ASHR SVAR Models

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Setup

Load packages & 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(sandwhich) #regression errors
require(stargazer) #nice req tables
require(tidytext) #text mining
require(textstem) #lemmatization
require(quanteda) #tokenization
require(texreg) #arima tables
require(vars) #VAR models
require(xts) #time series objects
require(tseries) #includes adf test
require(quantmod)
require(TSA)
require(aTSA)
require(tibble)
require(FinTS)
require(kableExtra)
require(writexl)
require(purrr)
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/armax_functions.R"))
source(here("helperfunctions/var_irf.R"))
```

Load Data

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

#select timeframe
Vdata = filter(data,between(timestamp, as.Date('2014-01-01'), as.Date('2025-05-07')))
```

Some SVAR estimations

Note that this is not an exhaustive list of our VAR estimations, you can find more by going on /modeling/VAR/VAR_SPY_TRUE or VAR_ASHR_TRUE or VAR_VGK_TRUE).

Dummy variable

Here we use a dummy variable which equal to one if Trump has made a post or 0 otherwise, taking into account the closed hour market posts.

```
y = cbind(Vdata$dummy, Vdata$ASHR_vol)
colnames(y)[1:2] <- c("dummy", "vol")
est.VAR <- VAR(y,p=6)
mod_vol <- est.VAR$varresult$vol
texreg(mod_vol, digits = 6)</pre>
```

```
Omega <- var(residuals(est.VAR))

#make the B matrix
loss <- function(param){
    #Define the restriction
    B <- matrix(c(param[1], param[2], 0, param[3]), ncol = 2)

    #Make BB' approximatively equal to omega
    X <- Omega - B %*% t(B)

    #loss function
    loss <- sum(X^2)
    return(loss)
}

res.opt <- optim(c(1, 0, 1), loss, method = "BFGS")
B.hat <- matrix(c(res.opt$par[1], res.opt$par[2], 0, res.opt$par[3]), ncol = 2)

print(cbind(Omega,B.hat %*% t(B.hat)))</pre>
```

	Model 1
dummy.l1	-0.000006***
J	(0.000001)
vol.l1	0.282482***
	(0.007140)
dummy.l2	-0.000005^{***}
v	(0.000001)
vol.l2	0.072926***
	(0.007410)
dummy.l3	-0.000006^{***}
v	(0.000001)
vol.l3	0.047892***
	(0.007418)
dummy.l4	-0.000004^{***}
· ·	(0.000001)
vol.l4	0.056084***
	(0.007416)
dummy.l5	-0.000006^{***}
	(0.000001)
vol.l5	0.059763***
	(0.007410)
dummy.l6	-0.000005***
	(0.000001)
vol.l6	0.109466^{***}
	(0.007136)
const	0.000095***
	(0.000005)
\mathbb{R}^2	0.178208
$Adj. R^2$	0.177714
Num. obs.	19965

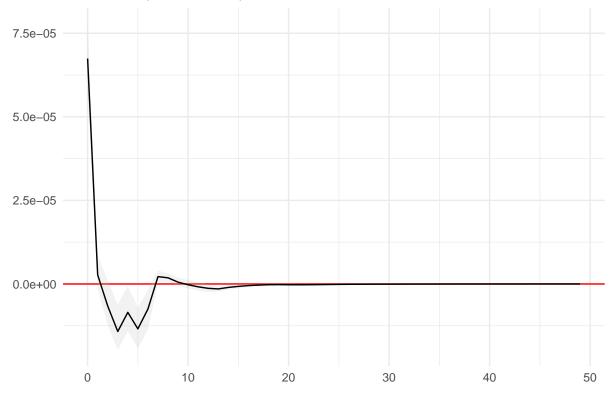
***p < 0.001; **p < 0.01; *p < 0.05

Table 1: Statistical models

```
#the plot
single_varirf <- extract_varirf(irf_res)

#the plot
single_varirf %>%
    ggplot(aes(x=period, y=irf_dummy_vol, ymin=lower_dummy_vol, ymax=upper_dummy_vol)) +
    geom_hline(yintercept = 0, color="red") +
    geom_ribbon(fill="grey", alpha=0.2) +
    geom_line() +
    theme_light() +
    ggtitle("IRF Dummy on Volatility")+
    ylab("") +
    xlab("") +
    theme_minimal()
```

