All R code in PDF

Sam Ansari, s3765571

4 juni 2020

ALL THE R CODE WE USED

Library

```
library("DataCombine")
library('ggplot2')
library("corrplot")
library("tidyverse")
library("dplyr")
library("openxlsx")
library("tseries")
library('fpp2')
                # For forecasting
library('dynlm') # To estimate ARDL models
library('urca') # For the Dickey Fuller test
library('corrplot')# For plotting correlation matrices
library('quadprog')# For quadratic optimization
library('forecast')
library('readxl') # To read Excel files
library('fpp2') # For forecasting
library('tseries') # To estimate ARMA models
library('dynlm') # To estimate ARDL models
library('urca') # For the Dickey Fuller test
library('corrplot')# For plotting correlation matrices
library('quadprog')# For quadratic optimization
library('forecast')# Lots of handy forecasting routines
library('vars')
                  # VARs
library('zoo')
library('lubridate')
```

Data loading

```
data_1 <- read.xlsx("WEI.xlsx", sheet = 2, detectDates = TRUE)
sp500_newdata <- read.csv("sp500newdata.csv")
CCI <- read.csv('CCI.csv')</pre>
```

Data manipulation

```
sp500_newdata <- sp500_newdata %>%
  mutate(average_open_close = (Open + Close) / 2)
data <- data %>% cbind(sp500data$average_open_close)
```

```
colnames(data)[11] <- "average_open_close"</pre>
sp500_52week_change <- PercChange(data = sp500_newdata, Var = "average_open_close", NewVar = "sp500_52w
sp500_52week_change <- sp500_52week_change$sp500_52week_change</pre>
sp500_52week_change <- sp500_52week_change[!is.na(sp500_52week_change)]</pre>
data_1$sp500_52week_change <- sp500_52week_change
sp_500_52week_diff <- diff(sp500_newdata$average_open_close, lag = 52)</pre>
data_1$sp_500_52week_diff <- sp_500_52week_diff
#read CSV file and obtain data from 2007-2020, with values around 0
CCI_data = CCI %>% slice(3:nrow(CCI)) %>% mutate(percentage = Value - 100)
               <- ts(CCI_data[,9],start = 2007,frequency=12)
#Take difference with respect to the value of last year
diff_CCI = diff(CCI_2007, 12)
diff_CCI = ts(as.vector(diff_CCI), start = 2008, frequency = 12)
# Merge low and high freq time series
           <- zoo(diff_CCI,time(diff_CCI))</pre>
lowfreq
highfreq
             <- zoo(WEI,time(WEI))
             <- merge(lowfreq, highfreq)
merged
# Approximate the NAs and output at the dates of the WEI
             <- na.approx(merged$lowfreq, xout = time(WEI),rule=2)</pre>
             <- ts(CCIw, start = 2008, frequency=52)
CCIw
data_1$CCIw =as.vector(CCIw)
#preparing all time series
              <- ts(data_1$WEI, decimal_date(ymd("2008-01-05")), frequency = 365.25/7)</pre>
WEI 365
               <- ts(data_1$CCIw, decimal_date(ymd("2008-01-05")), frequency = 365.25/7)
CCIw 365
                                <- ts(data_1$sp500_52week_change, decimal_date(ymd("2008-01-05")), frequ</pre>
sp500_52week_change_365
                              <- ts(data_1$sp_500_52week_diff, decimal_date(ymd("2008-01-05")), frequenc</pre>
sp_500_52week_diff_365
noise<-ts(rnorm(length(CCIw_365))*sqrt(sd((CCIw_365)/100)),decimal_date(ymd("2008-01-05")),frequency=36
CCIn <- CCIw_365+noise
```

