

ARMA-X Figures

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Full Timeframe (Jan 2024 to May 2025)

```
#load final dataset
source(here("helperfunctions/full_data.R"))

#backup
backup = data

#select timeframe
data = filter(data,between(timestamp, as.Date('2014-01-01'), as.Date('2025-05-07')))

#for interpretation
mean1 = mean(data$SPY_vol)
```

SPY Models

We choose the specification in the `armax_models` file. In this file, we will just run said specifications to produce nice tables and graphs to include in our final paper. This is also why there are specification differences in the separate timeframes. We always use the best fit we found earlier.

```
models <- list()

# ARMA-X(3,3,1) with Tweet Dummy as Exogenous
models[["Model 1"]] <- armax(data$SPY_vol, xreg = data$dummy, latex = F,
                             nb.lags = 1, p = 3, q = 3)

# ARMA-X(3,3,1) with Tweet Count as Exogenous
models[["Model 2"]] <- armax(data$SPY_vol, xreg = data$N, latex = F,
                             nb.lags = 1, p = 3, q = 3)

# ARMA-X(3,2,3) with Tariff Mentions as Exogenous
models[["Model 3"]] <- armax(data$SPY_vol, xreg = data$tariff, latex = F,
                             nb.lags = 3, p = 3, q = 2)

# ARMA-X(3,2,1) with Trade Mentions as Exogenous
models[["Model 4"]] <- armax(data$SPY_vol, xreg = data$trade, latex = F,
                             nb.lags = 1, p = 3, q = 2)

# ARMA-X(3,2,0) with China Mentions as Exogenous
models[["Model 5"]] <- armax(data$SPY_vol, xreg = data$china, latex = F,
                             nb.lags = 0, p = 3, q = 2)
```

SPY Table

```
names = list( "ar1" = "AR(1)",
               "ar2" = "AR(2)",
               "ar3" = "AR(3)",
               "ma1" = "MA(1)",
               "ma2" = "MA(2)",
```

```

    "ma3" = "MA(3)",
    "(Intercept)" = "Constant",
    "dummy_lag_0" = "$TweetDummy_{t}$",
    "dummy_lag_1" = "$TweetDummy_{t-1}$",
    "N_lag_0" = "$TweetCount_{t}$",
    "N_lag_1" = "$TweetCount_{t-1}$",
    "tariff_lag_0" = "$Tariff_{t}$",
    "tariff_lag_1" = "$Tariff_{t-1}$",
    "tariff_lag_2" = "$Tariff_{t-2}$",
    "tariff_lag_3" = "$Tariff_{t-3}$",
    "trade_lag_0" = "$Trade_{t}$",
    "trade_lag_1" = "$Trade_{t-1}$",
    "china_lag_0" = "$China_{t}$")

table1 = texreg(models,
  custom.model.names = names(models),
  custom.coef.map = names,
  caption = "ARMAX Models of Average Hourly Volatility",
  caption.above = TRUE,
  label = "tab:armax",
  digits = 4)

table1

write(table1, file = "armax_table1.tex")

