

ARMA-X Analysis Tutorial

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Data

Load Base Data

```
# 1. Load Political Social Media

#contains posts from Twitter & TruthSocial
social <- read.csv(here("data/mothership", "social.csv"))

social_hourly <- read.csv(here("data/mothership", "socialhourly.csv"))

# 2. Load Financial

#SP500
SPY <- read.csv(here("data/mothership", "SPY.csv"))

#STOXX50
VGK <- read.csv(here("data/mothership", "VGK.csv"))

#CSI 300 (China)
ASHR <- read.csv(here("data/mothership", "ASHR.CSV"))

#make posixct
SPY$timestamp = as.POSIXct(SPY$timestamp,format = "%Y-%m-%d %H:%M:%S")
VGK$timestamp = as.POSIXct(VGK$timestamp,format = "%Y-%m-%d %H:%M:%S")
ASHR$timestamp = as.POSIXct(ASHR$timestamp,format = "%Y-%m-%d %H:%M:%S")
social$timestamp = as.POSIXct(social$timestamp,format = "%Y-%m-%d %H:%M:%S")
social_hourly$timestamp = as.POSIXct(social_hourly$timestamp,format = "%Y-%m-%d %H:%M:%S")
```

Volatility

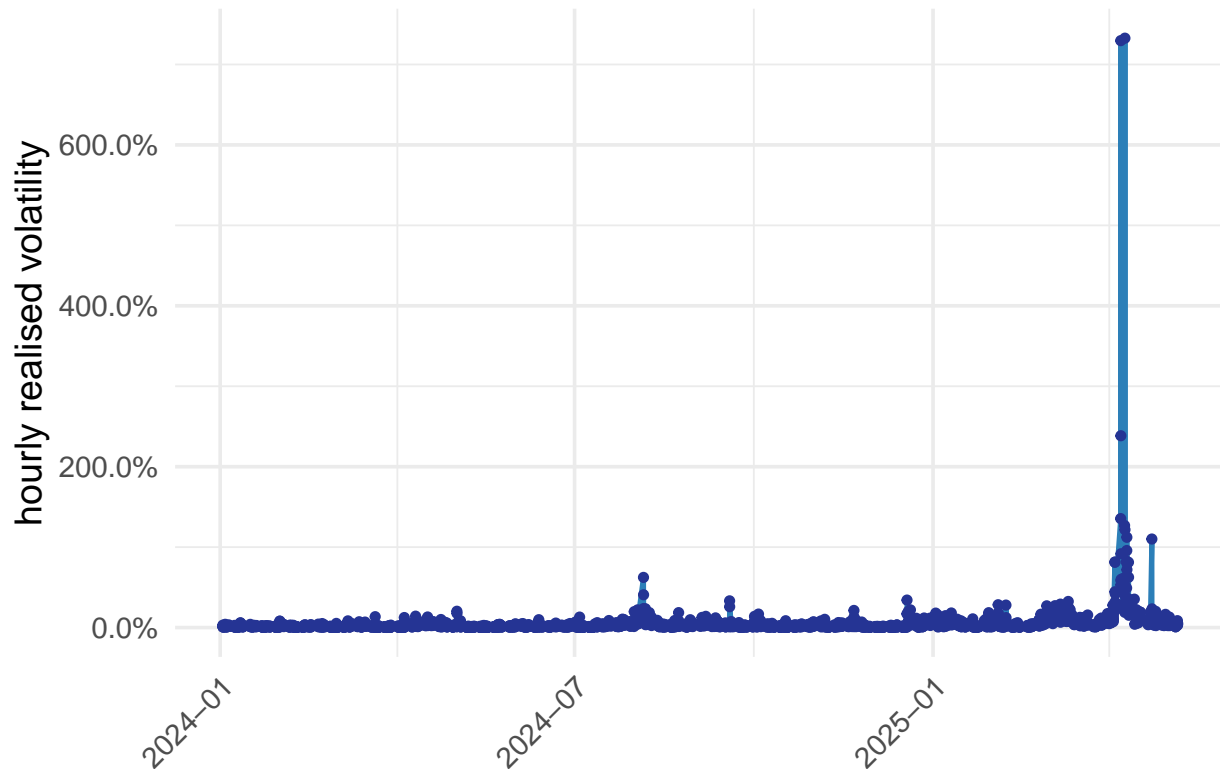
```
#find hourly volatility
#NOTE: this ignores tweets made outside trading hours!!
SPY_volatility_alltime = dplyr::select(SPY,timestamp,r_vol_h)

#aggregating per hour
SPY_volatility_alltime = SPY_volatility_alltime %>%
  mutate(timestamp = floor_date(timestamp, unit = "hour")) %>%
  distinct(timestamp, .keep_all = TRUE)

#select time period
SPY_volatility = filter(SPY_volatility_alltime,
  between(timestamp,
    as.Date('2024-01-01'),
    as.Date('2025-05-07')))

#plot
hvol_plotter(SPY_volatility,breaks="3 month",
  title="Realised Volatility - SPY")
```

