor(date), fill = factor(clst_num), color=factor(data_type)

#n3 <- ggplot(df_final_mkt_val,aes(y = mkt_val, x = date ))
#
# # Data + aes + Geometries
n3 <- n3 + geom_bar(stat="identity", width = 0.6)
#n3 <- n3 + geom_line()
#
# # Data + Aesthetic Mapping + Geom + Facets
# #n <- n + facet_grid(. ~ Department.Title )
#
# #Data + Aesthetics + Geoms + Facets + Statistics
# # ... add a linear regression model here
#n3 <- n3 + geom_smooth(method="gam" )
# #Data + Aesthetics + Geoms + Facets + Statistics + Co-ordinates
# #n <- n + scale_y_discrete(limits=c("00000","120000"), breaks=seq(00000,120000,10000))
n3 <- n3 + scale_x_discrete()
n3 <- n3 + xlab("Years")
n3 <- n3 + ylab("MKT_TOTAL_VAL ")
#
# #Data + Aesthetics + Geoms + Facets + Statistics + Co-ordinates + Theme
n3 <- n3 + theme(axis.text.x = element_text(angle=305))
#
n3 <- n3 + ggtitle(title)
n3 <- n3 +facet_wrap( ~ zone)
print(n3)

```

MKT TOTAL VALUE in different ZONEs CORE and non-Buffer



```

# Identify clusters with growth in CENTER
cluster_max_val_parcel_center <- remove_char_X(rownames(partCENTER100$var[[2]]))
# Identify clusters with decline in CENTER
cluster_declining_val_parcel_center <- remove_char_X(rownames(partCENTER100$var[[1]]))

# Identify clusters with growth
cluster_max_val_parcel_EDGE <- remove_char_X(rownames(partEDGE100$var[[8]]))
# Identify clusters with growth
cluster_declining_val_parcel_EDGE <- remove_char_X(rownames(partEDGE100$var[[7]]))

# Cluster1 in Core showing declining trend
cluster_max_val_parcel_CORE <- remove_char_X(rownames(partCORE100$var[[1]] ))
# Cluster1 in Core showing declining trend
cluster_growing_val_parcel_CORE <- remove_char_X(rownames(partCORE100$var[[11]]))

cluster_max_val_parcel_NON_BFR <- remove_char_X(rownames(part1500$var[[5]]))

#####

# PLOTS on the MAP for EDGE

#####
library(ggmap)

#####
#Street Car Line : Convert shape files to data
#####
shp_file_streetcar <- readOGR('/Users/rajesh/Desktop/Coursera/SpringBoardGithub/StreetCar0719/streetcar'
                             layer="StreetCar" )

## OGR data source with driver: ESRI Shapefile
## Source: "/Users/rajesh/Desktop/Coursera/SpringBoardGithub/StreetCar0719/streetcarbuffer_parcel/gis_
## with 25 features
## It has 11 fields

proj4string(shp_file_streetcar)

## [1] "+proj=lcc +lat_1=38.73333333333333 +lat_2=40.03333333333333 +lat_0=38 +lon_0=-82.5 +x_0=600000 -
shp_file_streetcar <- spTransform(shp_file_streetcar, CRS("+proj=longlat +datum=WGS84"))

# convert to a data.frame for use with ggplot2/ggmap and plot
data <- fortify(shp_file_streetcar)

#####

# PLOTS on the MAP for EDGE

#####

myMap <- get_map(location="1208 Sycamore st, Cincinnati,OH", source="google", maptype="roadmap", crop=F

```

```
## Map from URL : http://maps.googleapis.com/maps/api/staticmap?center=1208+Sycamore+st,+Cincinnati,OH&
## Information from URL : http://maps.googleapis.com/maps/api/geocode/json?address=1208%20Sycamore%20st
title <- paste0("Density of Clusters of parcels with growing MKT_VAL in EDGE ")
```

```
# Growth
df_cluster_max_val_EDGE <- as.data.frame(cluster_max_val_parcels_EDGE)
colnames(df_cluster_max_val_EDGE) <- c("PARCEL_ID")
df_cluster_max_val_EDGE <- semi_join(df_buffer_EDGE, df_cluster_max_val_EDGE, by="PARCEL_ID")

