

Econometria IV - Trabalho Final

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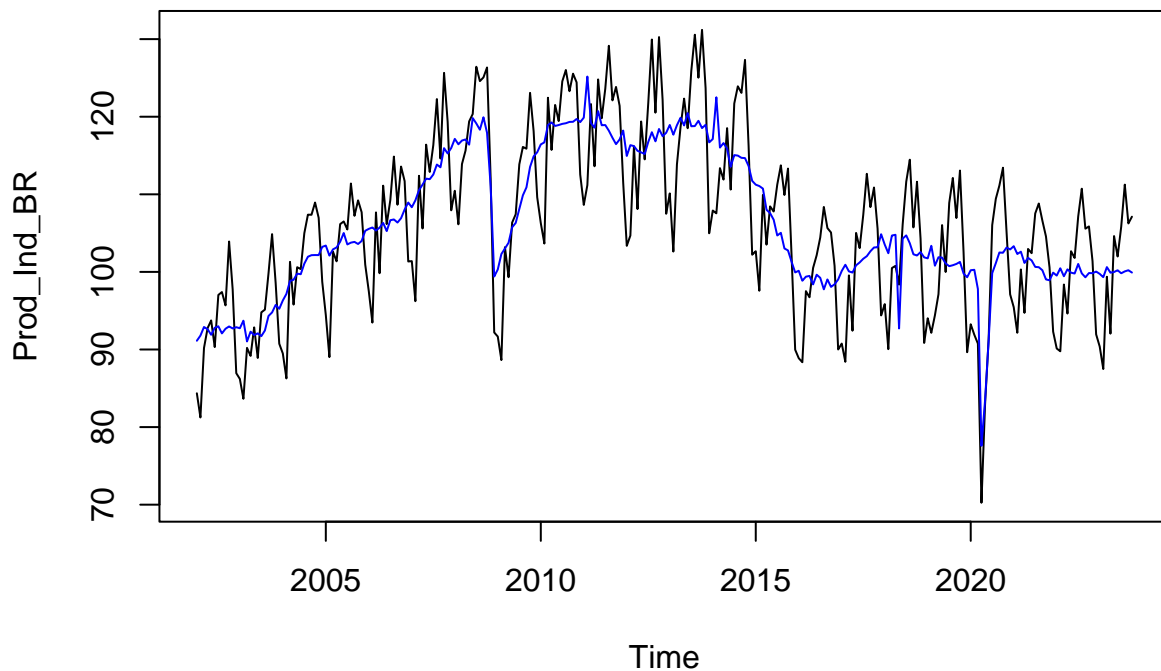
Carregando o ambiente necessário para análise:

Carregando a base de dados a ser utilizada:

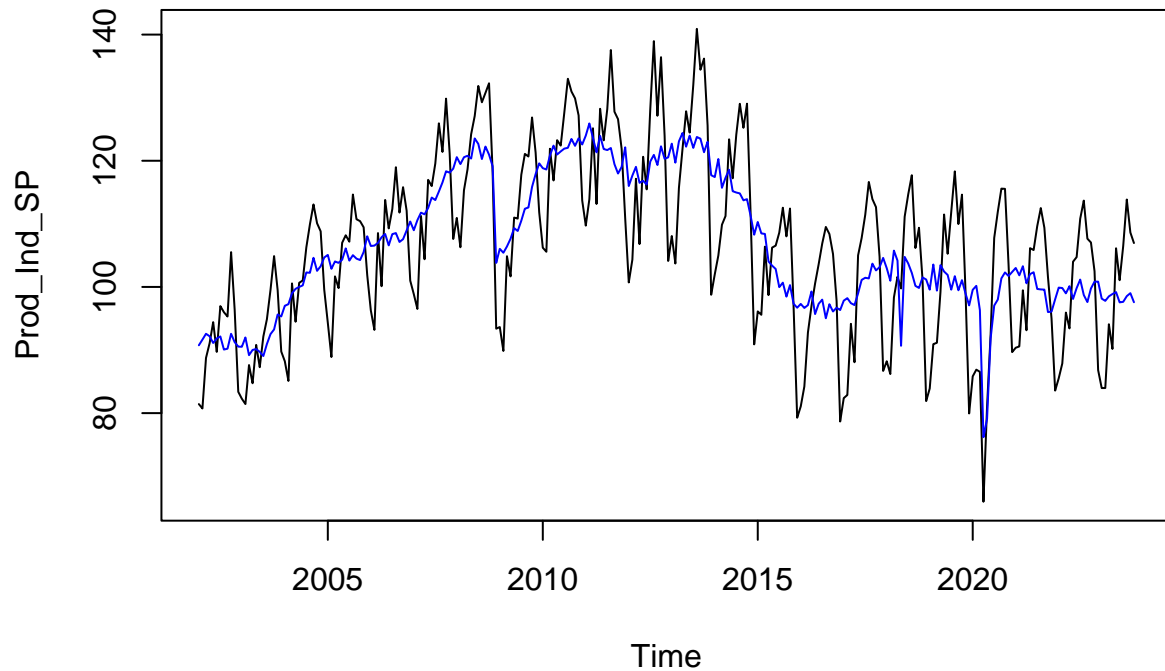
```
setwd("D:/")  
base_SVAR <- read_excel("D:/Base_Trabalho ECIV.xlsx")  
base_SVAR_ts <- ts(base_SVAR[, -1], start = c(2002, 1), frequency = 12)
```

Tratando a base de dados a ser utilizada: 1. dessazonalizando as séries de produção industrial 2. aplicando logaritmo natural em todas colunas menos selic-over 3. tomando a primeira diferença para estacionarizar as séries

```
#Dessazonalizando as séries de produção industrial  
dessaz <- seas(base_SVAR_ts[, 1:14])  
base_SVAR_ts_dessaz <- final(dessaz)  
ts.plot(base_SVAR_ts[, 1], base_SVAR_ts_dessaz[, 1], gpars = list(ylab = "Prod_Ind_BR", col = c("black", "blue")))
```



```
ts.plot(base_SVAR_ts[,14],base_SVAR_ts_dessaz [,14], gpars = list(ylab = "Prod_Ind_SP", col = c ("black"
```