Plots we used

```
plot_WEI_SP500_CCI <- ggplot(data = data_1, aes(x = Date)) +
    geom_line(aes(y = WEI, colour = "WEI")) +
    geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500")) +
    geom_line(aes(y = CCIw * 1.5, colour = "CCI")) +
    geom_hline(yintercept = 0, colour = 'black') +
    scale_color_manual("", values = c("WEI" = "green", "S&P500" = "blue", "CCI" = "red")) +
    ggtitle("WEI vs S&P500 52 week % change scaled by 10 vs CCI scaled by 1.5") +
    ylab("WEI, S&P500 and CCI") +
    xlab("Date")
    plot_WEI_SP500_CCI

plot_WEI_SP500_CCI_2008_2010 <- ggplot(data = data_1[1:105, ], aes(x = Date)) +
    geom_line(aes(y = WEI, colour = "WEI")) +
    geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500")) +
    geom_line(aes(y = CCIw * 1.5, colour = "CCI")) +
    geom_hline(yintercept = 0, colour = 'black') +</pre>
```

```
scale_color_manual("", values = c("WEI" = "green", "S&P500" = "blue", "CCI" = "red")) +
  ggtitle("WEI vs S&P500 52 week % change scaled by 10 vs CCI scaled by 1.5 during 2008-2010") +
  ylab("WEI, S&P500 and CCI") +
  xlab("Date")
plot_WEI_SP500_CCI_2008_2010
plot_WEI_SP500_CCI_before_covid <- ggplot(data = data_1[560:630,], aes(x = Date)) +
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500")) +
  geom_line(aes(y = CCIw * 1.5, colour = "CCI")) +
  geom_hline(yintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500" = "blue", "CCI" = "red")) +
  ggtitle("WEI vs S&P500 52 week % change scaled by 10 vs CCI scaled by 1.5 before covid") +
  ylab("WEI, S&P500 and CCI") +
  xlab("Date")
plot_WEI_SP500_CCI_before_covid
plot_WEI_SP500_CCI_during_covid <- ggplot(data = data_1[630:639, ], aes(x = Date)) +
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500")) +
  geom_line(aes(y = CCIw * 1.5, colour = "CCI")) +
  geom_hline(yintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500" = "blue", "CCI" = "red")) +
  ggtitle("WEI vs S&P500 52 week % change scaled by 10 vs CCI scaled by 1.5 during covid") +
  ylab("WEI, S&P500 and CCI") +
  xlab("Date")
plot_WEI_SP500_CCI_during_covid
Corrplot
data 2 <- data 1
colnames(data_2)[11:13] <- c("S&P500 %change", "S&P500 difference", "CCI")
z <- cbind(data_2[4:8], data_2[11:13])
Z \leftarrow cor(z)
corrplot(Z, method = "color")
(P)ACF plots and Dicky-Fuller tests
acf(data_1$WEI)
pacf(data 1$WEI)
# From the ACF plots we suspected non covariance stationarity, so we conducted some Dicky-Fuller tests.
# For the WEI
        <- floor(12*((length(data_1$WEI)/100)^0.25))
        <- ur.df(data_1$WEI,type=c("drift"),lags=pmax,selectlags=c("BIC"))</pre>
summdf <- summary(dft)</pre>
print(summdf@test.name)
print(c("Test statistic: ", summdf@teststat[1]))
print(c("Crit. vals", summdf@cval[1,]))
# For the S&P500
pmax <- floor(12*((length(data_1$sp500_52week_change)/100)^0.25))
```

BIC and AIC

```
# Information criteria for the WEI.
bic_WEI = matrix(NA,7,6)
aic_WEI = matrix(NA,7,6)
T_{est} = matrix(NA, 7, 6)
for (i in seq(2,8)){
  for (j in seq(0,5)){
    fit = Arima(WEI, order = c(i,0,j))
    T_est[i-1, j+1] = length(fit$residuals)
    bic_WEI[i-1,j+1] = fit$bic
    aic_WEI[i-1,j+1] = fit$aic
  }
}
T_est
colnames(bic_WEI) <- c("MA(0)","MA(1)","MA(2)","MA(3)","MA(4)",'MA(5)')</pre>
rownames(bic_WEI) <- c('AR(2)',"AR(3)","AR(4)","AR(5)","AR(6)","AR(7)",'AR(8)')
bic WEI #table 1
min_values_bic= sort(bic_WEI)[1:3]
min_index_bic=c()
for (i in 1:3){
  min_index_bic[i] = which(bic_WEI==min_values_bic[i])
}
min_index_bic
```

```
# Information criteria for the VAR model.
colnames(aic_WEI) <- c("MA(0)","MA(1)","MA(2)","MA(3)","MA(4)",'MA(5)')
rownames(aic_WEI) <- c('AR(2)',"AR(3)","AR(4)","AR(5)","AR(6)","AR(7)",'AR(8)')
aic_WEI #table 2
min_values_aic= sort(aic_WEI)[1:3]
min_index_aic=c()
for (i in 1:3){
    min_index_aic[i] = which(aic_WEI==min_values_aic[i])
}
min_index_aic</pre>
```

```
# BIC graph
           <- cbind(CCIw_365, sp500_52week_change_365 , WEI_365)</pre>
Y
colnames(Y) <- c('CCI', 'SP500', 'WEI')</pre>
VARmodel_ic <- VARselect(Y,type=c("const"),lag.max=8)</pre>
           <- as.data.frame(t(VARmodel ic$criteria))
ic
ggplot(data=ic, aes(x=seq(1,8),y=`SC(n)`))+geom_line()+ylab("BIC")+xlab("VAR(p)")
ggplot(data=ic, aes(x=seq(1,8),y=`AIC(n)`))+geom_line()+ylab("AIC")+xlab("VAR(p)")
WEI residual analysis
#residual autocorrelation
fit_1 \leftarrow Arima(WEI, order = c(2,0,3)) #figure 4
checkresiduals(fit 1)
fit_2 \leftarrow Arima(WEI, order = c(3,0,0))
checkresiduals(fit_2)
fit_3 \leftarrow Arima(WEI, order = c(2,0,0))
checkresiduals(fit_3)
fit_4 <- Arima(WEI, order = c(5,0,4)) #figure 5</pre>
checkresiduals(fit_4)
fit_5 \leftarrow Arima(WEI, order = c(6,0,4))
checkresiduals(fit_5)
fit_6 \leftarrow Arima(WEI, order = c(7,0,5))
checkresiduals(fit 6)
0,0,0,0,0,0,0,0,0,0,0,0,NA,NA,NA,NA,NA,NA))
checkresiduals(fit 7)
autoplot(fit_1)
fit_1
autoplot(fit_2)
fit_2
autoplot(fit_3)
fit_3
autoplot(fit 4)
fit 4
autoplot(fit_5)
fit 5
autoplot(fit_6)
fit 6
```

```
autoplot(fit_7)
fit_7
```

