ARMA-X Figures

Contents

Full Timeframe (Jan 2024 to May 2025)	2
SPY Models	2
SPY Table	2
SPY IRFs	3
SPY Residuals	9
First Term	9
SPY Models	9
SPY Residuals	10
Second Term	10
SPY Models	10
SPY IRFs	11
SPY Residuals	12
SPY Table (both terms)	12

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')))
```

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.

```
models <- list()
# ARMA-X(3,3,1) with Tweet Dummy as Exogenous
models[["Model 1"]] <- armax(data$SPY_vol, xreg = data$dummy, latex = F,</pre>
                              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,</pre>
                              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,</pre>
                              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,</pre>
                              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,</pre>
                              nb.lags = 0, p = 3, q = 2)
```

SPY Table

```
"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" = "$China_{t}$")
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)
```

SPY IRFs

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

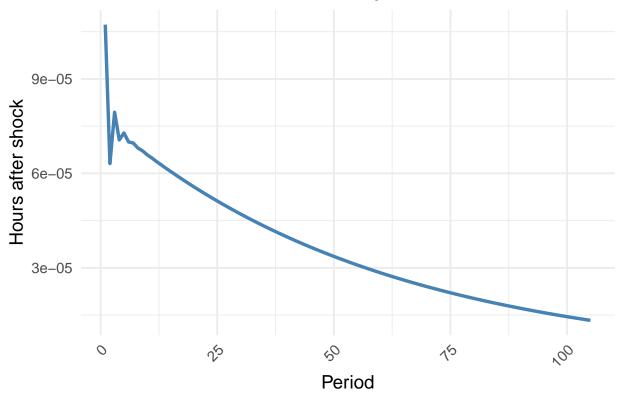
irf.plot(models[["Model 1"]],nb.periods,title="Tweet Dummy Shock")
```

Table 1: ARMAX Models of Average Hourly Volatility

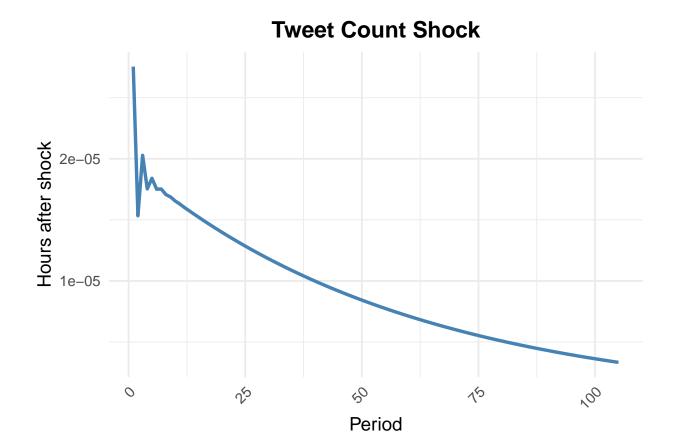
	Model 1	Model 2	Model 3	Model 4	Model 5
$\overline{AR(1)}$	0.0300	0.0278	0.2200***	2.1903***	0.2209***
	(0.0510)	(0.0510)	(0.0084)	(0.0096)	(0.0084)
AR(2)	0.7229***	0.7210***	0.9388***	-1.4727^{***}	0.9382***
17 (2)	(0.0397)	(0.0399)	(0.0037)	(0.0173)	(0.0037)
AR(3)	0.2110***	0.2148***	-0.1837***	0.2784***	-0.1837***
λπλ (1)	(0.0287) 0.2751^{***}	(0.0284) 0.2779^{***}	$(0.0079) \\ 0.0870^{***}$	(0.0082) $-1.8955***$	$(0.0079) \\ 0.0878^{***}$
MA(1)	(0.2731) (0.0496)	(0.0496)	(0.0042)	-1.8955 (0.0062)	(0.0042)
MA(2)	-0.6445^{***}	-0.6430^{***}	-0.8960^{***}	0.9165***	-0.8950^{***}
WIII(2)	(0.0284)	(0.0285)	(0.0042)	(0.0063)	(0.0042)
MA(3)	-0.3527***	-0.3563***	(0.0012)	(0.0000)	(0.0012)
(-)	(0.0256)	(0.0253)			
$TweetDummy_t$	0.0014***	,			
	(0.0002)				
$TweetDummy_{t-1}$	0.0008***				
	(0.0002)				
$TweetCount_t \\$		0.0004^{***}			
		(0.0001)			
$TweetCount_{t-1} \\$		0.0002**			
T		(0.0001)	0.0005*		
$Tariff_t$			0.0035*		
$Tariff_{t-1}$			$(0.0014) \\ 0.0191^{***}$		
$I \text{ at } iJ J_{t-1}$			(0.0015)		
$Tariff_{t-2}$			0.0103***		
$t \text{ ar } v_{J} j t=2$			(0.0015)		
$Tariff_{t-3}$			-0.0045**		
v v t=3			(0.0014)		
$Trade_t$,	0.0032	
•				(0.0018)	
$Trade_{t-1}$				0.0016	
				(0.0018)	
$China_t$					0.0026*
170		1000000000	10000000000	180101811	(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 22890.6081	-45658.6497 22878.8348	-45934.0340 23021.4774	-45745.0361 22917.0770	-45777.3186 22928.2675
Log Likelihood Num. obs.	19970	22878.8348 19970	19968	19970	22928.2675 19971
Num. obs.	19970	19910	19900	19910	13311