Realised Volatility – SPY



```
#find hourly volatility
#NOTE: this ignores tweets made outside trading hours!!
VGK_volatility_alltime = dplyr::select(VGK,timestamp,r_vol_h)

#aggregating per hour
VGK_volatility_alltime = VGK_volatility_alltime %>%
  mutate(timestamp = floor_date(timestamp, unit = "hour")) %>%
  distinct(timestamp, .keep_all = TRUE)

#select time period
VGK_volatility = filter(VGK_volatility_alltime,
  between(timestamp,
    as.Date('2024-01-01'),
    as.Date('2025-05-07')))
```

```
#find hourly volatility
#NOTE: this ignores tweets made outside trading hours!!
ASHR_volatility_alltime = dplyr::select(ASHR,timestamp,r_vol_h)

#aggregating per hour
ASHR_volatility_alltime = ASHR_volatility_alltime %>%
  mutate(timestamp = floor_date(timestamp, unit = "hour")) %>%
  distinct(timestamp, .keep_all = TRUE)

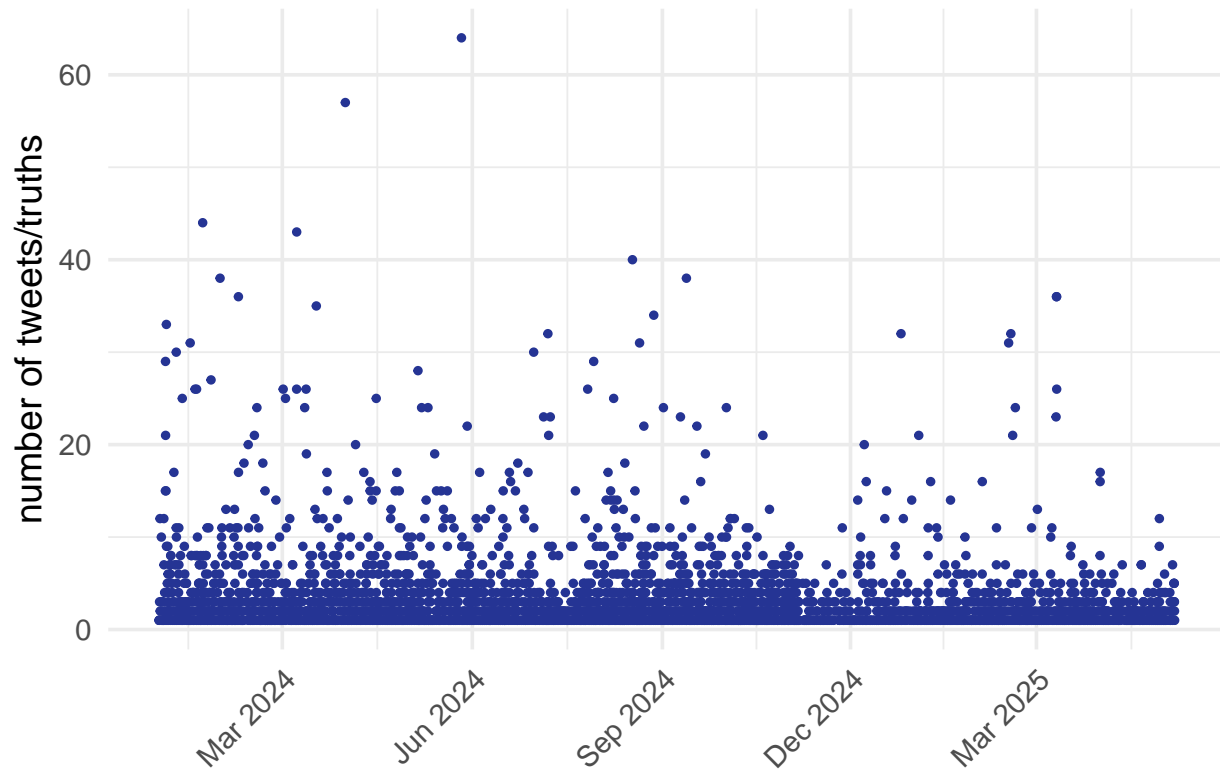
#select time period
ASHR_volatility = filter(ASHR_volatility_alltime,
```

```
between(timestamp,  
          as.Date('2024-01-01'),  
          as.Date('2025-05-07')))
```

