

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

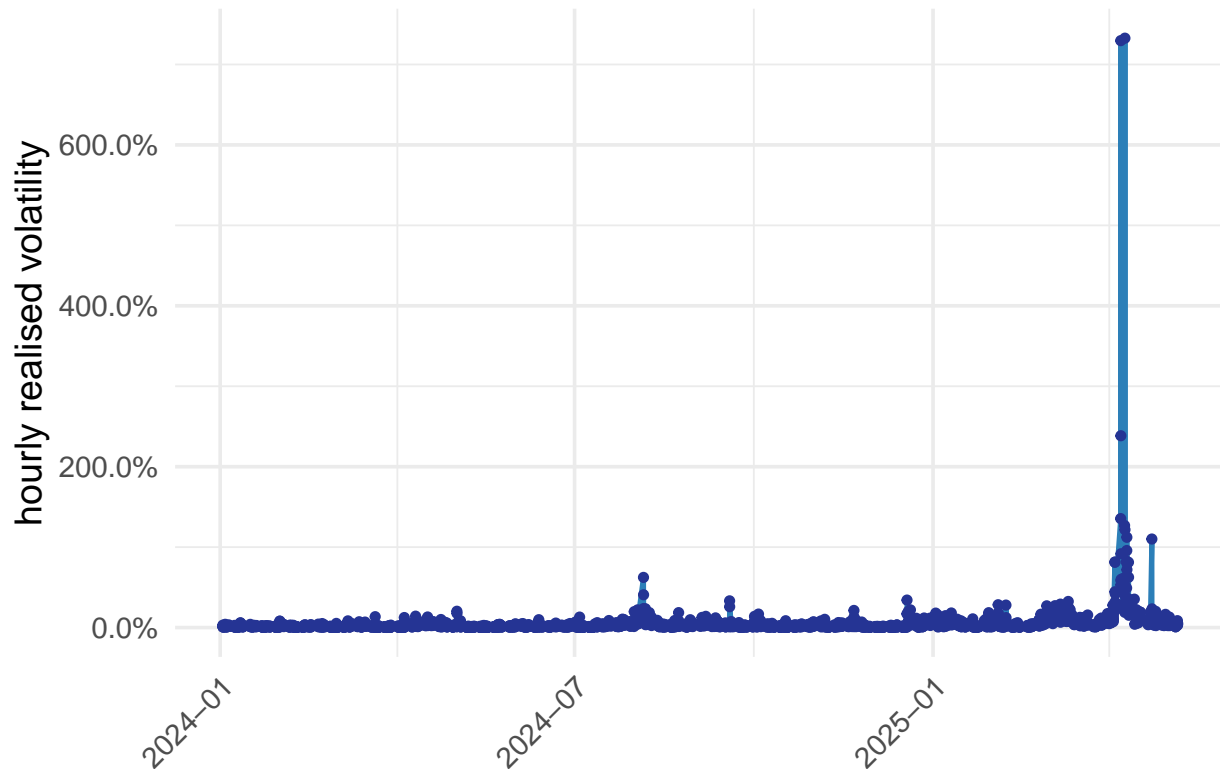
```
#find hourly volatility
#NOTE: this ignores tweets made outside trading hours!!
SPY_volatility_alltime = dplyr::select(SPY,timestamp,r_vol_h)

#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

