

```
library("readxl")
library("lubridate")
library("proto")
library("gsubfn")
library("RSQLite")
library("sqldf")
library("tidyverse")
library("ggplot2")
library("mice")
library("caret")
library("dplyr")
library("reshape2")
library("janitor")
library("tidyr")
library("MASS")
library("car")
library("Metrics")
library("glmnet")
library("xgboost")
library("gbm")
library("mboost")
library("elasticnet")
library("tseries")
library("zoo")
library("forecast")
library("chron")
library("RColorBrewer")
library("lattice")
library("ncdf4")
library("fpp")
library("Rmisc")
```

```
#NASA
```

```
nasa <- read.csv("./Assignment//Team/Assignment 1//Dataset//nasa_clean.csv")
```

```
summary(nasa)
```

```
head(nasa)
```

```
#changing all to numeric
```

```
nasa$Date <- as.Date(nasa$Date, format = "%Y-%m-%d")
```

```
class(nasa$Date )
```

```
nasa$Temp_Diff <- as.numeric(as.character(nasa$Temp_Diff))
```

```
#nasa$Avg_temp <- as.numeric(as.character(nasa$Avg_temp))
```

```
nasa <- drop_na(nasa)
```

```
#adding Temp_Diff to baseline (14)
```

```
nasa$Avg_Temp <- 14 + nasa$Temp_Diff
```

```
#nasa$Avg_Temp <- 14 + nasa$Avg_temp
```

```
nasa = nasa[,!(names(nasa) %in% c("Temp_Diff"))]
```

```
#nasa = nasa[,!(names(nasa) %in% c("Avg_temp"))]
```

```
#nasa = nasa[,!(names(nasa) %in% c("X"))]
```

```
nasa$Avg_Temp<-as.numeric(nasa$Avg_Temp)
```

```
#frequency: 1=annual, 4=quarterly, 12=monthly
```

```
#creating timeseries
```

```
#save a numeric vector of 2040 observations (from 1850 jan - 2020 feb)
```

```
nasa <- nasa[order(as.Date(nasa$Date, format="%Y-%m-%d")),]
```

```
nasats <- ts(nasa$Avg_Temp, start=c(1880,1), end=c(2020,03), frequency=12)
```

```
tail(nasats)
```

```
plot(nasats)
```

```
#plot various decompositions into error/noise, trend and seasonality
```

```
fit <- decompose(nasats, type="multiplicative") #decompose using "classical" method, multiplicative form
plot(fit)
```

```
fit <- decompose(nasats, type="additive") #decompose using "classical" method, additive form
plot(fit)
```

```
#gives you the confident interval: there is a trend, high confidence interval for seasonality

fit <- stl(nasats, t.window=12, s.window="periodic") #decompose using STL (Season and trend using Loess)
plot(fit)
```

```
#creating a test and train
```

```
#Split the data into train and test sets at "2007-01-01"
```

```
train_tbl <- nasa %>% filter(nasa$Date < ymd("2007-01-01"))
```

```
test_tbl <- nasa %>% filter(nasa$Date >= ymd("2007-01-01"))
```

```
train_tbl <- train_tbl %>% filter(train_tbl$Date >= ymd("1940-01-01"))
```

```
#to order by date, not necessary, but good to know
```

```
#train_tbl <- train_tbl[order(as.Date(train_tbl$Date, format="%Y-%m-%d")),]
```

```
#test_tbl <- test_tbl[order(as.Date(test_tbl$Date, format="%Y-%m-%d")),]
```

```
tail(train_tbl)
```

```
#creating timeseries for the data
```

```
traints <- ts(train_tbl$Avg_Temp, start=c(1940,1), end=c(2006,12), frequency=12)
```

```
testts <- ts(test_tbl$Avg_Temp, start=c(2007,1), end=c(2020,3), frequency=12)
```

```
tail(traints)
```

```
#Qn 1 & 2: Forecast the global average temperatures & 90% confidence intervals
```

```
# Create exponential smoothing models: additive vs multiplicative noise (first A vs M), additive vs multiplicative trend (second A vs M) and no seasonality vs automatic detection (third N vs Z) trend and no seasonality (AAN), multiplicative (MMN)
```

```
nasa_AAN <- ets(traints, model="AAN", damped=TRUE)
```

```
nasa_MMN <- ets(traints, model="MMN", damped=TRUE)
```

```
nasa_AAA <- ets(traints, model="AAA", damped=TRUE)
```

```
nasa_MMM <- ets(traints, model="MMM", damped=TRUE)
```

```
nasa_MAZ <- ets(traints, model="MAZ", damped=TRUE)
```

```
nasa_AAN
```

```
nasa_MMN
```

```
nasa_AAA
```

```
nasa_MMM
```

```
nasa_MAZ
```

```
# Create their prediction "cones" for 159 months (2007-01-01 to 2020-03-01) into the future with quintile confidence intervals
```

```
nasa_AAN_pred <- forecast(nasa_AAN, h=159, level=c(0.8, 0.95))
```

```
nasa_MMN_pred <- forecast(nasa_MMN, h=159, level=c(0.8, 0.95))
```

```
nasa_AAA_pred <- forecast(nasa_AAA, h=159, level=c(0.8, 0.95))
```

```
nasa_MMM_pred <- forecast(nasa_MMM, h=159, level=c(0.8, 0.95))
```

```
nasa_MAZ_pred <- forecast(nasa_MAZ, h=159, level=c(0.8, 0.95))
```

```
#check mape of prediction on test dataset