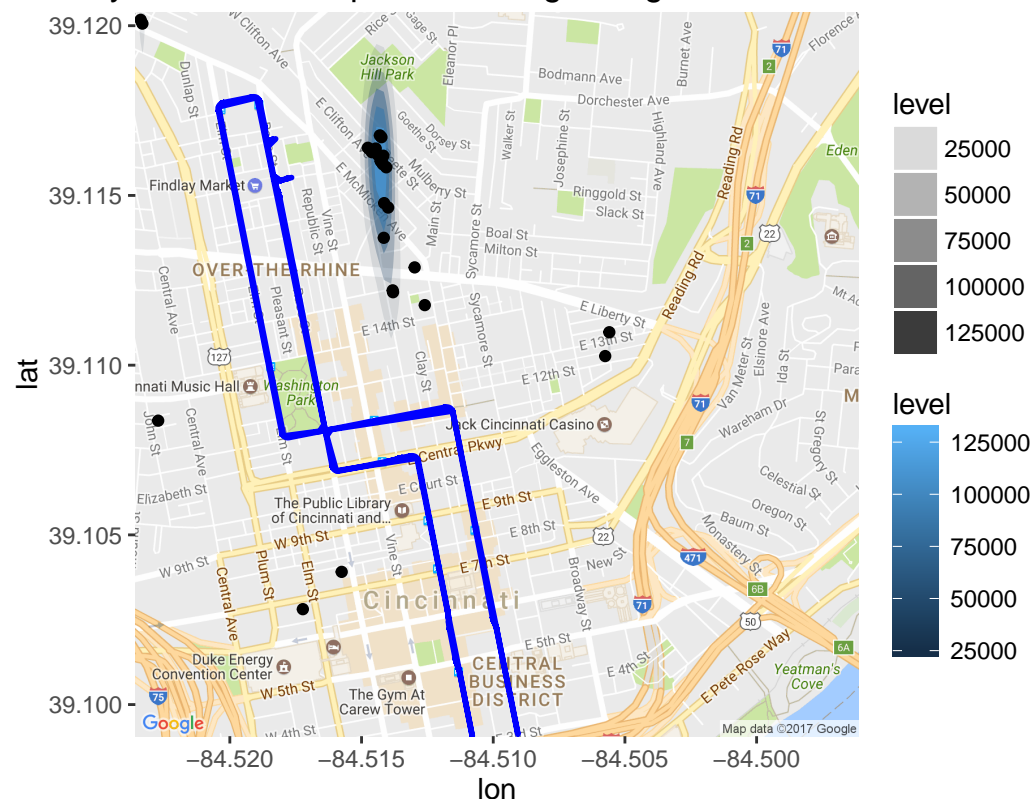
CinciCAGISDensityMap <- ggmap(myMap)
CinciCAGISDensityMap <- CinciCAGISDensityMap + stat_density2d(aes(x = as.numeric(cent_long), y = as.nu
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_point(data = df_cluster_max_val_EDGE, aes(x = as.n
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data
                                alpha = 1.0, size = 1.0, fill = 'red', color

CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
print(CinciCAGISDensityMap)
```

```
## Warning: Removed 2 rows containing non-finite values (stat_density2d).
```

```
## Warning: Removed 2 rows containing missing values (geom_point).
```

## Density of Clusters of parcels with growing MKT\_VAL in EDGE



### #Decline

```
myMap <- get_map(location="1208 Sycamore st, Cincinnati,OH", source="google", maptype="roadmap", crop=F

## Map from URL : http://maps.googleapis.com/maps/api/staticmap?center=1208+Sycamore+st,+Cincinnati,OH&
## Information from URL : http://maps.googleapis.com/maps/api/geocode/json?address=1208%20Sycamore%20st
```

```

title <- paste0("Density of Clusters of parcels with declining MKT_VAL in EDGE ")

df_cluster_declining_val_EDGE <- as.data.frame(cluster_declining_val_parcel_EDGE)
colnames(df_cluster_declining_val_EDGE)<- c("PARCEL_ID")
df_cluster_declining_val_EDGE <- semi_join(df_buffer_EDGE,df_cluster_declining_val_EDGE, by="PARCEL_ID")

CinciCAGISDensityMap <- ggmap(myMap)
CinciCAGISDensityMap <- CinciCAGISDensityMap + stat_density2d(aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), data = df_cluster_declining_val_EDGE, aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), kernel = "kernel4", fill = "blue", alpha = 1.0, size = 1.0, fill = 'red', color = 'red')
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_point(data = df_cluster_declining_val_EDGE, aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), size = 1.0, fill = 'red', color = 'red')
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data = df_cluster_declining_val_EDGE, fill = 'red', alpha = 1.0, size = 1.0, fill = 'red', color = 'red')

