VGK ARMA-X Analysis

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	Model 1
ar1	0.9844***
	(0.0025)
ma1	-0.7428***
	(0.0075)
ma2	-0.1699****
	(0.0087)
ma3	-0.0270^{***}
	(0.0072)
intercept	0.0003***
_	(0.0001)
$dummy_lag_0$	0.0000^*
	(0.0000)
dummy_lag_1	0.0000
	(0.0000)
$dummy_lag_2$	0.0000
v — 0—	(0.0000)
AIC	-200264.8032
AICc	-200264.7942
BIC	-200193.6858
Log Likelihood	100141.4016
Num. obs.	19969
*** p < 0.001: ** p < 0.0	01: *p < 0.05

p < 0.001; p < 0.01; p < 0.05

Table 1: ARMAX Model Results

Stationarity

```
adf.test(data$VGK_vol)
adf.test(data$N)
adf.test(data$tariff)
adf.test(data$china)
```

FTSE Developed Europe All Cap Index Univariate ARMA-X Models

Tweet Dummy as Exogenous

```
#auto.armax selects the lowest AIC value given r (exogenous variable lags)
res1 = auto.armax(data$VGK_vol,xreg=data$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$VGK_vol, xreg=data$dummy, nb.lags=2,
                  p=5, q=0, d=0, latex=T)
```

	Model 1
ar1	0.2612***
	(0.0071)
ar2	0.0243^{***}
	(0.0073)
ar3	0.0306^{***}
	(0.0073)
ar4	0.0391***
	(0.0073)
ar5	0.0471***
	(0.0071)
intercept	0.0003***
	(0.0000)
$dummy_lag_0$	0.0000^{**}
	(0.0000)
$dummy_lag_1$	0.0000
	(0.0000)
$dummy_lag_2$	0.0000
	(0.0000)
AIC	-199889.4166
AICc	-199889.4056
BIC	-199810.3972
Log Likelihood	99954.7083
Num. obs.	19969
*** p < 0.001: ** p < 0.0	01: *p < 0.05

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

Table 2: ARMAX Model Results

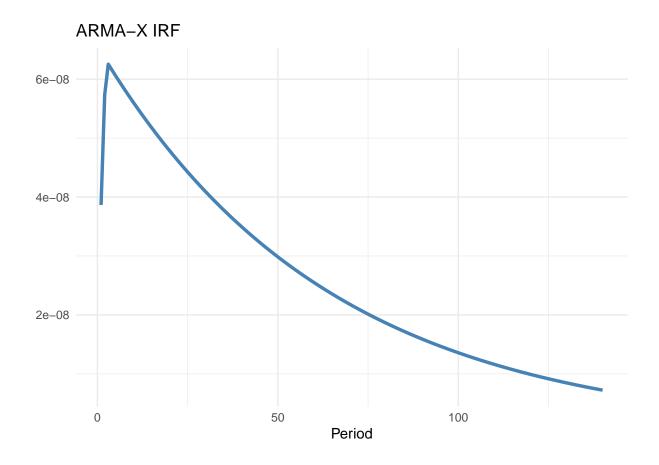
	Model 1
ar1	0.9843***
	(0.0025)
ma1	-0.7428^{***}
	(0.0075)
ma2	-0.1703^{***}
	(0.0087)
ma3	-0.0265^{***}
	(0.0072)
intercept	0.0004
dummy_lag_0	0.0000
AIC	-200279.9599
AICc	-200279.9543
BIC	-200224.6457
Log Likelihood	100146.9800
Num. obs.	19971
*** p < 0.001: ** p < 0.0	0.05

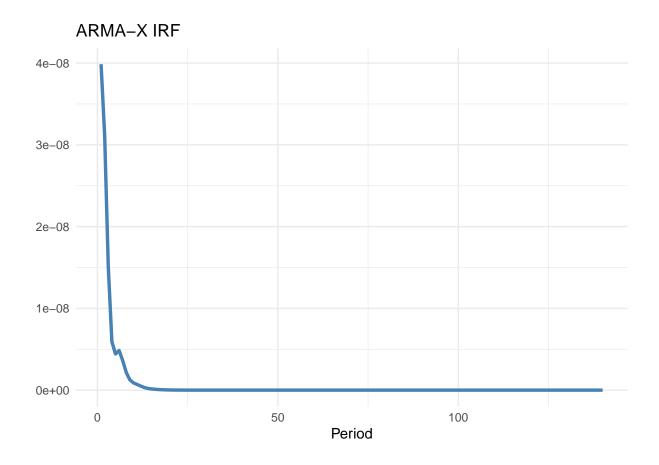
 $^{^{***}}p < 0.001; \ ^{**}p < 0.01; \ ^*p < 0.05$