Number of Posts

```
#find count  
tweetcount_alltime = dplyr::select(social_hourly,timestamp,N)  
  
#select time period  
tweetcount = filter(tweetcount_alltime,  
                     between(timestamp,  
                               as.Date('2024-01-01'),  
                               as.Date('2025-05-07')))  
  
#plot  
ggplot(tweetcount, aes(x = timestamp, y = N)) +  
  geom_point(color = "#253494", size = 1) +  
  scale_x_datetime(date_labels = "%b %Y", date_breaks = "3 month") +  
  labs(title = "Trump Social Media Count",  
        x = NULL,  
        y = "number of tweets/truths") +  
  theme_minimal(base_size = 14) +  
  theme(axis.text.x = element_text(angle = 45, hjust = 1),  
        plot.title = element_text(face = "bold", hjust = 0.5))
```

Trump Social Media Count



Dummy for Social Media Post

```
#find dummy
tweetdummy_alltime = dplyr::select(social_hourly,timestamp,dummy)

#select time period
tweetdummy = filter(tweetdummy_alltime,
                     between(timestamp,
                              as.Date('2024-01-01'),
                              as.Date('2025-05-07')))
```

Number of Tweets Mentioning Tariffs

```
#find count
tariff_alltime = dplyr::select(social_hourly,timestamp,total_tariff)

#select time period
tariff = filter(tariff_alltime,
                between(timestamp,
                        as.Date('2024-01-01'),
                        as.Date('2025-05-07')))
```

Number of Tweets Mentioning Trade

```
#find count
trade_alltime = dplyr::select(social_hourly,timestamp,total_trade)

#select time period
trade = filter(trade_alltime,
               between(timestamp,
                       as.Date('2024-01-01'),
                       as.Date('2025-05-07')))
```

Proportion of Positive

```
#find count
positive_alltime = dplyr::select(social_hourly,timestamp,prop_positive)

#select time period
positive = filter(positive_alltime,
                  between(timestamp,
                          as.Date('2024-01-01'),
                          as.Date('2025-05-07')))
```

Proportion of Negative

```
#find count
negative_alltime = dplyr::select(social_hourly,timestamp,prop_negative)

#select time period
negative = filter(negative_alltime,
                  between(timestamp,
                          as.Date('2024-01-01'),
                          as.Date('2025-05-07')))
```

Merge

```
#merge our dependant and independant vars
armax_data = left_join(SPY_volatility, VGK_volatility, by="timestamp")
armax_data = left_join(armax_data, ASHR_volatility, by="timestamp")
armax_data = left_join(armax_data, tweetdummy, by="timestamp")
armax_data = left_join(armax_data, tweetcount, by="timestamp")
armax_data = left_join(armax_data, tariff, by="timestamp")
armax_data = left_join(armax_data, trade, by="timestamp")
armax_data = left_join(armax_data, positive, by="timestamp")
armax_data = left_join(armax_data, negative, by="timestamp")

#rename volatility columns
names(armax_data)[2] <- "SPY_vol"
```

```
names(armax_data)[3] <- "VGK_vol"
names(armax_data)[4] <- "ASHR_vol"

#convert NA to zeroes
armax_data$N[is.na(armax_data$N)] = 0
armax_data$dummy[is.na(armax_data$dummy)] = 0
armax_data$total_tariff[is.na(armax_data$total_tariff)] = 0
armax_data$total_trade[is.na(armax_data$total_trade)] = 0
armax_data$prop_positive[is.na(armax_data$prop_positive)] = 0
armax_data$prop_negative[is.na(armax_data$prop_negative)] = 0
```