Forecasting

```
## ARMA forecasts:
# ARMA(2,3)
fit_1 <- Arima(WEI_365, order = c(2,0,3))
fARMA_1 <- forecast(fit_1,h=208)
autoplot(fARMA_1)

# VAR(3)
Y <- cbind(WEI_365, CCIw_365 , sp500_52week_change_365 )
VAR3 <- VAR(Y,p=3,type = c('const'))
fVAR3 <- forecast(VAR3, h=208)
autoplot(fVAR3$forecast$WEI)
VAR3$varresult$WEI$coefficients</pre>
```

MSE/MAE

```
#MSE of the ARMA models
      <- as.Date("2008/1/5") # Estimation start
fs
       <- as.Date("2016/1/2") # First forecast
       <- as.Date("2020/03/21")# Final forecast
convert_date <- function(date){</pre>
  c(as.numeric(format(date, '%Y')),
    ceiling(as.numeric(format(date,'%W'))))
  # Use %W for weeks and do not divide by 3.
dates <- seq(fs,fe,by="week") # (or "week"...)</pre>
       <- length(dates)
                                        # number of forecasts
qF
       <- convert_date(fs)</pre>
       <- convert date(fe)
target <- window(WEI_365,start=qF,end=qL)</pre>
in_out_ARMA = function(hor, p, q){
        <- ts(data=matrix(NA,n,1),start=qF,frequency=365.25/7)</pre>
       <- ts(data=matrix(NA,n,1),start=qF,frequency=365.25/7)
  fce
  for (i_d in seq(1,n)){
    # Define estimation sample (ends h periods before 1st forecast)
    # Start at the first forecast date,
    # Then move back h+1 quarters back in time
   est <- seq(dates[i_d],length=hor+1, by = "-1 week")[hor+1]
    # Now define the data we can use to estimate the model
   yest <- window(WEI_365,end=convert_date(est))</pre>
    # Fit the AR models using Arima
   fit
               <- Arima(yest, order=c(p,0,q)) #Fit model
   fc[i_d,1]
                 <- forecast(fit,h=hor)$mean[hor]#Get forecast</pre>
   fce[i_d,1] <- fc[i_d,1]-target[i_d]
                                              #Get forecast error
```

```
results
                  <- list()
  results$fc
                  <- fc
  results$fce
                  <- fce
  results$target <- target
  return(results)
}
          <-c(26,52,104)
h all
                              # Which horizons to consider
          <- length(h_all)
mseARMA <- matrix(NA,1h,3) # Full sample
p = c(2,3,5)
q = c(3,0,4)
parameters = as.data.frame(cbind(p,q))
for (p in 1:3){
 for (i in seq(1,1h)){
                        <- in_out_ARMA(h_all[i],parameters[p,1],parameters[p,2])</pre>
    fcARMA
                    <- colMeans(fcARMA$fce^2, na.rm = T)
    mseARMA[i,p]
  }
}
rownames(mseARMA) <- c("26-step","52-step","104-step")</pre>
colnames(mseARMA) \leftarrow c('ARMA(2,3)', 'ARMA(3,0)', 'ARMA(5,4)')
mseARMA
```

```
# Absolute error/MAE
          <-c(26,52,104)
                                # Which horizons to consider
h_all
lh
          <- length(h_all)
abeARMA <- matrix(NA,lh,3)
p = c(2,3,5)
q = c(3,0,4)
parameters = as.data.frame(cbind(p,q))
for (p in 1:3){
  for (i in seq(1,lh)){
    fcARMA
                        <- in_out_ARMA(h_all[i],parameters[p,1],parameters[p,2])</pre>
    abeARMA[i,p]
                    <- colMeans(abs(fcARMA$fce), na.rm = T)
  }
}
rownames(abeARMA) <- c("26-step","52-step","104-step")</pre>
colnames(abeARMA) \leftarrow c('ARMA(2,3)', 'ARMA(3,0)', 'ARMA(5,4)')
abeARMA
```

