

# ARMA-X Analysis Tutorial

## Contents

<b>Setup</b>	<b>2</b>
Load Libraries & Functions . . . . .	2
Load Data . . . . .	2
<b>Stationarity</b>	<b>3</b>
<b>Univariate ARMA-X Models</b>	<b>4</b>
Custom ARMA-X Specification . . . . .	4
Best ARMA-X Given r . . . . .	4
Best ARMA-X . . . . .	6
IRF . . . . .	6

# Setup

## Load Libraries & Functions

```
rm(list=ls())
require(tinytex) #LaTeX
require(ggplot2) #plots
require(AEC) #JP-Renne functions
require(AER) #NW formula
require(forecast) #time series stuff
require(expm) #matrix exponents
require(here) #directory finder
require(stringr) #analysis of strings, important for the detection in tweets
require(dplyr) #data management
require(lubridate) #data dates management
require(zoo) #for lagging
require(jtools) #tables
require(huxtable) #tables
require(lmtest) #reg tests
require(vroom) #for loading data
require(data.table) #for data filtering
require(sysid) #for ARMA-X modeling
require(sandwich) #regression errors
require(stargazer) #nice reg tables
require(tidytext) #text mining
require(textstem) #lemmatization
require(quanteda) #tokenization
require(texreg) #arima tables
require(future.apply) #parallel computation (speed)
require(aTSA) #adf test

getwd()
#setwd("../") -> set wd at base repo folder

#load helper functions
source(here("helperfunctions/data_loaders.R"))
source(here("helperfunctions/date_selector.R"))
source(here("helperfunctions/plotters.R"))
source(here("helperfunctions/quick_arma.R"))
source(here("helperfunctions/r.vol_calculators.R"))
source(here("helperfunctions/truths_cleaning_function.R"))
source(here("helperfunctions/arimax_functions.R"))
```

## Load Data

Use the full\_data script to load our final dataset. Then select the timeframe.

```
#load final dataset
source(here("helperfunctions/full_data.R"))

#select timeframe
```

```
data = filter(data,between(timestamp, as.Date('2014-01-01'), as.Date('2025-05-07')))  
  
#first term  
#data = filter(data,between(timestamp, as.Date('2017-01-20'), as.Date('2021-01-20')))  
  
#second term  
#data = filter(data,between(timestamp, as.Date('2025-01-20'), as.Date('2025-05-07')))
```

## Stationarity

Having a p.value of 0.01 for our main variables seems to indicate stationarity.

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

	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

\*\*\* $p < 0.001$ ; \*\* $p < 0.01$ ; \* $p < 0.05$

Table 1: ARMAX Model Results

## Univariate ARMA-X Models

### Custom ARMA-X Specification

This first function allows to run a certain specification of an ARMA-X model. We needed to create this function due to the creation of the lags for the exogenous regressor.

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

### Best ARMA-X Given r

This second function uses the base auto.arima function in order to look for the best (lowest AIC) specification given a certain number of x lags. Once again it was necessary to create a custom function for this due to the creation of the lags for the exogenous regressor.

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

	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*** (0.0040)
tariff_lag_0	0.0042** (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
*** $p < 0.001$ ; ** $p < 0.01$ ; * $p < 0.05$	

Table 2: 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** (0.0014)
AIC	-46020.9547
AICc	-46020.9415
BIC	-45934.0340
Log Likelihood	23021.4774
Num. obs.	19968

\*\*\* $p < 0.001$ ; \*\* $p < 0.01$ ; \* $p < 0.05$

Table 3: ARMAX selected by AIC

## Best ARMA-X

This third function is the most sophisticated. We've had to essentially custom code the `auto.arima` function in order to loop for  $p$ ,  $q$  and  $r$ . It also provides many useful details such as a plot of the AIC value for the different number of lags. Note that the output of this function is a list, so in order get the model output, it is needed to specifically call it (see how the call works for this third model in the IRF section). The `latex` option in all functions determines whether the output is console-friendly, or output-friendly.