IRF Dummy on Volatility



```
#does volatility Granger cause dummy mentions
grangertest(y[,c("vol","dummy")], order = 6)
```

Res.Df	\mathbf{Df}	${f F}$	$\Pr(>F)$
2e + 04			
2e + 04	-6	4.06	0.000446

```
#does dummy mentions Granger cause volatility
grangertest(y[,c("dummy", "vol")], order = 6)
```

Res.Df	Df	F	Pr(>F)
2e+04			
2e+04	-6	22.2	3.64e-26

Number of Post

```
y2 = cbind(Vdata$N , Vdata$ASHR_vol)
colnames(y2)[1:2] <- c("N", "vol")
est.VAR2 <- VAR(y2,p=6)
mod_vol2 <- est.VAR2$varresult$vol
texreg(mod_vol2, digits = 6)</pre>
```

```
Omega2 <- var(residuals(est.VAR2))

#make the B matrix
loss2 <- function(param2){
    #Define the restriction
    B2 <- matrix(c(param2[1], param2[2], 0, param2[3]), ncol = 2)

    #Make BB' approximatively equal to omega
    X2 <- Omega2 - B2 %*% t(B2)

    #loss function
    loss2 <- sum(X2^2)
    return(loss2)
}

res.opt2 <- optim(c(1, 0, 1), loss2, method = "BFGS")
B.hat2 <- matrix(c(res.opt2$par[1], res.opt2$par[2], 0, res.opt2$par[3]), ncol = 2)

print(cbind(Omega2,B.hat2 %*% t(B.hat2)))</pre>
```

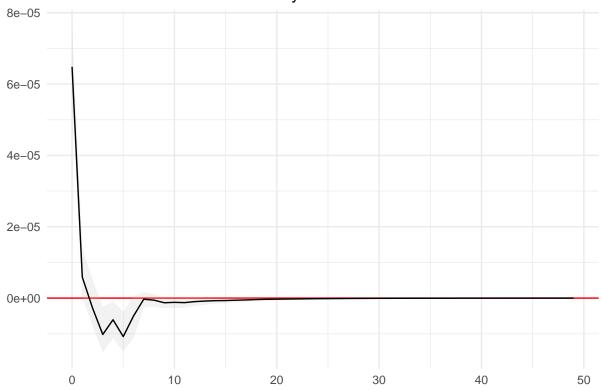
	Model 1
N.l1	-0.000001^{***}
	(0.000000)
vol.l1	0.282497^{***}
	(0.007132)
N.12	-0.000001^{***}
	(0.000000)
vol.l2	0.072640^{***}
	(0.007403)
N.13	-0.000002^{***}
	(0.000000)
vol.l3	0.047738^{***}
	(0.007409)
N.14	-0.000001^{**}
	(0.000000)
vol.l4	0.056237^{***}
	(0.007408)
N.l5	-0.000002^{***}
	(0.000000)
vol.l5	0.059528***
	(0.007403)
N.16	-0.000001^{***}
	(0.000000)
vol.l6	0.109380^{***}
	(0.007128)
const	0.000081^{***}
	(0.000004)
\mathbb{R}^2	0.176168
$Adj. R^2$	0.175673
Num. obs.	19965

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

Table 2: Statistical models

```
geom_ribbon(fill="grey", alpha=0.2) +
geom_line() +
theme_light() +
ggtitle("IRF Number of Posts on Volatility")+
ylab("")+
xlab("") +
theme_minimal()
```