IRF

```
plot(irf_WEI,plot.type=c("single"))
irf_WEI_CCI <- irf(VARmodel,impulse=c("CCI"),</pre>
                   response=c("WEI"), ortho=T, n.ahead = 208)
plot(irf_WEI_CCI,plot.type=c("single"))
Combined forecast
fit 1 <- Arima(WEI 365, order = c(2,0,3))
fARMA_1 <- forecast(fit_1,h=208)</pre>
autoplot(fARMA_1)
Y <- cbind(WEI_365, CCIw_365 , sp500_52week_change_365 )
VAR3 <- VAR(Y,p=3,type = c('const'))</pre>
fVAR3 <- forecast(VAR3, h=208)
autoplot(fVAR3$forecast$WEI)
VAR3$varresult$WEI$coefficients
fcombined = matrix(0,length(fARMA 1$mean),6)
for (i in 1:208){
  fcombined[i,2] = 0.5*as.numeric(fVAR3$forecast$WEI_365$mean[i])+0.5*as.numeric(fARMA_1$mean[i])
  fcombined[i,3] = 0.5*as.numeric(fVAR3$forecast$WEI_365$lower[i,1])+0.5*as.numeric(fARMA_1$lower[i,1])
  fcombined[i,4] = 0.5*as.numeric(fVAR3$forecast$WEI_365$lower[i,2])+0.5*as.numeric(fARMA_1$lower[i,2])
  fcombined[i,5] = 0.5*as.numeric(fVAR3$forecast$WEI_365$upper[i,1])+0.5*as.numeric(fARMA_1$upper[i,1])
  fcombined[i,6] = 0.5*as.numeric(fVAR3$forecast$WEI_365$upper[i,2])+0.5*as.numeric(fARMA_1$upper[i,2])
}
combinedForecast_1 = ts(c(as.vector(WEI_365),fcombined[,2]), decimal_date(ymd("2008-01-05")), frequence
combinedForecast_low1 = ts( c(as.vector(WEI_365),fcombined[,3]), decimal_date(ymd("2008-01-05")), frequ
combinedForecast_low2 = ts( c(as.vector(WEI_365),fcombined[,4]), decimal_date(ymd("2008-01-05")), frequ
combinedForecast_high1 = ts( c(as.vector(WEI_365),fcombined[,5]), decimal_date(ymd("2008-01-05")), freq
combinedForecast_high2 = ts( c(as.vector(WEI_365),fcombined[,6]), decimal_date(ymd("2008-01-05")), freq
ts.plot(combinedForecast_low1, combinedForecast_low2, combinedForecast_high1, combinedForecast_high2, c
        col= c('#4842f5','#00b5af','#4842f5', '#00b5af','#000000'), ylab = 'WEI', main = 'Combined Var(
legend('bottomleft', legend = c('95% low', '80 low', '95% high', '80% high', 'forecast'), col = c('#484
# The calculation of the SSR of the combined model.
fcombined2 = matrix(0,636,2)
for (i in 4:639){
  fcombined2[i-3,2] = 0.5*as.numeric(VAR3$varresult$WEI_365$fitted.values[i-3])+0.5*as.numeric(fit_1$fi
residuals_combined = c()
for(i in 4:639){
  residuals_combined[i-3] = as.vector(WEI_365)[i] - fcombined2[i-3,2]
}
SSR_c = sum(residuals_combined^2)
SSR_VAR = sum(as.numeric(VAR3$varresult$WEI_365$residuals)^2)
SSR_ARMA = sum(as.numeric(fit_1$residuals)[4:639]^2)
SSR = matrix(c(SSR c, SSR VAR, SSR ARMA),1,3)
rownames(SSR) <- c("SSR")</pre>
colnames(SSR) <- c('Combined','VAR(3)','ARMA(2,3)')</pre>
SSR
```

This is code we did not use in the paper, but gave different insights

```
# From these plots we suspected non covariance stationarity
acf(data_1$sp500_52week_change)
pacf(data_1$sp500_52week_change)
acf(data_1$CCIw)
pacf(data_1$CCIw)
```

Restricted Var

```
Y <- cbind(CCIw_365 , sp500_52week_change_365 , WEI_365)

colnames(Y) <- c('CCI', 'SP500', 'WEI')

VARmodel_ic <- VARselect(Y,type=c("const"),lag.max=8)

ic <- as.data.frame(t(VARmodel_ic$criteria))

ic

ggplot(data=ic, aes(x=seq(1,8),y=`SC(n)`))+geom_line()+ylab("BIC")+xlab("VAR(p)")

ggplot(data=ic, aes(x=seq(1,8),y=`AIC(n)`))+geom_line()+ylab("AIC")+xlab("VAR(p)")
```

```
#restricted VAR
         <- 6;
р1
         <- VAR( Y,p=p1,type=c("const"))
VARr
nseries <- 3;
        <- matrix(1,nrow = nseries,ncol=nseries)
#mzero
         <- matrix(0, nrow = nseries, ncol=nseries)
        <- matrix(1,nrow = nseries,ncol=1)
vones
lag1mat <- matrix(c(1, 1, 1,</pre>
                     1, 1, 1,
                     1, 1, 1)
                   ,nrow = nseries,ncol=nseries, byrow = TRUE) # lag matrix cols = cci, sp500 and WEI. R
lag2mat <- matrix(c(0, 0, 0,
                     0, 0, 0,
                     0, 0, 0)
                   ,nrow = nseries,ncol=nseries, byrow = TRUE)
lag3mat <- matrix(c(1, 1, 1,</pre>
                     1, 1, 1,
                     1, 1, 1)
                   ,nrow = nseries,ncol=nseries, byrow = TRUE)
lag4mat \leftarrow matrix(c(0, 0, 0,
                     0, 0, 0,
                     0, 0, 0)
                   ,nrow = nseries,ncol=nseries, byrow = TRUE)
lag5mat <- matrix(c(1, 1, 1,</pre>
                     1, 1, 1,
                     1, 1, 1)
                   ,nrow = nseries,ncol=nseries, byrow = TRUE)
lag6mat <- matrix(c(0, 0, 0,
                     0, 0, 0,
                     0, 0, 0)
                   ,nrow = nseries,ncol=nseries, byrow = TRUE)
lag7mat <- matrix(c(1, 1, 1,</pre>
                     1, 1, 1,
                     1, 1, 1)
```

```
,nrow = nseries,ncol=nseries, byrow = TRUE)
lag8mat \leftarrow matrix(c(0, 0, 0,
                   0, 0, 0,
                   0, 0, 0)
                  ,nrow = nseries,ncol=nseries, byrow = TRUE)
lag9mat <- matrix(c(1, 1, 1,</pre>
                   1, 1, 1,
                   1, 1, 1)
                  ,nrow = nseries,ncol=nseries, byrow = TRUE)
restrict <- matrix(cbind(lag1mat, lag2mat, lag3mat, lag4mat, lag5mat, lag6mat, vones), nrow = 3, ncol
        <- restrict(VARr, method = "man", resmat = restrict)</pre>
VARr
# Somehow BIC has to be calculated by hand
resid <- residuals(VARr)</pre>
        <- length(resid[,1])
        <- log(det(t(resid)%*%resid/T)) + (log(T)/T)*sum(restrict)
BIC
BIC
fVARr <- forecast(VARr, h=200)</pre>
autoplot(fVARr$forecast$WEI)
VARr$varresult$WEI$coefficients
# You can check that now the third lag is omitted by typing
summary(VARr)
roots(VARr)
irf_WEI <- irf(VARr,impulse=c("SP500"),</pre>
              response=c("WEI"),ortho=T, n.ahead = 300)
plot(irf_WEI,plot.type=c("single"))
irf_CCI <- irf(VARr,impulse=c("SP500"),</pre>
              response=c("CCI"), ortho=T, n.ahead = 300)
plot(irf_CCI,plot.type=c("single"))
irf_WEI_CCI <- irf(VARr,impulse=c("CCI"),</pre>
                  response=c("WEI"), ortho=T, n.ahead = 300)
plot(irf_WEI_CCI,plot.type=c("single"))
rejected ARMA
fit_2 \leftarrow Arima(WEI, order = c(5,0,4)) #figure 5
fARMA_2 <- forecast(fit_2,h=208)</pre>
autoplot(fARMA_2)
0,0,0,0,0,0,0,0,0,0,0,0,NA,NA,NA,NA,NA,NA))
fARMA_3 <- forecast(fit_3,h=208)</pre>
autoplot(fARMA_3)
```