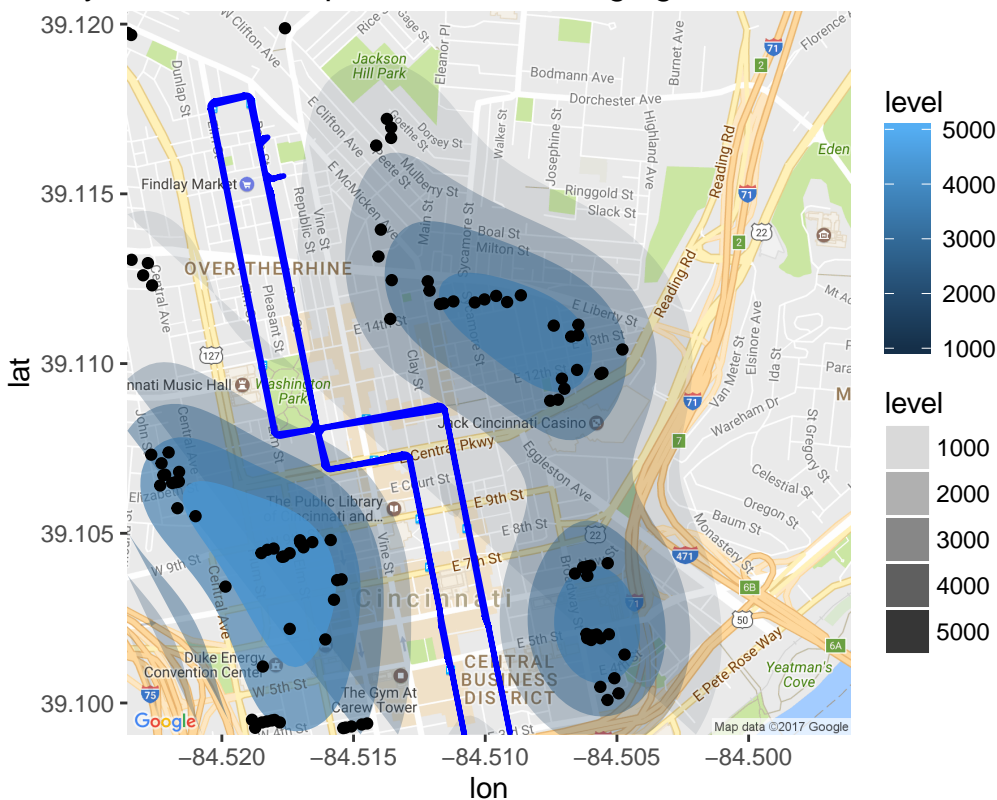
CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
print(CinciCAGISDensityMap)

```

## Warning: Removed 20 rows containing non-finite values (stat\_density2d).

## Warning: Removed 20 rows containing missing values (geom\_point).

### Density of Clusters of parcels with declining MKT\_VAL in EDGE



*# plot on MAPS for CENTER*

```

df_cluster_max_val_CENTER <- as.data.frame(cluster_max_val_parcel_CENTER)
colnames(df_cluster_max_val_CENTER)<- c("PARCEL_ID")
df_cluster_max_val_CENTER <- semi_join(df_buffer_CENTER,df_cluster_max_val_CENTER, by="PARCEL_ID")

myMap <- get_map(location="1208 Sycamore st, Cincinnati,OH", source="google", matype="roadmap", crop=F)

