ARMA-X Analysis Tutorial

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	Model 1	
ar1	0.9828***	
	(0.0017)	
ma1	-0.6786^{***}	
	(0.0073)	
ma2	-0.2118^{***}	
	(0.0087)	
ma3	-0.0120	
	(0.0080)	
ma4	0.0331***	
	(0.0071)	
intercept	0.0202***	
	(0.0041)	
$dummy_lag_0$	0.0013***	
	(0.0002)	
$dummy_lag_1$	0.0007^{***}	
	(0.0002)	
$dummy_lag_2$	0.0001	
	(0.0002)	
AIC	-45719.7236	
AICc	-45719.7126	
BIC	-45640.7043	
Log Likelihood	22869.8618	
Num. obs.	19969	
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$		

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

Table 1: ARMAX Model Results

S&P500 Univariate ARMA-X Models

Tweet Dummy as Exogenous

```
#auto.armax selects the lowest AIC value given r (exogenous variable lags)
res1 = auto.armax(data$SPY_vol,xreg=data$dummy,nb.lags=2,
                latex=T, max.p = 6, max.q = 6, max.d=0)
```

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

```
#auto.armax.r selects the lowest AIC checking all 3 p,q,r values
res3 = auto.armax.r(data$SPY_vol, x=data$dummy,
               max_p = 6, max_q = 6, max_r = 6, criterion = "AIC", latex=T)
```

```
#we want to plot the IRFs of these models
nb.periods = 7
irf.plot(res1,nb.periods)
```

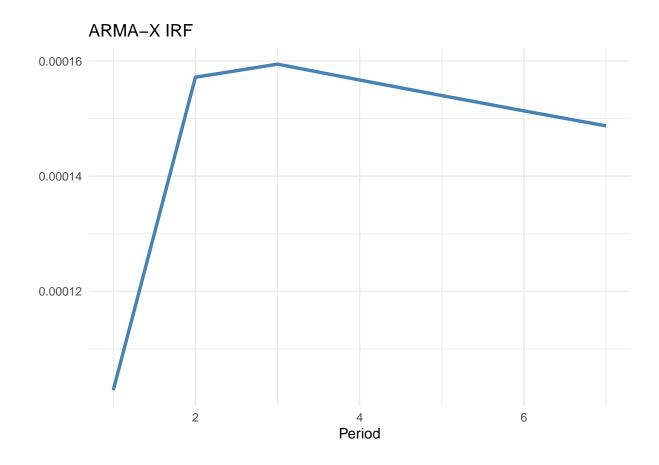
	Model 1
ar1	0.3576***
	(0.0071)
ar2	0.0416^{***}
	(0.0075)
ar3	0.0994^{***}
	(0.0074)
ar4	0.1045***
	(0.0075)
ar5	0.0816***
	(0.0071)
intercept	0.0199***
	(0.0018)
dummy_lag_0	0.0015***
	(0.0002)
$dummy_lag_1$	0.0009***
	(0.0002)
$dummy_lag_2$	0.0001
	(0.0002)
AIC	-44706.1942
AICc	-44706.1832
BIC	-44627.1749
Log Likelihood	22363.0971
Num. obs.	19969
*** n < 0.001: ** n < 0.0	$0.1 \cdot *n < 0.05$

