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 = 3, max_q = 3, max_r = 3, criterion = "AIC", latex=T)
```

```
#we want to plot the IRFs of these models
nb.periods = 7 * 20
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	$0.1 \cdot *n < 0.05$

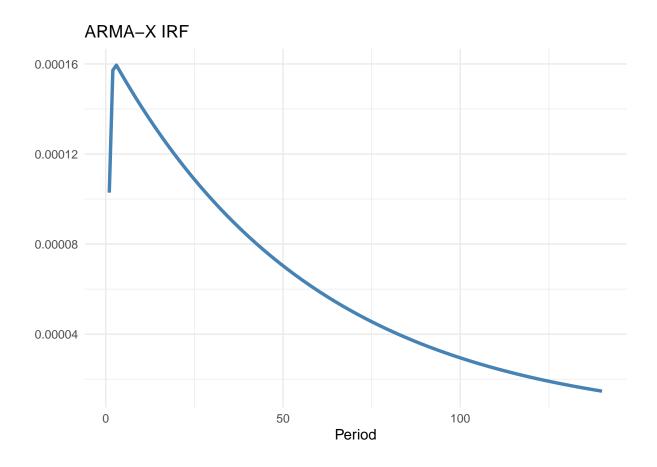
^{***}p < 0.001; **p < 0.01; *p < 0.05

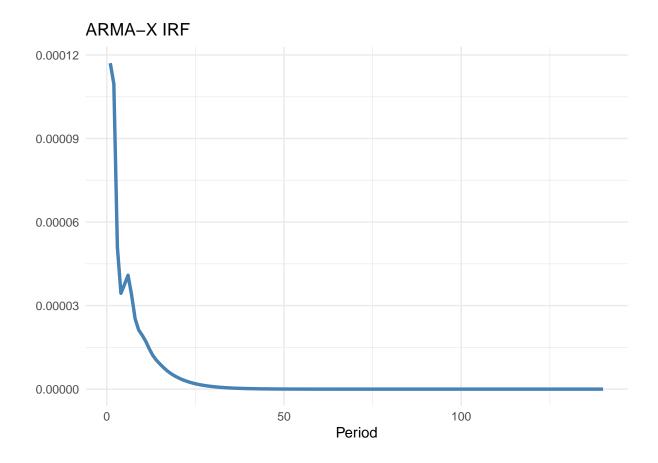
Table 2: ARMAX Model Results

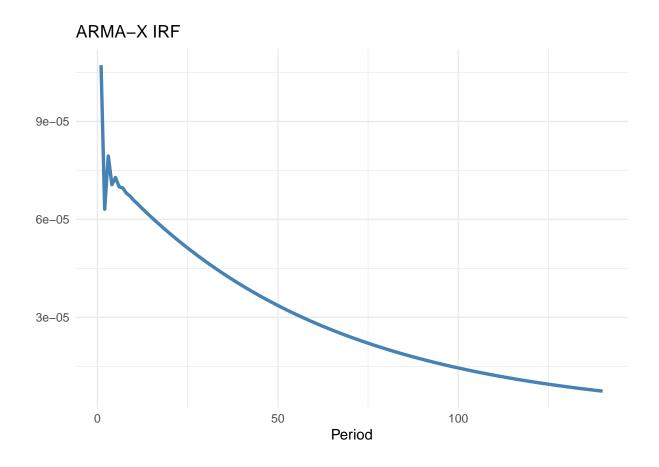
	Model 1
ar1	0.0300
	(0.0510)
ar2	0.7229***
	(0.0397)
ar3	0.2110^{***}
	(0.0287)
ma1	0.2751^{***}
	(0.0496)
ma2	-0.6445^{***}
	(0.0284)
ma3	-0.3527***
	(0.0256)
intercept	0.0202^{***}
	(0.0042)
$dummy_lag_0$	0.0014^{***}
	(0.0002)
$dummy_lag_1$	0.0008***
	(0.0002)
AIC	-45761.2161
AICc	-45761.2051
BIC	-45682.1963
Log Likelihood	22890.6081
Num. obs.	19970
***n < 0.001: **n < 0.0	01: *n < 0.05