## Map from URL : http://maps.googleapis.com/maps/api/staticmap?center=1208+Sycamore+st,+Cincinnati,OH&

```

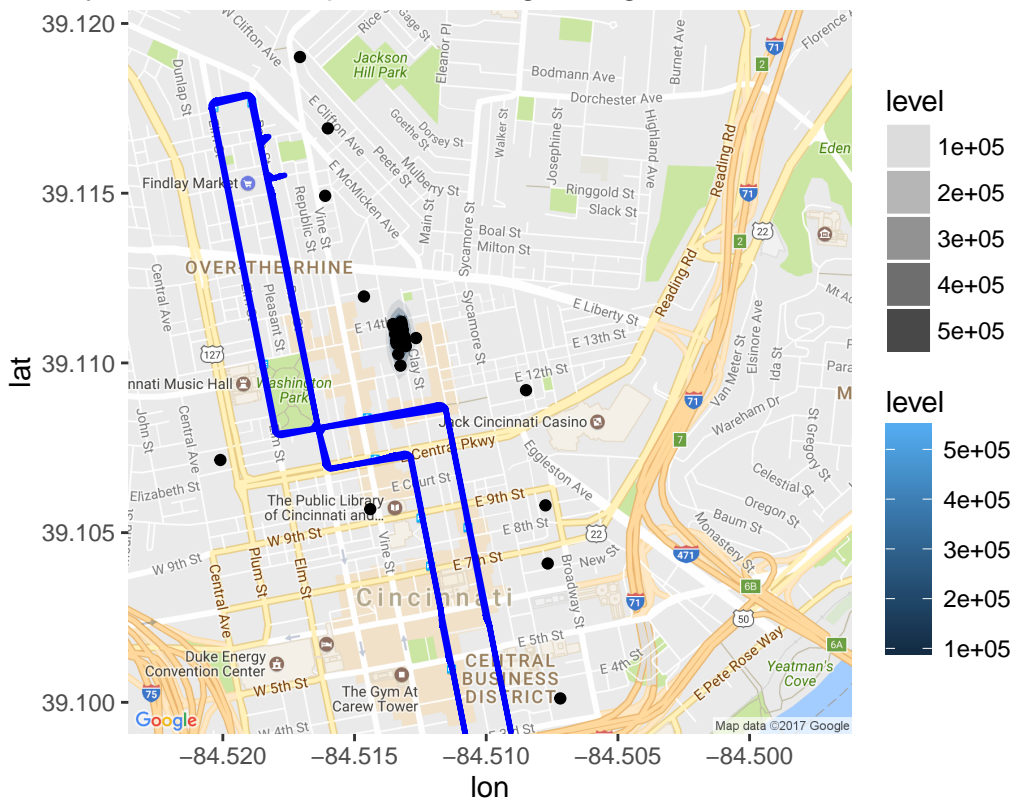
```
## Information from URL : http://maps.googleapis.com/maps/api/geocode/json?address=1208%20Sycamore%20st
title <- paste0("Density of Clusters of parcels with growingMKT_VAL in CENTER ")

CinciCAGISDensityMap <- ggmap(myMap)
CinciCAGISDensityMap <- CinciCAGISDensityMap + stat_density2d(aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), data = df_cluster_max_val_CENTER, geom_point(data = df_cluster_max_val_CENTER, aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), alpha = 1.0, size = 1.0, fill = 'red', color = 'red'))
CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data = df_buffer_CENTER, alpha = 1.0, size = 1.0, fill = 'red', color = 'red')
print(CinciCAGISDensityMap)
```

```
## Warning: Removed 1 rows containing non-finite values (stat_density2d).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