IRF Number of Posts on Volatility



```
#does volatility Granger cause N mentions
grangertest(y2[,c("vol","N")], order = 6)
```

Res.Df	\mathbf{Df}	\mathbf{F}	$\Pr(>F)$
2e+04			
2e+04	-6	3.17	0.00413

```
#does N mentions Granger cause volatility
grangertest(y2[,c("N", "vol")], order = 6)
```

Res.Df	Df	\mathbf{F}	$\Pr(>F)$
2e+04			
2e+04	-6	13.9	8.02e-16

Tariff

```
y3 = cbind(Vdata$tariff , Vdata$ASHR_vol)
colnames(y3)[1:2] <- c("tariff", "vol")
est.VAR3 <- VAR(y3,p=6)
mod_vol3 <- est.VAR3$varresult$vol
texreg(mod_vol3, digits = 6)</pre>
```

```
Omega3 <- var(residuals(est.VAR3))

#make the B matrix
loss3 <- function(param3){
    #Define the restriction
    B3 <- matrix(c(param3[1], param3[2], 0, param3[3]), ncol = 2)

    #Make BB' approximatively equal to omega
    X3 <- Omega3 - B3 %*% t(B3)

    #loss function
    loss3 <- sum(X3^2)
    return(loss3)
}

res.opt3 <- optim(c(1, 0, 1), loss3, method = "BFGS")
B.hat3 <- matrix(c(res.opt3$par[1], res.opt3$par[2], 0, res.opt3$par[3]), ncol = 2)

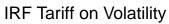
print(cbind(Omega3,B.hat3 %*% t(B.hat3)))</pre>
```

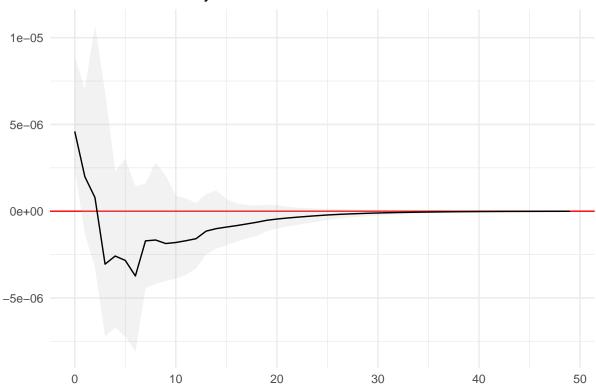
	Model 1
tariff.l1	0.000002
	(0.000008)
vol.l1	0.280773***
	(0.007038)
tariff.l2	-0.000000
	(0.000008)
vol.l2	0.072461^{***}
	(0.007302)
tariff.l3	-0.000010
	(0.000008)
vol.l3	0.045252^{***}
	(0.007309)
tariff.l4	-0.000005
	(0.000008)
vol.l4	0.056243^{***}
	(0.007309)
tariff.l5	-0.000005
	(0.000008)
vol.l5	0.056910^{***}
	(0.007302)
tariff.l6	-0.000008
	(0.000008)
vol.l6	0.109191^{***}
	(0.007037)
const	0.000061^{***}
	(0.000003)
\mathbb{R}^2	0.172906
$Adj. R^2$	0.172409
Num. obs.	19965

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

Table 3: Statistical models

```
xlab("") +
theme_minimal()
```





#does volatility Granger cause tariff mentions
grangertest(y3[,c("vol","tariff")], order = 6)

Res.Df	Df	\mathbf{F}	$\Pr(>F)$
2e+04			
2e+04	-6	0.722	0.632

#does tariff mentions Granger cause volatility
grangertest(y3[,c("tariff", "vol")], order = 6)