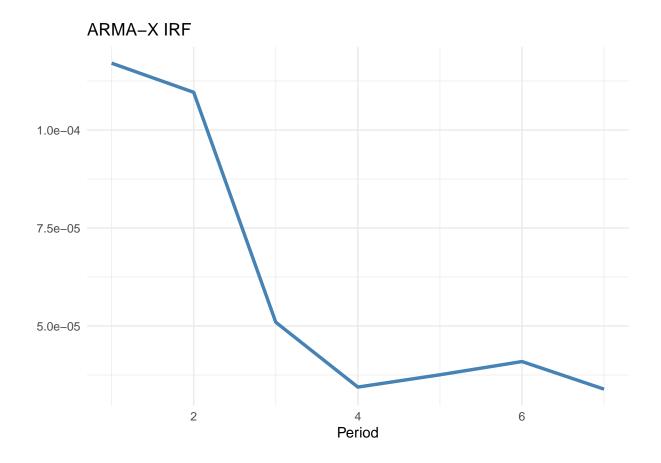
^{***}p < 0.001; **p < 0.01; *p < 0.05

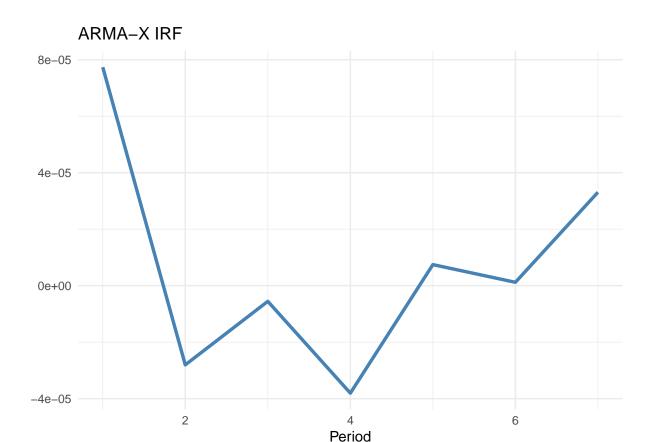
Table 2: ARMAX Model Results

	Model 1	
ar1	-0.6515***	
arr	(0.0104)	
ar2	0.0295^*	
a1 2		
9	(0.0133)	
ar3	0.0193*	
,	(0.0085)	
ar4	0.1384***	
	(0.0122) 0.6404^{***}	
ar5		
	(0.0111)	
ar6	0.6864^{***}	
	(0.0075)	
ma1	0.9508***	
	(0.0078)	
ma2	0.2802***	
	(0.0137)	
ma3	0.1832***	
	(0.0117)	
ma4	0.0692***	
	(0.0101)	
ma5	-0.6142^{***}	
111000	(0.0107)	
ma6	-0.8002^{***}	
11140	(0.0056)	
intercept	0.0231***	
шестесри	(0.0042)	
dummer lag 0	0.0042)	
dummy_lag_0		
dumanas la m 1	$(0.0002) \\ 0.0003$	
dummy_lag_1		
1 1 0	(0.0002)	
dummy_lag_2	-0.0004	
1 1 0	(0.0002)	
$dummy_lag_3$	-0.0006**	
	(0.0002)	
$dummy_lag_4$	-0.0004	
	(0.0002)	
$dummy_lag_5$	-0.0005^{**}	
	(0.0002)	
AIC	-47081.1839	
AICc	-47081.1418	
BIC	-46923.1482	
Log Likelihood	23560.5920	
Num. obs.	19966	
***p < 0.001; **p < 0.0		
p < 0.001, p < 0.01, p < 0.00		

Table 3: ARMAX selected by AIC







Tweet Count as Exogenous

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

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

	Model 1
ar1	0.9828***
	(0.0017)
ma1	-0.6780^{***}
	(0.0073)
ma2	-0.2129^{***}
	(0.0086)
ma3	-0.0118
	(0.0080)
ma4	0.0335***
	(0.0071)
intercept	0.0210***
	(0.0041)
N_{lag_0}	0.0003***
	(0.0001)
N_lag_1	0.0002**
	(0.0001)
N_{lag_2}	0.0000
	(0.0001)
AIC	-45696.1228
AICc	-45696.1118
BIC	-45617.1034
Log Likelihood	22858.0614
Num. obs.	19969
*** n < 0.001, ** n < 0.0	01. *n < 0.05

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

Table 4: ARMAX Model Results

	Model 1
ar1	0.3584***
	(0.0071)
ar2	0.0410^{***}
	(0.0075)
ar3	0.0991^{***}
	(0.0074)
ar4	0.1040***
	(0.0075)
ar5	0.0815***
	(0.0071)
intercept	0.0208***
	(0.0018)
N_lag_0	0.0004***
	(0.0001)
N_lag_1	0.0002***
	(0.0001)
N_lag_2	0.0000
	(0.0001)
AIC	-44677.6875
AICc	-44677.6765
BIC	-44598.6682
Log Likelihood	22348.8438
Num. obs.	19969
*** n < 0.001 · ** n < 0.0	$0.1 \cdot *n < 0.05$

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

Table 5: ARMAX Model Results

)*** 1) 4* 9) 5* 5) *** 7)
1) 1* 9) 5* 5) *** 7)
9) 5* 5) *** 7)
5* 5) *** 7)
5* 5) *** 7)
7)
7)

2)
9)

5)

6)

2)

5)

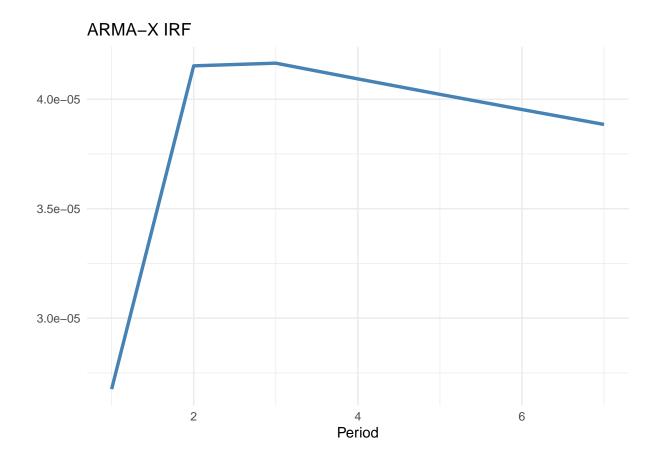
9)
2***
4)
2***
5)