S&P500 ARMA-X Tariff Models

Finding Model

```
#auto.armax selects the lowest AIC value given r (exogenous variable lags)
res1 = auto.armax(armax_data$SPY_vol,xreg=armax_data$total_tariff,nb.lags=7,
                  latex=T,max.p = 7, max.q = 7, max.d=0)

##
## \begin{table}
## \begin{center}
## \begin{tabular}{l c}
## \hline
## & Model 1 \\\
## \hline
## ar1                & $0.9758^{***}$ \\\
##                    & $(0.0063)$ \\\
## ma1                & $-0.6906^{***}$ \\\
##                    & $(0.0217)$ \\\
## ma2                & $-0.1800^{***}$ \\\
##                    & $(0.0214)$ \\\
## intercept          & $0.0543^{*}$ \\\
##                    & $(0.0228)$ \\\
## total\_tariff\_lag\_0 & $-0.0066$ \\\
##                    & $(0.0113)$ \\\
## total\_tariff\_lag\_1 & $-0.0131$ \\\
##                    & $(0.0116)$ \\\
## total\_tariff\_lag\_2 & $0.0359^{**}$ \\\
##                    & $(0.0117)$ \\\
## total\_tariff\_lag\_3 & $-0.0049$ \\\
##                    & $(0.0117)$ \\\
## total\_tariff\_lag\_4 & $0.0044$ \\\
##                    & $(0.0117)$ \\\
## total\_tariff\_lag\_5 & $0.0037$ \\\
##                    & $(0.0116)$ \\\
## total\_tariff\_lag\_6 & $-0.0188$ \\\
##                    & $(0.0115)$ \\\
## total\_tariff\_lag\_7 & $-0.0141$ \\\
##                    & $(0.0112)$ \\\
## \hline
## AIC                & $-674.3212$ \\\
## AICc               & $-674.1655$ \\\
## BIC                & $-599.4019$ \\\
## Log Likelihood     & $350.1606$ \\\
## Num. obs.         & $2352$ \\\
## \hline
## \multicolumn{2}{l}{\scriptsize$^{***}$p<0.001$; $^{**}$p<0.01$; $^{*}$p<0.05$}}
## \end{tabular}
## \caption{ARMAX Model Results}
## \label{tab:armax}
## \end{center}
## \end{table}
```