check ARDL and VAR

uin sample ARDL test

```
forecastARDL <- function(y,X,es,fs,fe,maxARp,hor){</pre>
  dates <- seq(fs,fe,by="week") # (or "week"...)
  n
          <- length(dates)
                                            # number of forecasts
  qΓ
          <- convert_date(fs)</pre>
 qL
          <- convert_date(fe)</pre>
  target <- window(y, start=qF, end=qL) # What we are forecasting.
  # Define ts objects where forecasts/forecast errors are saved.
  # (Note that frequency=4 applies to quarterly data!)
        <- ts(data=matrix(NA,n,maxARp),start=qF,frequency=365.25/7)</pre>
  fce
        <- ts(data=matrix(NA,n,maxARp),start=qF,frequency=365.25/7)</pre>
  for (i d in seq(1,n)){
    # Define estimation sample (ends h periods before 1st forecast)
    # Start at the first forecast date,
    # Then move back h+1 quarters back in time
    est <- seq(dates[i_d],length=hor+1, by = "-1 week")[hor+1]</pre>
    # Now define the data we can use to estimate the model
    Y = cbind(y, X)
    Yest <- window(Y,end=convert_date(est))</pre>
    # Fit the AR models using Arima
    for (j in seq(1,maxARp)){
                     <- VAR(Yest,p=j,type=c('const')) #Fit model
                     <- forecast(fit,h=hor)$forecast$y$mean[hor]#Get forecast</pre>
      fc[i_d,j]
      fce[i_d,j]
                    <- fc[i_d,j]-target[i_d]
                                                      #Get forecast error
    }
  }
                  <- list()
  results
  results$fc
                  <- fc
 results$fce
                  <- fce
 results$target <- target
  return(results)
}
```

```
# Get forecasts
X_SP <- cbind(WEI_365,sp500_52week_change_365 )</pre>
X CCI <- cbind(WEI 365,CCIw 365)</pre>
fcARDLh1_SP <- forecastARDL(WEI_365,X_SP,es,fs,fe,maxARp,1)</pre>
fcARDLh1 CCI <- forecastARDL(WEI 365,X CCI,es,fs,fe,maxARp,1)</pre>
# Calculate MSE and compare
mseARDL_SP <- colMeans(fcARDLh1_SP$fce^2)</pre>
mseARDL CCI
               <- colMeans(fcARDLh1 CCI$fce^2)</pre>
              <- rbind(mseARMA[1,],mseARDL_SP)
compare_SP
               <- rbind(mseARMA[1,],mseARDL_CCI)
compare_CCI
rownames(compare_SP) <- c("AR", "ARDL")</pre>
colnames(compare_SP) <- c("p=1","p=2","p=3","p=4",'p=5','p=6')</pre>
rownames(compare_CCI) <- c("AR","ARDL")</pre>
colnames(compare_CCI) <- c("p=1","p=2","p=3","p=4",'p=5','p=6')</pre>
round(compare_SP,digits=3)
round(compare_CCI,digits=3)
```

leading indicator analysis

```
plot240 <- ggplot(data = data_1) +
    geom_line(mapping = aes(x = Date, y = sp_500_52week_diff / 100), colour = "red") +
    geom_line(mapping = aes(x = Date, y = WEI), colour = "blue") +
    geom_line(aes(x = Date, y = 0), colour = "black")
plot240

plot239 <- ggplot(data = data_1) +
    geom_line(mapping = aes(x = Date, y = sp500_52week_change / 10), colour = "red") +
    geom_line(mapping = aes(x = Date, y = WEI), colour = "blue") +
    geom_hline(yintercept = 0, colour = 'black') +
    ggtitle("WEI vs S&P500 52 week percentage change") +
    ylab("S&P500 scaled by 10")
plot239</pre>
```

```
correlation <- cor(select(data_1, 4:8, 11:12))
print(correlation)
corrplot(correlation, method = "color", na.remove = TRUE)</pre>
```

SP500 zorgen dat je een 52 weken percentage change difference neemt, die plotten tegenover WEI zonder veranderingen.