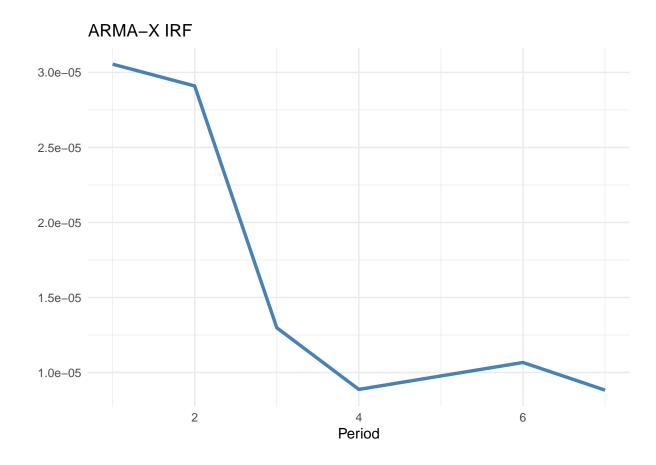
1)

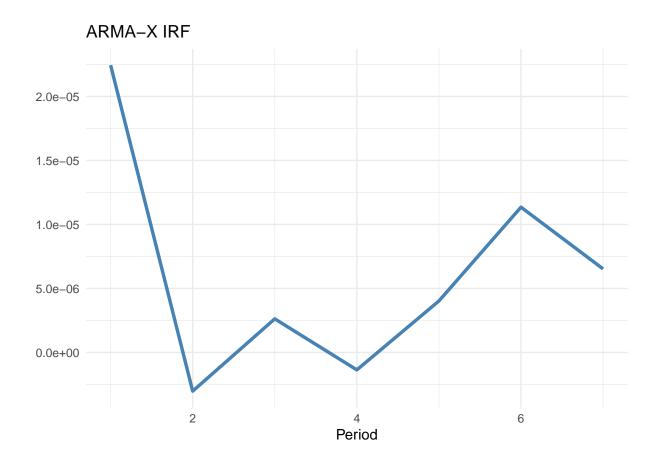
1)
**
1)
1226
3953
9908
113
)

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

Table 6: ARMAX selected by AIC







Tariff as Exogenous

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

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

	Model 1
ar1	1.7000***
	(0.1313)
ar2	-0.8772***
	(0.1518)
ar3	0.1689***
	(0.0232)
ma1	-1.3999****
	(0.1327)
ma2	0.4605***
	(0.1192)
intercept	0.0217***
1	(0.0040)
$tariff_lag_0$	0.0042^{**}
_ 0_	(0.0014)
tariff lag 1	0.0199***
	(0.0015)
$tariff_lag_2$	0.0112***
	(0.0014)
AIC	-45860.5245
AICc	-45860.5134
BIC	-45781.5051
Log Likelihood	22940.2622
Num. obs.	19969
*** n < 0.001: ** n < 0.0	$0.1 \cdot *n < 0.05$

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

Table 7: ARMAX Model Results

	Model 1
ar1	0.3572***
	(0.0071)
ar2	0.0427^{***}
	(0.0075)
ar3	0.0903***
	(0.0075)
ar4	0.0978***
	(0.0075)
ar5	0.0859***
	(0.0071)
intercept	0.0217***
-	(0.0017)
$tariff_lag_0$	0.0047**
	(0.0015)
$tariff_lag_1$	0.0201***
	(0.0015)
$tariff_lag_2$	0.0109***
	(0.0015)
AIC	-44818.4470
AICc	-44818.4359
BIC	-44739.4276
Log Likelihood	22419.2235
Num. obs.	19969
*** n < 0.001 · ** n < 0.0	$0.1 \cdot *n < 0.05$

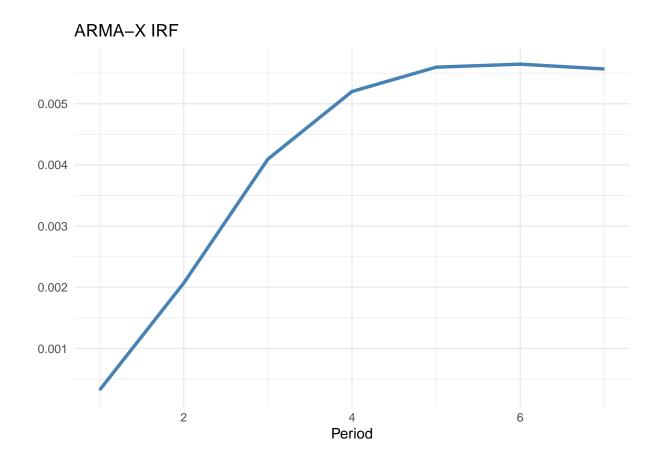
 $^{***}p < 0.001; \ ^{**}p < 0.01; \ ^*p < 0.05$