```
#armax enables a custom armax specification with p,q,r
res2 = armax(armax_data$SPY_vol, xreg=armax_data$total_tariff, nb.lags=2,
             p=5, q=0, d=0, latex=T)
```

```
##
## \begin{table}
## \begin{center}
## \begin{tabular}{l c}
## \hline
## & Model 1 \\\
## \hline
## ar1 &  $0.3224^{***}$  \\\
## &  $(0.0206)$  \\\
## ar2 &  $0.0329$  \\\
## &  $(0.0219)$  \\\
## ar3 &  $0.1113^{***}$  \\\
## &  $(0.0224)$  \\\
## ar4 &  $0.0896^{***}$  \\\
## &  $(0.0223)$  \\\
## ar5 &  $0.0460^{*}$  \\\
## &  $(0.0208)$  \\\
## intercept &  $0.0539^{***}$  \\\
## &  $(0.0110)$  \\\
## total\_tariff\_lag\_0 &  $-0.0128$  \\\
## &  $(0.0115)$  \\\
## total\_tariff\_lag\_1 &  $-0.0250^{*}$  \\\
## &  $(0.0122)$  \\\
## total\_tariff\_lag\_2 &  $0.0312^{**}$  \\\
## &  $(0.0114)$  \\\
## \hline
## AIC &  $-597.9298$  \\\
## AICc &  $-597.8360$  \\\
## BIC &  $-540.2783$  \\\
## Log Likelihood &  $308.9649$  \\\
## Num. obs. &  $2357$  \\\
## \hline
## \multicolumn{2}{l}{ $^{***}p<0.001$ ;  $^{**}p<0.01$ ;  $^{*}p<0.05$ }}
## \end{tabular}
## \caption{ARMAX Model Results}
## \label{tab:armax}
## \end{center}
## \end{table}
```

```
#auto.armax.r selects the lowest AIC checking all 3 p,q,r values
res3 = auto.armax.r(armax_data$SPY_vol, x=armax_data$total_tariff,
                   max_p = 7, max_q = 7, max_r = 3, criterion = "AIC", latex=T)
```

```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

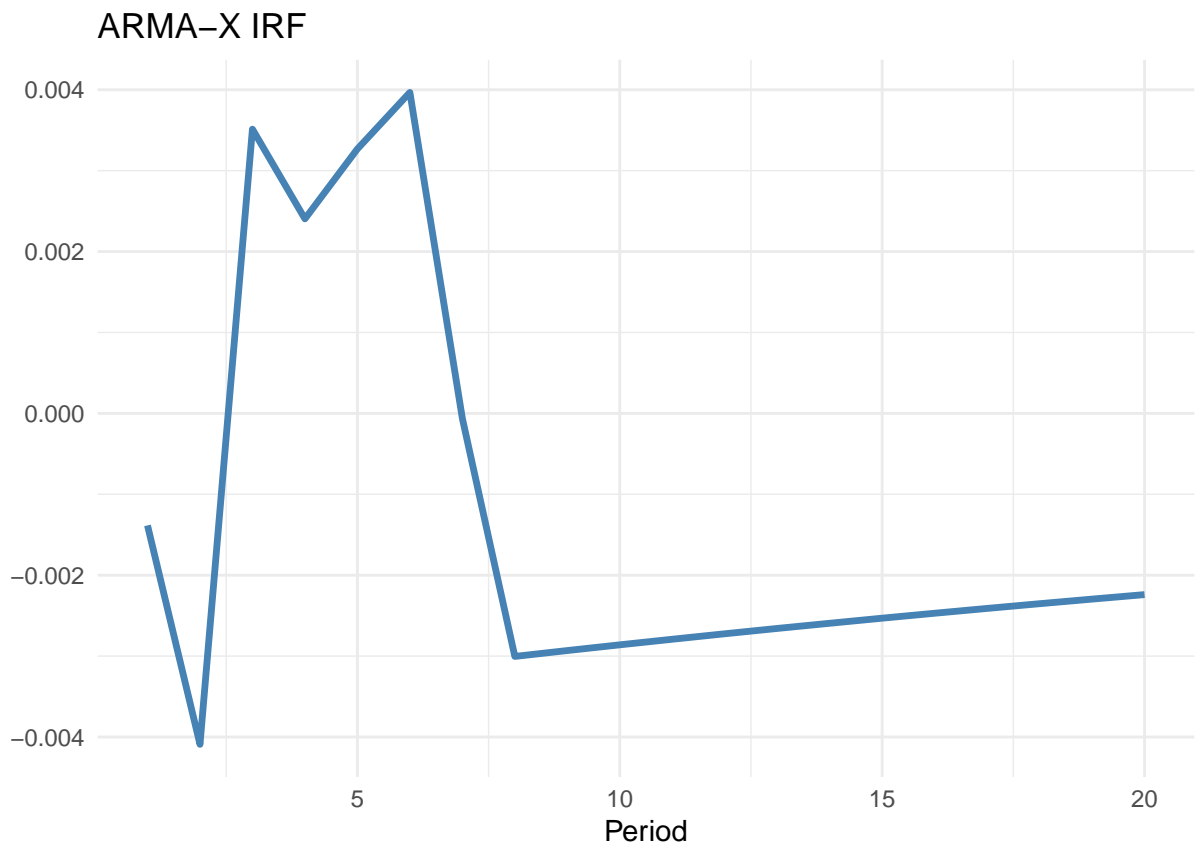
```
## Warning in sqrt(diag(model$var.coef)): NaNs produced
```