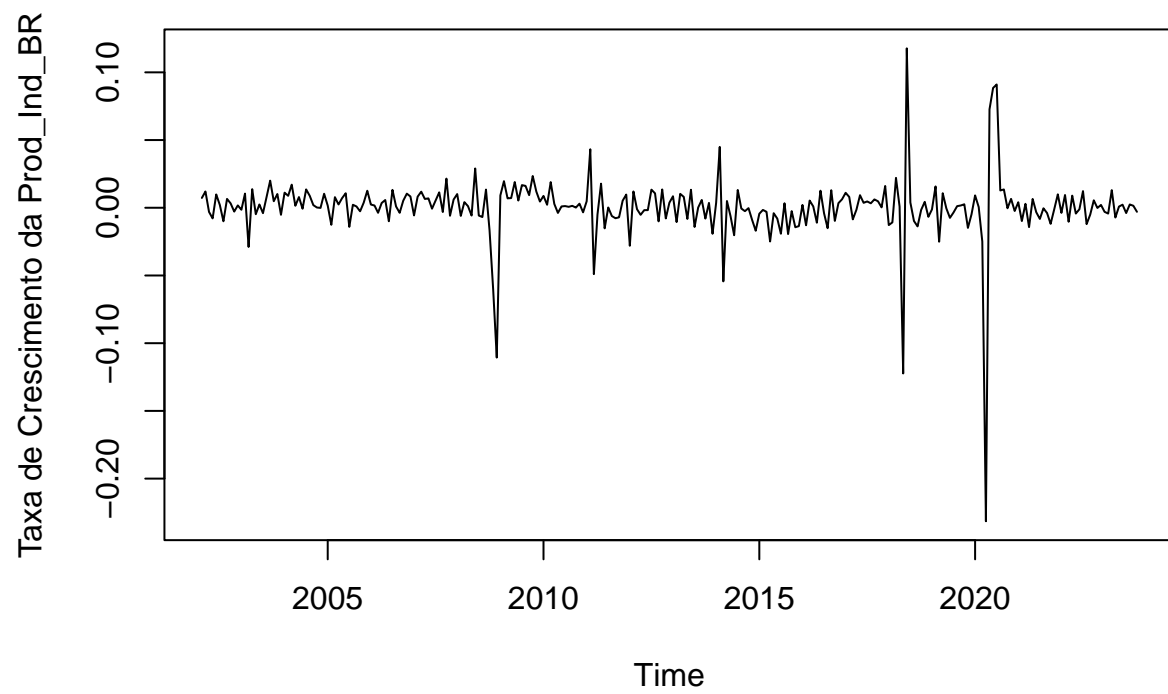


```
#Juntando os dados de IPCA
base_SVAR_ts_dessaz <- cbind(base_SVAR_ts_dessaz, base_SVAR_ts[,15])

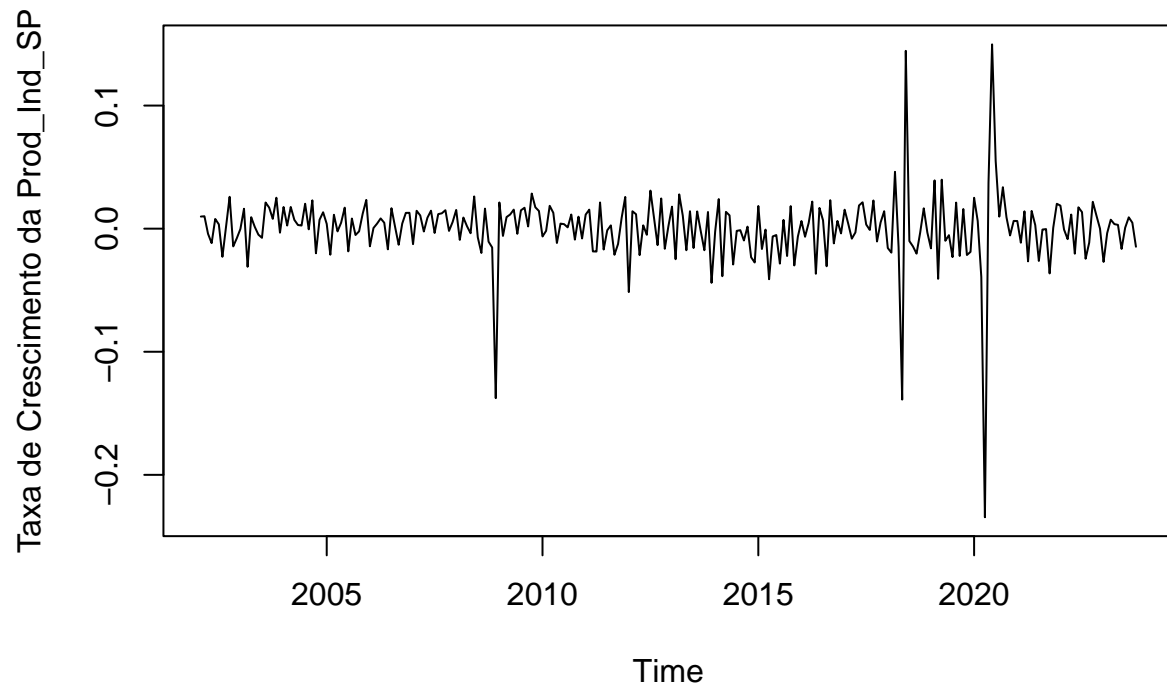
#aplicando logaritmo em todos os dados (com exceção da taxa de juros selic-over)
base_SVAR_ts_dessaz_ln <- log(base_SVAR_ts_dessaz)

#Tomando a primeira diferença do log dos dados (com exceção da taxa de juros selic-over)
base_SVAR_ts_dessaz_ln_diff <- diff(base_SVAR_ts_dessaz_ln)

ts.plot(base_SVAR_ts_dessaz_ln_diff[,1],gpars = list(ylab = "Taxa de Crescimento da Prod_Ind_BR", col =
```



```
ts.plot(base_SVAR_ts_dessaz_ln_diff[,14],gpars = list(ylab = "Taxa de Crescimento da Prod_Ind_SP", col = "blue", lty = 1))
```



```
#dif_selic <- diff(base_SVAR_ts[,16])

selic_over <- base_SVAR_ts[-1,16]

#base de dados final após tratamentos
base_SVAR_final <- cbind(base_SVAR_ts_dessaz_ln_diff, selic_over)
colnames(base_SVAR_final) <- cbind ("g_BR", "g_AM", "g_BA", "g_CE", "g_ES", "g_GO", "g_MG", "g_PA", "g_PE", "g_
#ts.plot(base_SVAR_final[,15],gpars = list(ylab = "Taxa de Juros Selic Over", col = c("black")))
```