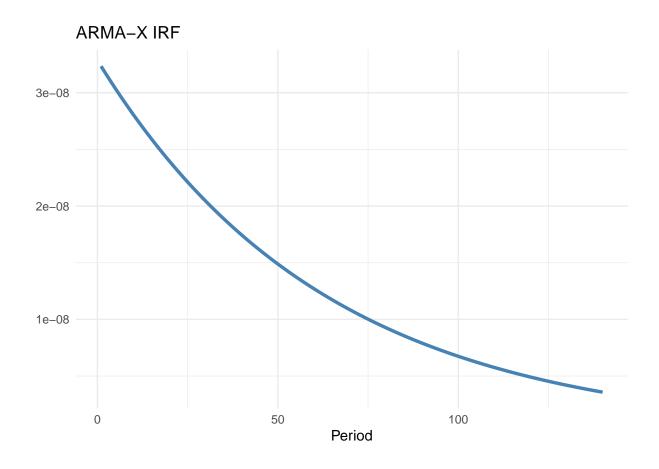
Table 3: ARMAX selected by AIC $\,$

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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







Tweet Count as Exogenous

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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

Model 1
0.2611***
(0.0071)
0.0237^{**}
(0.0073)
0.0309***
(0.0074)
0.0395***
(0.0073)
0.0466***
(0.0071)
0.0004***
(0.0000)
0.0000
(0.0000)
0.0000
(0.0000)
0.0000
(0.0000)
-199875.0009
-199874.9899
-199795.9815
99947.5004
19969

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

Table 4: ARMAX Model Results

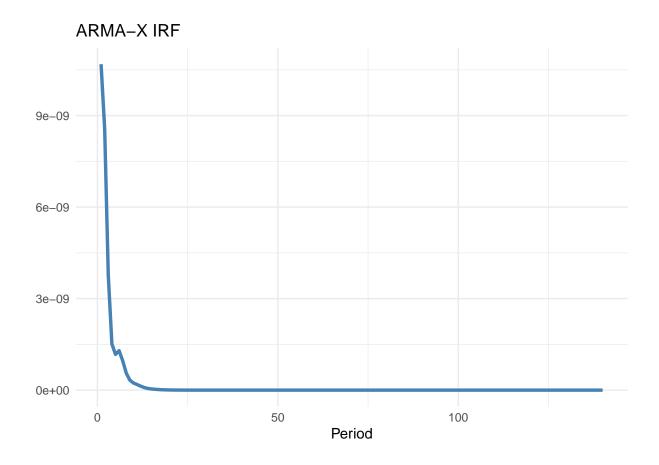
	Model 1
ar1	0.2611***
	(0.0071)
ar2	0.0237^{**}
	(0.0073)
ar3	0.0309***
	(0.0074)
ar4	0.0395***
	(0.0073)
ar5	0.0466***
	(0.0071)
intercept	0.0004***
	(0.0000)
N_lag_0	0.0000
	(0.0000)
N_lag_1	0.0000
	(0.0000)
N_{lag_2}	0.0000
	(0.0000)
AIC	-199875.0009
AICc	-199874.9899
BIC	-199795.9815
Log Likelihood	99947.5004
Num. obs.	19969
*** n < 0.001 · ** n < 0.0	01·*n < 0.05