Res.Df	Df	F	Pr(>F)
2e+04			
2e+04	-6	0.715	0.638

	Model 1
trade.l1	-0.000025^*
	(0.000010)
vol.l1	0.281371^{***}
	(0.007039)
trade.l2	0.000012
	(0.000010)
vol.l2	0.071544^{***}
	(0.007306)
trade.l3	-0.000019
	(0.000010)
vol.l3	0.045746^{***}
	(0.007312)
trade.l4	-0.000009
	(0.000010)
vol.l4	0.056738^{***}
	(0.007316)
trade.l5	-0.000013
	(0.000010)
vol.l5	0.056226^{***}
	(0.007311)
trade.l6	-0.000013
	(0.000010)
vol.l6	0.109845^{***}
	(0.007043)
const	0.000062^{***}
	(0.000003)
\mathbb{R}^2	0.173422
$Adj. R^2$	0.172925
Num. obs.	19965
*** p < 0.001: ** p	p < 0.01: * $p < 0.05$

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

Table 4: Statistical models

Trade

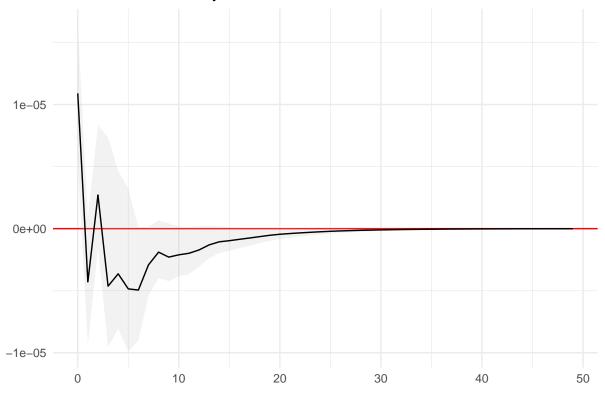
```
y4 = cbind(Vdata$trade , Vdata$ASHR_vol)
colnames(y4)[1:2] <- c("trade", "vol")
est.VAR4 <- VAR(y4,p=6)
mod_vol4 <- est.VAR4$varresult$vol
texreg(mod_vol4, digits = 6)</pre>
```

```
Omega4 <- var(residuals(est.VAR4))

#make the B matrix
loss4 <- function(param4){
    #Define the restriction
    B4 <- matrix(c(param4[1], param4[2], 0, param4[3]), ncol = 2)

#Make BB' approximatively equal to omega
X4 <- Omega4 - B4 %*% t(B4)</pre>
```

IRF Trade on Volatility



#does volatility Granger cause trade mentions
grangertest(y4[,c("vol","trade")], order = 6)

Res.Df	\mathbf{Df}	\mathbf{F}	$\Pr(>F)$
2e+04			
2e+04	-6	5.36	1.55e-05

#does trade mentions Granger cause volatility
grangertest(y4[,c("trade", "vol")], order = 6)

Res.Df	\mathbf{Df}	\mathbf{F}	$\Pr(>F)$
2e+04			
2e+04	-6	2.79	0.0103

China

	Model 1
china.l1	-0.000005
	(0.000006)
vol.l1	0.280637^{***}
	(0.007040)
china.l2	-0.000004
	(0.000006)
vol.l2	0.072261^{***}
	(0.007305)
china.l3	-0.000011
	(0.000006)
vol.l3	0.045298^{***}
	(0.007312)
china.l4	-0.000007
	(0.000006)
vol.l4	0.056264^{***}
	(0.007312)
china.l5	-0.000007
	(0.000006)
vol.l5	0.056857^{***}
	(0.007305)
china.l6	-0.000010
	(0.000006)
vol.l6	0.109272^{***}
	(0.007040)
const	0.000063^{***}
	(0.000003)
\mathbb{R}^2	0.173206
$Adj. R^2$	0.172709
Num. obs.	19965
**** n < 0.001. *** n	< 0.01· *n < 0.05