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

Table 3: ARMAX selected by AIC







Tweet Count as Exogenous

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

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
*** $p < 0.001$: ** $p < 0.01$: * $p < 0.05$	

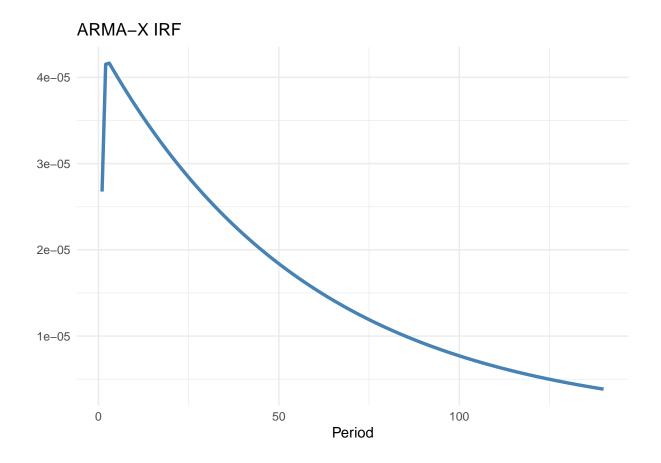
^{***}p < 0.001; **p < 0.01; *p < 0.05

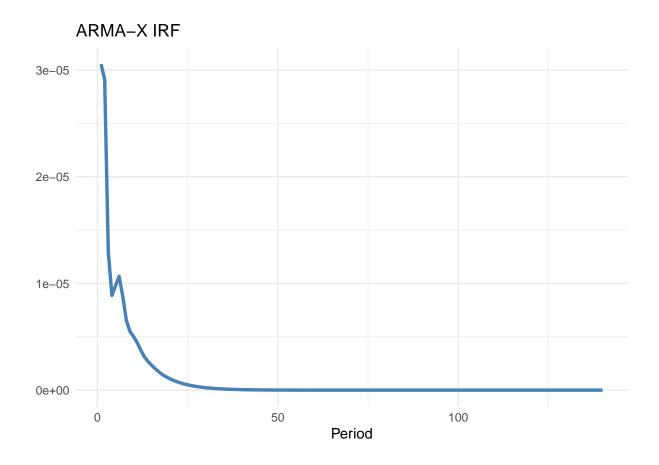
Table 5: ARMAX Model Results

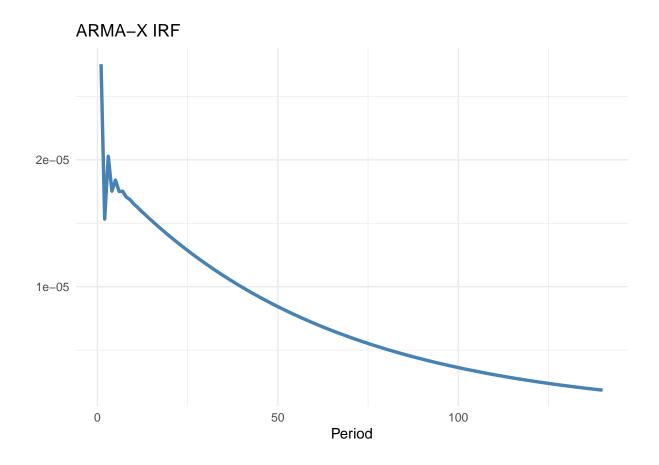
	Model 1
ar1	0.0278
	(0.0510)
ar2	0.7210***
	(0.0399)
ar3	0.2148***
	(0.0284)
ma1	0.2779***
	(0.0496)
ma2	-0.6430^{***}
	(0.0285)
ma3	-0.3563^{***}
	(0.0253)
intercept	0.0211***
-	(0.0042)
N_{lag_0}	0.0004***
_ 0_	(0.0001)
N_lag_1	0.0002^{**}
	(0.0001)
AIC	-45737.6695
AICc	-45737.6585
BIC	-45658.6497
Log Likelihood	22878.8348
Num. obs.	19970

***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	1.7000***
	(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 < 0.01: * n < 0.05	