Verificando a estacionariedade das séries

```
adf.test(base_SVAR_final[,1], alternative = 'stationary')
```

```
##
## Augmented Dickey-Fuller Test
##
## data: base_SVAR_final[, 1]
## Dickey-Fuller = -7.1136, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,2], alternative = 'stationary')
```

```
##
## Augmented Dickey-Fuller Test
##
## data: base_SVAR_final[, 2]
## Dickey-Fuller = -9.2582, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,3], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: base_SVAR_final[, 3]  
## Dickey-Fuller = -9.0257, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,4], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: base_SVAR_final[, 4]  
## Dickey-Fuller = -8.5445, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,5], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: base_SVAR_final[, 5]  
## Dickey-Fuller = -6.9899, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,6], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: base_SVAR_final[, 6]  
## Dickey-Fuller = -9.6125, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,7], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: base_SVAR_final[, 7]  
## Dickey-Fuller = -7.1026, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,8], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: base_SVAR_final[, 8]  
## Dickey-Fuller = -9.0733, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[,9], alternative = 'stationary')
```

```
##  
## Augmented Dickey-Fuller Test
```

```

##
## data:  base_SVAR_final[, 9]
## Dickey-Fuller = -9.2832, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
adf.test(base_SVAR_final[,10], alternative = 'stationary')

##
## Augmented Dickey-Fuller Test
##
## data:  base_SVAR_final[, 10]
## Dickey-Fuller = -7.493, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
adf.test(base_SVAR_final[,11], alternative = 'stationary')

##
## Augmented Dickey-Fuller Test
##
## data:  base_SVAR_final[, 11]
## Dickey-Fuller = -7.7711, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
adf.test(base_SVAR_final[,12], alternative = 'stationary')

##
## Augmented Dickey-Fuller Test
##
## data:  base_SVAR_final[, 12]
## Dickey-Fuller = -7.2981, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
adf.test(base_SVAR_final[,13], alternative = 'stationary')

##
## Augmented Dickey-Fuller Test
##
## data:  base_SVAR_final[, 13]
## Dickey-Fuller = -7.7467, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
adf.test(base_SVAR_final[,14], alternative = 'stationary')

##
## Augmented Dickey-Fuller Test
##
## data:  base_SVAR_final[, 14]
## Dickey-Fuller = -6.6528, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
adf.test(base_SVAR_final[,15], alternative = 'stationary')

##
## Augmented Dickey-Fuller Test
##
## data:  base_SVAR_final[, 15]
## Dickey-Fuller = -5.2698, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary

```

```
adf.test(base_SVAR_final[,16], alternative = 'stationary')
```

```
##
## Augmented Dickey-Fuller Test
##
## data: base_SVAR_final[, 16]
## Dickey-Fuller = -3.0612, Lag order = 6, p-value = 0.129
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[1:106,16], alternative = 'stationary')
```

```
##
## Augmented Dickey-Fuller Test
##
## data: base_SVAR_final[1:106, 16]
## Dickey-Fuller = -4.1021, Lag order = 4, p-value = 0.01
## alternative hypothesis: stationary
```

```
adf.test(base_SVAR_final[107:261,16], alternative = 'stationary')
```

```
##
## Augmented Dickey-Fuller Test
##
## data: base_SVAR_final[107:261, 16]
## Dickey-Fuller = -2.4957, Lag order = 5, p-value = 0.3698
## alternative hypothesis: stationary
```

Agora, iremos estimar o modelo VAR estrutural. Lembrando que temos que estimar um VAR para cada estado para os quais temos dados e iremos impor a restrição de Cholesky seguindo a ordem BR-Estado-IPCA-Juros das variáveis

Primeiro, criando a restrição de Cholesky

```
#Restrição - Cholesky
```

```
amat <- diag(4)
amat [2,1] <- NA
amat [3,1] <- NA
amat [4,1] <- NA
amat [3,2] <- NA
amat [4,2] <- NA
amat [4,3] <- NA
amat
```

```
##      [,1] [,2] [,3] [,4]
## [1,]    1    0    0    0
## [2,]   NA    1    0    0
## [3,]   NA   NA    1    0
## [4,]   NA   NA   NA    1
```

Agora, vamos estimar o SVAR para cada estado, considerando:

- 1) Toda amostra
- 2) Período dos Autores
- 3) Período estendido (de dezembro 2010 até outubro de 2023)

```
#####
#####AMAZONAS#####
#####
#Selecionando o Lag
```

```
lagselect_AM <- VARselect(base_SVAR_final[,c(1,2,15,16)],lag.max=10, type="both")
lagselect_AM$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4
```

```
lagselect_AM$criteria
```

```
##           1           2           3           4           5
## AIC(n) -2.920815e+01 -2.954923e+01 -2.956284e+01 -2.963024e+01 -2.961025e+01
## HQ(n)  -2.907250e+01 -2.932314e+01 -2.924631e+01 -2.922328e+01 -2.911284e+01
## SC(n)  -2.887106e+01 -2.898740e+01 -2.877629e+01 -2.861896e+01 -2.837423e+01
## FPE(n)  2.065743e-13  1.468948e-13  1.449513e-13  1.355743e-13  1.384259e-13
##           6           7           8           9          10
## AIC(n) -2.958992e+01 -2.958746e+01 -2.952443e+01 -2.948516e+01 -2.943300e+01
## HQ(n)  -2.900208e+01 -2.890919e+01 -2.875572e+01 -2.862601e+01 -2.848342e+01
## SC(n)  -2.812918e+01 -2.790199e+01 -2.761423e+01 -2.735022e+01 -2.707334e+01
## FPE(n)  1.414345e-13  1.420112e-13  1.515695e-13  1.580640e-13  1.670813e-13
```

```
#Estimando o VAR Reduzido
```

```
VAR_AM <- VAR(base_SVAR_final[-261,c(1,2,15,16)],p=5, season = NULL, exog = NULL, type = "both")
VAR_AM
```

```
##
```

```
## VAR Estimation Results:
```

```
## =====
```

```
##
```

```
## Estimated coefficients for equation g_BR:
```

```
## =====
```

```
## Call:
```

```
## g_BR = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -2.302016e-01  5.441923e-02  3.807603e-01 -1.087479e-02 -1.701960e-01
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## -2.035160e-03 -3.528290e-01  1.485380e-03  1.465316e-02 -4.343662e-02
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_AM.l4      IPCA.l4
##  3.717517e-01 -1.184162e-02  6.167391e-02 -4.554015e-02 -1.187370e+00
## Selic.Over.l4      g_BR.l5      g_AM.l5      IPCA.l5 Selic.Over.l5
##  8.674594e-03  3.520210e-02 -3.247112e-02  1.621523e-01  1.110495e-02
##      const      trend
##  8.821540e-03 -2.926016e-05
```

```
##
```

```
##
```

```
## Estimated coefficients for equation g_AM:
```

```
## =====
```

```
## Call:
```

```
## g_AM = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  9.788117e-01 -3.712107e-01  2.718921e+00 -4.560965e-03 -1.502431e-02
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## -3.053491e-01 -3.696897e+00  6.807303e-02  5.232980e-01 -3.177889e-01
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_AM.l4      IPCA.l4
##  9.055906e-01 -7.324536e-02  4.543694e-01 -2.475100e-01 -1.408222e+00
```



```

## Selic.Over.l4      g_BR.l5      g_AM.l5      IPCA.l5 Selic.Over.l5
## -3.123772e-02  1.843982e-02 -6.952064e-02  5.325946e-02  4.211988e-02
##      const      trend
## 1.000504e-02 -1.434046e-05
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -2.129862e-02  8.455681e-03  6.641692e-01 -5.713561e-04 -4.477467e-03
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## -9.530582e-04 -7.449304e-03  4.884434e-03  1.500004e-02 -2.123427e-03
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_AM.l4      IPCA.l4
## 4.242447e-02  4.762374e-04  2.375596e-03 -7.629183e-04 -3.176358e-02
## Selic.Over.l4      g_BR.l5      g_AM.l5      IPCA.l5 Selic.Over.l5
## 2.312946e-03 -1.612388e-02  3.236457e-03 -1.606127e-01 -7.252050e-03
##      const      trend
## 3.342730e-03 -5.359446e-06
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 6.192326e-03  5.859813e-02  2.346182e+00  3.273548e-01 -1.043312e-01
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## 3.717761e-02  2.351678e+00  5.669463e-01 -5.762913e-04  7.075741e-02
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_AM.l4      IPCA.l4
## 3.946351e+00  3.800501e-01  2.259619e-01 -7.534420e-02 -8.369834e-01
## Selic.Over.l4      g_BR.l5      g_AM.l5      IPCA.l5 Selic.Over.l5
## -2.605672e-01  2.310004e-01 -5.024437e-02  3.624162e+00 -8.349351e-02
##      const      trend
## 1.659985e-02 -9.385015e-05

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_AM, type="BG")
```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_AM
## Chi-squared = 104.87, df = 80, p-value = 0.03258

```

```
arch.test(VAR_AM, multivariate.only = FALSE)
```

```

## $g_BR
##
## ARCH test (univariate)
##

```

```

## data: Residual of g_BR equation
## Chi-squared = 3.7565, df = 16, p-value = 0.9993
##
##
## $g_AM
##
## ARCH test (univariate)
##
## data: Residual of g_AM equation
## Chi-squared = 9.9923, df = 16, p-value = 0.867
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 22.477, df = 16, p-value = 0.1284
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 56.207, df = 16, p-value = 2.249e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_AM
## Chi-squared = 739.05, df = 500, p-value = 1.691e-11

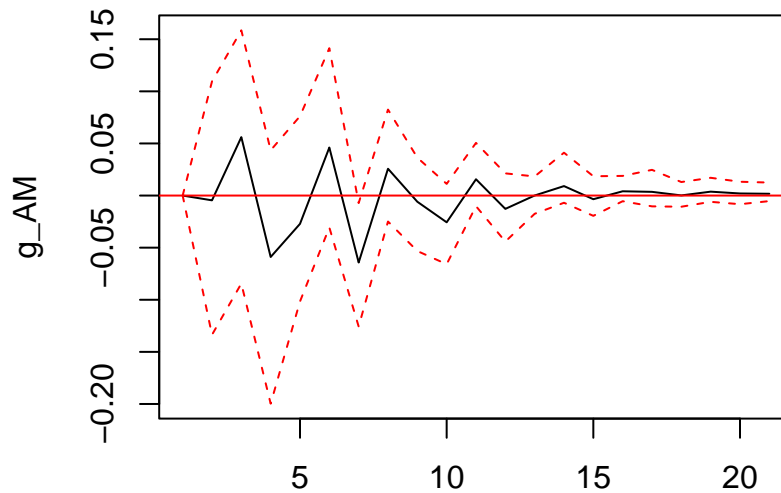
#Estimando o VAR Estrutural
SVAR_AM <- SVAR(VAR_AM, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod = "scoring")
SVAR_AM

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##      g_BR      g_AM      IPCA Selic.Over
## g_BR      1.00000 0.000000 0.00000      0
## g_AM      -0.01754 1.000000 0.00000      0
## IPCA       0.08374 0.047061 1.00000      0
## Selic.Over 0.07816 0.009562 0.09047      1

#Função Impulso-Resposta
SVAR_AM_irf <- irf(SVAR_AM, impulse = "Selic.Over", response = "g_AM", n.ahead=20, ortho = TRUE)
plot(SVAR_AM_irf)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