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

Table 5: Statistical models

```
y5 = cbind(Vdata$china , Vdata$ASHR_vol)
colnames(y5)[1:2] <- c("china", "vol")
est.VAR5 <- VAR(y5,p=6)
mod_vol5 <- est.VAR5$varresult$vol
texreg(mod_vol5, digits = 6)</pre>
```

```
Omega5 <- var(residuals(est.VAR5))

#make the B matrix
loss5 <- function(param5){
    #Define the restriction
    B5 <- matrix(c(param5[1], param5[2], 0, param5[3]), ncol = 2)

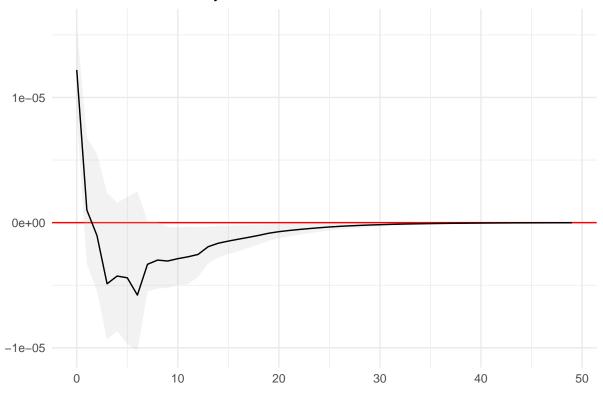
#Make BB' approximatively equal to omega
X5 <- Omega5 - B5 %*% t(B5)

#loss function</pre>
```

```
loss5 <- sum(X5^2)
  return(loss5)
}
res.opt5 <- optim(c(1, 0, 1), loss5, method = "BFGS")
B.hat5 \leftarrow matrix(c(res.opt5$par[1], res.opt5$par[2], 0, res.opt5$par[3]), ncol = 2)
print(cbind(Omega5,B.hat5 %*% t(B.hat5)))
#irf creation
irf_res5 <- irf(est.VAR5, impulse = "china", response = "vol",</pre>
                  bmat=b.hat5, n.ahead = 7 * 7, boot = TRUE, ci = 0.95)
#function to extract relevant objects for plotting
single_varirf5 <- extract_varirf(irf_res5)</pre>
#the plot
single_varirf5 %>%
  ggplot(aes(x=period, y=irf_china_vol, ymin=lower_china_vol, ymax=upper_china_vol)) +
  geom_hline(yintercept = 0, color="red") +
  geom_ribbon(fill="grey", alpha=0.2) +
  geom_line() +
  theme_light() +
  ggtitle("IRF China on Volatility")+
  ylab("")+
```

xlab("") +
theme_minimal()

IRF China on Volatility



#does volatility Granger cause china mentions
grangertest(y5[,c("vol","china")], order = 6)

Res.Df	\mathbf{Df}	${f F}$	$\Pr(>F)$
2e+04			
2e+04	-6	1.08	0.372

#does china mentions Granger cause volatility
grangertest(y5[,c("china", "vol")], order = 6)

Res.Df	\mathbf{Df}	\mathbf{F}	$\Pr(>F)$
2e+04			
2e+04	-6	1.92	0.0733

Interaction number of post and tariff

here is an example of our interaction

	Model 1
interaction.l1	-0.000000
interaction.11	
1.14	(0.000000)
vol.l1	0.280904***
	(0.007037)
interaction.l2	-0.000000
	(0.000000)
vol.l2	0.072496^{***}
	(0.007302)
interaction.l3	-0.000000
	(0.000000)
vol.l3	0.045218^{***}
	(0.007309)
interaction.l4	-0.000000
	(0.000000)
vol.l4	0.056276***
	(0.007309)
interaction.l5	-0.000000
	(0.000000)
vol.l5	0.056856***
	(0.007302)
interaction.l6	-0.000000
	(0.000000)
vol.l6	0.109078***
	(0.007037)
const	0.000060***
	(0.000003)
\mathbb{R}^2	0.172766
$Adj. R^2$	0.172269
Num. obs.	19965
*** .0.001 **	0.01 * .0.05