Oil price, 52 weken percentage change dan verschillen over 52 weken (misschien doet deze stap te weinig voor verlies aan data, dit zelf bekijken)

(vanwege inflatie)

midas modellen

```
SP500change <- PercChange(data = data, Var = "S&P500", NewVar = "SP500change")
SP500change <- SP500change $$SP500change
data$$SP500change <- SP500change

sp500_perc_change<- PercChange(data = data, Var = "average_open_close", NewVar = "sp500_perc_change")
sp500_perc_change <- sp500_perc_change$sp500_perc_change
data$sp500_perc_change <- sp500_perc_change
data$lnSP500 <- log(data$S&P500")
data$lnBB <- log(data$BB)
data$lnM1 <- log(data$M1)
data$lnM1 <- log(data$Oil)

WEI_time_series_change <- diff(data$WEI)
WEI_time_series_change <- append(WEI_time_series_change, 0, after = 0)
data <- data %>% cbind(WEI_time_series_change)
```

Plotting several variables against the WEI to identify some correlation

```
plot1 <- ggplot(data = data, mapping = aes(x = BB, y = WEI)) +
    geom_point()
plot2 <- ggplot(data = data, mapping = aes(x = T10Y3M, y = WEI)) +
    geom_point()
plot3 <- ggplot(data = data, mapping = aes(x = M1, y = WEI)) +
    geom_point()
plot3

plot4 <- ggplot(data = data) +
    geom_line(mapping = aes(x = Date, y = WEI), color = "blue") +
    geom_line(mapping = aes(x = Date, y = T10Y3M, color = "red"))
plot4</pre>
```

```
plot5 <- ggplot(data = data) +</pre>
  geom_line(mapping = aes(x = Date, y = WEI), color = "blue", label = "WEI") +
  geom_line(mapping = aes(x = Date, y = sp500_perc_change, color = "red", label = "Change in BB")) +
  geom_line(mapping = aes(x = Date, y = T10Y3M, color = "black", label = "Bond rates")) +
  geom_line(aes(x = Date, y = 0))
plot5
plot6 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = ln0il, color = "red")) +
  geom_line(aes(x = Date, y = lnSP500, color = "blue")) +
  geom_line(aes(x = Date, y = lnBB, color = "black")) +
  geom_line(aes(x = Date, y = lnM1, color = "white"))
plot6
plot7 <- ggplot(data = data, aes(x = Date)) +
  geom_line(aes(y = M1change, color = "Money supply")) +
  geom_line(aes(y =WEIchange, colour = "WEI"))+
  geom_hline(yintercept = 0, color = 'black') + scale_colour_manual("",
                                                                      values = c("Money supply"="blue", ""
  ggtitle("WEI vs Money supply percentage change") +
  ylab("WEI and Money supply percentage change")
plot7
plot8 <- ggplot(data = data, aes(x = Date)) +</pre>
  geom_line(aes(y =BBchange, colour = "Bank borrowings")) +
  geom_line(aes(y =WEI, colour = "WEI"))+
  geom_hline(yintercept = 0, color = 'black') + scale_colour_manual("",
                                                                      values = c("Bank borrowings"="blue"
  ggtitle("WEI vs Bank borrowings percentage change") +
  ylab("WEI and the BB percentage percentage change")
plot8
plot9 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = BBchange, color = "darkred")) +
  geom_line(aes(x = Date, y = M1change, color = "lightblue"))
plot9
plot10 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = WEI, color = "Red")) +
  geom_line(aes(x = Date, y = SP500change, color = "Blue"))
plot10
plot11 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = Oil*10, color = "red")) +
  geom_line(aes(x = Date, y = WEI*100, color = "blue"))
plot11
plot12 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = WEI*1000, color = "red")) +
  geom_line(aes(x = Date, y = `S&P500`*10, color = "blue"))
plot12
plot13 <- ggplot(data = data) +</pre>
```

```
geom_line(mapping = aes(x = Date, y = BBchange, color = "red"))
plot13
plot14 <- ggplot(data = sp500data) +
  geom_line(aes(x = Date, y = average_high_low, group = 1, color = "darkred")) +
  geom_line(aes(x = Date, y = average_open_close, group = 1, color = "lightblue"))
plot14
plot15 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = SP500change, color = "darkred")) +
  geom_line(aes(x = Date, y = sp500_perc_change, color = "lightblue"))
plot15
plot16 <- ggplot(data = data, aes(x= Date)) +</pre>
  geom_line(aes(y = sp500_perc_change, colour = "S&P500")) +
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_hline(yintercept = 0, colour = 'black') + scale_colour_manual("", values = c("S&P500"="blue", "W
  ggtitle("The WEI vs S&P500 percentage changes") +
  ylab("WEI and S&P500 percentage changes")
plot16
data[(which.min(data$WEIchange) - 1):(which.min(data$WEIchange) + 1), ]
plot(x = data$Date, y = data$WEI, type = "1")
data(length(data$WEIchange - 10))
par(mfcol = c(2,2))
plot(data$BB ,data$WEI)
plot(data$T10Y3M, data$WEI)
plot(data$M1, data$WEI)
plot18 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = WEI_time_series_change, color = "darkred")) +
  geom_line(aes(x = Date, y = BBchange / 10, color = "lightblue")) +
  geom_line(aes(x = Date, y = 0 , color = "green")) +
  ggtitle("Change in WEI vs BB growth rate")
plot18
plot19 <- ggplot(data = data) +</pre>
  geom_line(aes(x = Date, y = WEI_time_series_change, color = "darkred")) +
  geom_line(aes(x = Date, y = M1change / 2, color = "lightblue")) +
  geom_line(aes(x = Date, y = 0 , color = "green")) +
  ggtitle("Change in WEI vs M1 growth rate")
plot19
plot20 <- ggplot(data = data[560:630, ]) +
  geom_line(aes(x = Date, y = WEI_time_series_change, color = "darkred")) +
  geom_line(aes(x = Date, y = diff_oil_price / 10, color = "lightblue")) +