###Comparando Diferentes Períodos###

#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#

#Selecionando o Lag

```
lagselect_AM_pa <- VARselect(base_SVAR_final[1:106,c(1,2,15,16)],lag.max=10, type="both")
lagselect_AM_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      2      1      1      2
```

```
lagselect_AM_pa$criteria
```

```
##          1          2          3          4          5
## AIC(n) -3.046568e+01 -3.049102e+01 -3.042297e+01 -3.034911e+01 -3.028624e+01
## HQ(n)  -3.020654e+01 -3.005913e+01 -2.981832e+01 -2.957169e+01 -2.933607e+01
## SC(n)  -2.982459e+01 -2.942254e+01 -2.892710e+01 -2.842584e+01 -2.793559e+01
## FPE(n)  5.877695e-14  5.744272e-14  6.181722e-14  6.719701e-14  7.265184e-14
##          6          7          8          9         10
## AIC(n) -3.019320e+01 -3.012869e+01 -3.010217e+01 -3.020499e+01 -3.009941e+01
## HQ(n)  -2.907027e+01 -2.883300e+01 -2.863372e+01 -2.856378e+01 -2.828545e+01
## SC(n)  -2.741516e+01 -2.692326e+01 -2.646934e+01 -2.614477e+01 -2.561180e+01
## FPE(n)  8.153313e-14  8.971088e-14  9.605036e-14  9.153385e-14  1.091192e-13
```

#Estimando o VAR Reduzido

```
VAR_AM_pa <- VAR(base_SVAR_final[1:106,c(1,2,15,16)], p = 3, season = NULL, exog = NULL, type = "const")
VAR_AM_pa
```

```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.339673302 -0.013993643  0.109067204 -0.021252421 -0.151916350
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## 0.041868458 -0.054211893  0.000445711  0.059512653 -0.030594611
##      IPCA.l3 Selic.Over.l3      const
## 0.114775596  0.017394211  0.004900439
##
##
## Estimated coefficients for equation g_AM:
## =====
## Call:
## g_AM = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.9216735744 -0.5220229787  1.5657017886 -0.0685445892  0.0624508934
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## -0.1145278120 -1.6648037040  0.0557088789  0.7526137096 -0.2334383598
##      IPCA.l3 Selic.Over.l3      const
## 0.7767888598  0.0124862344 -0.0008121605
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -2.528530e-03  3.656259e-03  7.578883e-01  1.255903e-03 -2.551473e-02
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## 3.199261e-05 -6.614555e-03  1.653334e-05  2.541489e-02 -4.464235e-03
##      IPCA.l3 Selic.Over.l3      const
## -6.361579e-02 -6.213736e-04  8.757014e-04
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.46644438  0.23085253  4.31830402  0.29479373 -0.49373624
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## 0.17295466  3.82621235  0.38762727  0.40474981  0.08692257
##      IPCA.l3 Selic.Over.l3      const
## 5.78165054  0.21873890  0.03309143

```

```
#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
```

```
serial.test(VAR_AM_pa, type="BG")
```

```
##  
## Breusch-Godfrey LM test  
##  
## data: Residuals of VAR object VAR_AM_pa  
## Chi-squared = 102.78, df = 80, p-value = 0.04407
```

```
arch.test(VAR_AM_pa, multivariate.only = FALSE)
```

```
## $g_BR  
##  
## ARCH test (univariate)  
##  
## data: Residual of g_BR equation  
## Chi-squared = 15.583, df = 16, p-value = 0.4824  
##  
##  
## $g_AM  
##  
## ARCH test (univariate)  
##  
## data: Residual of g_AM equation  
## Chi-squared = 17.777, df = 16, p-value = 0.3371  
##  
##  
## $IPCA  
##  
## ARCH test (univariate)  
##  
## data: Residual of IPCA equation  
## Chi-squared = 12.494, df = 16, p-value = 0.7094  
##  
##  
## $Selic.Over  
##  
## ARCH test (univariate)  
##  
## data: Residual of Selic.Over equation  
## Chi-squared = 23.751, df = 16, p-value = 0.09509  
##  
##  
## ARCH (multivariate)  
##  
## data: Residuals of VAR object VAR_AM_pa  
## Chi-squared = 619.89, df = 500, p-value = 0.000196
```

```
#Estimando o VAR Estrutural
```

```
SVAR_AM_pa <- SVAR(VAR_AM_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")  
SVAR_AM_pa
```

```
##
```

```
## SVAR Estimation Results:
```

```
## =====
```

```
##
```

```
##
```

