HW3

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```
#library Import
library(caret)
library(tidyverse)

library(fable)
library(feasts)
library(fredr)
library(tsibble)
library(kableExtra)
library(plyr)
set.seed(23)
```

Problem 1

```
#creating the tibble/dataframe with 30 observations
tsdf <- tibble(ts_index = c(1:30), r = rnorm(30))</pre>
tsdf$y <- 0
#logic that doesn't really work, but sets the y values according to a function with 3 lags
tsdf <- tsdf %>%
  mutate(y = ifelse(ts_index < 4, r,</pre>
            y = 0.5 + 0.5*lag(y,1) - 0.1*lag(y, 2) + 0.25*lag(y, 3) + r)
#replacing rows 1 through 3 with the associated r values
tsdf[1:3,3] <- tsdf[1:3,2]
#making 7 different possible Autoregressive models (1, 2, 3) using loocv in caret
train_ct1 <- trainControl(method = "LOOCV")</pre>
# 1,2,3 lag structure
lm_model1 \leftarrow train(y \sim lag(y) + lag(y, 2) + lag(y, 3), data = tsdf,
                    na.action = na.pass,
                    trControl = train_ct1,
                    method = "lm")
# 1 2 lag structure
lm_{model2} \leftarrow train(y \sim lag(y) + lag(y, 2), data = tsdf,
                   na.action = na.pass,
                   trControl = train_ct1,
```

```
# 1 3 lag structure
lm_{model3} \leftarrow train(y \sim lag(y) + lag(y, 3), data = tsdf,
                   na.action = na.pass,
                    trControl = train_ct1,
                   method = "lm")
lm_model4 \leftarrow train(y \sim lag(y, 2) + lag(y, 3), data = tsdf,
                   na.action = na.pass,
                    trControl = train_ct1,
                   method = "lm")
lm_model5 <- train(y ~ lag(y), data = tsdf,</pre>
                   na.action = na.pass,
                   trControl = train_ct1,
                   method = "lm")
lm_model6 <- train(y ~ lag(y, 2), data = tsdf,</pre>
                   na.action = na.pass,
                   trControl = train_ct1,
                   method = "lm")
lm_model7 <- train(y ~ lag(y, 3), data = tsdf,</pre>
                   na.action = na.pass,
                    trControl = train_ct1,
                   method = "lm")
#binding all of the results by rows into one matrix
results_matrix <- as_tibble(bind_rows(lm_model1$results,</pre>
                               lm_model2$results,
                               lm_model3$results,
                               lm_model4$results,
                               lm_model5$results,
                               lm_model6$results,
                               lm_model7$results))
results_matrix[,1] <- NULL
#Matrix of the AIC values
AIC_matrix <- AIC(lm_model1$finalModel, lm_model2$finalModel, lm_model3$finalModel,
                  lm model4$finalModel, lm model5$finalModel, lm model6$finalModel, lm model7$finalMode
Warning in AIC.default(lm_model1$finalModel, lm_model2$finalModel,
lm_model3$finalModel, : models are not all fitted to the same number of
observations
#Removing the first column
AIC_matrix[,1] <- NULL
```

Warning in BIC.default(lm_model1\$finalModel, lm_model2\$finalModel,

method = "lm")

lm_model4\$finalModel, lm_model5\$finalModel, lm_model6\$finalModel, lm_model7\$finalMode

BIC_matrix <- BIC(lm_model1\$finalModel, lm_model2\$finalModel, lm_model3\$finalModel,

lm_model3\$finalModel, : models are not all fitted to the same number of
observations

```
#removing the first column
BIC_matrix[,1] <- NULL

#appending them to the results matrix
results_matrix['AIC'] <- AIC_matrix
results_matrix['BIC'] <- BIC_matrix</pre>
```

```
#similar cross-validation to above, except using k-folds
#making 7 different possible Autoregressive models (1, 2, 3) using loocv in caret
train ct2 <- trainControl(method = "cv", number = 10)</pre>
# Creating a training set using 80% of the data
inTrain2 <- createDataPartition(y = tsdf$y, p = 0.8, list = FALSE)</pre>
#training data
train_set2 <- tsdf[inTrain2, ]</pre>
#test data (with the other 20 percent)
test_set2 <- tsdf[-inTrain2, ]</pre>
# 1,2,3 lag structure
lm_model_cv1 \leftarrow train(y \sim lag(y) + lag(y, 2) + lag(y, 3), data = train_set2,
                    na.action = na.pass,
                    trControl = train ct2,
                    method = "lm")
# 1 2 lag structure
lm_model_cv2 <- train(y ~ lag(y) + lag(y, 2), data = train_set2,</pre>
                    na.action = na.pass,
                    trControl = train_ct2,
                    method = "lm")
```

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

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