```

```
#https://stackoverflow.com/questions/43048018/accuracy-function
```

```
print(accuracy(nasa_AAN_pred, testts)) #RMSE: 0.1934915, MAPE: 1.0244427
```

```
print(accuracy(nasa_MMN_pred, testts)) #RMSE: 0.1941361, MAPE: 1.0259993
```

```
print(accuracy(nasa_AAA_pred, testts)) #RMSE: 0.1906236, MAPE: 1.0182961
```

```
print(accuracy(nasa_MMM_pred, testts)) #RMSE: 0.191294, MAPE: 1.0168532
```

```
print(accuracy(nasa_MAZ_pred, testts)) #RMSE: 0.1946198, MAPE: 1.0282960
```

```
#Create a trigonometric box-cox autoregressive trend seasonality (TBATS) model
```

```
nasa_tbats <- tbats(traints)
```

```
nasa_tbats
```

```
#forecast for 159 months (2007-01-01 to 2020-03-01)
```

```
nasa_tbats_pred <- forecast(nasa_tbats, h=159, level=c(0.8, 0.95))
```

```
par(mfrow=c(1,1))
```

```
plot(nasa_tbats_pred, xlab="Time", ylab="Predicted Global Avg Temp")
```

```
#check accuracy with test dataset
```

```
print(accuracy(nasa_tbats_pred, testts)) #RMSE: 0.2325858, MAPE: 1.2098155
```

```
#Creating ARIMA model
```

```
fit <- auto.arima(traints, seasonal=TRUE)
```

```
fit
```

```
#forecast for 159 months (2007-01-01 to 2020-03-01)
```

```
nasa_arima_pred <- forecast(fit, h=159, level=c(0.8, 0.95))
```

```
plot(nasa_arima_pred, xlab="Time", ylab="Predicted Global Avg Temp")
```

```
#check accuracy with test dataset
```

```
print(accuracy(nasa_arima_pred, testts)) #RMSE: 0.2305829, MAPE: 1.1754087
```

```
#Creating ARIMA model
```

```
fit <- auto.arima(traints, seasonal=FALSE)
```

```
fit
```

```
#forecast for 159 months (2007-01-01 to 2020-03-01)
```

```
nasa_arima_pred <- forecast(fit, h=159, level=c(0.8, 0.95))
```

```
plot(nasa_arima_pred, xlab="Time", ylab="Predicted Global Avg Temp")
```

```
#check accuracy with test dataset
```

```
print(accuracy(nasa_arima_pred,testts)) #RMSE: 0.2305829, MAPE: 1.1754087 #worst model also  
omg
```

```
#####
```

```
##### ARIMA with regressors
```

```
#####
```

```
month <- as.numeric(format(train_tbl$Date, "%m"))
```

```
monthMatrix <- cbind(Months=model.matrix(~as.factor(month))) # Create dummies for each day-of-  
week
```

```
monthMatrix <- monthMatrix[,-1]# Remove "intercept" (12th month) dummy
```

```
colnames(monthMatrix) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep", "Oct", "Nov",  
"Dec") # Rename columns
```

```
matrix_of_regressors <- monthMatrix
```

```
nasatrain_arima <- auto.arima(traints, xreg=matrix_of_regressors,seasonal=TRUE, approximation =  
FALSE) # Train a model
```

```
nasatrain_arima # See what it is
```

```
#forecast for 159 months (2007-01-01 to 2020-03-01)
```

```
monthTestPred <- 1
```

```
monthTestPred <- rbind(monthTestPred, 2)
```

```
monthTestPred <- rbind(monthTestPred, 3)
```

```
for (i in 1:12){
```

```
  for(j in 1:13)
```

```
  {
```

```
    monthTestPred <- rbind(monthTestPred, i)
```

```
  }}
```

```
monthTestPred
```

```
#remove first row
```

```
#monthPred <- monthPred[-1,]
```

```
#making the matrix
```

```

monthTestPredMatrix <- cbind(Months=model.matrix(~as.factor(monthTestPred))) # Create
dummies for each month of year

monthTestPredMatrix

monthTestPredMatrix <- monthTestPredMatrix[,-1]# Remove "intercept" dummy

colnames(monthTestPredMatrix) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep", "Oct",
"Nov", "Dec") # Rename columns


matrix_of_test_pred_regressors <- monthTestPredMatrix


nasatest_arima_pred <- forecast(nasatrain_arima, h=159, xreg = matrix_of_test_pred_regressors,
level=c(0.8, 0.95))

plot(nasatest_arima_pred, xlab="Time", ylab="Predicted Avg Temp")

print(accuracy(nasatest_arima_pred,testts)) #RMSE: 0.1897208, MAPE: 0.9724306 #best so far


#####

##### Forecast predictions using ARIMA on regressors for 969 months (2020 March to 2100
December)

#####

#month Matrix on full NASA dataset

month <- as.numeric(format(nasa$Date, "%m"))


monthMatrix <- cbind(Months=model.matrix(~as.factor(month))) # Create dummies for each month
of year

monthMatrix <- monthMatrix[,-1]# Remove "intercept" dummy

colnames(monthMatrix) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep", "Oct", "Nov",
"Dec") # Rename columns


matrix_of_regressors <- monthMatrix

nasatemp_arima <- auto.arima(nasats, xreg=matrix_of_regressors,seasonal=TRUE, approximation =
FALSE) # Train a model

nasatemp_arima # See what it is