```

SPY IRFs

```

#we want to plot the IRFs of these models
nb.periods = 7 * 15

#irf.plot(models[["Model 1"]],nb.periods,title="Tweet Dummy Shock")
#irf.plot(models[["Model 2"]],nb.periods,title="Tweet Count Shock")
plot1 = irf.plot(models[["Model 3"]],nb.periods,
  title="Tariff Mention Shock - Full Timeframe")

plot1

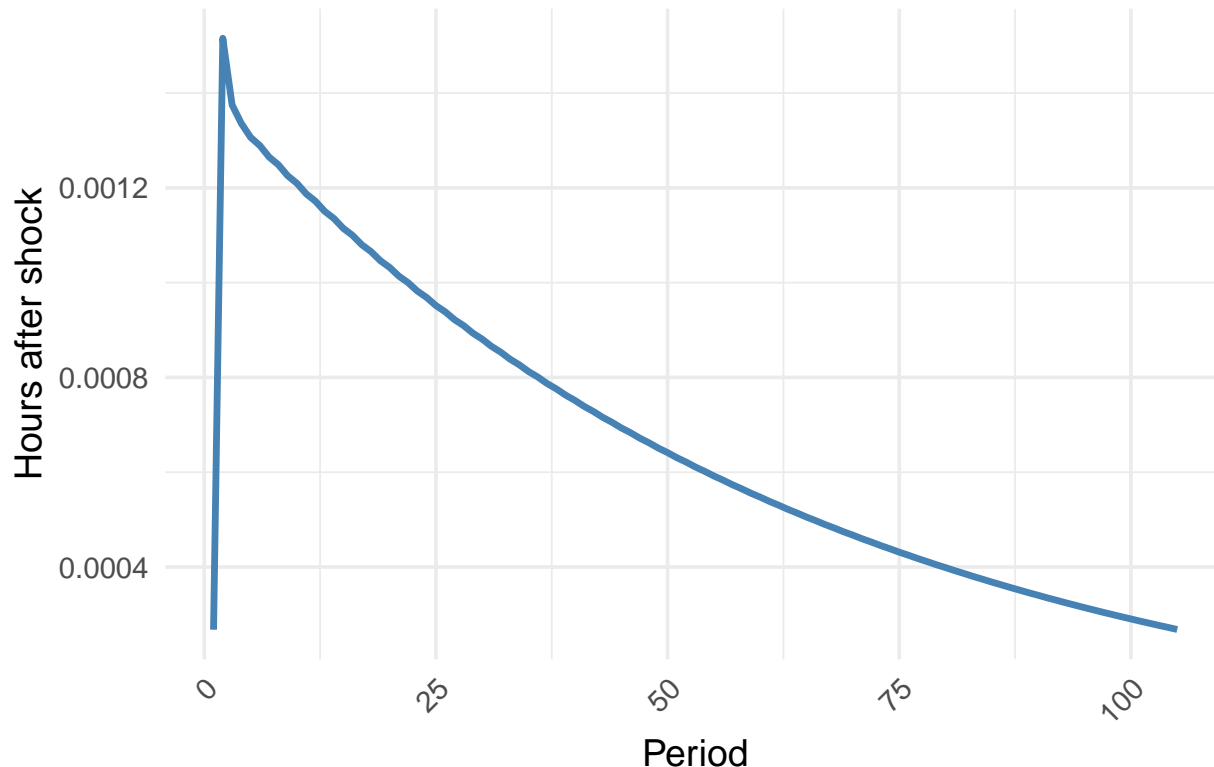
```

Table 1: ARMAX Models of Average Hourly Volatility

	Model 1	Model 2	Model 3	Model 4	Model 5
AR(1)	0.0300 (0.0510)	0.0278 (0.0510)	0.2200*** (0.0084)	2.1903*** (0.0096)	0.2209*** (0.0084)
AR(2)	0.7229*** (0.0397)	0.7210*** (0.0399)	0.9388*** (0.0037)	-1.4727*** (0.0173)	0.9382*** (0.0037)
AR(3)	0.2110*** (0.0287)	0.2148*** (0.0284)	-0.1837*** (0.0079)	0.2784*** (0.0082)	-0.1837*** (0.0079)
MA(1)	0.2751*** (0.0496)	0.2779*** (0.0496)	0.0870*** (0.0042)	-1.8955*** (0.0062)	0.0878*** (0.0042)
MA(2)	-0.6445*** (0.0284)	-0.6430*** (0.0285)	-0.8960*** (0.0042)	0.9165*** (0.0063)	-0.8950*** (0.0042)
MA(3)	-0.3527*** (0.0256)	-0.3563*** (0.0253)			
<i>TweetDummy_t</i>	0.0014*** (0.0002)				
<i>TweetDummy_{t-1}</i>	0.0008*** (0.0002)				
<i>TweetCount_t</i>		0.0004*** (0.0001)			
<i>TweetCount_{t-1}</i>		0.0002** (0.0001)			
<i>Tariff_t</i>			0.0035* (0.0014)		
<i>Tariff_{t-1}</i>			0.0191*** (0.0015)		
<i>Tariff_{t-2}</i>			0.0103*** (0.0015)		
<i>Tariff_{t-3}</i>			-0.0045** (0.0014)		
<i>Trade_t</i>				0.0032 (0.0018)	
<i>Trade_{t-1}</i>				0.0016 (0.0018)	
<i>China_t</i>					0.0026* (0.0012)
AIC	-45761.2161	-45737.6695	-46020.9547	-45816.1540	-45840.5349
AICc	-45761.2051	-45737.6585	-46020.9415	-45816.1449	-45840.5277
BIC	-45682.1963	-45658.6497	-45934.0340	-45745.0361	-45777.3186
Log Likelihood	22890.6081	22878.8348	23021.4774	22917.0770	22928.2675
Num. obs.	19970	19970	19968	19970	19971