```
#Getting the results and appending them to the same matrix as before
models <- c(lm_model_cv1, lm_model_cv2, lm_model_cv3,</pre>
             lm_model_cv4, lm_model_cv5, lm_model_cv6, lm_model_cv7)
predictions1 <- predict(lm_model_cv1, test_set2)</pre>
predictions2 <- predict(lm_model_cv2, test_set2)</pre>
predictions3 <- predict(lm_model_cv3, test_set2)</pre>
predictions4 <- predict(lm_model_cv4, test_set2)</pre>
predictions5 <- predict(lm_model_cv5, test_set2)</pre>
predictions6 <- predict(lm_model_cv6, test_set2)</pre>
predictions7 <- predict(lm_model_cv7, test_set2)</pre>
#post prediction resampling
newdata1 <- postResample(predictions1, test_set2$y)</pre>
newdata2 <- postResample(predictions2, test_set2$y)</pre>
newdata3 <- postResample(predictions3, test_set2$y)</pre>
newdata4 <- postResample(predictions4, test_set2$y)</pre>
newdata5 <- postResample(predictions5, test_set2$y)</pre>
```

Warning in pred - obs: longer object length is not a multiple of shorter object length

Warning in pred - obs: longer object length is not a multiple of shorter object length

```
newdata6 <- postResample(predictions6, test_set2$y)
newdata7 <- postResample(predictions7, test_set2$y)

#creating a new matrix and binding the RMSE values
results2 <- rbind(newdata1, newdata2, newdata3, newdata4, newdata5, newdata6, newdata7) %>%
    as.data.frame() %>%
    subset(select = RMSE)

#Binding it to the original matrix
results_matrix <- cbind(results_matrix, results2)

#renaming the columns
colnames(results_matrix)[6] <- "k_RMSE"</pre>
```

Here is the section with 300 observations instead

```
#making 7 different possible Autoregressive models (1, 2, 3) using loocv in caret
train_ct1 <- trainControl(method = "LOOCV")</pre>
# 1,2,3 lag structure
lm_{model1} \leftarrow train(y \sim lag(y) + lag(y, 2) + lag(y, 3), data = tsdf,
                    na.action = na.pass,
                    trControl = train_ct1,
                    method = "lm")
# 1 2 lag structure
lm_model2 \leftarrow train(y \sim lag(y) + lag(y, 2), data = tsdf,
                    na.action = na.pass,
                    trControl = train_ct1,
                    method = "lm")
# 1 3 lag structure
lm_model3 \leftarrow train(y \sim lag(y) + lag(y, 3), data = tsdf,
                    na.action = na.pass,
                    trControl = train_ct1,
                    method = "lm")
```

```
na.action = na.pass,
                   trControl = train_ct1,
                   method = "lm")
lm_model5 <- train(y ~ lag(y), data = tsdf,</pre>
                   na.action = na.pass,
                   trControl = train ct1,
                   method = "lm")
lm_model6 <- train(y ~ lag(y, 2), data = tsdf,</pre>
                   na.action = na.pass,
                   trControl = train_ct1,
                   method = "lm")
lm_model7 <- train(y ~ lag(y, 3), data = tsdf,</pre>
                   na.action = na.pass,
                   trControl = train_ct1,
                   method = "lm")
#binding all of the results by rows into one matrix
results_matrix2 <- as_tibble(bind_rows(lm_model1$results,</pre>
                               lm model2$results,
                               lm_model3$results,
                               lm_model4$results,
                               lm_model5$results,
                               lm_model6$results,
                               lm_model7$results))
results_matrix2[,1] <- NULL</pre>
#Matrix of the AIC values
AIC_matrix <- AIC(lm_model1$finalModel, lm_model2$finalModel, lm_model3$finalModel,
                  lm_model4$finalModel, lm_model5$finalModel, lm_model6$finalModel, lm_model7$finalMode
Warning in AIC.default(lm_model1$finalModel, lm_model2$finalModel,
lm_model3$finalModel, : models are not all fitted to the same number of
observations
#Removing the first column
AIC_matrix[,1] <- NULL
BIC_matrix <- BIC(lm_model1$finalModel, lm_model2$finalModel, lm_model3$finalModel,
                  lm_model4$finalModel, lm_model5$finalModel, lm_model6$finalModel, lm_model7$finalMode
Warning in BIC.default(lm_model1$finalModel, lm_model2$finalModel,
lm model3$finalModel, : models are not all fitted to the same number of
observations
#removing the first column
BIC_matrix[,1] <- NULL</pre>
#appending them to the results matrix
```

 $lm_model4 \leftarrow train(y \sim lag(y, 2) + lag(y, 3), data = tsdf,$