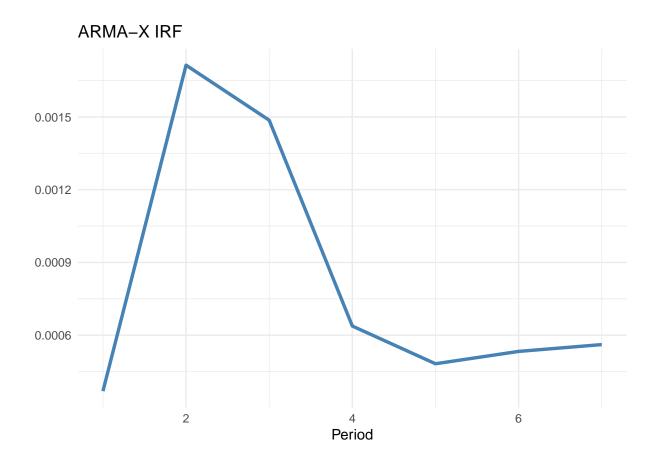
Table 8: ARMAX Model Results

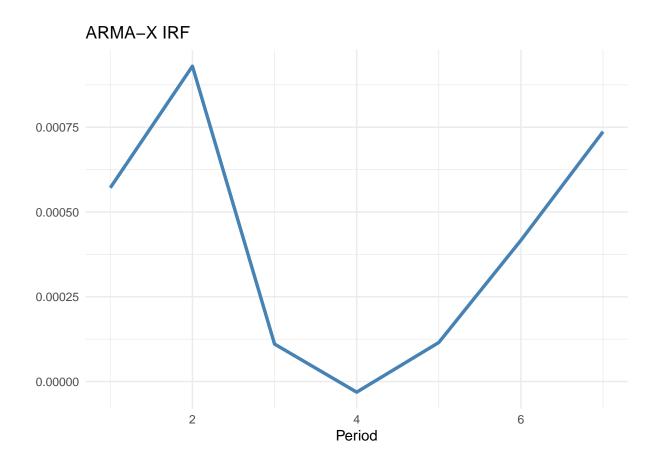
	Model 1
ar1	-0.6474***
	(0.0099)
ar2	0.0332**
	(0.0128)
ar3	$0.0164^{'}$
	(0.0089)
ar4	0.1330***
	(0.0116)
ar5	0.6417***
	(0.0113)
ar6	0.6856***
	(0.0076)
ma1	0.9487***
	(0.0075)
ma2	0.2737***
	(0.0130)
ma3	0.1743***
	(0.0118)
ma4	0.0618***
	(0.0097)
ma5	-0.6178***
	(0.0107)
ma6	-0.7993^{***}
	(0.0058)
intercept	0.0217^{***}
	(0.0040)
$tariff_lag_0$	0.0077***
	(0.0013)
$tariff_lag_1$	0.0175***
	(0.0014)
$tariff_lag_2$	0.0093***
	(0.0013)
AIC	-47180.9824
AICc	-47180.9517
BIC	-47046.6495
Log Likelihood	23607.4912
Num. obs.	19969
p < 0.001; **p <	$01; \frac{p}{< 0.05}$

p < 0.001; p < 0.01; p < 0.01; p < 0.05

Table 9: ARMAX selected by AIC







Trade Mention as Exogenous

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

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

	Model 1
ar1	0.9828***
	(0.0017)
ma1	-0.6774^{***}
	(0.0073)
ma2	-0.2149^{***}
	(0.0087)
ma3	-0.0126
	(0.0080)
ma4	0.0352***
	(0.0071)
intercept	0.0222***
•	(0.0041)
$trade_lag_0$	0.0019
	(0.0019)
$trade_lag_1$	0.0042^{*}
	(0.0019)
$trade_lag_2$	0.0071***
	(0.0019)
AIC	-45677.4427
AICc	-45677.4317
BIC	-45598.4233
Log Likelihood	22848.7213
Num. obs.	19969
*** n < 0.001 · ** n < 0.0	$0.1 \cdot *n < 0.05$