```

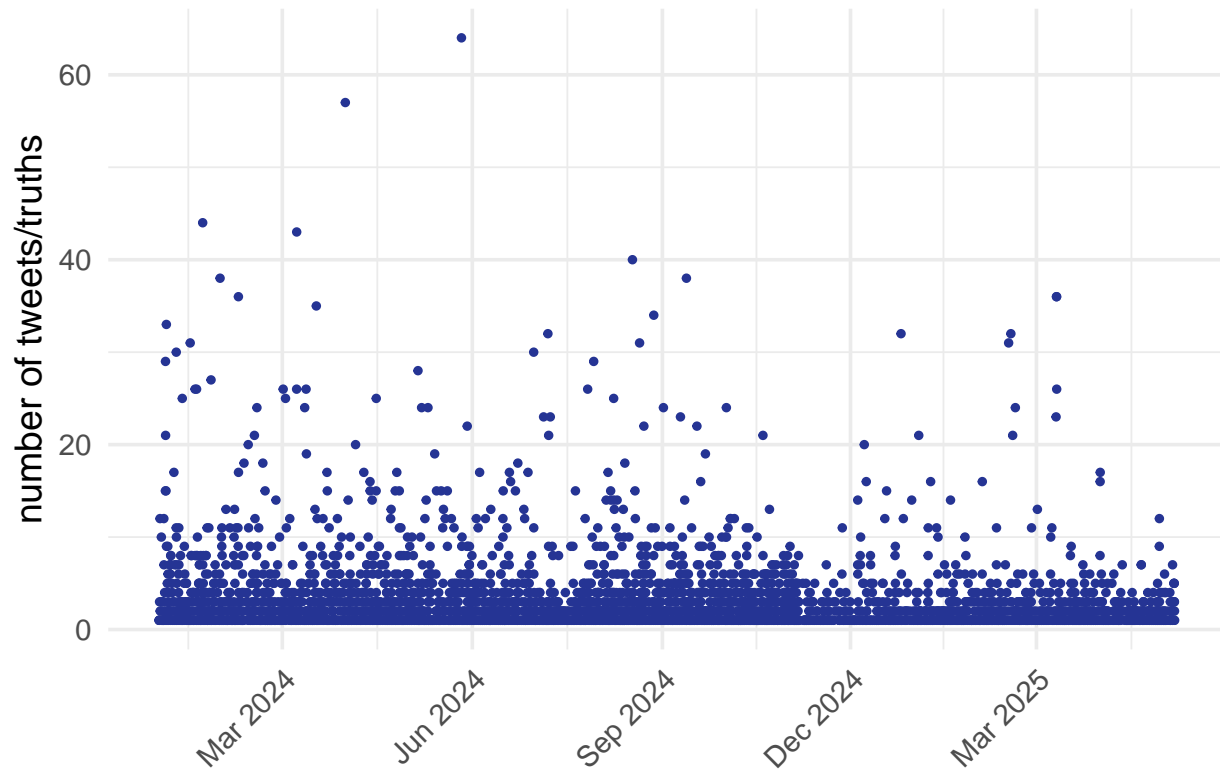
#find count
tweetcount_alltime = dplyr::select(social_hourly,timestamp,N)

#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=F,max.p = 7, max.q = 7, max.d=0)
```

```
##
## =====
##                               Model 1
## -----
## 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)
## -----
## AIC                -674.3212
## AICc               -674.1655
## BIC                -599.4019
## Log Likelihood      350.1606
## Num. obs.          2352
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
#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=F)
```

```
##
## =====
```



```
##                               Model 1
## -----
## 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)
## -----
## AIC                          -597.9298
## AICc                         -597.8360
## BIC                          -540.2783
## Log Likelihood               308.9649
## Num. obs.                    2357
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
#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=F)
```

```
## 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
```

```
##
## =====
##                               Model 1
## -----
## 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)
## -----
## AIC          -940.4859
## AICc         -940.1933
## BIC          -836.7133
## Log Likelihood 488.2429
## Num. obs.    2357
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05

```

Plotting IRFs

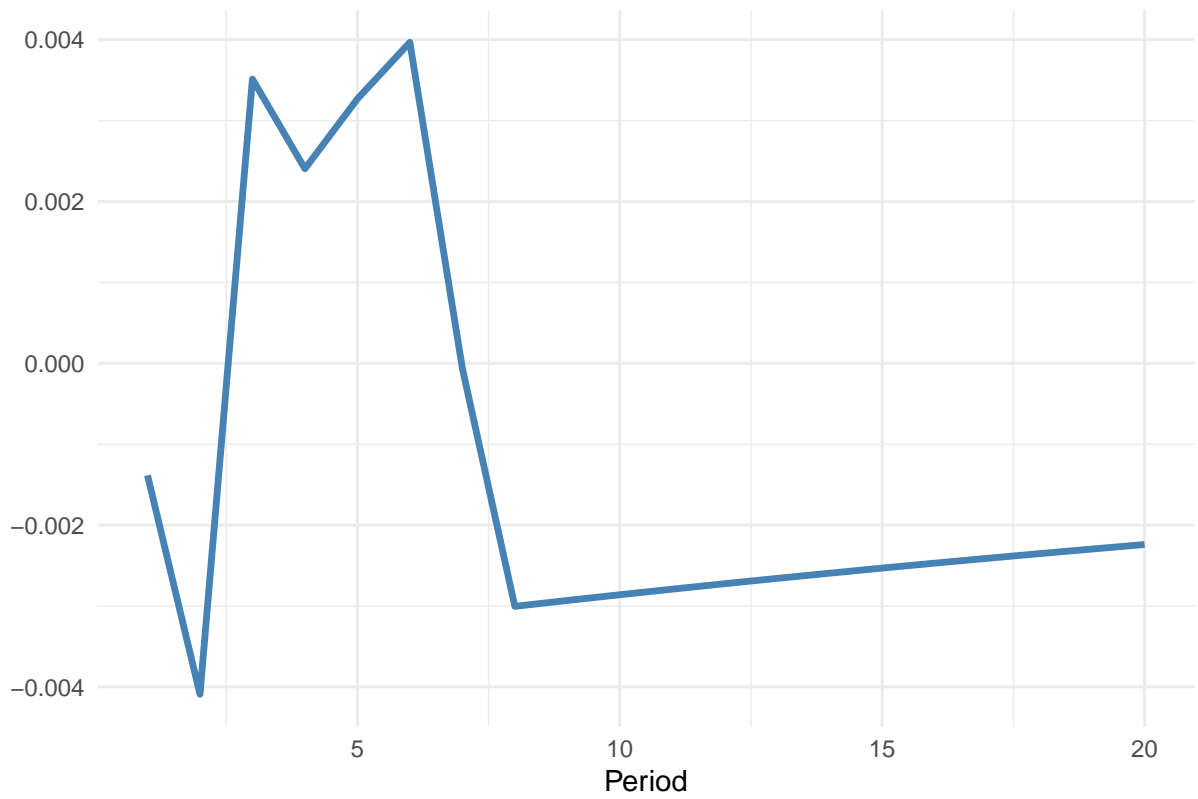
```