```
##
## \begin{table}
## \begin{center}
## \begin{tabular}{l c}
## \hline
## & Model 1 \\
## \hline
## ar1 & $1.0801$ & \\
## & $$ & \\
## ar2 & $-0.4288$ & \\
## & $$ & \\
## ar3 & $0.4976^{\{***\}}$ & \\
## & $(0.0488)$ & \\
## ar4 & $-0.3019^{\{***\}}$ & \\
## & $(0.0281)$ & \\
## ar5 & $0.7924^{\{***\}}$ & \\
## & $(0.0293)$ & \\
## ar6 & $-0.6814^{\{***\}}$ & \\
## & $(0.0150)$ & \\
## ma1 & $-0.7735^{\{***\}}$ & \\
## & $(0.0094)$ & \\
## ma2 & $0.1292^{\{***\}}$ & \\
## & $(0.0303)$ & \\
## ma3 & $-0.3723^{\{***\}}$ & \\
## & $(0.0240)$ & \\
## ma4 & $0.2371^{\{***\}}$ & \\
## & $(0.0211)$ & \\
## ma5 & $-0.9602^{\{***\}}$ & \\
## & $(0.0095)$ & \\
## ma6 & $0.6554$ & \\
## & $$ & \\
## ma7 & $0.2287^{\{***\}}$ & \\
## & $(0.0181)$ & \\
## intercept & $0.0507^{\{***\}}$ & \\
## & $(0.0138)$ & \\
## total\_tariff\_lag\_0 & $0.0030$ & \\
## & $(0.0086)$ & \\
## total\_tariff\_lag\_1 & $-0.0101$ & \\
## & $(0.0094)$ & \\
## total\_tariff\_lag\_2 & $0.0112$ & \\
## & $(0.0088)$ & \\
## \hline
## AIC & $-940.4859$ & \\
## AICc & $-940.1933$ & \\
## BIC & $-836.7133$ & \\
## Log Likelihood & $488.2429$ & \\
## Num. obs. & $2357$ & \\
## \hline
## \multicolumn{2}{l}{\scriptsize$^{\{***\}}p<0.001$; $^{\{**\}}p<0.01$; $^{\{*\}}p<0.05$}}
## \end{tabular}
## \caption{ARMAX selected by AIC}
## \label{tab:armax_select}
```

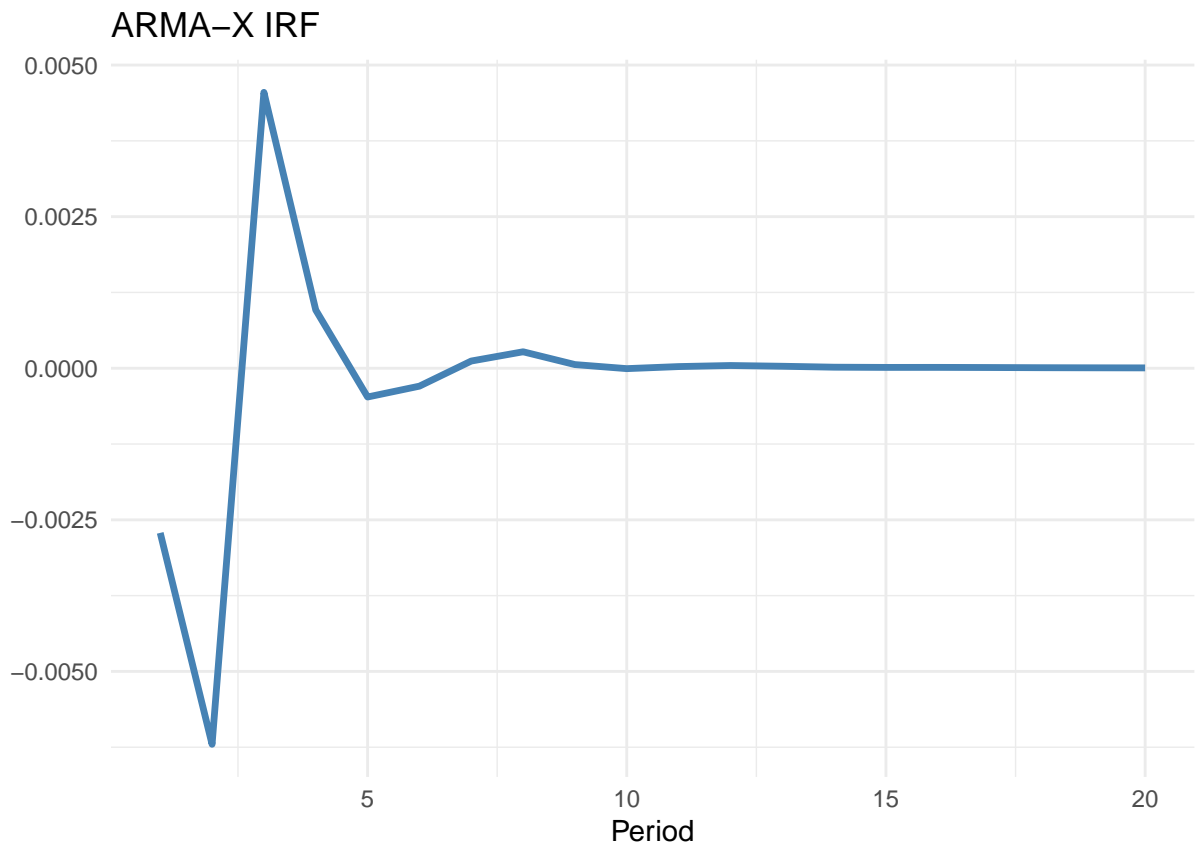
```
## \end{center}  
## \end{table}
```

Plotting IRFs

```
nb.periods = 20  
  
irf.plot(res1,nb.periods)
```



```
irf.plot(res2,nb.periods)
```



```
irf.plot(res3$model,nb.periods)
```