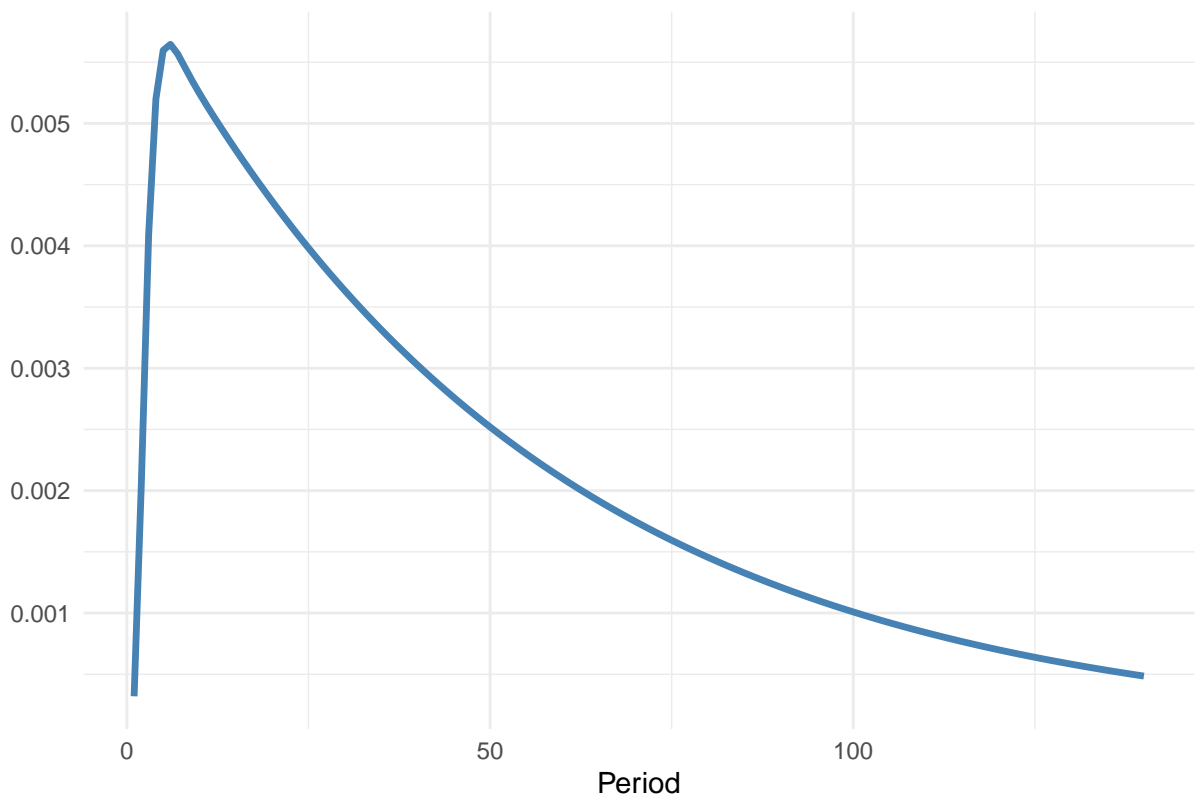
```
#auto.armax.r selects the lowest AIC checking all 3 p,q,r values
res3 = auto.armax.r(data$SPY_vol, x=data$tariff,
                    max_p = 3, max_q = 3, max_r = 3, criterion = "AIC", latex=T)
```

## IRF

Finally, we have our personal twist on the IRF plot function from the AEC package. It enables automatic extraction of the  $\phi$ 's and  $\beta$ 's of a fitted ARMA-X model. It then uses the provided `sim.arma` function where it inputs the  $\beta$ 's as  $\theta$ 's. Finally, it automatically plots it, using a fancy `ggplot`.

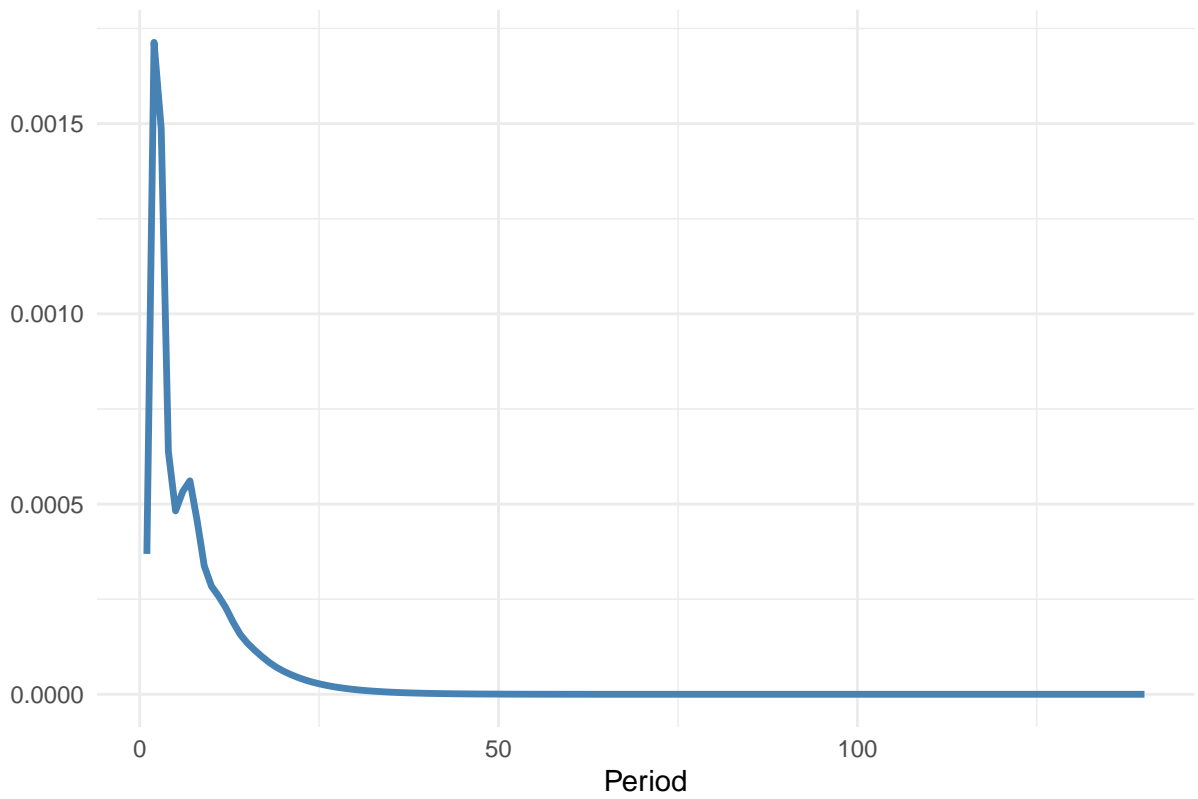
```
#we want to plot the IRFs of these models
nb.periods = 7 * 20
irf.plot(res1,nb.periods)
```

ARMA-X IRF



```
irf.plot(res2,nb.periods)
```

ARMA-X IRF



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



ARMA-X IRF