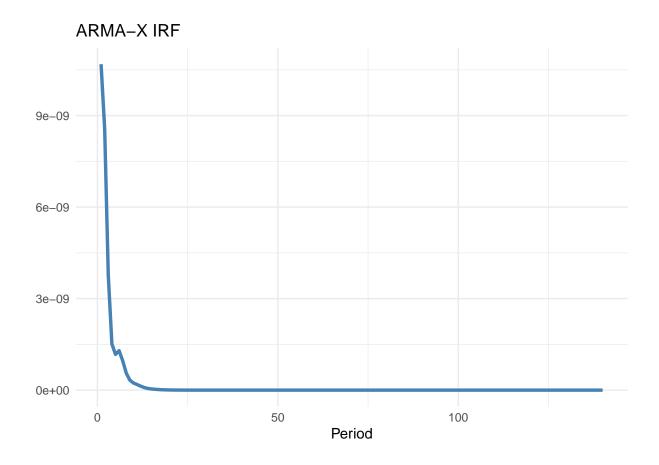
^{***}p < 0.001; **p < 0.01; *p < 0.05

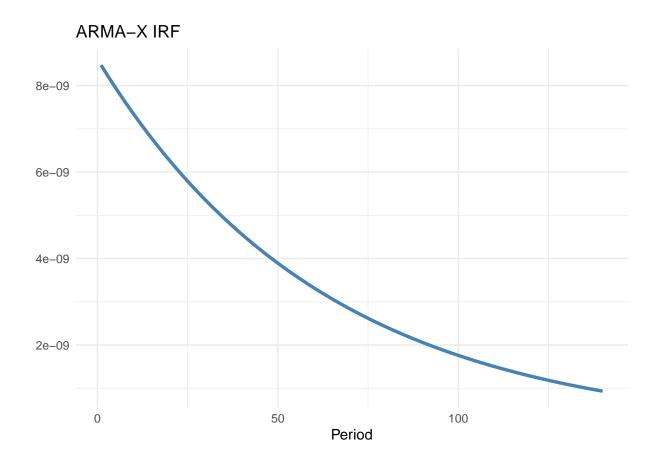
Table 5: ARMAX Model Results

	Model 1
ar1	0.9843***
	(0.0025)
ma1	-0.7427^{***}
	(0.0075)
ma2	-0.1708***
	(0.0087)
ma3	-0.0260***
	(0.0072)
intercept	0.0004
$N_{lag}0$	0.0000
AIC	-200269.0091
AICc	-200269.0035
BIC	-200213.6949
Log Likelihood	100141.5046
Num. obs.	19971
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$	

Table 6: ARMAX selected by AIC







Tariff as Exogenous

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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

	Model 1
ar1	1.1886***
	(0.0089)
ar2	-0.2208***
	(0.0111)
ar3	0.0049
	(0.0110)
ar4	$0.0131^{'}$
	(0.0075)
ma1	-0.9477^{***}
	(0.0054)
intercept	0.0004***
_	(0.0000)
$tariff_lag_0$	0.0000
_ 0_	(0.0000)
$tariff_lag_1$	0.0002***
	(0.0000)
$tariff_lag_2$	0.0001^{*}
	(0.0000)
AIC	-200247.2535
AICc	-200247.2424
BIC	-200168.2341
Log Likelihood	100133.6267
Num. obs.	19969
*** n < 0.001: ** n < 0.0	$0.1 \cdot *n < 0.05$