```
results_matrix2['AIC'] <- AIC_matrix
results_matrix2['BIC'] <- BIC_matrix
```

```
\#similar\ cross-validation\ to\ above,\ except\ using\ k-folds
#making 7 different possible Autoregressive models (1, 2, 3) using loocv in caret
train_ct2 <- trainControl(method = "cv", number = 10)</pre>
# Creating a training set using 80% of the data
inTrain2 <- createDataPartition(y = tsdf$y, p = 0.8, list = FALSE)</pre>
#training data
train_set2 <- tsdf[inTrain2, ]</pre>
#test data (with the other 20 percent)
test_set2 <- tsdf[-inTrain2, ]</pre>
# 1,2,3 lag structure
lm_model_cv1 \leftarrow train(y \sim lag(y) + lag(y, 2) + lag(y, 3), data = train_set2,
                    na.action = na.pass,
                    trControl = train ct2,
                    method = "lm")
# 1 2 lag structure
lm_model_cv2 <- train(y ~ lag(y) + lag(y, 2), data = train_set2,</pre>
                    na.action = na.pass,
                    trControl = train_ct2,
                    method = "lm")
# 1 3 lag structure
lm_model_cv3 \leftarrow train(y \sim lag(y) + lag(y, 3), data = train_set2,
                    na.action = na.pass,
                    trControl = train_ct2,
                    method = "lm")
lm_model_cv4 \leftarrow train(y \sim lag(y, 2) + lag(y, 3), data = train_set2,
                    na.action = na.pass,
                    trControl = train ct2,
                    method = "lm")
lm_model_cv5 <- train(y ~ lag(y), data = train_set2,</pre>
                    na.action = na.pass,
                    trControl = train ct2,
                    method = "lm")
lm_model_cv6 <- train(y ~ lag(y, 2), data = train_set2,</pre>
                    na.action = na.pass,
                    trControl = train_ct2,
                    method = "lm")
lm_model_cv7 <- train(y ~ lag(y, 3), data = train_set2,</pre>
                    na.action = na.pass,
```

```
trControl = train_ct2,
                    method = "lm")
beepr::beep("coin")
#Getting the results and appending them to the same matrix as before
models <- c(lm_model_cv1, lm_model_cv2, lm_model_cv3,</pre>
            lm_model_cv4, lm_model_cv5, lm_model_cv6, lm_model_cv7)
predictions1 <- predict(lm model cv1, test set2)</pre>
predictions2 <- predict(lm_model_cv2, test_set2)</pre>
predictions3 <- predict(lm_model_cv3, test_set2)</pre>
predictions4 <- predict(lm_model_cv4, test_set2)</pre>
predictions5 <- predict(lm_model_cv5, test_set2)</pre>
predictions6 <- predict(lm_model_cv6, test_set2)</pre>
predictions7 <- predict(lm_model_cv7, test_set2)</pre>
#post prediction resampling
model123 <- postResample(predictions1, test_set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
length
model12 <- postResample(predictions2, test_set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
length
model13 <- postResample(predictions3, test_set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
length
model23 <- postResample(predictions4, test set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
```

length

```
model1 <- postResample(predictions5, test_set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
length
model2 <- postResample(predictions6, test set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
length
model3 <- postResample(predictions7, test set2$y)</pre>
Warning in pred - obs: longer object length is not a multiple of shorter object
length
Warning in pred - obs: longer object length is not a multiple of shorter object
length
#creating a new matrix and binding the RMSE values
results3 <- rbind(model123, model12, model13, model23, model2, model3) %>%
  as.data.frame() %>%
  subset(select = RMSE)
#Binding it to the original matrix
results_matrix2 <- cbind(results_matrix2, results3)</pre>
#renaming the columns
colnames(results_matrix2)[6] <- "k_RMSE"</pre>
```

Here is the section with the Model Selection using the four models

```
#importing the data
data <- read_csv("data.csv") %>% na.omit()

Warning: Missing column names filled in: 'X1' [1]

Parsed with column specification:
cols(
    X1 = col_double(),
    DATE = col_date(format = ""),
    fl_nonfarm = col_double(),
```