### Density of Clusters of parcels with growingMKT\_VAL in CENTER



```
# Cluster 1 has declining values
df_cluster_declining_val_CENTER <- as.data.frame(cluster_declining_val_parcel CENTER)
colnames(df_cluster_declining_val_CENTER) <- c("PARCEL_ID")
df_cluster_declining_val_CENTER <- semi_join(df_buffer_CENTER, df_cluster_declining_val_CENTER, by="PARCEL_ID")

myMap <- get_map(location="1208 Sycamore st, Cincinnati, OH", source="google", maptype="roadmap", crop=FALSE)

## Map from URL : http://maps.googleapis.com/maps/api/staticmap?center=1208+Sycamore+st,+Cincinnati,OH&zoom=15
## Information from URL : http://maps.googleapis.com/maps/api/geocode/json?address=1208%20Sycamore%20st
title <- paste0("Density of Clusters of parcels with declining MKT_VAL in CENTER ")
```



```

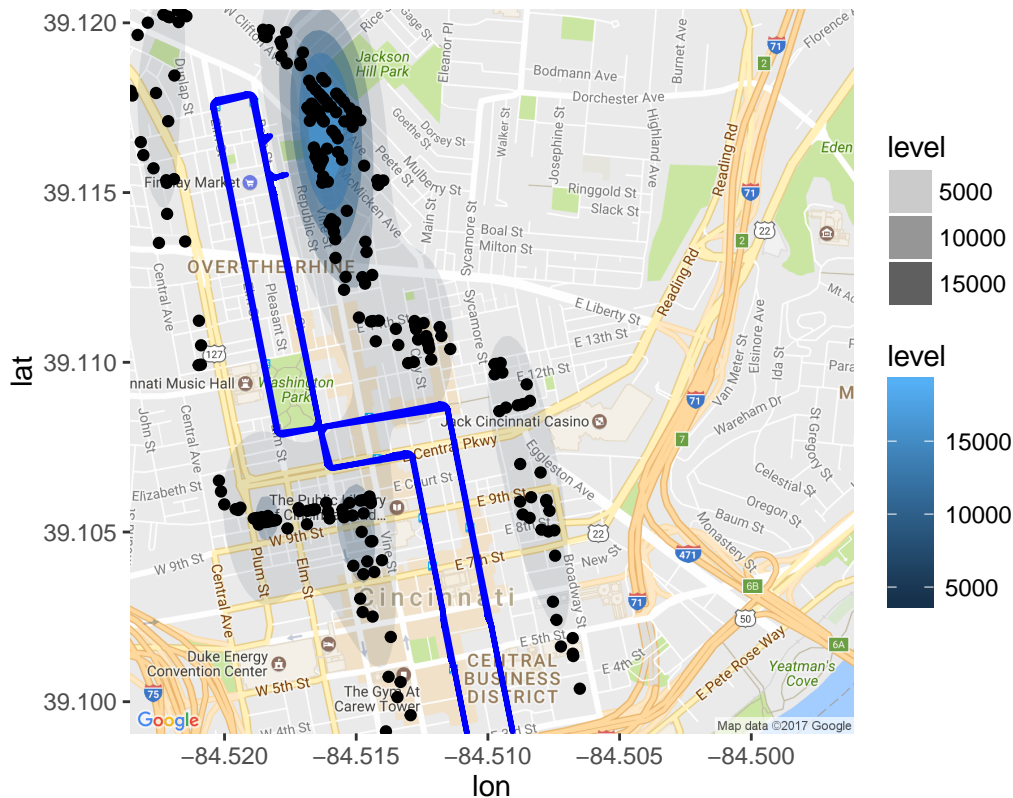
CinciCAGISDensityMap <- ggmap(myMap)
CinciCAGISDensityMap <- CinciCAGISDensityMap + stat_density2d(aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), data = df_cluster_declining_val_CENTER, aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), alpha = 1.0, size = 1.0, fill = 'red', color = 'black')
CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data = df_cluster_declining_val_CENTER, alpha = 1.0, size = 1.0, fill = 'red', color = 'black')
print(CinciCAGISDensityMap)