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

Table 10: ARMAX Model Results

	Model 1	
ar1	0.3598^{***}	
	(0.0071)	
ar2	0.0395^{***}	
	(0.0075)	
ar3	0.0974***	
	(0.0074)	
ar4	0.1024***	
	(0.0075)	
ar5	0.0827***	
	(0.0071)	
intercept	0.0221***	
_	(0.0018)	
$trade_lag_0$	0.0027	
	(0.0019)	
$trade_lag_1$	0.0045^*	
	(0.0020)	
$trade_lag_2$	0.0075***	
	(0.0019)	
AIC	-44647.3628	
AICc	-44647.3518	
BIC	-44568.3435	
Log Likelihood	22333.6814	
Num. obs.	19969	
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$		

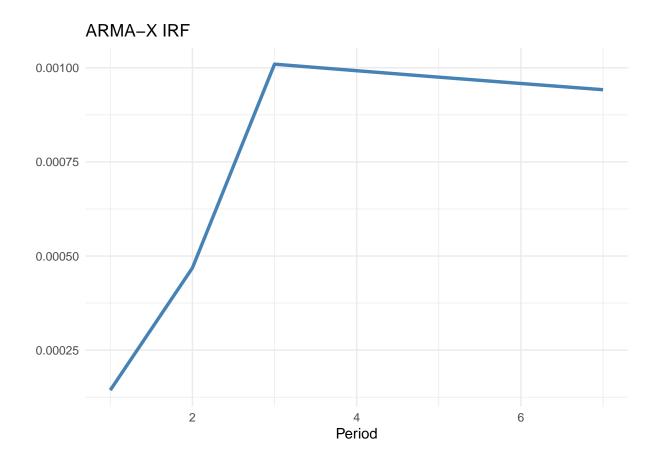
p < 0.001; p < 0.01; p < 0.05

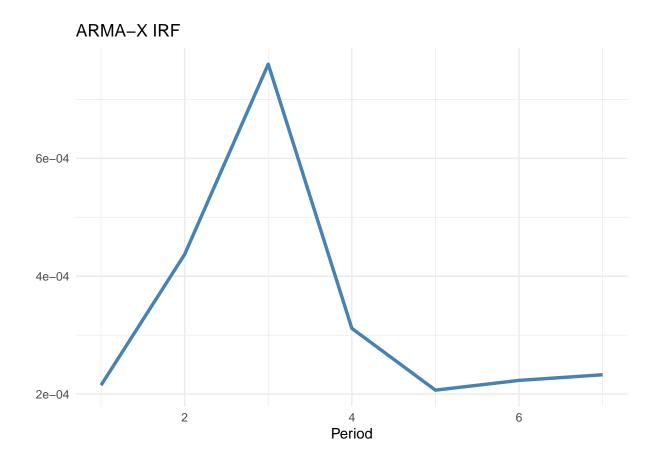
Table 11: ARMAX Model Results

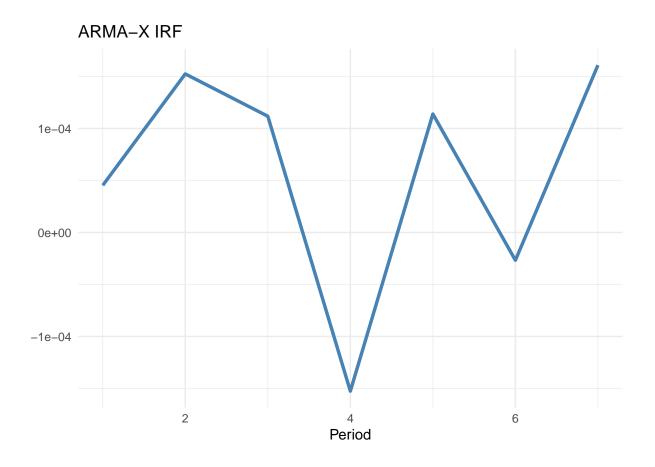
	Model 1
ar1	-0.6480^{***}
	(0.0101)
ar2	0.0399**
	(0.0127)
ar3	0.0285^{***}
	(0.0085)
ar4	0.1394^{***}
	(0.0119)
ar5	0.6351^{***}
	(0.0111)
ar6	0.6831^{***}
	(0.0076)
ma1	0.9475^{***}
	$(0.0075) \\ 0.2704***$
ma2	
	(0.0130)
ma3	0.1722^{***}
	(0.0112)
ma4	0.0623***
	(0.0097)
ma5	-0.6173^{***}
	(0.0105)
ma6	-0.8010^{***}
	(0.0056)
intercept	0.0225***
. 1 1 0	(0.0045)
$trade_lag_0$	0.0006
. 1 1 1	(0.0017)
$trade_lag_1$	0.0024
. 1 1 0	(0.0018)
$trade_lag_2$	0.0028
. 1 1 0	(0.0018)
$trade_lag_3$	-0.0012
ATC	(0.0017)
AIC	-47030.2624 -47030.2281
AICc	-47030.2281 -46888.0285
BIC	-46888.0285 23533.1312
Log Likelihood	23533.1312 19968
Num. obs. ***p < 0.001; ***p < 0.00	