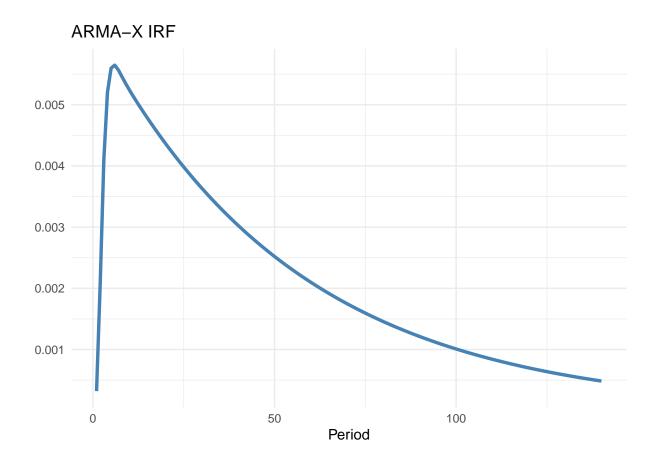
 $^{***}p < 0.001; \ ^{**}p < 0.01; \ ^*p < 0.05$

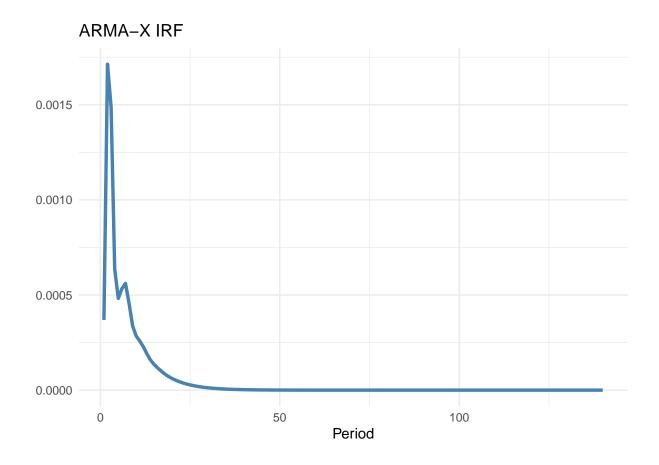
Table 8: ARMAX Model Results

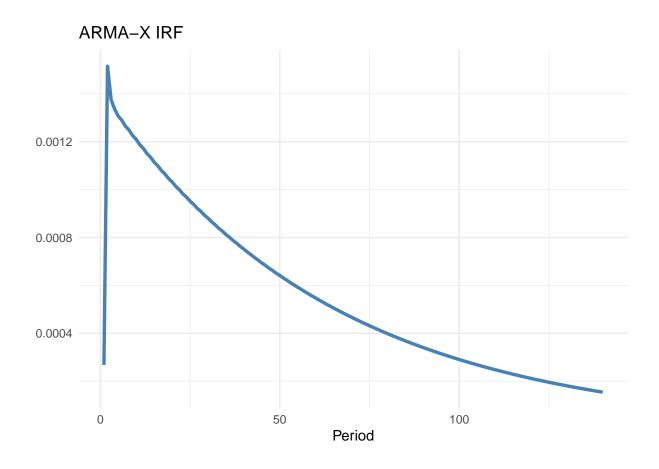
	Model 1
ar1	0.2200***
	(0.0084)
ar2	0.9388***
	(0.0037)
ar3	-0.1837***
	(0.0079)
ma1	0.0870***
	(0.0042)
ma2	-0.8960^{***}
	(0.0042)
intercept	0.0219***
-	(0.0042)
$tariff_lag_0$	0.0035^{*}
_	(0.0014)
$tariff_lag_1$	0.0191***
_	(0.0015)
$tariff_lag_2$	0.0103***
_	(0.0015)
$tariff_lag_3$	-0.0045^{**}
_ 0_	(0.0014)
AIC	-46020.9547
AICc	-46020.9415
BIC	-45934.0340
Log Likelihood	23021.4774
Num. obs.	19968
***p < 0.001; **p < 0.0	01: *p < 0.05

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	0.9828***
	(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
*** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$	