```

## Warning: Removed 18 rows containing non-finite values (stat\_density2d).

## Warning: Removed 18 rows containing missing values (geom\_point).

## Density of Clusters of parcels with declining MKT\_VAL in CENTER



*# plot on MAPS for CORE*

```

df_cluster_max_val_CORE <- as.data.frame(cluster_max_val_parcel CORE)
colnames(df_cluster_max_val_CORE) <- c("PARCEL_ID")
df_cluster_max_val_CORE <- semi_join(df_buffer_CORE, df_cluster_max_val_CORE, by="PARCEL_ID")
myMap <- get_map(location="1208 Sycamore st, Cincinnati, OH", source="google", maptype="roadmap", crop=FALSE)

## Map from URL : http://maps.googleapis.com/maps/api/staticmap?center=1208+Sycamore+st,+Cincinnati,OH&size=400x400&maptype=roadmap
## Information from URL : http://maps.googleapis.com/maps/api/geocode/json?address=1208%20Sycamore%20st,+Cincinnati,OH&key=AIzaSyB39...
title <- paste0("Density of Clusters of parcels in with declining MKT_VAL in CORE ")

CinciCAGISDensityMap <- ggmap(myMap)
CinciCAGISDensityMap <- CinciCAGISDensityMap + stat_density2d(aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), data = df_cluster_max_val_CORE, aes(x = as.numeric(cent_long), y = as.numeric(cent_lat)), alpha = 1.0, size = 1.0, fill = 'red', color = 'black')
CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data = df_cluster_max_val_CORE, alpha = 1.0, size = 1.0, fill = 'red', color = 'black')
print(CinciCAGISDensityMap)

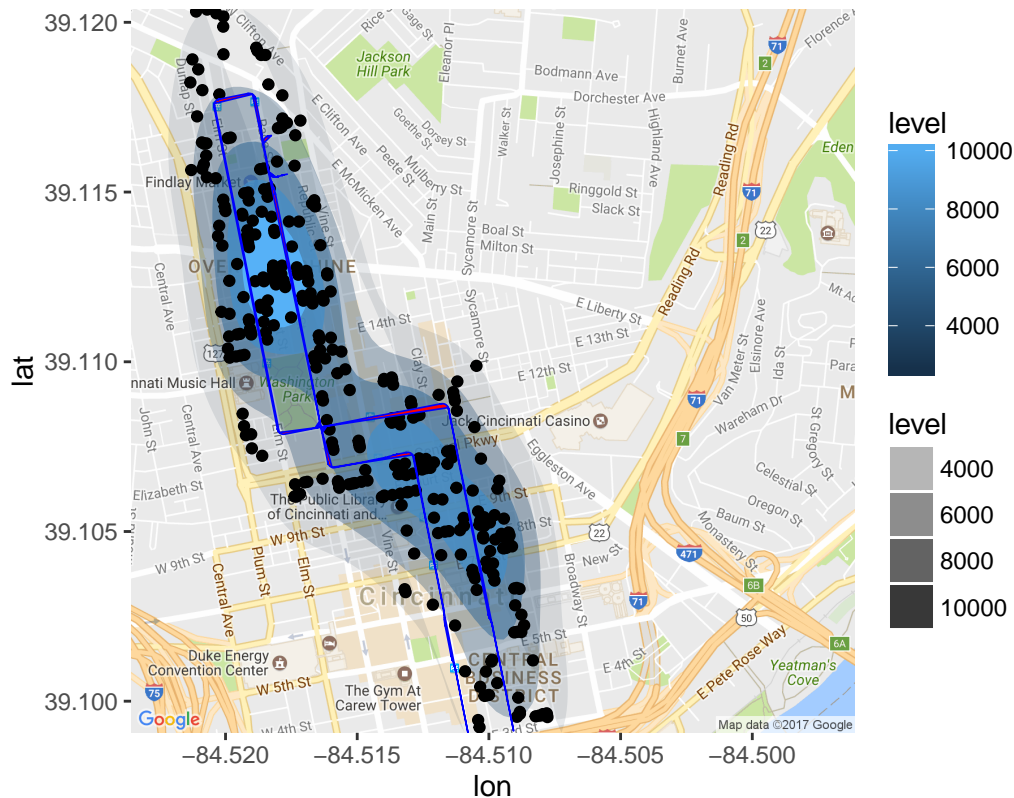
```

```
CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data = df_cluster_growing_val_CORE,
alpha = 1.0, size = 0.3, fill = 'red', color = 'black')
print(CinciCAGISDensityMap)
```

## Warning: Removed 2 rows containing non-finite values (stat\_density2d).

## Warning: Removed 2 rows containing missing values (geom\_point).

## Density of Clusters of parcels in with declining MKT\_VAL in CORE



```
df_cluster_growing_val_CORE <- as.data.frame(cluster_growing_val_parcels_CORE)
colnames(df_cluster_growing_val_CORE) <- c("PARCEL_ID")
df_cluster_max_val_CORE <- semi_join(df_buffer_CORE, df_cluster_growing_val_CORE, by="PARCEL_ID")
myMap <- get_map(location="1208 Sycamore st, Cincinnati, OH", source="google", maptype="roadmap", crop=F)
```

## Map from URL : <http://maps.googleapis.com/maps/api/staticmap?center=1208+Sycamore+st,+Cincinnati,OH&zoom=15>

## Information from URL : <http://maps.googleapis.com/maps/api/geocode/json?address=1208%20Sycamore%20st,+Cincinnati,OH>

```
title <- paste0("Density of Clusters of parcels with increasing MKT_VAL in CORE ")
CinciCAGISDensityMap <- ggmap(myMap)
CinciCAGISDensityMap <- CinciCAGISDensityMap + stat_density2d(aes(x = as.numeric(cent_long), y = as.numeric(cent_lat), group = group ), data = df_cluster_max_val_CORE,
alpha = 1.0, size = 0.3, fill = 'red', color = 'black')
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_point(data = df_cluster_max_val_CORE, aes(x = as.numeric(cent_long), y = as.numeric(cent_lat), group = group ), data = df_cluster_max_val_CORE,
alpha = 1.0, size = 0.3, fill = 'red', color = 'black')
CinciCAGISDensityMap <- CinciCAGISDensityMap + ggtitle(title)
CinciCAGISDensityMap <- CinciCAGISDensityMap + geom_polygon(aes(x = long, y = lat, group = group ), data = df_cluster_growing_val_CORE,
alpha = 1.0, size = 0.3, fill = 'red', color = 'black')
print(CinciCAGISDensityMap)
```



Density of Clusters of parcels with increasing MKT\_VAL in CORE