```

```

#forecast for 969 months (2020-04-01 to 2100-12-01)
monthPred <- 4
monthPred <- rbind(monthPred, 5)
monthPred <- rbind(monthPred, 6)
monthPred <- rbind(monthPred, 7)
monthPred <- rbind(monthPred, 8)
monthPred <- rbind(monthPred, 9)
monthPred <- rbind(monthPred, 10)
monthPred <- rbind(monthPred, 11)
monthPred <- rbind(monthPred, 12)
for (i in 1:12){
  for(j in 1:80)
  {
    monthPred <- rbind(monthPred, i)
  }
}
monthPred

#remove first row
#monthPred <- monthPred[-1,]

#making the matrix
monthPredMatrix <- cbind(Months=model.matrix(~as.factor(monthPred))) # Create dummies for
each month of year

monthPredMatrix

monthPredMatrix <- monthPredMatrix[,-1]# Remove "intercept" dummy

colnames(monthPredMatrix) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep", "Oct", "Nov",
"Dec") # Rename columns

matrix_of_pred_regressors <- monthPredMatrix

nasatemp_arma_pred <- forecast(nasatemp_arma, h=969, xreg = matrix_of_pred_regressors,
level=c(0.8, 0.90))

plot(nasatemp_arma_pred, xlab="Time", ylab="Predicted Avg Temp")