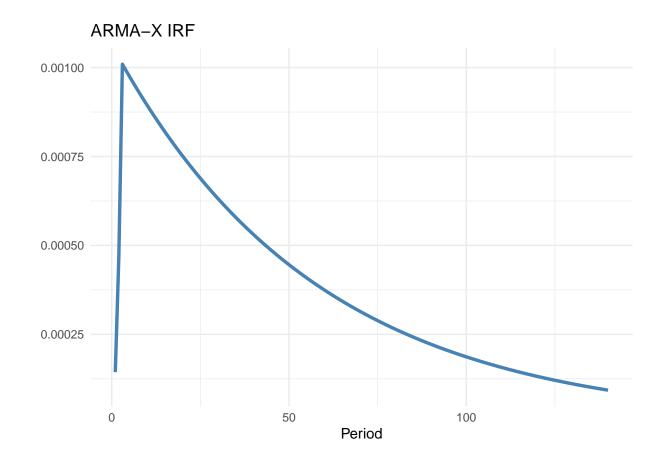
p < 0.001; p < 0.01; p < 0.01; p < 0.05

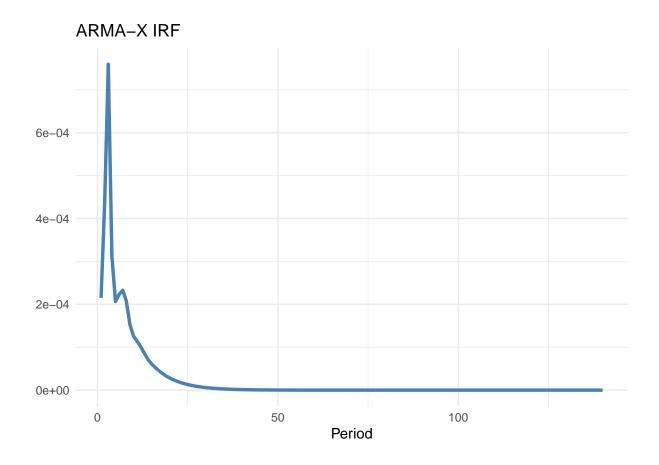
Table 11: ARMAX Model Results

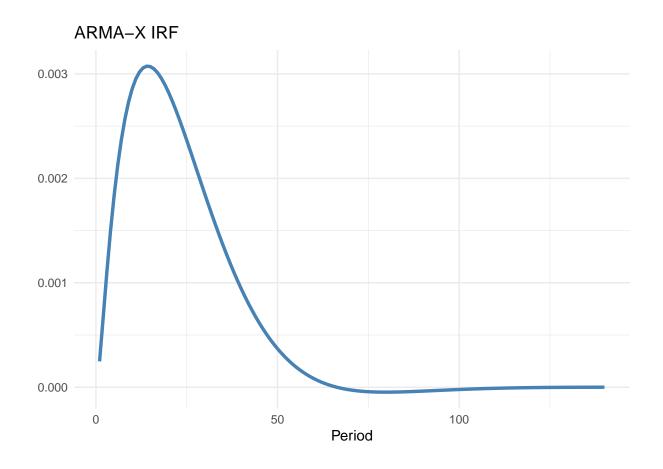
	Model 1
ar1	2.1903***
	(0.0096)
ar2	-1.4727^{***}
	(0.0173)
ar3	0.2784^{***}
	(0.0082)
ma1	-1.8955***
	(0.0062)
ma2	0.9165^{***}
	(0.0063)
intercept	0.0225***
	(0.0028)
$trade_lag_0$	0.0032
	(0.0018)
$trade_lag_1$	0.0016
	(0.0018)
AIC	-45816.1540
AICc	-45816.1449
BIC	-45745.0361
Log Likelihood	22917.0770
Num. obs.	19970
***	04 * .005