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

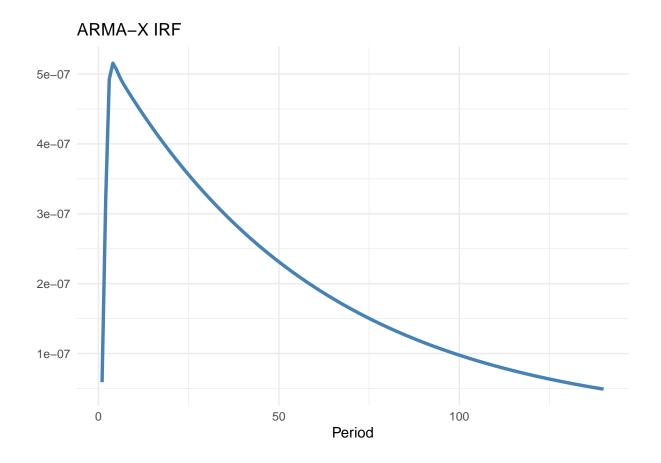
Table 7: ARMAX Model Results

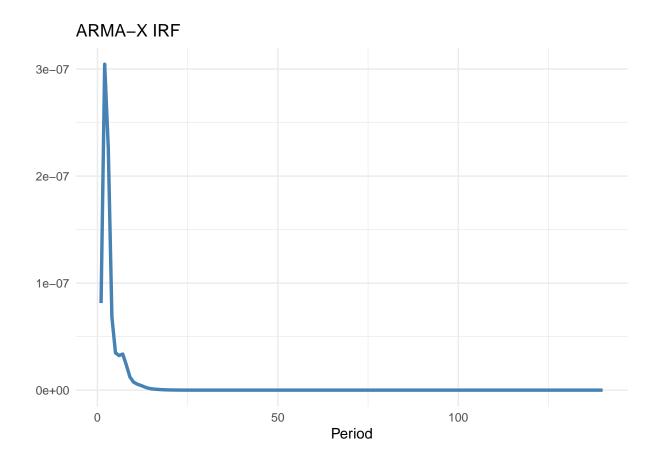
	Model 1
ar1	0.2606***
	(0.0071)
ar2	0.0232^{**}
	(0.0073)
ar3	0.0284^{***}
	(0.0073)
ar4	0.0376^{***}
	(0.0073)
ar5	0.0461^{***}
	(0.0071)
intercept	0.0004***
	(0.0000)
$tariff_lag_0$	0.0001
	(0.0000)
tariff_lag_1	0.0002^{***}
	(0.0000)
$tariff_lag_2$	0.0001**
	(0.0000)
AIC	-199877.9797
AICc	-199877.9687
BIC	-199798.9603
Log Likelihood	99948.9898
Num. obs.	19969
***p < 0.001; **p < 0.0	01; *p < 0.05

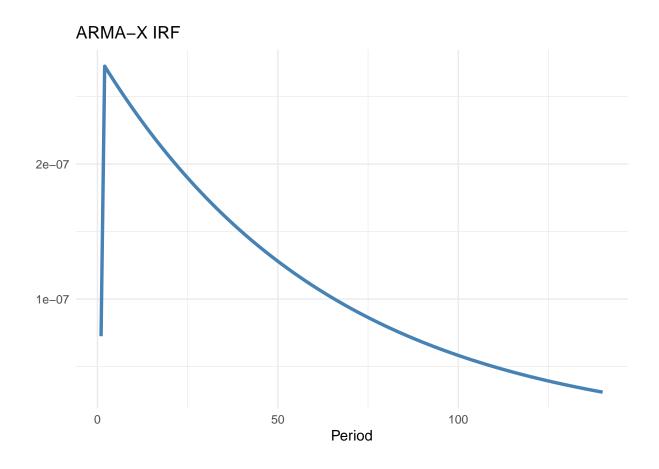
Table 8: ARMAX Model Results

	Model 1
ar1	0.9844***
	(0.0025)
ma1	-0.7432^{***}
	(0.0075)
ma2	-0.1701^{***}
	(0.0087)
ma3	-0.0270^{***}
	(0.0073)
intercept	0.0004***
	(0.0000)
$tariff_lag_0$	0.0000
	(0.0000)
$tariff_lag_1$	0.0001***
	(0.0000)
AIC	-200253.6847
AICc	-200253.6775
BIC	-200190.4688
Log Likelihood	100134.8424
Num. obs.	19970
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$	

Table 9: ARMAX selected by AIC







Trade Mention as Exogenous

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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

	Model 1
ar1	0.9844***
	(0.0025)
ma1	-0.7425***
	(0.0075)
ma2	-0.1709^{***}
	(0.0087)
ma3	-0.0266****
	(0.0072)
intercept	0.0004***
	(0.0000)
$trade_lag_0$	-0.0000
	(0.0000)
$trade_lag_1$	0.0000
	(0.0000)
$trade_lag_2$	0.0001
	(0.0000)
AIC	-200224.4642
AICc	-200224.4552
BIC	-200153.3468
Log Likelihood	100121.2321
Num. obs.	19969
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$	