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

Table 12: ARMAX selected by AIC







China Mention as Exogenous

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

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

	Model 1
ar1	0.9828***
	(0.0017)
ma1	-0.6793^{***}
	(0.0073)
ma2	-0.2139***
	(0.0087)
ma3	-0.0126
	(0.0080)
ma4	0.0355***
	(0.0071)
intercept	0.0215***
•	(0.0041)
$china_lag_0$	0.0047***
	(0.0012)
china_lag_1	0.0084***
	(0.0012)
$china_lag_2$	0.0054***
	(0.0012)
AIC	-45721.8164
AICc	-45721.8054
BIC	-45642.7971
Log Likelihood	22870.9082
Num. obs.	19969
***n < 0.001: **n < 0.0	01·*n < 0.05

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

Table 13: ARMAX Model Results

	Model 1
ar1	0.3584***
	(0.0071)
ar2	0.0405^{***}
	(0.0075)
ar3	0.0972^{***}
	(0.0074)
ar4	0.1022***
	(0.0075)
ar5	0.0825^{***}
	(0.0071)
intercept	0.0215^{***}
	(0.0018)
$china_lag_0$	0.0047^{***}
	(0.0012)
$china_lag_1$	0.0079***
	(0.0013)
$china_lag_2$	0.0051^{***}
	(0.0012)
AIC	-44680.9095
AICc	-44680.8985
BIC	-44601.8902
Log Likelihood	22350.4548
Num. obs.	19969
*** $p < 0.001$: ** $p < 0.01$: * $p < 0.05$	