*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$

Tariff Mention Shock – Full Timeframe



```
#irf.plot(models[["Model 4"]],nb.periods,title="Trade Mention Shock")
#irf.plot(models[["Model 5"]],nb.periods,title="China Mention Shock")

ggsave("armax_plot1.png",plot=plot1,bg="white")
```

SPY Residuals

```
res1 = checkresiduals(models[["Model 1"]], plot = FALSE)
res2 = checkresiduals(models[["Model 2"]], plot = FALSE)
res3 = checkresiduals(models[["Model 3"]], plot = FALSE)
res4 = checkresiduals(models[["Model 4"]], plot = FALSE)
res5 = checkresiduals(models[["Model 5"]], plot = FALSE)

resnames = c("Twitter Dummy", "Twitter Count", "Tariff", "Trade", "China")

#extract p-values directly from checkresiduals results
pvals <- data.frame("X-Regressor" = resnames,
                    `Full Timeframe` = c(
                        res1$p.value,
                        res2$p.value,
                        res3$p.value,
                        res4$p.value,
                        res5$p.value))
```

First Term

```
#load final dataset
data = backup

#first term
data = filter(data,between(timestamp, as.Date('2017-01-20'), as.Date('2021-01-20')))

#for interpretation
mean2 = mean(data$SPY_vol)
```

SPY Models

```
models <- list()

# ARMA-X(3,3,0) with Tariff Mentions as Exogenous
models[["First Term (1)"]] <- armax(data$SPY_vol, xreg = data$tariff, latex = F,
                                     nb.lags = 0, p = 3, q = 3)

# ARMA-X(3,3,0) with Trade Mentions as Exogenous
models[["First Term (2)"]] <- armax(data$SPY_vol, xreg = data$trade, latex = F,
                                     nb.lags = 0, p = 3, q = 3)

# ARMA-X(3,3,0) with Trade Mentions as Exogenous
models[["First Term (3)"]] <- armax(data$SPY_vol, xreg = data$china, latex = F,
                                     nb.lags = 0, p = 3, q = 3)
```

SPY Residuals

```
res6 = checkresiduals(models[["First Term (1)"]], plot = FALSE)
res7 = checkresiduals(models[["First Term (2)"]], plot = FALSE)
res8 = checkresiduals(models[["First Term (3)"]], plot = FALSE)

pvals_new1 <- data.frame(
  "First-Term" = c(
    NA,
    NA,
    res6$p.value,
    res7$p.value,
    res8$p.value))
```

Second Term

```
#load final dataset
data = backup
```

```

#second term
data = filter(data,between(timestamp, as.Date('2025-01-20'), as.Date('2025-05-07')))

#for interpretation
mean3 = mean(data$SPY_vol)

```

SPY Models

```

# ARMA-X(3,2,3) with Tariff Mentions as Exogenous
models[["Second Term (1)"]] <- armax(data$SPY_vol, xreg = data$tariff, latex = F,
                                     nb.lags = 2, p = 1, q = 2)

# ARMA-X(3,2,1) with Trade Mentions as Exogenous
models[["Second Term (2)"]] <- armax(data$SPY_vol, xreg = data$trade, latex = F,
                                     nb.lags = 0, p = 1, q = 2)

# ARMA-X(3,2,0) with China Mentions as Exogenous
models[["Second Term (3)"]] <- armax(data$SPY_vol, xreg = data$china, latex = F,
                                     nb.lags = 2, p = 1, q = 2)