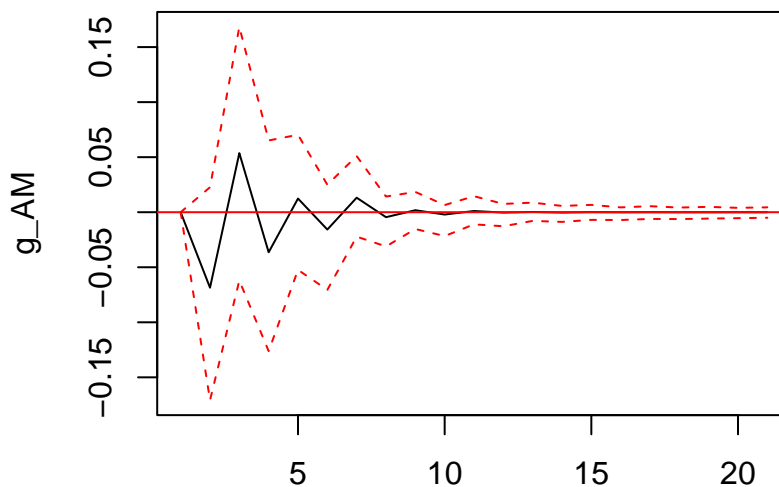
```
## Estimated A matrix:
```

```
##           g_BR    g_AM    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_AM      0.07114  1.00000  0.00000         0
## IPCA      0.09276  0.07115  1.00000         0
## Selic.Over 0.09302  0.04580  0.09197         1
```

```
#Função Impulso-Resposta
```

```
SVAR_AM_irf_pa <- irf(SVAR_AM_pa, impulse = "Selic.Over", response = "g_AM", n.ahead=20, ortho = TRUE)
plot(SVAR_AM_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#
```

```
lagselect_AM_pe <- VARselect(base_SVAR_final[107:261,c(1,2,15,16)],lag.max=10, type="both")
lagselect_AM_pe$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4
```

```
lagselect_AM_pe$criteria
```

```
##           1           2           3           4           5
## AIC(n) -2.882047e+01 -2.924502e+01 -2.928007e+01 -2.930525e+01 -2.923350e+01
## HQ(n)  -2.862027e+01 -2.891135e+01 -2.881293e+01 -2.870464e+01 -2.849943e+01
```

```

## SC(n) -2.832777e+01 -2.842385e+01 -2.813043e+01 -2.782715e+01 -2.742693e+01
## FPE(n) 3.044459e-13 1.992654e-13 1.926973e-13 1.884208e-13 2.033072e-13
##          6          7          8          9         10
## AIC(n) -2.922505e+01 -2.922856e+01 -2.909455e+01 -2.897886e+01 -2.885833e+01
## HQ(n) -2.835752e+01 -2.822756e+01 -2.796008e+01 -2.771092e+01 -2.745692e+01
## SC(n) -2.709002e+01 -2.676506e+01 -2.630258e+01 -2.585843e+01 -2.540943e+01
## FPE(n) 2.063144e-13 2.073604e-13 2.397953e-13 2.731414e-13 3.137897e-13

#Estimando o VAR Reduzido
VAR_AM_pe <- VAR(base_SVAR_final[107:261,c(1,2,15,16)], p = lagselect_AM_pe$selection[1], season = NULL)
VAR_AM_pe

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.521619409  0.127313837  0.615213994 -0.004759807 -0.397704186
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
##  0.035531735 -0.621294294  0.016347891 -0.118259826  0.002912650
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_AM.l4      IPCA.l4
##  0.388932758 -0.008539746 -0.030105535 -0.032525295 -2.039678208
## Selic.Over.l4      const
## -0.008124699  0.009443917
##
##
## Estimated coefficients for equation g_AM:
## =====
## Call:
## g_AM = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.80208594 -0.24316908  4.10736702  0.04316201 -0.17392372
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## -0.34808050 -6.18965259  0.09294132  0.27458924 -0.17242510
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_AM.l4      IPCA.l4
##  1.35461227 -0.11302819  0.22364231 -0.22114305 -3.05172280
## Selic.Over.l4      const
## -0.03663464  0.02812336
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.020822706  0.009226820  0.601672856  0.001003592  0.003053378
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
## -0.003553527 -0.017074369  0.006703345  0.019207349 -0.001074159

```

```
##      IPCA.13 Selic.Over.13      g_BR.14      g_AM.14      IPCA.14
##      0.096055577 -0.004369091  0.010387750 -0.002946113 -0.190614229
## Selic.Over.14      const
##      -0.003688553  0.002677512
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_AM.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_AM.l2 + IPCA.l2 + Selic.Over.l3
##
##      g_BR.l1      g_AM.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      -0.18199569  0.09142564  2.37300440  0.31568894 -0.43807184
##      g_AM.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_AM.l3
##      0.09831419  1.24756682  0.62216759 -0.64956980  0.22861484
##      IPCA.13 Selic.Over.13      g_BR.14      g_AM.14      IPCA.14
##      4.55001601  0.31583538 -0.33806731  0.07450517  3.14977839
## Selic.Over.14      const
##      -0.29184722 -0.02754447
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_AM_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_AM_pe
## Chi-squared = 98.612, df = 80, p-value = 0.07745
```

```
arch.test(VAR_AM_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.1644, df = 16, p-value = 1
##
##
## $g_AM
##
## ARCH test (univariate)
##
## data: Residual of g_AM equation
## Chi-squared = 3.246, df = 16, p-value = 0.9997
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 14.532, df = 16, p-value = 0.5591
##
```



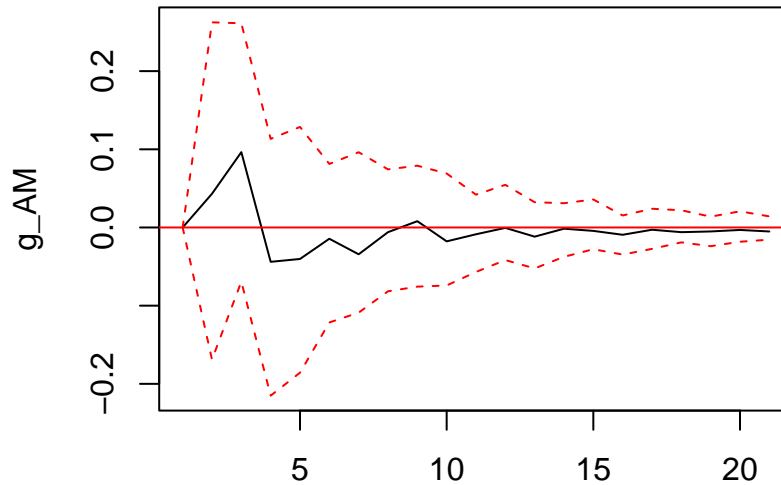
```
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 41.641, df = 16, p-value = 0.0004462
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_AM_pe
## Chi-squared = 587.2, df = 500, p-value = 0.004238

#Estimando o VAR Estrutural
SVAR_AM_pe <- SVAR(VAR_AM_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_AM_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR      g_AM  IPCA Selic.Over
## g_BR      1.00000  0.000000 0.000         0
## g_AM     -0.06967  1.000000 0.000         0
## IPCA      0.08112  0.035789 1.000         0
## Selic.Over 0.07677 -0.003225 0.092         1

#Função Impulso-Resposta
SVAR_AM_irf_pe <- irf(SVAR_AM_pe, impulse = "Selic.Over", response = "g_AM", n.ahead=20, ortho = TRUE)
plot(SVAR_AM_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####BAHIA#####
#####
#Seleccionando o Lag
lagselect_BA <- VARselect(base_SVAR_final[,c(1,3,15,16)],lag.max=10, type="both")
lagselect_BA$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_BA$criteria

##           1           2           3           4           5
## AIC(n) -2.955785e+01 -2.987436e+01 -2.987219e+01 -2.992931e+01 -2.991058e+01
## HQ(n)  -2.942219e+01 -2.964827e+01 -2.955566e+01 -2.952234e+01 -2.941318e+01
## SC(n)  -2.922075e+01 -2.931254e+01 -2.908563e+01 -2.891802e+01 -2.867457e+01
## FPE(n)  1.456148e-13  1.061212e-13  1.063836e-13  1.005301e-13  1.025140e-13
##           6           7           8           9          10
## AIC(n) -2.986459e+01 -2.979820e+01 -2.979060e+01 -2.976301e+01 -2.970408e+01
## HQ(n)  -2.927675e+01 -2.911992e+01 -2.902188e+01 -2.890386e+01 -2.875449e+01
## SC(n)  -2.840385e+01 -2.811272e+01 -2.788039e+01 -2.762807e+01 -2.734441e+01
## FPE(n)  1.074649e-13  1.150276e-13  1.161498e-13  1.197193e-13  1.274093e-13

#Estimando o VAR Reduzido
VAR_BA <- VAR(base_SVAR_final[,c(1,3,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_BA

##
```

```

## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.108914169   0.001896894   0.352250173 -0.008622137 -0.138688744
##      g_BA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_BA.l3
## -0.012097337 -0.268616192 -0.005442204 -0.088264600 -0.027036577
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_BA.l4      IPCA.l4
##  0.268507353 -0.009236810 -0.036730283 -0.001692393 -1.179068572
## Selic.Over.l4      g_BR.l5      g_BA.l5      IPCA.l5 Selic.Over.l5
##  0.008573894 -0.026740835   0.019755040   0.309651235   0.016848260
##      const
##  0.001001320
##
##
## Estimated coefficients for equation g_BA:
## =====
## Call:
## g_BA = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.194031493 -0.348105354   1.485908673 -0.045115290   0.367248905
##      g_BA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_BA.l3
## -0.350492656   0.308452238   0.007590089   0.022123793 -0.204071630
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_BA.l4      IPCA.l4
## -0.895390465   0.029049729   0.185848982 -0.107025754   1.568112611
## Selic.Over.l4      g_BR.l5      g_BA.l5      IPCA.l5 Selic.Over.l5
## -0.009724018 -0.018741795 -0.048329082 -1.587155861   0.023850668
##      const
## -0.010539369
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0049331368  0.0001341405  0.6550001041 -0.0003041579  0.0016209284
##      g_BA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_BA.l3
## -0.0058700534  0.0198373854  0.0045017551  0.0057577273  0.0014501312
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_BA.l4      IPCA.l4
##  0.0202880366  0.0005569764  0.0049691909 -0.0002743999 -0.0131840991
## Selic.Over.l4      g_BR.l5      g_BA.l5      IPCA.l5 Selic.Over.l5
##  0.0024145819 -0.0123022931  0.0050753002 -0.1710497976 -0.0065961109
##      const
##  0.0019450834
##
##