  geom_line(aes(x = Date, y = 0 , color = "green")) +
  ggtitle("diff in WEI vs diff in Oil price")
plot20
```

```
plot17 <- ggplot(data = data[560:628, ]) +</pre>
  geom_line(aes(x = Date, y = WEI_time_series_change, color = "darkred")) +
  geom_line(aes(x = Date, y = SP500_diff_change / 50, color = "lightblue")) +
  geom_line(aes(x = Date, y = 0 , color = "green")) +
  ggtitle("Change in WEI vs S&P500 growth rate")
plot17
plottest <- ggplot(data = data[560:628, ]) +</pre>
  geom_line(aes(x = Date, y = WEI_time_series_change, color = "darkred")) +
  geom_line(aes(x = Date, y = sp500_perc_change / 5, color = "lightblue")) +
  geom_line(aes(x = Date, y = 0 , color = "green")) +
  ggtitle("Change in WEI vs S&P500 growth rate")
plottest
plothoertje <- ggplot(data = data[500:600, ]) +</pre>
  geom_line(aes(x = Date, y = sp500_perc_change, color = "darkred")) +
  geom_line(aes(x = Date, y = SP500_diff_change / 50, color = 'lightblue'))
plothoertje
plot17 <- ggplot(data = data, aes(x = Date)) +</pre>
  geom_line(aes(y =WEI_difference100, colour = "WEI difference scaled by 100")) +
  geom_line(aes(y =SP500_time_series_change, colour = "S&P500 difference")) +
  geom_hline(yintercept = 0, colour = 'black') + scale_colour_manual("", values = c("WEI difference sca
                                                                                      "S&P500 difference"
  ggtitle("Difference within WEI vs S&P500 2008-2020") + ylab("Difference with WEI scaled by 100")
plot21
plot_WEI_SP500 <- ggplot(data = data_1, aes(x = Date)) +</pre>
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500 52 week %change scaled by 10")) +
  geom_hline(yintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500 52 week %change scaled by 10" = "blue")) +
  ggtitle("WEI vs 52 week % change of S&P500") +
  ylab("WEI and 52 week % change S&P500") +
  xlab("Date")
plot_WEI_SP500
plot WEI SP500 2008 2010 <- ggplot(data = data 1[1:105, ], aes(x = Date)) +
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500 52 week %change scaled by 10")) +
  geom_hline(yintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500 52 week %change scaled by 10" = "blue")) +
  ggtitle("WEI vs 52 week % change of S&P500 in 2008-2010") +
  ylab("WEI and 52 week % change S&P500") +
  xlab("Date")
plot_WEI_SP500_2008_2010
plot_WEI_SP500_before_covid <- ggplot(data = data_1[560:630, ], aes(x = Date)) +
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500 52 week %change")) +
  geom_hline(yintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500 52 week %change" = "blue")) +
  ggtitle("WEI vs 52 week % change of S&P500 before COVID-19") +
```

```
ylab("WEI and 52 week % change S&P500") +
  xlab("Date")
plot_WEI_SP500_before_covid
plot_WEI_SP500_during_covid <- ggplot(data = data_1[630:639, ], aes(x = Date)) +
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500 52 week %change")) +
  geom hline(vintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500 52 week %change" = "blue")) +
  ggtitle("WEI vs 52 week % change of S&P500 before COVID-19") +
  ylab("WEI and 52 week % change S&P500") +
  xlab("Date")
plot_WEI_SP500_during_covid
plot_WEI_SP500_CCI <- ggplot(data = data_1[1:105, ], aes(x = Date)) +</pre>
  geom_line(aes(y = WEI, colour = "WEI")) +
  geom_line(aes(y = sp500_52week_change / 10, colour = "S&P500")) +
  geom_line(aes(y = CCIw * 1.5, colour = "CCI")) +
  geom_hline(yintercept = 0, colour = 'black') +
  scale_color_manual("", values = c("WEI" = "green", "S&P500" = "blue", "CCI" = "red")) +
  ggtitle("WEI vs S&P500 52 week % change scaled by 10 vs CCI scaled by 1.5") +
  ylab("WEI, S&P500 and CCI") +
  xlab("Date")
plot_WEI_SP500_CCI
plot17_2008_2010 <- ggplot(data = data[1:63,], aes(x = Date)) +
  geom_line(aes(y =WEI_difference100, colour = "WEI difference scaled by 100")) +
  geom_line(aes(y =SP500_time_series_change, colour = "S&P500 difference")) +
  geom_hline(yintercept = 0, colour = 'black') + scale_colour_manual("", values = c("WEI difference sca
                                                                                     "S&P500 difference"
  ggtitle("Difference within WEI vs S&P500 2008-2010") + ylab("Difference with WEI scaled by 100")
plot17_2008_2010
plot17_before_covid <- ggplot(data = data[560:630,], aes(x = Date)) +</pre>
  geom_line(aes(y =WEI_difference100, colour = "WEI difference scaled by 100")) +
  geom_line(aes(y =SP500_time_series_change, colour = "S&P500 difference")) +
  geom_hline(yintercept = 0, colour = 'black') + scale_colour_manual("", values = c("WEI difference sca
                                                                                     "S&P500 difference"
  ggtitle("Difference within WEI vs S&P500 before COVID-19") + ylab("Difference with WEI scaled by 100"
plot17_before_covid
plot17_during_covid <- ggplot(data = data[630:639,], aes(x = Date)) +</pre>
  geom_line(aes(y =WEI_difference100, colour = "WEI difference scaled by 100")) +
  geom_line(aes(y =SP500_time_series_change, colour = "S&P500 difference")) +
  geom_hline(yintercept = 0, colour = 'black') + scale_colour_manual("", values = c("WEI difference sca
                                                                                     "S&P500 difference"
  ggtitle("Difference within WEI vs S&P500 during COVID-2019") + ylab("Difference with WEI scaled by 10
plot17_during_covid
```