```

SPY IRFs

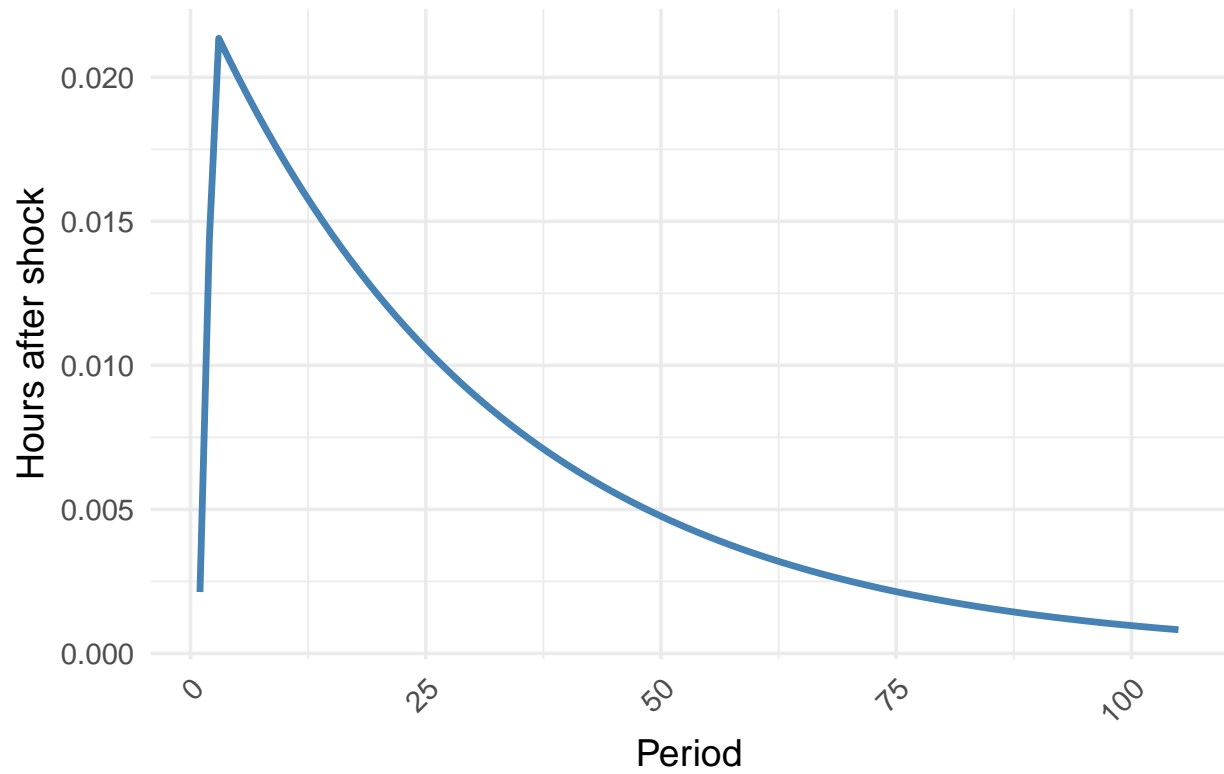
```

#we want to plot the IRFs of these models
nb.periods = 7 * 15

plot2 = irf.plot(models[["Second Term (1)"]],nb.periods,
                 title="Tariff Mention Shock - Second Term")
plot2