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

Table 10: ARMAX Model Results

	Model 1
ar1	0.2615***
	(0.0071)
ar2	0.0232^{**}
	(0.0073)
ar3	0.0298^{***}
	(0.0073)
ar4	0.0386^{***}
	(0.0073)
ar5	0.0458^{***}
	(0.0071)
intercept	0.0004***
	(0.0000)
$trade_lag_0$	-0.0000
	(0.0000)
${\rm trade_lag_1}$	0.0000
	(0.0000)
$trade_lag_2$	0.0001
	(0.0000)
AIC	-199846.8393
AICc	-199846.8283
BIC	-199767.8200
Log Likelihood	99933.4197
Num. obs.	19969
*** < 0.001. ** < 0.0	01 * .0.05

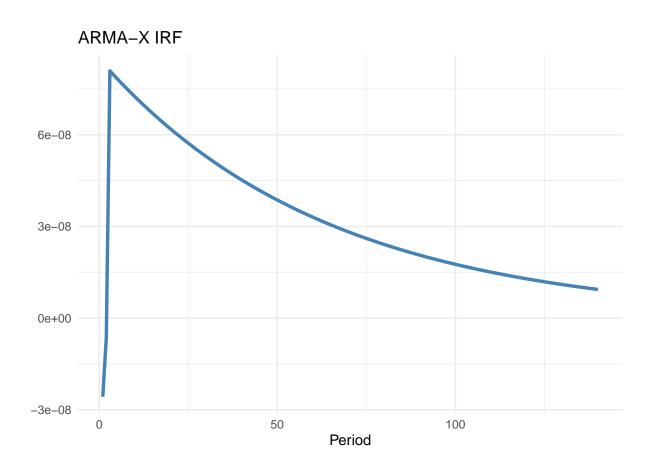
^{***}p < 0.001; **p < 0.01; *p < 0.05

Table 11: ARMAX Model Results

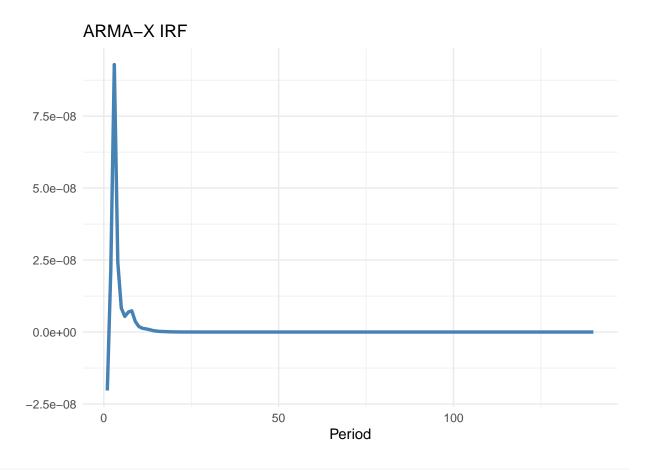
	Model 1
ar1	0.9843***
	(0.0025)
ma1	-0.7424^{***}
	(0.0075)
ma2	-0.1706^{***}
	(0.0087)
ma3	-0.0267^{***}
	(0.0072)
intercept	0.0004***
	(0.0000)
$trade_lag_0$	-0.0000
	(0.0000)
AIC	-200248.5095
AICc	-200248.5039
BIC	-200193.1952
Log Likelihood	100131.2548
Num. obs.	19971
***n < 0.001 · **n < 0.01 · *n < 0.05	

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

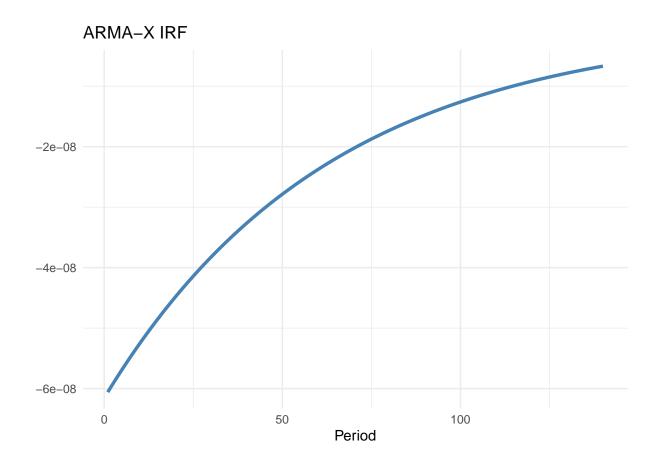
Table 12: ARMAX selected by AIC



irf.plot(res2,nb.periods)