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

Table 6: Statistical models

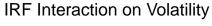
```
#interaction
##N and tariff, 2 variables

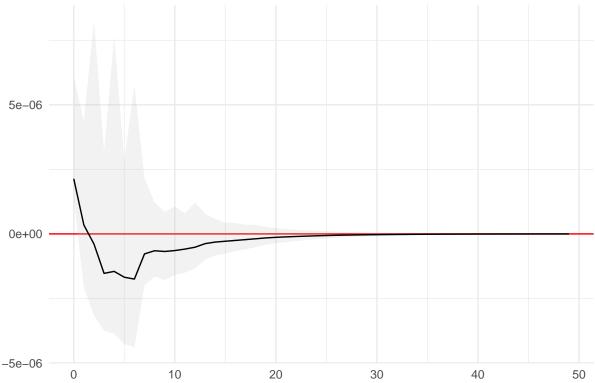
int1 = Vdata$tariff * Vdata$N

y12 = cbind(int1, Vdata$ASHR_vol)
colnames(y12)[1:2] <- c("interaction", "vol")
est.VAR12 <- VAR(y12,p=6)
mod_vol12 <- est.VAR12$varresult$vol
texreg(mod_vol12, digits = 6)</pre>
```

```
Omega12 <- var(residuals(est.VAR12))
#make the B matrix
loss12 <- function(param12){
    #Define the restriction
    B12 <- matrix(c(param12[1], param12[2], 0, param12[3]), ncol = 2)</pre>
```

```
#Make BB' approximatively equal to omega
  X12 <- Omega12 - B12 %*% t(B12)</pre>
  #loss function
  loss12 <- sum(X12^2)
  return(loss12)
res.opt12 <- optim(c(1, 0, 1), loss12, method = "BFGS")
B.hat12 \leftarrow matrix(c(res.opt12\$par[1], res.opt12\$par[2], 0, res.opt12\$par[3]), ncol = 2)
print(cbind(Omega12,B.hat12 %*% t(B.hat12)))
#irf creation
irf_res12 <- irf(est.VAR12, impulse = "interaction", response = "vol",</pre>
                  bmat=b.hat12, n.ahead = 7 * 7, boot = TRUE, ci = 0.95)
#function to extract relevant objects for plotting
single_varirf12 <- extract_varirf(irf_res12)</pre>
#the plot
single_varirf12 %>%
  ggplot(aes(x=period, y=irf_interaction_vol, ymin=lower_interaction_vol, ymax=upper_interaction_vol))
  geom_hline(yintercept = 0, color="red") +
  geom_ribbon(fill="grey", alpha=0.2) +
  geom_line() +
  theme_light() +
  ggtitle("IRF Interaction on Volatility")+
  ylab("")+
  xlab("") +
  theme_minimal()
```





Terms

Here we look for the first and second mandate effect of posts. We will use the tariff variable as a proxy for the posts

First mandate

```
# First and Second Mandate

#first term

Vdata_f = filter(data,between(timestamp, as.Date('2017-01-20'), as.Date('2021-01-20')))

#second term

Vdata_s = filter(data,between(timestamp, as.Date('2025-01-20'), as.Date('2025-05-07')))

y_f_d = cbind(Vdata_f$tariff, Vdata_f$ASHR_vol)

colnames(y_f_d)[1:2] <- c("tariff", "vol")

est.VAR_f_d <- VAR(y_f_d,p=6)

mod_vol_f_d <- est.VAR_f_d$varresult$vol

texreg(mod_vol_f_d, digits = 6)</pre>
```