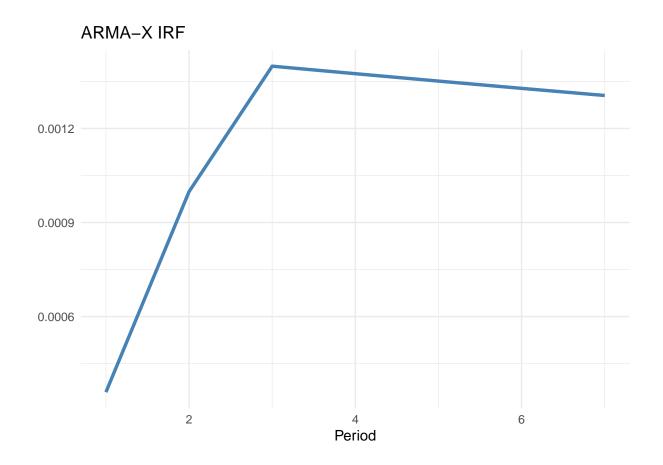
^{***}p < 0.001; **p < 0.01; *p < 0.05

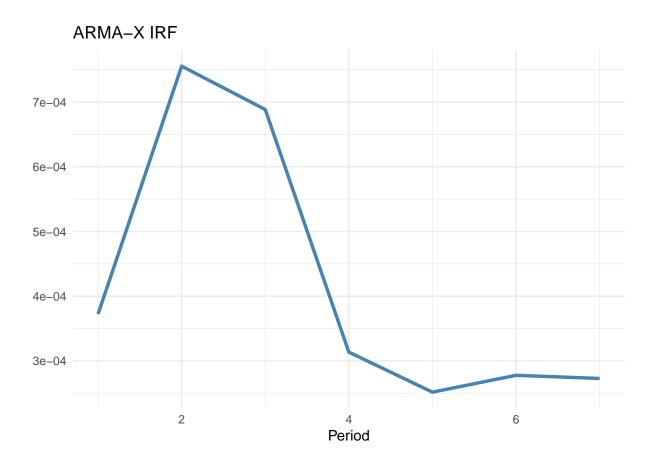
Table 14: ARMAX Model Results

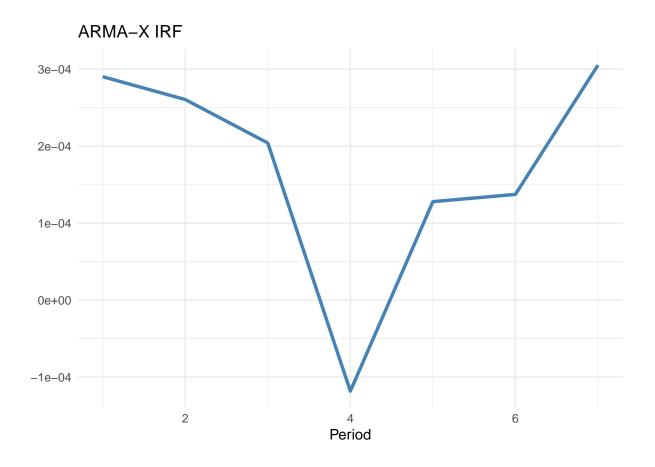
	Model 1
ar1	-0.6497***
	(0.0103)
ar2	0.0325*
	(0.0132)
ar3	0.0203^{*}
	(0.0086)
ar4	0.1350***
	(0.0120)
ar5	0.6374^{***}
	(0.0110)
ar6	0.6859^{***}
	(0.0076)
ma1	0.9502^{***}
	(0.0078)
ma2	0.2775^{***}
	(0.0136)
ma3	0.1797^{***}
	(0.0118)
ma4	0.0682***
	(0.0101)
ma5	-0.6136^{***}
	(0.0106)
ma6	-0.7998***
	(0.0056)
intercept	0.0217^{***}
	(0.0040)
$china_lag_0$	0.0039***
	(0.0011)
$china_lag_1$	0.0060***
	(0.0011)
$china_lag_2$	0.0049***
	(0.0011)
AIC	-47085.9168
AICc	-47085.8861
BIC	-46951.5839
Log Likelihood	23559.9584
Num. obs.	19969
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$	

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

Table 15: ARMAX selected by AIC







Positive Vibe as Exogenous

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

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

	Model 1
ar1	0.9828***
	(0.0017)
ma1	-0.6777^{***}
	(0.0073)
ma2	-0.2146^{***}
	(0.0087)
ma3	-0.0117
	(0.0080)
ma4	0.0345^{***}
	(0.0071)
intercept	0.0195^{***}
	(0.0042)
$prop_positive_lag_0$	0.0068***
	(0.0017)
$prop_positive_lag_1$	0.0027
	(0.0017)
$prop_positive_lag_2$	0.0045^{**}
	(0.0017)
AIC	-45686.3386
AICc	-45686.3275
BIC	-45607.3192
Log Likelihood	22853.1693
Num. obs.	19969

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

Table 16: ARMAX Model Results

	Model 1
ar1	0.3594***
	(0.0071)
ar2	0.0398^{***}
	(0.0075)
ar3	0.0987***
	(0.0074)
ar4	0.1032***
	(0.0075)
ar5	0.0820***
	(0.0071)
intercept	0.0194***
	(0.0019)
prop_positive_lag_0	0.0073***
	(0.0017)
$prop_positive_lag_1$	0.0030
	(0.0017)
prop_positive_lag_2	0.0045^{**}
	(0.0017)
AIC	-44656.4789
AICc	-44656.4679
BIC	-44577.4595
Log Likelihood	22338.2394
Num. obs.	19969
***	0.05