^{***}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 * 20

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	$0.1 \cdot *n < 0.05$

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

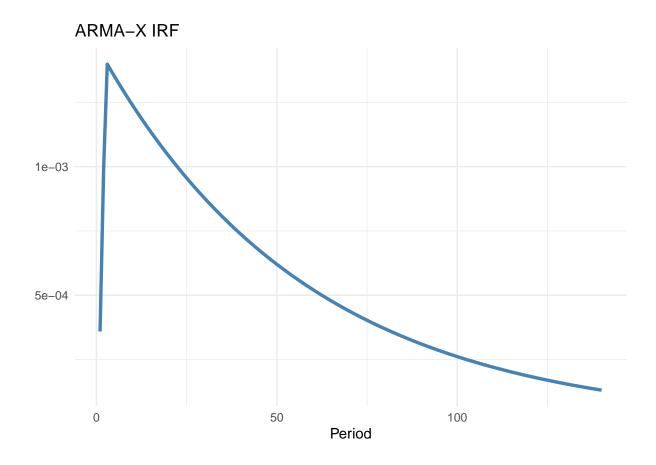
Table 13: ARMAX Model Results

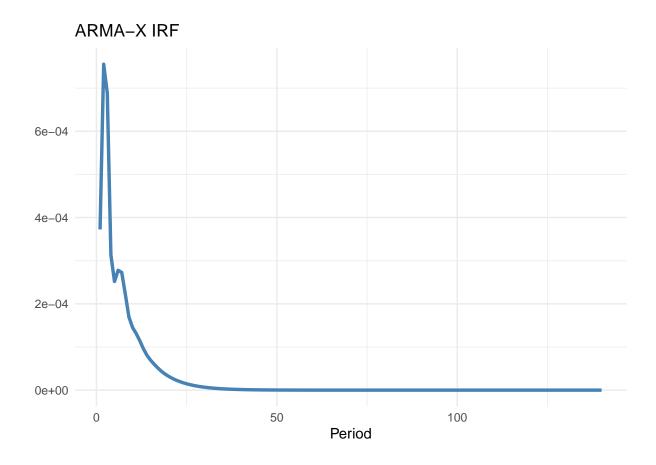
	35 114
	Model 1
ar1	0.3584***
	(0.0071)
ar2	0.0405^{***}
	(0.0075)
ar3	0.0972^{***}
	(0.0074)
ar4	0.1022^{***}
	(0.0075)
ar5	0.0825***
	(0.0071)
intercept	0.0215***
	(0.0018)
$china_lag_0$	0.0047^{***}
	(0.0012)
$china_lag_1$	0.0079^{***}
	(0.0013)
$china_lag_2$	0.0051***
	(0.0012)
AIC	-44680.9095
AICc	-44680.8985
BIC	-44601.8902
Log Likelihood	22350.4548
Num. obs.	19969
*** $p < 0.001$; ** $p < 0.001$	01; *p < 0.05

Table 14: ARMAX Model Results

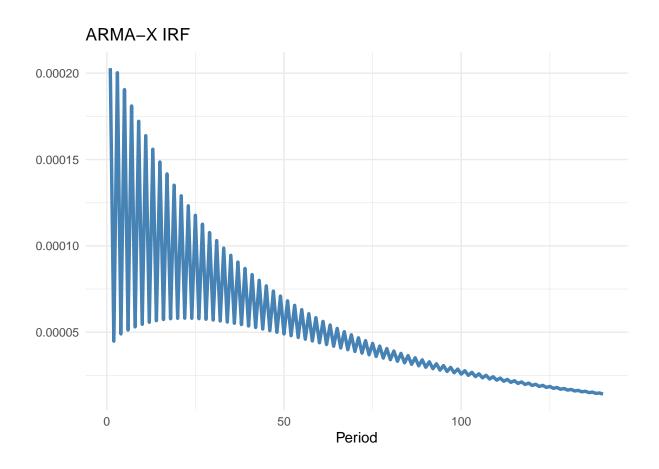
	Model 1
ar1	0.2209***
	(0.0084)
ar2	0.9382^{***}
	(0.0037)
ar3	-0.1837^{***}
	(0.0079)
ma1	0.0878***
	(0.0042)
ma2	-0.8950***
	(0.0042)
intercept	0.0225^{***}
	(0.0042)
$china_lag_0$	0.0026^{*}
	(0.0012)
AIC	-45840.5349
AICc	-45840.5277
BIC	-45777.3186
Log Likelihood	22928.2675
Num. obs.	19971
***p < 0.001; **p < 0.0	01; *p < 0.05

Table 15: ARMAX selected by AIC





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



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	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
*** < 0.001. ** < 0.01. * <	0.05