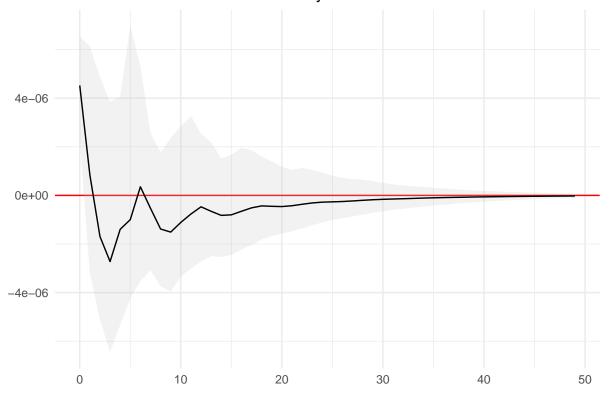
	Model 1
tariff.l1	-0.000001
	(0.000008)
vol.l1	0.245435^{***}
	(0.011754)
tariff.l2	-0.000007
	(0.000008)
vol.l2	0.075223^{***}
	(0.012070)
tariff.l3	-0.000007
	(0.000008)
vol.l3	0.060102^{***}
	(0.012082)
tariff.l4	-0.000002
	(0.000008)
vol.l4	0.060634^{***}
	(0.012082)
tariff.l5	-0.000002
	(0.000008)
vol.l5	0.085747^{***}
	(0.012070)
tariff.l6	0.000001
	(0.000008)
vol.l6	0.173709^{***}
	(0.011754)
const	0.000033^{***}
	(0.000003)
\mathbb{R}^2	0.218236
$Adj. R^2$	0.216900
Num. obs.	7036

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

Table 7: Statistical models

```
Omega_f_d <- var(residuals(est.VAR_f_d))</pre>
#make the B matrix
loss_f_d <- function(param_f_d){</pre>
  #Define the restriction
 B_f_d \leftarrow matrix(c(param_f_d[1], param_f_d[2], 0, param_f_d[3]), ncol = 2)
  #Make BB' approximatively equal to omega
 X_f_d \leftarrow Omega_f_d - B_f_d %*% t(B_f_d)
 #loss function
 loss_f_d \leftarrow sum(X_f_d^2)
 return(loss_f_d)
}
res.opt_f_d \leftarrow optim(c(1, 0, 1), loss_f_d, method = "BFGS")
B.hat_f_d \leftarrow matrix(c(res.opt_f_dpar[1], res.opt_f_dpar[2], 0, res.opt_f_dpar[3]), ncol = 2)
print(cbind(Omega_f_d,B.hat_f_d %*% t(B.hat_f_d)))
#irf creation
irf_res_f_d <- irf(est.VAR_f_d, impulse = "tariff", response = "vol",</pre>
                   bmat=b.hat_f_d, n.ahead = 7 * 7, boot = TRUE, ci = 0.95)
#function to extract relevant objects for plotting
single_varirf_f_d <- extract_varirf(irf_res_f_d)</pre>
#the plot
single_varirf_f_d %>%
  ggplot(aes(x=period, y=irf_tariff_vol, ymin=lower_tariff_vol, ymax=upper_tariff_vol)) +
  geom_hline(yintercept = 0, color="red") +
  geom_ribbon(fill="grey", alpha=0.2) +
 geom_line() +
 theme_light() +
  ggtitle("IRF First Mandate tariff on Volatility")+
 vlab("")+
 xlab("") +
 theme_minimal()
```

IRF First Mandate tariff on Volatility



#does vol granger cause tariff
grangertest(y_f_d[,c("vol","tariff")], order = 6)

Res.Df	Df	\mathbf{F}	$\Pr(>F)$
7.02e+03			
7.03e+03	-6	0.384	0.889

#does tariff granger cause vol
grangertest(y_f_d[,c("tariff", "vol")], order = 6)

Res.Df	Df	\mathbf{F}	$\Pr(>F)$
7.02e+03			
7.03e+03	-6	0.346	0.912

second mandate

	N. 1.1.1
	Model 1
tariff.l1	0.000008
	(0.000005)
vol.l1	0.443019^{***}
	(0.044731)
tariff.l2	0.000006
	(0.000005)
vol.l2	0.075886
	(0.049172)
tariff.l3	-0.000005
	(0.000005)
vol.l3	-0.062004
	(0.049326)
tariff.l4	-0.000001
	(0.000005)
vol.l4	0.075354
	(0.049365)
tariff.l5	-0.000001
	(0.000005)
vol.l5	-0.003760
	(0.049339)
tariff.l6	-0.000006
	(0.000005)
vol.16	0.123405**
	(0.044686)
const	0.000033**
	(0.000011)
\mathbb{R}^2	0.287673
$Adj. R^2$	0.270543
Num. obs.	512
	.0.01 * .0.05