Gekloot met time series

```
BB_time_series = ts(data[, 9], start = c(2008), frequency = 365.25/7)
BB_ts_change = diff(BB_time_series)
data$BB t
y \leftarrow data.frame(rep(c(0), times = 639))
WEI_time_series = ts(data[,4], start= c(2008), frequency = 365.25/7)
WEI_ts_change = diff(WEI_time_series)
time_series_data <- data.frame(WEI_ts_change)</pre>
data$WEI_time_series <- WEI_time_series</pre>
WEI time series change <- diff(data$WEI)
WEI_time_series_change <- append(WEI_time_series_change, 0, after = 0)</pre>
data <- data %>% cbind(WEI_time_series_change)
ggplot(data = data) +
  geom_line(aes(x = Date, y = diff(WEI)))
autoplot(diff(BB))
p1 <- autoplot(diff(WEI))</pre>
plot(x = data$Date, y = data$WEI)
plot(x = data$Date, y = diff(data$WEI))
p1$layers +
  geom_line(data = data,
            mapping = aes(x = Date, y = sp500_perc_change),
            inherit.aes = F)
een_variabele_naam <- cbind(diff(WEI), diff(BB_time_series / 100000))</pre>
plot.ts(een_variabele_naam, plot.type = "single", col = c("blue", "red"))
```

Correlation matrix and plots

```
x <- c(cor(data$WEI, data$T10Y3M), cor(data$WEI, data$BB), cor(data$WEI, data$M1))
print(x)
cor(data[4:10])
corrplot(data[4:10], type = "upper", order = "hclust", tl.col = "black", tl.srt = 45)
cor(data[-1])
cor(data[-1])
cor.test(x = data$WEI, y = data$`S&P500`, method=c("pearson", "kendall", "spearman"))

data_2 <- data_1
colnames(data_2)[12:13] <- c("S&P500 52 week difference", "CCI")
cor_all <- cor(select(data_2, 4:8, 11:13))
corrplot(cor_all, method = "color", na.rm = T)
x <- as.data.frame(cor(data_2[11:13]))
y <- as.data.frame(cor(data_2[4:8]))
z <- as.data.frame(rbind(x, y))
z <- cbind(data_2[4:8], data_2[11:13])</pre>
```

```
Z <- cor(z)
corrplot(Z, method = "color")</pre>
```

Regressing several variables to identify statsitical significance

```
model1 <- lm(WEI ~ T10Y3M + BBchange, data = data)</pre>
summary(model1)
anova(model1)
plot(model1)
summary(data)
model2 <- lm(WEI ~ T10Y3M, data = data)</pre>
summary(model2)
model3 <- lm(WEI ~ `S&P500` + Oil + FFR, data = data)</pre>
summary(model3)
plot(model3)
model4 <- lm(WEI ~ lnSP500 + Oil + FFR, data = data)</pre>
summary(model4)
plot(model4)
model5 <- lm(WEI ~ lnSP500 + lnBB + lnM1 + lnOil, data = data)</pre>
summary(model5)
model6 <- lm(WEI ~ lnOil + FFR + T10Y3M, data = data)</pre>
summary(model5)
```

Autocorrelation models (acf and pacf)

```
acf(data$sp500_perc_change, na.action = na.pass)
pacf(data$sp500_perc_change, na.action = na.pass)
```