```



```
write.csv(nasatemp_arima_pred, "NASA ARIMA Predictions.csv", row.names = FALSE, na = "")
```

```
#Qn 6: constant temperature prediction per Mr. Armstrong
```

```
#getting 2006 dataset
```

```
data_2006 <- nasa %>% filter(nasa$Date < ymd("2007-01-01"))
```

```
data_2006 <- data_2006 %>% filter(data_2006$Date >= ymd("2006-01-01"))
```

```
data_2006
```

```
#getting the average temp of 2016
```

```
avg_temp_2006<- mean(data_2006$Avg_Temp)
```

```
avg_temp_2006
```

```
#Create dataset for 2007-01-01 to 2017-12-01 with constant average temp of 2016
```

```
constant_temp <- test_tbl %>% filter(test_tbl$Date <= ymd("2017-12-01"))
```

```
constant_temp$Avg_Temp <- avg_temp_2006
```

```
constant_temp
```

```
#Create dataset for 2007-01-01 to 2017-12-01 for NASA dataset
```

```
test_tbl_actual <- test_tbl %>% filter(test_tbl$Date <= ymd("2017-12-01"))
```

```
#creating timeseries for the data
```

```
head(train_tbl)
```

```
traints <- ts(train_tbl$Avg_Temp, start=c(1940,1), end=c(2006,12), frequency=12) #train
```

```
constantts <- ts(constant_temp$Avg_Temp, start=c(2007,1), end=c(2017,12), frequency=12)
```

```
#"actual"
```

```
testts_act <- ts(test_tbl_actual$Avg_Temp, start=c(2007,1), end=c(2017,12), frequency=12) #actual  
NASA
```

```
plot(traints)
```

```
plot(constantts)
```

```
plot(testts_act)
```

```
#####
```

```
##### ARIMA with regressors [train]
```

```
#####
```

```
month <- as.numeric(format(train_tbl$Date, "%m"))
```

```
monthMatrix <- cbind(Months=model.matrix(~as.factor(month))) # Create dummies for each month  
of year
```

```
monthMatrix <- monthMatrix[,-1]# Remove "intercept" (12th month) dummy
```

```
colnames(monthMatrix) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep", "Oct", "Nov",  
"Dec") # Rename columns
```

```
matrix_of_regressors <- monthMatrix
```

```
nasatrain_arima <- auto.arima(traints, xreg=matrix_of_regressors,seasonal=TRUE, approximation =  
FALSE) # Train a model
```

```
nasatrain_arima # See what it is
```

```
#forecast for 132 months (2007-01-01 to 2017-12-01)
```

```
month_constant_temp <- as.numeric(format(test_tbl_actual$Date, "%m"))
```

```
monthMatrix_constant_temp <- cbind(Months=model.matrix(~as.factor(month_constant_temp))) #  
Create dummies for each day-of-week
```

```
monthMatrix_constant_temp <- monthMatrix_constant_temp[,-1]# Remove "intercept" (12th  
month) dummy
```

```
colnames(monthMatrix_constant_temp) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep",  
"Oct", "Nov", "Dec") # Rename columns
```

```
matrix_of_regressors_constant_temp <- monthMatrix_constant_temp
```

```
nasa_arima_pred <- forecast(nasatrain_arima, h=132, xreg = matrix_of_regressors_constant_temp,  
level=c(0.8, 0.95))
```

```
print(accuracy(nasa_arima_pred,testts_act)) #predicted vs actual
```

```
print(accuracy(constantts,testts_act)) #constant vs actual
```

```

par(mfrow=c(1,1))

plot(nasa_arima_pred,xlim=c(2007,2017), ylim=c(14.3, 15.4), main='NASA - Actual temperature
against forecasted and constant')

par(new=TRUE)

plot(testts_act, ylab='Avg Global Monthly Temp', xlim=c(2007,2017), ylim=c(14.3, 15.4), main='NASA
- Actual temperature against forecasted and constant')

abline(constant_temp$Avg_Temp,0, col="red")

