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')))
```

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 different timeframes. We also use the best fit we found earlier.

```
models <- list()</pre>
# 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

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

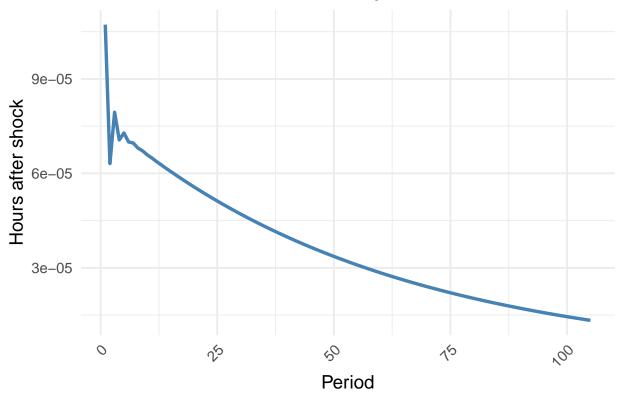
irf.plot(models[["Model 1"]],nb.periods,title="Tweet Dummy Shock")
```

Table 1: ARMAX Models of Average Hourly Volatility

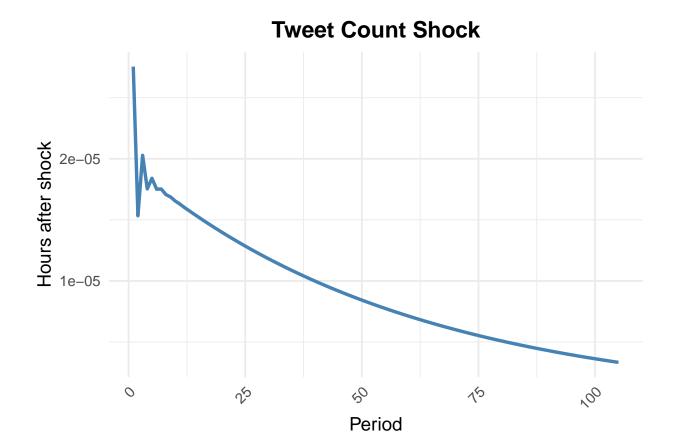
	Model 1	Model 2	Model 3	Model 4	Model 5
$\overline{AR(1)}$	0.0300	0.0278	0.2200***	2.1903***	0.2209***
	(0.0510)	(0.0510)	(0.0084)	(0.0096)	(0.0084)
AR(2)	0.7229***	0.7210***	0.9388***	-1.4727^{***}	0.9382***
17 (2)	(0.0397)	(0.0399)	(0.0037)	(0.0173)	(0.0037)
AR(3)	0.2110***	0.2148***	-0.1837***	0.2784***	-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



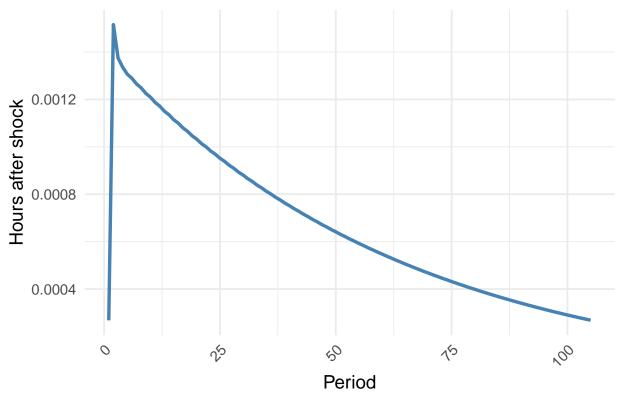


irf.plot(models[["Model 2"]],nb.periods,title="Tweet Count Shock")

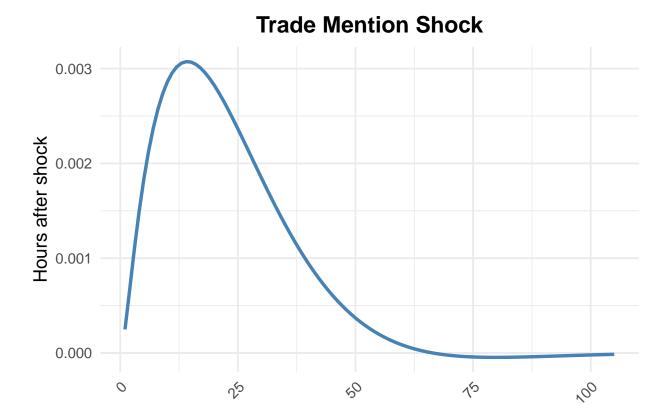


irf.plot(models[["Model 3"]],nb.periods,title="Tariff Mention Shock")



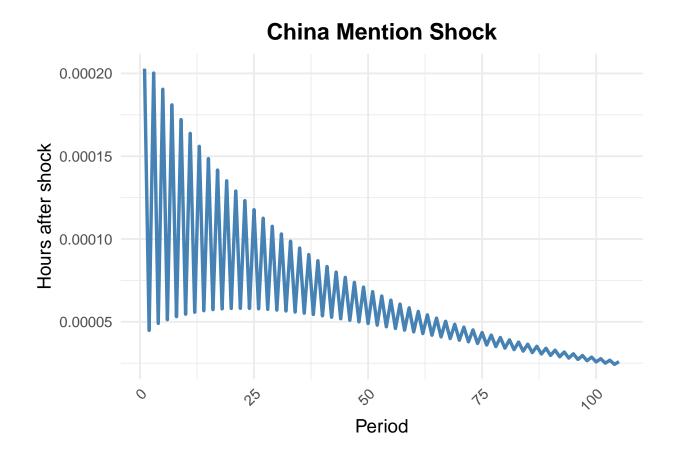


irf.plot(models[["Model 4"]],nb.periods,title="Trade Mention Shock")



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

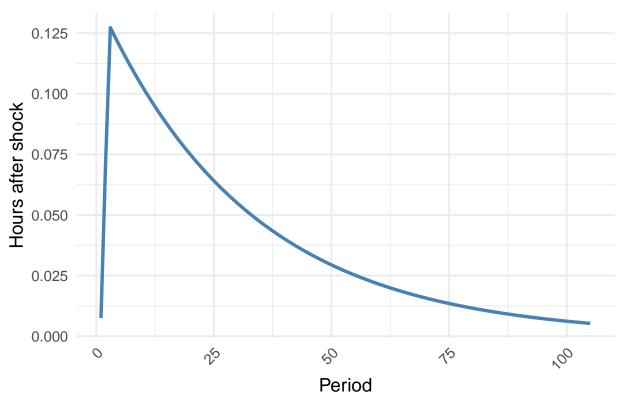
```
#we want to plot the IRFs of these models
nb.periods = 7 * 15

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