```



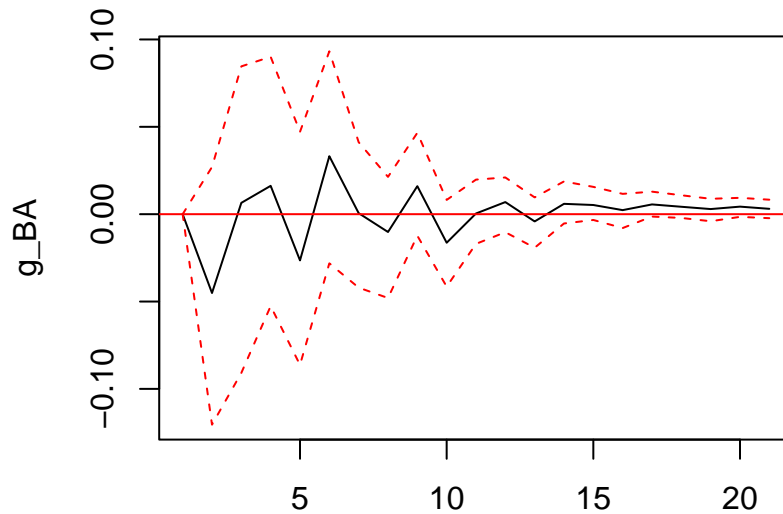
```

##
## data: Residual of Selic.Over equation
## Chi-squared = 53.831, df = 16, p-value = 5.525e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_BA
## Chi-squared = 637.26, df = 500, p-value = 2.941e-05
#Estimando o VAR Estrutural
SVAR_BA <- SVAR(VAR_BA, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_BA

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_BA    IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_BA      0.05144 1.00000 0.00000         0
## IPCA      0.08780 0.07032 1.00000         0
## Selic.Over 0.07617 0.03720 0.09111         1
#Função Impulso-Resposta
SVAR_BA_irf <- irf(SVAR_BA, impulse = "Selic.Over", response = "g_BA", n.ahead=20, ortho = TRUE)
plot(SVAR_BA_irf)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
```

```
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
```

```
#Selecionando o Lag
```

```
lagselect_BA_pa <- VARselect(base_SVAR_final[1:106,c(1,3,15,16)],lag.max=10, type="both")
lagselect_BA_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      9      2      1      2
```

```
lagselect_BA_pa$criteria
```

```
##           1           2           3           4           5
## AIC(n) -3.059156e+01 -3.091119e+01 -3.080477e+01 -3.081507e+01 -3.082705e+01
## HQ(n)  -3.033242e+01 -3.047930e+01 -3.020012e+01 -3.003766e+01 -2.987687e+01
## SC(n)  -2.995047e+01 -2.984272e+01 -2.930890e+01 -2.889181e+01 -2.847639e+01
## FPE(n)  5.182497e-14  3.773606e-14  4.219839e-14  4.216819e-14  4.230370e-14
##           6           7           8           9          10
## AIC(n) -3.079766e+01 -3.081546e+01 -3.094593e+01 -3.102468e+01 -3.098492e+01
## HQ(n)  -2.967473e+01 -2.951977e+01 -2.947748e+01 -2.938347e+01 -2.917095e+01
## SC(n)  -2.801962e+01 -2.761002e+01 -2.731310e+01 -2.696446e+01 -2.649731e+01
## FPE(n)  4.454725e-14  4.514274e-14  4.131048e-14  4.032696e-14  4.501233e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_BA_pa <- VAR(base_SVAR_final[1:106,c(1,3,15,16)], p = 2, season = NULL, exog = NULL, type = "const")
VAR_BA_pa
```

```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.269622882 0.014398070 0.015640657 -0.012939215 -0.095606147
##      g_BA.l2      IPCA.l2 Selic.Over.l2      const
## 0.037995147 -0.078914134 0.009891732 0.005803841
##
##
## Estimated coefficients for equation g_BA:
## =====
## Call:
## g_BA = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.965724265 -0.343277104 2.883318630 -0.036855216 -0.304723889
##      g_BA.l2      IPCA.l2 Selic.Over.l2      const
## -0.456757649 -2.829003556 0.032084609 0.006769642
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.002954226 0.003370470 0.761964157 0.001491596 -0.013288206
##      g_BA.l2      IPCA.l2 Selic.Over.l2      const
## -0.010740895 -0.040739209 -0.001093684 0.001062620
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_BA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.13462060 0.27533585 2.61679875 0.44742314 -0.48924441
##      g_BA.l2      IPCA.l2 Selic.Over.l2      const
## 0.05599305 7.26579028 0.47038176 0.03936732
##
## Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
##
## serial.test(VAR_BA_pa, type="BG")
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_BA_pa