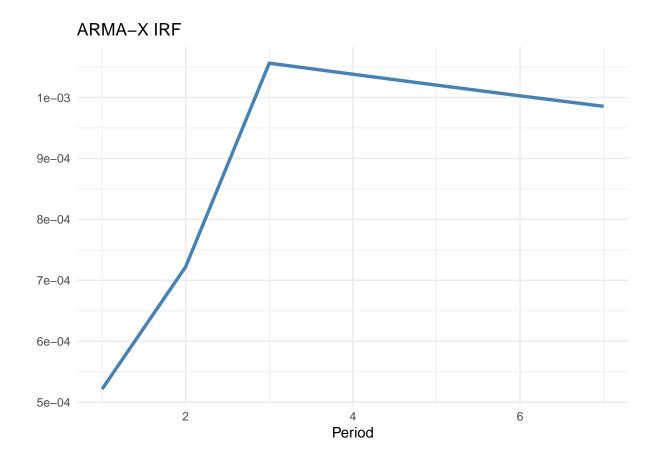
^{***}p < 0.001; **p < 0.01; *p < 0.05

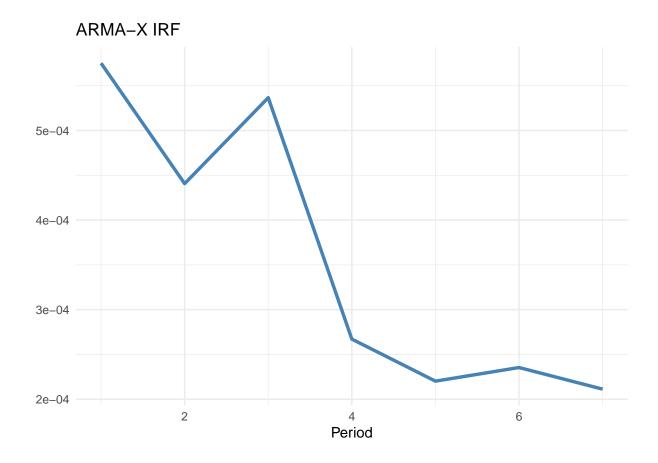
Table 17: ARMAX Model Results

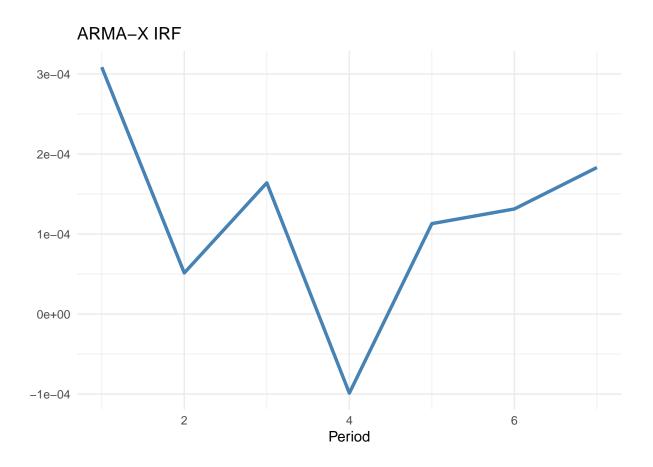
	Model 1
ar1	-0.6498***
	(0.0102)
ar2	0.0305^{*}
	(0.0129)
ar3	0.0197^{*}
	(0.0085)
ar4	0.1384***
	(0.0118)
ar5	0.6398***
	(0.0109)
ar6	0.6852***
	(0.0076)
ma1	0.9498***
	(0.0077)
ma2	0.2786***
	(0.0132)
ma3	0.1814***
	(0.0115)
ma4	0.0670^{***}
	(0.0099)
ma5	-0.6165^{***}
	(0.0104)
ma6	-0.8008***
	(0.0055)
intercept	0.0204^{***}
	(0.0042)
$prop_positive_lag_0$	0.0041**
	(0.0015)
prop_positive_lag_1	0.0034^{*}
	(0.0016)
prop_positive_lag_2	0.0025
	(0.0015)
AIC	-47042.9836
AICc	-47042.9529
BIC	-46908.6507
Log Likelihood	23538.4918
Num. obs.	19969
***n < 0.001, **n < 0.01, *n < 0.05	

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

Table 18: ARMAX selected by AIC







Negative Vibe as Exogenous

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

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

	Model 1
ar1	0.9828***
	(0.0017)
ma1	-0.6776^{***}
	(0.0073)
ma2	-0.2142^{***}
	(0.0087)
ma3	-0.0126
	(0.0080)
ma4	0.0348^{***}
	(0.0071)
intercept	0.0213^{***}
	(0.0042)
$prop_negative_lag_0$	0.0068^{**}
	(0.0023)
$prop_negative_lag_1$	0.0030
	(0.0023)
$prop_negative_lag_2$	-0.0007
	(0.0023)
AIC	-45673.0125
AICc	-45673.0015
BIC	-45593.9932
Log Likelihood	22846.5063
Num. obs.	19969

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

Table 19: ARMAX Model Results

	Model 1
ar1	0.3594***
	(0.0071)
ar2	0.0403^{***}
	(0.0075)
ar3	0.0978^{***}
	(0.0074)
ar4	0.1033^{***}
	(0.0075)
ar5	0.0819^{***}
	(0.0071)
intercept	0.0212^{***}
	(0.0019)
$prop_negative_lag_0$	0.0075^{**}
	(0.0023)
$prop_negative_lag_1$	0.0036
	(0.0023)
$prop_negative_lag_2$	-0.0012
	(0.0023)
AIC	-44643.1033
AICc	-44643.0923
BIC	-44564.0840
Log Likelihood	22331.5517
Num. obs.	19969
**** : 0 001 ** : 0 01 * : 0	

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

Table 20: ARMAX Model Results

	Model 1
ar1	-0.6518***
CII	(0.0101)
ar2	0.0293^*
ai 2	(0.0129)
ar3	0.0196*
aro	(0.0084)
ar4	0.1367***
	(0.0117)
ar5	0.6404***
W10	(0.0108)
ar6	0.6871***
V V	(0.0075)
ma1	0.9511***
	(0.0076)
ma2	0.2807***
	(0.0132)
ma3	0.1829***
	(0.0115)
ma4	0.0688***
	(0.0100)
ma5	-0.6155^{***}
	(0.0104)
ma6	-0.8013^{***}
	(0.0055)
intercept	0.0213***
	(0.0041)
$prop_negative_lag_0$	0.0061**
	(0.0020)
$prop_negative_lag_1$	0.0034
	(0.0020)
AIC	-47047.3129
AICc	-47047.2856
BIC	-46920.8811
Log Likelihood	23539.6564
Num. obs.	19970
***n < 0.001: **n < 0.01: *n < 0	0.05

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

Table 21: ARMAX selected by AIC