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

Table 8: Statistical models

```
y_s_d = cbind(Vdata_s$tariff, Vdata_s$ASHR_vol)
colnames(y_s_d)[1:2] <- c("tariff", "vol")
est.VAR_s_d <- VAR(y_s_d,p=6)
mod_vol_s_d <- est.VAR_s_d$varresult$vol
texreg(mod_vol_s_d, digits = 6)</pre>
```

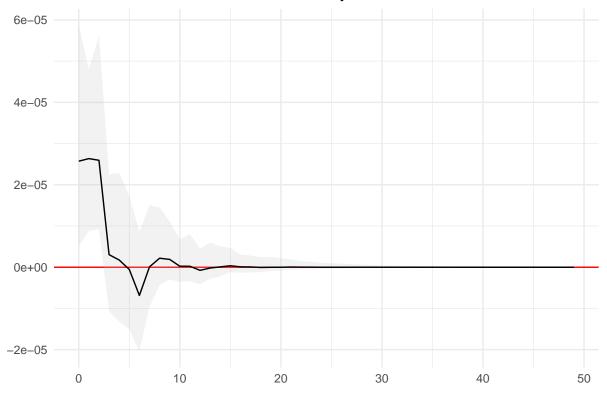
```
Omega_s_d <- var(residuals(est.VAR_s_d))
#make the B matrix
loss_s_d <- function(param_s_d){
    #Define the restriction
B_s_d <- matrix(c(param_s_d[1], param_s_d[2], 0, param_s_d[3]), ncol = 2)

#Make BB' approximatively equal to omega
X_s_d <- Omega_s_d - B_s_d %*% t(B_s_d)

#loss function
loss_s_d <- sum(X_s_d^2)</pre>
```

```
return(loss_s_d)
}
res.opt_s_d \leftarrow optim(c(1, 0, 1), loss_s_d, method = "BFGS")
B.hat_s_d \leftarrow matrix(c(res.opt_s_dpar[1], res.opt_s_dpar[2], 0, res.opt_s_dpar[3]), ncol = 2)
print(cbind(Omega_s_d,B.hat_s_d %*% t(B.hat_s_d)))
#irf creation
irf_res_s_d <- irf(est.VAR_s_d, impulse = "tariff", response = "vol",</pre>
                  bmat=b.hat_s_d, n.ahead = 7 * 7, boot = TRUE, ci = 0.95)
#function to extract relevant objects for plotting
single_varirf_s_d <- extract_varirf(irf_res_s_d)</pre>
#the plot
single_varirf_s_d %>%
  ggplot(aes(x=period, y=irf_tariff_vol, ymin=lower_tariff_vol, ymax=upper_tariff_vol)) +
  geom_hline(yintercept = 0, color="red") +
  geom_ribbon(fill="grey", alpha=0.2) +
  geom_line() +
  theme light() +
  ggtitle("IRF Second Mandate tariff on Volatility")+
  ylab("")+
  xlab("") +
  theme_minimal()
```

IRF Second Mandate tariff on Volatility



#does vol granger cause tariff
grangertest(y_s_d[,c("vol","tariff")], order = 6)

Res.Df	Df	${f F}$	$\Pr(>F)$
499			
505	-6	1.49	0.179

#does tariff granger cause vol
grangertest(y_s_d[,c("tariff", "vol")], order = 6)

Res.Df	\mathbf{Df}	\mathbf{F}	$\Pr(>F)$
499			
505	-6	1.18	0.313