nb.periods = 20

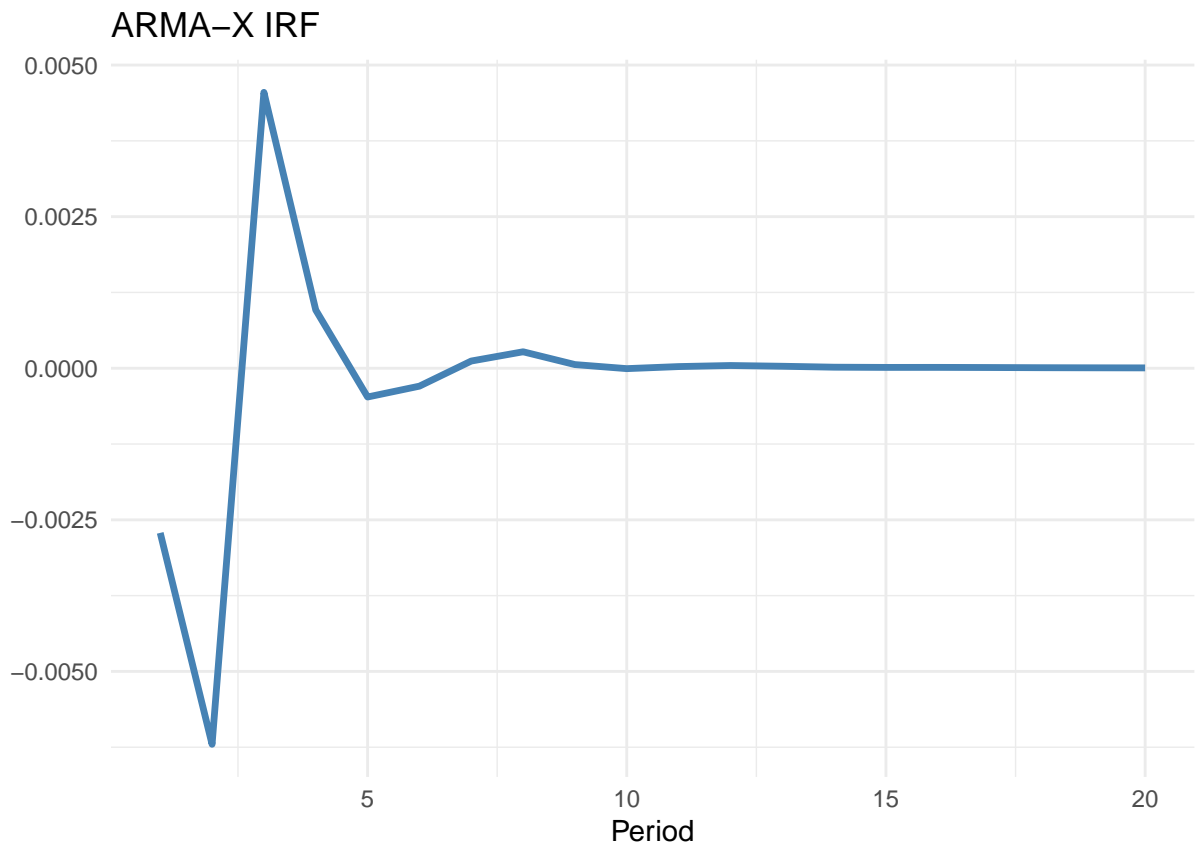
irf.plot(res1,nb.periods)

```

ARMA-X IRF



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



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