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



China Mention as Exogenous

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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

	Model 1
ar1	1.1894***
	(0.0088)
ar2	-0.2214^{***}
	(0.0111)
ar3	0.0064
	(0.0110)
ar4	0.0119
	(0.0075)
ma1	-0.9482^{***}
	(0.0053)
intercept	0.0004***
•	(0.0000)
china_lag_0	0.0001*
_ 0_	(0.0000)
china lag 1	0.0001*
_ 0_	(0.0000)
china_lag_2	0.0000
_ 0_	(0.0000)
AIC	-200232.3259
AICc	-200232.3148
BIC	-200153.3065
Log Likelihood	100126.1629
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

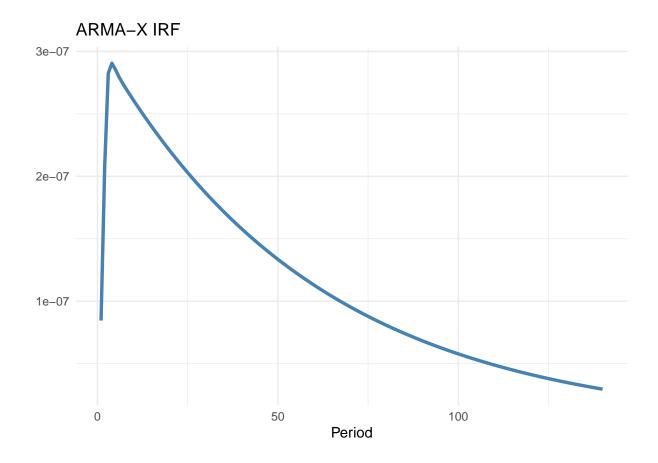
	Model 1
ar1	0.2610***
	(0.0071)
ar2	0.0232**
	(0.0073)
ar3	0.0300^{***}
	(0.0073)
ar4	0.0389^{***}
	(0.0073)
ar5	0.0460***
	(0.0071)
intercept	0.0004^{***}
	(0.0000)
$china_lag_0$	0.0001^*
	(0.0000)
$china_lag_1$	0.0001^{**}
	(0.0000)
$china_lag_2$	0.0000
	(0.0000)
AIC	-199856.3346
AICc	-199856.3236
BIC	-199777.3152
Log Likelihood	99938.1673
Num. obs.	19969
*** $p < 0.001$; *** $p < 0.01$; * $p < 0.05$	