```

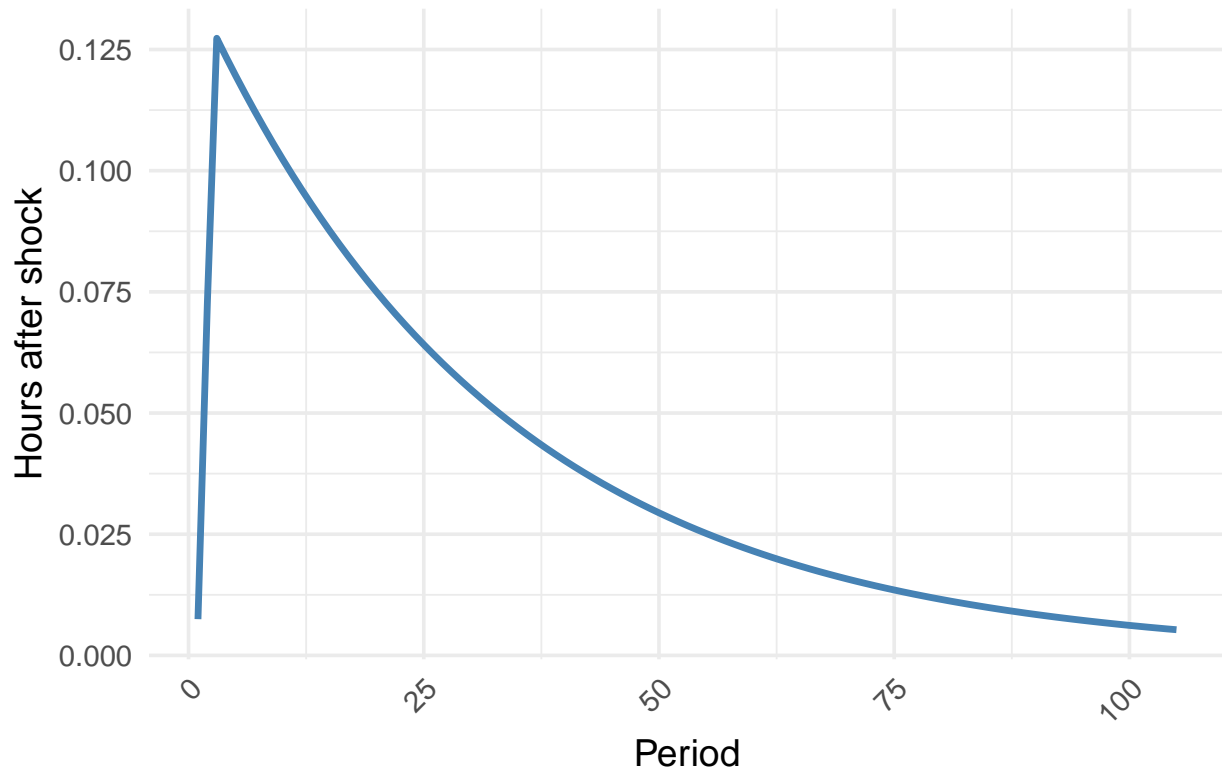
Tariff Mention Shock – Second Term



```
ggsave("arimax_plot2.png",plot=plot2,bg="white")

plot3 = irf.plot(models[["Second Term (3)"]],nb.periods,
                 title="China Mention Shock - Second Term")
plot3
```


China Mention Shock – Second Term



```
ggsave("armax_plot3.png",plot=plot3,bg="white")
```

SPY Residuals

```
res9 = checkresiduals(models[["Second Term (1)"]], plot = FALSE)
res10 = checkresiduals(models[["Second Term (2)"]], plot = FALSE)
res11 = checkresiduals(models[["Second Term (3)"]], plot = FALSE)

pvals_new2 <- data.frame(
  "Second-Term" = c(
    NA,
    NA,
    res9$p.value,
    res10$p.value,
    res11$p.value))

#combine with other term
pvals_combined <- cbind(pvals,pvals_new1)
pvals_combined <- cbind(pvals_combined, pvals_new2)
```