```
fl_lf = col_double(),
 us_epr_25to54 = col_double(),
 fl_bp = col_double()
)
data[3:6] <- log(data[3:6])
colnames(data)[3:6] <- c("ln_fl_nonfarm", "ln_fl_lf", "ln_us_epr", "ln_fl_bp")</pre>
head(data)
# A tibble: 6 x 6
    X1 DATE
                 ln_fl_nonfarm ln_fl_lf ln_us_epr ln_fl_bp
 <dbl> <date>
                         <dbl>
                                 <dbl>
                                          <dbl>
                                                   <dbl>
  589 1988-01-01
                          8.51
                                            4.11
                                                    9.37
                                  15.6
1
2 590 1988-02-01
                         8.52
                                 15.6
                                            4.11
                                                    9.41
3 591 1988-03-01
                         8.53
                                 15.6
                                           4.12
                                                   9.66
  592 1988-04-01
                         8.53
                                  15.6
                                           4.12
                                                    9.54
5 593 1988-05-01
                         8.53 15.6
                                           4.13
                                                  9.61
6 594 1988-06-01
                          8.53 15.6
                                           4.14 9.87
```

First Model

```
#Making the four different models
#creating a new dataframe
#FIRST MODEL

data['d.nonfarm'] <- difference(data$ln_fl_nonfarm, differences = 1)

data['d.nonfarm_lag'] <- difference(data$ln_fl_nonfarm, lag = 12, difference = 1)

data['d.lf_lag'] <- difference(data$ln_fl_lf, lag = 12, differences = 1)

data['d.fl_bp_lag'] <- difference(data$ln_fl_bp, lag = 12, differences = 1)

data['d.usepr'] <- difference(data$ln_us_epr, lag = 12, differences = 1)

months <- yearmonth(data$DATE) %>%
    format(format = "%m") %>%
    as.factor()

data['months'] <- months</pre>
```

Second Model

Third Model

Fourth Model

```
#FOURTH MODEL
data['d.lf_lag'] <- difference(data$ln_fl_lf, lag = 2, differences = 1)
data['d.lf_lag_12'] <- difference(data$ln_fl_lf, lag = 12, differences = 1)
data['d.lf_lag_24'] <- difference(data$ln_fl_lf, lag = 24, differences = 1)
data['d.fl_bp_lag'] <- difference(data$ln_fl_bp, lag = 2, differences = 1)
data['d.fl_bp_12'] <- difference(data$ln_fl_bp, lag = 12, differences = 1)
data['d.fl_bp_24'] <- difference(data$ln_fl_bp, lag = 24, differences = 1)
data['d.usepr'] <- difference(data$ln_us_epr, lag = 2, differences = 1)
data['d.usepr_12'] <- difference(data$ln_us_epr, lag = 12, differences = 1)</pre>
```

```
data['d.usepr_24'] <- difference(data$ln_us_epr, lag = 24, differences = 1)
#LOOCV
model_4 <- train(d.nonfarm ~ d.nonfarm_lag + d.lf_lag + d.lf_lag_12 + d.lf_lag_24 + d.fl_bp_lag +
                   d.fl_bp_12 + d.fl_bp_24 + d.usepr + d.usepr_12 + d.usepr_24 + months,
                 na.action = na.exclude,
                 data = data,
                 trControl = trainControl(method = "LOOCV"),
                 method = "lm")
#writing results to final table
final_results <- rbind(final_results, model_4$results)</pre>
final results[,1] <- NULL</pre>
#Matrix of the AIC values
AIC_final <- AIC(model_1\finalModel, model_2\finalModel, model_3\finalModel,
                  model_4$finalModel)
Warning in AIC.default(model_1$finalModel, model_2$finalModel,
model_3$finalModel, : models are not all fitted to the same number of
observations
#Removing the first column
AIC_final[,1] <- NULL
BIC_final <- BIC(model_1$finalModel, model_2$finalModel, model_3$finalModel,
                  model_4$finalModel)
Warning in BIC.default(model_1$finalModel, model_2$finalModel,
model_3$finalModel, : models are not all fitted to the same number of
observations
#removing the first column
BIC_final[,1] <- NULL
#appending them to the results matrix
final_results['AIC'] <- AIC_final</pre>
```

RMSE	Rsquared	MAE	AIC	BIC
0.0047814	0.7733409	0.0035833	-2910.616	-2840.124
0.0047430	0.7769618	0.0035699	-2917.164	-2846.672
0.0047580	0.7755934	0.0035853	-2915.047	-2832.807
0.0047374	0.7768743	0.0035409	-2822.839	-2733.522

final_results['BIC'] <- BIC_final</pre>

kable(final_results, format = "latex") %>%

kable_styling(position = "center", latex_options = "striped")

beepr::beep("coin")