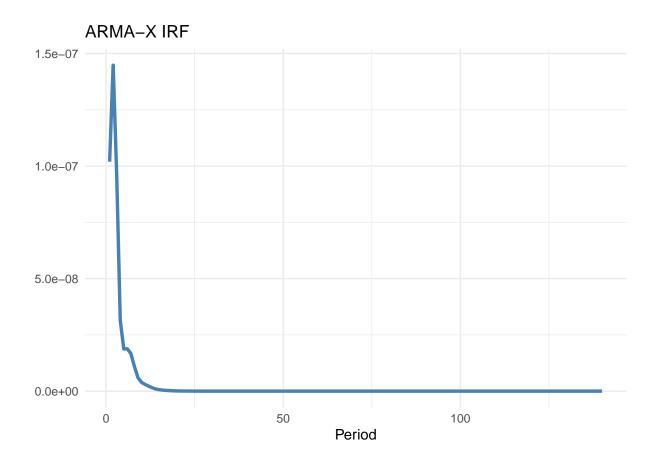
p < 0.001; p < 0.01; p < 0.01; p < 0.05

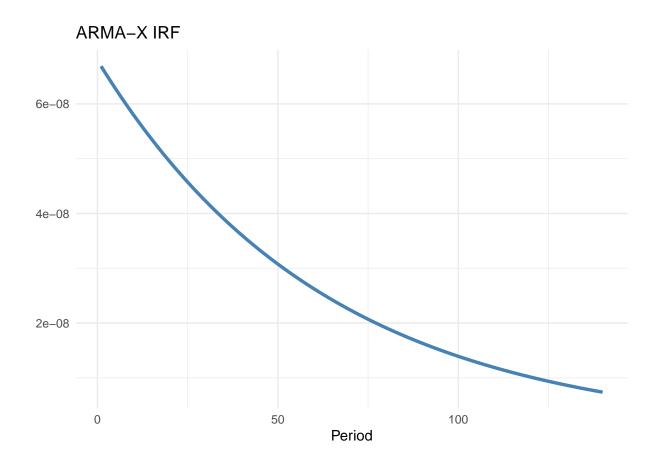
Table 14: ARMAX Model Results

	Model 1
ar1	0.9843***
	(0.0025)
ma1	-0.7427^{***}
	(0.0075)
ma2	-0.1708^{***}
	(0.0087)
ma3	-0.0261****
	(0.0072)
intercept	0.0004***
	(0.0000)
$china_lag_0$	0.0000
_	(0.0000)
AIC	-200251.1835
AICc	-200251.1779
BIC	-200195.8692
Log Likelihood	100132.5917
Num. obs.	19971
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$	

Table 15: ARMAX selected by AIC







Positive Vibe as Exogenous

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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

	Model 1
ar1	1.1896***
	(0.0088)
ar2	-0.2216^{***}
	(0.0111)
ar3	0.0068
	(0.0110)
ar4	0.0116
	(0.0075)
ma1	-0.9483^{***}
	(0.0053)
intercept	0.0003***
	(0.0000)
prop_positive_lag_0	0.0001**
	(0.0000)
prop_positive_lag_1	0.0001
	(0.0000)
prop_positive_lag_2	0.0000
	(0.0000)
AIC	-200235.6804
AICc	-200235.6694
BIC	-200156.6611
Log Likelihood	100127.8402
Num. obs.	19969

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

Table 16: ARMAX Model Results

	Model 1
ar1	0.2612***
	(0.0071)
ar2	0.0233^{**}
	(0.0073)
ar3	0.0304^{***}
	(0.0073)
ar4	0.0392***
	(0.0073)
ar5	0.0465***
	(0.0071)
intercept	0.0003***
	(0.0000)
$prop_positive_lag_0$	0.0001***
	(0.0000)
prop_positive_lag_1	0.0001
	(0.0000)
$prop_positive_lag_2$	0.0000
	(0.0000)
AIC	-199857.8911
AICc	-199857.8800
BIC	-199778.8717
Log Likelihood	99938.9455
Num. obs.	19969
*** n < 0.001, ** n < 0.01, * n < 0.01	0.05