^{***}p < 0.001; **p < 0.01; *p < 0.05



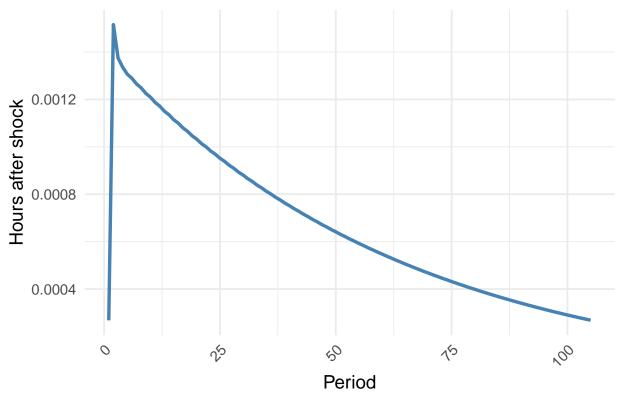


irf.plot(models[["Model 2"]],nb.periods,title="Tweet Count Shock")

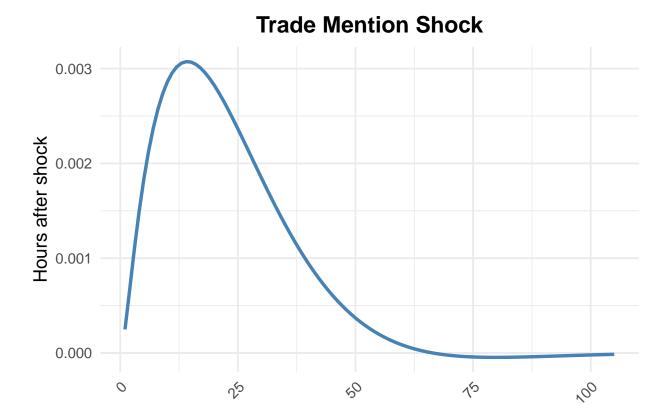


irf.plot(models[["Model 3"]],nb.periods,title="Tariff Mention Shock")



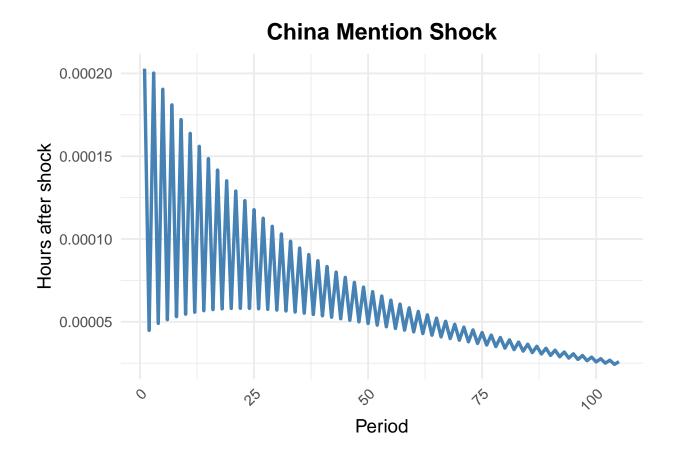


irf.plot(models[["Model 4"]],nb.periods,title="Trade Mention Shock")



Period

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



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)
```

First Term

```
#load final dataset
data = backup

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

SPY Models

SPY Residuals

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

Second Term

```
#load final dataset
data = backup

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

SPY Models

SPY IRFs

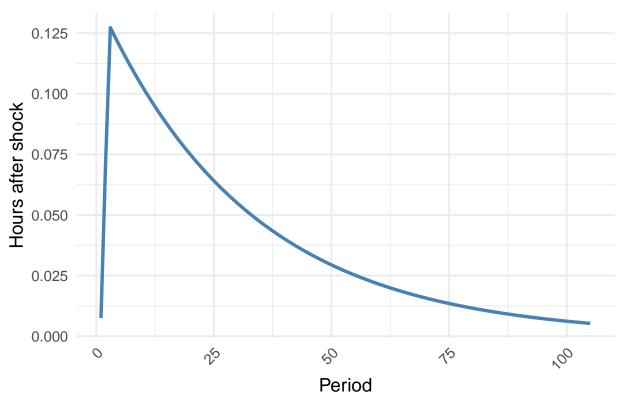
```
#we want to plot the IRFs of these models
nb.periods = 7 * 15

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

Tariff Mention Shock 0.020 0.015 0.000 0.000 Period

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





SPY Residuals

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

SPY Table (both terms)

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
AR(1)	0.2953***	0.2943***	0.2927***	0.9686***	0.9683***	0.9693***
	(0.0225)	(0.0224)	(0.0224)	(0.0163)	(0.0163)	(0.0161)
AR(2)	0.1434^{***}	0.1439^{***}	0.1438^{***}			
	(0.0220)	(0.0220)	(0.0219)			
AR(3)	0.5456^{***}	0.5462^{***}	0.5480^{***}			
	(0.0223)	(0.0222)	(0.0222)			
MA(1)	0.1854^{***}	0.1863^{***}	0.1866***	-0.6965***	-0.6905***	-0.7207***
	(0.0180)	(0.0179)	(0.0179)	(0.0469)	(0.0469)	(0.0467)
MA(2)	-0.1707^{***}	-0.1706***	-0.1695^{***}	-0.1732^{***}	-0.1755***	-0.1609^{***}
	(0.0169)	(0.0169)	(0.0168)	(0.0437)	(0.0438)	(0.0434)
MA(3)	-0.6557^{***}	-0.6564^{***}	-0.6575^{***}			
	(0.0162)	(0.0161)	(0.0161)			
$Tariff_t$	0.0011			0.0048		
	(0.0010)			(0.0099)		
$Tariff_{t-1}$				0.0278**		
				(0.0102)		
$Tariff_{t-2}$				0.0168		
				(0.0099)		
$Trade_t$		0.0023^{**}			-0.0074	
		(0.0009)			(0.0297)	
$China_t$			0.0018**			0.0173
			(0.0006)			(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