SPY Table (both terms)

```
xnames = list("ar1" = "AR(1)",
              "ar2" = "AR(2)",
              "ar3" = "AR(3)",
              "ma1" = "MA(1)",
              "ma2" = "MA(2)",
              "ma3" = "MA(3)",
              "(Intercept)" = "Constant",
              "tariff_lag_0" = "$Tariff_{t}$",
              "tariff_lag_1" = "$Tariff_{t-1}$",
              "tariff_lag_2" = "$Tariff_{t-2}$",
              "trade_lag_0" = "$Trade_{t}$",
              "china_lag_0" = "$China_{t}$",
              "china_lag_1" = "$China_{t-1}$",
              "china_lag_2" = "$China_{t-2}$")

table2 = texreg(models,
  custom.model.names = names(models),
  custom.coef.map = xnames,
  caption = "Split-Term ARMAX Models of Average Hourly Volatility",
  caption.above = TRUE,
  label = "tab:armax_term",
  digits = 4)
table2
```

```
write(table2, file = "armax_table2.tex")
```

SPY Residuals Table

```
table3 = knitr::kable(pvals_combined, digits = 100, format="latex",
  caption = "Ljung-Box Test p-values for Residuals")
table3
```

```
write(table3, file = "armax_table3.tex")
```

Descriptive Stats

```
means <- data.frame(
  Model = c("Full Time Mean", "First Term Mean", "Second Term Mean"),
  `SPY Volatility Mean` = c(
    mean1,
    mean2,
    mean3))
```

Table 2: Split-Term ARMAX Models of Average Hourly Volatility

	First Term (1)	First Term (2)	First Term (3)	Second Term (1)	Second Term (2)	Second Term (3)
AR(1)	0.2953*** (0.0225)	0.2943*** (0.0224)	0.2927*** (0.0224)	0.9686*** (0.0163)	0.9683*** (0.0163)	0.9693*** (0.0161)
AR(2)	0.1434*** (0.0220)	0.1439*** (0.0220)	0.1438*** (0.0219)			
AR(3)	0.5456*** (0.0223)	0.5462*** (0.0222)	0.5480*** (0.0222)			
MA(1)	0.1854*** (0.0180)	0.1863*** (0.0179)	0.1866*** (0.0179)	−0.6965*** (0.0469)	−0.6905*** (0.0469)	−0.7207*** (0.0467)
MA(2)	−0.1707*** (0.0169)	−0.1706*** (0.0169)	−0.1695*** (0.0168)	−0.1732*** (0.0437)	−0.1755*** (0.0438)	−0.1609*** (0.0434)
MA(3)	−0.6557*** (0.0162)	−0.6564*** (0.0161)	−0.6575*** (0.0161)			
$Tariff_t$	0.0011 (0.0010)			0.0048 (0.0099)		
$Tariff_{t-1}$				0.0278** (0.0102)		
$Tariff_{t-2}$				0.0168 (0.0099)		
$Trade_t$		0.0023** (0.0009)			−0.0074 (0.0297)	
$China_t$			0.0018** (0.0006)			0.0173 (0.0319)
$China_{t-1}$						0.1515*** (0.0324)
$China_{t-2}$						0.1309*** (0.0319)
AIC	−28604.6559	−28610.2269	−28613.1693	633.4836	638.2093	610.2140
AICc	−28604.6303	−28610.2013	−28613.1437	633.7676	638.3737	610.4980
BIC	−28542.9191	−28548.4901	−28551.4325	667.4525	663.7092	644.1829
Log Likelihood	14311.3279	14314.1134	14315.5847	−308.7418	−313.1047	−297.1070
Num. obs.	7042	7042	7042	516	518	516

*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$

Table 3: Ljung-Box Test p-values for Residuals

X.Regressor	Full.Timeframe	First.Term	Second.Term
Twitter Dummy	0	NA	NA
Twitter Count	0	NA	NA
Tariff	0	0	0.8489828
Trade	0	0	0.8322070
China	0	0	0.5122385

Table 4: Summary Statistics of SPY Volatility

Model	SPY.Volatility.Mean
Full Time Mean	0.022621
First Term Mean	0.017486
Second Term Mean	0.144248

```
table4 = knitr::kable(means, digits = 6, format="latex",
  caption = "Summary Statistics of SPY Volatility")
```

```
table4
```

```
write(table4, file = "armax_table4.tex")
```