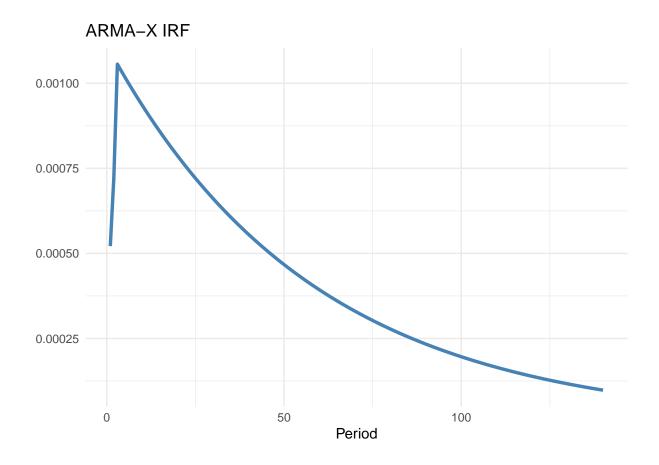
^{***}p < 0.001; **p < 0.01; *p < 0.05

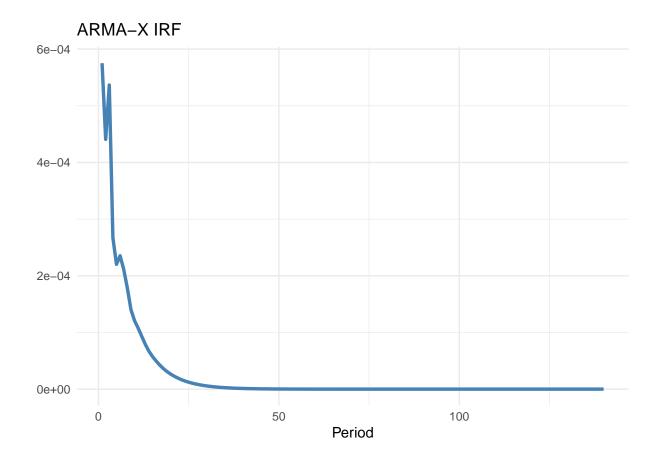
Table 17: ARMAX Model Results

	Model 1
ar1	0.0262
	(0.0503)
ar2	0.7230***
	(0.0390)
ar3	0.2146***
	(0.0283)
ma1	0.2800***
	(0.0489)
ma2	-0.6451^{***}
	(0.0277)
ma3	-0.3571***
	(0.0252)
intercept	0.0212^{***}
	(0.0042)
$prop_positive_lag_0$	0.0063^{***}
	(0.0016)
AIC	-45722.7625
AICc	-45722.7534
BIC	-45651.6441
Log Likelihood	22870.3812
Num. obs.	19971

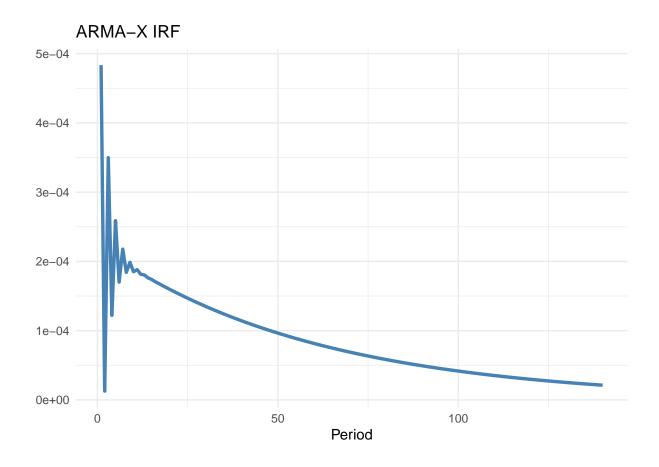
^{***}p < 0.001; **p < 0.01; *p < 0.05

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	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
***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	0.0237
	(0.0495)
ar2	0.7250^{***}
	(0.0379)
ar3	0.2150^{***}
	(0.0283)
ma1	0.2824^{***}
	(0.0481)
ma2	-0.6460^{***}
	(0.0270)
ma3	-0.3581***
	(0.0251)
intercept	0.0216^{***}
	(0.0042)
$prop_negative_lag_0$	0.0070^{**}
	(0.0022)
AIC	-45716.8054
AICc	-45716.7964
BIC	-45645.6871
Log Likelihood	22867.4027
Num. obs.	19971

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

Table 21: ARMAX selected by AIC