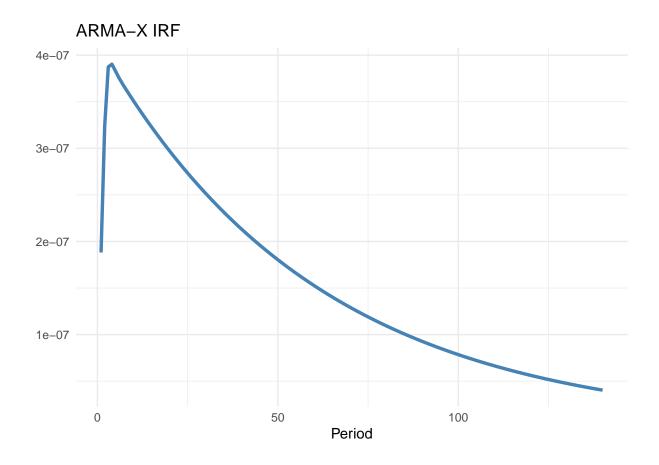
^{***}p < 0.001; **p < 0.01; *p < 0.05

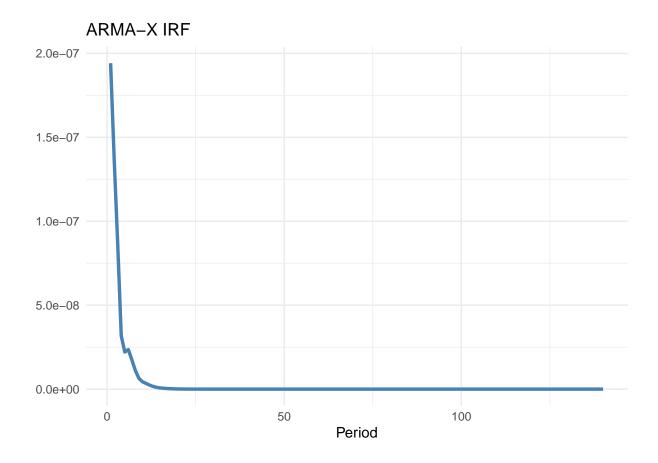
Table 17: ARMAX Model Results

	Model 1
ar1	0.9843***
	(0.0025)
ma1	-0.7426^{***}
	(0.0075)
ma2	-0.1708^{***}
	(0.0087)
ma3	-0.0262^{***}
	(0.0072)
intercept	0.0004***
	(0.0000)
$prop_positive_lag_0$	0.0001**
	(0.0000)
AIC	-200258.3763
AICc	-200258.3707
BIC	-200203.0620
Log Likelihood	100136.1881
Num. obs.	19971

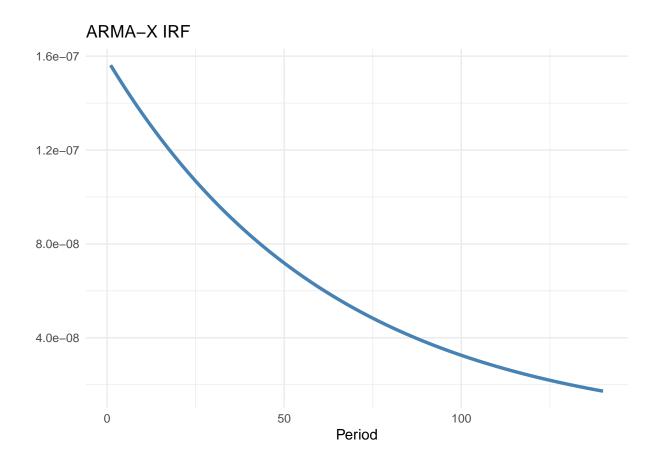
 $^{^{***}}p < 0.001; \ ^{**}p < 0.01; \ ^*p < 0.05$

Table 18: ARMAX selected by AIC





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



Negative Vibe as Exogenous

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20

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

	Model 1
ar1	1.1897***
	(0.0088)
ar2	-0.2216^{***}
	(0.0111)
ar3	0.0063
	(0.0110)
ar4	0.0119
	(0.0075)
ma1	-0.9482^{***}
	(0.0053)
intercept	0.0004***
	(0.0000)
prop_negative_lag_0	0.0001
	(0.0000)
$prop_negative_lag_1$	0.0001
	(0.0000)
$prop_negative_lag_2$	0.0000
	(0.0000)
AIC	-200224.5140
AICc	-200224.5030
BIC	-200145.4947
Log Likelihood	100122.2570
Num. obs.	19969

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

Table 19: ARMAX Model Results

	Model 1
ar1	0.2614***
	(0.0071)
ar2	0.0233^{**}
	(0.0073)
ar3	0.0301***
	(0.0073)
ar4	0.0389***
	(0.0073)
ar5	0.0459***
	(0.0071)
intercept	0.0004***
	(0.0000)
$prop_negative_lag_0$	0.0001
	(0.0000)
prop_negative_lag_1	0.0000
	(0.0000)
$prop_negative_lag_2$	-0.0000
	(0.0000)
AIC	-199846.3333
AICc	-199846.3223
BIC	-199767.3140
Log Likelihood	99933.1667
Num. obs.	19969
*** n < 0.001, ** n < 0.01, * n < 0	0.05

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

Table 20: ARMAX Model Results

	Model 1
ar1	0.9843***
COL 1	(0.0025)
ma1	-0.7425^{***}
maı	(0.0075)
ma2	-0.1707***
maz	00.
9	(0.0087)
ma3	-0.0265***
	(0.0072)
intercept	0.0004***
	(0.0000)
$prop_negative_lag_0$	0.0001
	(0.0000)
AIC	-200250.1442
AICc	-200250.1386
BIC	-200194.8300
Log Likelihood	100132.0721
Num. obs.	19971

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

Table 21: ARMAX selected by AIC