```

#Qn7: same analyses with another starting time and long-enough time-interval

#Split the data into train and test sets at "1996-01-01"

```

train_tbl <- nasa %>% filter(nasa$Date < ymd("1996-01-01"))
test_tbl <- nasa %>% filter(nasa$Date >= ymd("1996-01-01"))
train_tbl <- train_tbl %>% filter(train_tbl$Date >= ymd("1970-01-01"))
test_tbl <- test_tbl %>% filter(test_tbl$Date <= ymd("2005-12-01"))

#getting 1995 dataset

data_1995 <- train_tbl %>% filter(train_tbl$Date >= ymd("1995-01-01"))

data_1995

```

#getting the average temp of 1996

```

data_1995<- mean(data_1995$Avg_Temp)

data_1995

```

#Create dataset for 1996-01-01 to 2005-12-01 with constant average temp of 2016

```

constant_temp <- test_tbl

constant_temp$Avg_Temp <- data_1995

constant_temp

```

#creating timeseries for the data

```

head(train_tbl)

traints <- ts(train_tbl$Avg_Temp, start=c(1970,1), end=c(1995,12), frequency=12) #train

constantts <- ts(constant_temp$Avg_Temp, start=c(1996,1), end=c(2005,12), frequency=12)
#constant

```

```
testts_act <- ts(test_tbl_actual$Avg_Temp, start=c(1996,1), end=c(2005,12), frequency=12) #actual  
NASA
```

```
#####
```

```
##### ARIMA with regressors [train]
```

```
#####
```

```
month <- as.numeric(format(train_tbl$Date, "%m"))
```

```
monthMatrix <- cbind(Months=model.matrix(~as.factor(month))) # Create dummies for each month  
of year
```

```
monthMatrix <- monthMatrix[,-1]# Remove "intercept" (12th month) dummy
```

```
colnames(monthMatrix) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep", "Oct", "Nov",  
"Dec") # Rename columns
```

```
matrix_of_regressors <- monthMatrix
```

```
nasatrain_arima <- auto.arima(traints, xreg=matrix_of_regressors,seasonal=TRUE, approximation =  
FALSE) # Train a model
```

```
nasatrain_arima # See what it is
```

```
#forecast for 120 months (1996-01-01 to 2005-12-01)
```

```
month_constant_temp <- as.numeric(format(test_tbl$Date, "%m"))
```

```
monthMatrix_constant_temp <- cbind(Months=model.matrix(~as.factor(month_constant_temp))) #  
Create dummies for each day-of-week
```

```
monthMatrix_constant_temp <- monthMatrix_constant_temp[,-1]# Remove "intercept" (12th  
month) dummy
```

```
colnames(monthMatrix_constant_temp) <- c("Feb","Mar","Apr","May","Jun", "Jul", "Aug", "Sep",  
"Oct", "Nov", "Dec") # Rename columns
```

```
matrix_of_regressors_constant_temp <- monthMatrix_constant_temp
```

```
nasa_arima_pred <- forecast(nasatrain_arima, h=120, xreg = matrix_of_regressors_constant_temp,  
level=c(0.8, 0.95))
```

```
print(accuracy(nasa_arima_pred,testts_act)) #predicted vs actual
```

```
print(accuracy(constantts,testts_act)) #constant vs actual
```

```
par(mfrow=c(1,1))
```

```
plot(nasa_arima_pred,xlim=c(1996.32,2005), ylim=c(14.10, 15.4), main='NASA - Actual temperature  
against forecasted and constant')
```

```
par(new=TRUE)
```

```
plot(testts_act, ylab='Avg Global Monthly Temp', xlim=c(1996.32,2005), ylim=c(14.10, 15.4),  
main='NASA - Actual temperature against forecasted and constant')
```

```
abline(constant_temp$Avg_Temp,0, col="red")
```