```

```

## Chi-squared = 116.4, df = 80, p-value = 0.004931
arch.test(VAR_BA_pa, multivariate.only = FALSE)

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 13.806, df = 16, p-value = 0.6132
##
##
## $g_BA
##
## ARCH test (univariate)
##
## data: Residual of g_BA equation
## Chi-squared = 17.902, df = 16, p-value = 0.3297
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 19.259, df = 16, p-value = 0.2555
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 28.68, df = 16, p-value = 0.02618
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_BA_pa
## Chi-squared = 596.83, df = 500, p-value = 0.001837
#Estimando o VAR Estrutural
SVAR_BA_pa <- SVAR(VAR_BA_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod = "scoring")
SVAR_BA_pa

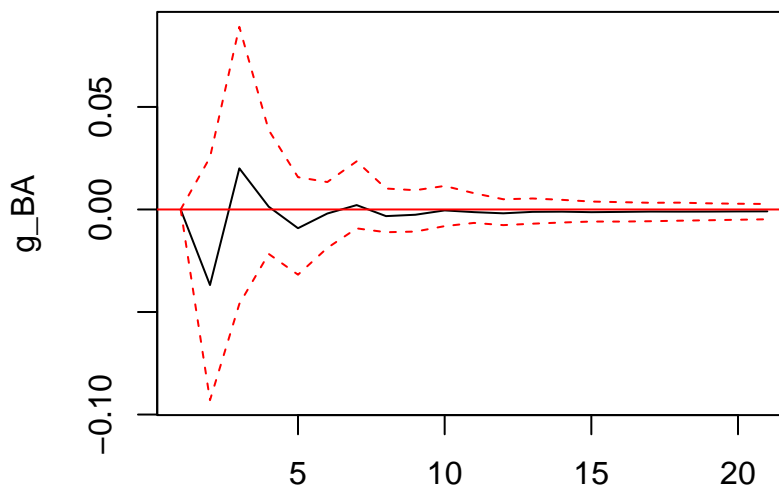
##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_BA    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_BA      0.08588  1.00000  0.00000      0

```



```
## IPCA      0.09430 0.07840 1.00000      0
## Selic.Over 0.09505 0.08543 0.09494      1
#Função Impulso-Resposta
SVAR_BA_irf_pa <- irf(SVAR_BA_pa, impulse = "Selic.Over", response = "g_BA", n.ahead=20, ortho = TRUE)
plot(SVAR_BA_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#
lagselect_BA_pe <- VARselect(base_SVAR_final[107:261,c(1,3,15,16)],lag.max=10, type="both")
lagselect_BA_pe$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      3      2      1      3
```

```
lagselect_BA_pe$criteria
```

	1	2	3	4	5
## AIC(n)	-2.923550e+01	-2.949477e+01	-2.951957e+01	-2.950186e+01	-2.941669e+01
## HQ(n)	-2.903530e+01	-2.916110e+01	-2.905243e+01	-2.890126e+01	-2.868262e+01
## SC(n)	-2.874280e+01	-2.867360e+01	-2.836994e+01	-2.802376e+01	-2.761012e+01
## FPE(n)	2.010327e-13	1.552262e-13	1.516567e-13	1.547893e-13	1.692769e-13
	6	7	8	9	10
## AIC(n)	-2.942175e+01	-2.933025e+01	-2.925985e+01	-2.916960e+01	-2.904551e+01
## HQ(n)	-2.855421e+01	-2.832924e+01	-2.812538e+01	-2.790166e+01	-2.764410e+01
## SC(n)	-2.728672e+01	-2.686674e+01	-2.646788e+01	-2.604917e+01	-2.559661e+01
## FPE(n)	1.694744e-13	1.873115e-13	2.032603e-13	2.257098e-13	2.602237e-13

#Estimando o VAR Reduzido

```
VAR_BA_pe <- VAR(base_SVAR_final[107:261,c(1,3,15,16)], p = lagselect_BA_pe$selection[1], season = NULL)
VAR_BA_pe
```

##

VAR Estimation Results:

=====

##

Estimated coefficients for equation g_BR:

=====

Call:

g_BR = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_BA.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.240181162	0.037774870	0.317945372	-0.025573898	-0.184500759
	g_BA.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_BA.l3
##	-0.024874564	-0.312329027	0.017388136	-0.139034354	-0.042320936
	IPCA.l3	Selic.Over.l3	const		
##	-0.957424311	0.001668794	0.007286306		

##

##

Estimated coefficients for equation g_BA:

=====

Call:

g_BA = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_BA.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	0.02480197	-0.26566676	1.34231385	-0.08329909	0.31005657
	g_BA.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_BA.l3
##	-0.19772032	2.48065579	0.02055305	-0.01032493	-0.19935772
	IPCA.l3	Selic.Over.l3	const		
##	-1.47994286	0.06804381	-0.01730712		

##

##

Estimated coefficients for equation IPCA:

=====

Call:

IPCA = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_BA.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	0.0001835422	-0.0025767828	0.5778387055	-0.0016971095	0.0042502014
	g_BA.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_BA.l3
##	-0.0046368622	0.0304449467	0.0051031911	0.0081868483	-0.0006105636
	IPCA.l3	Selic.Over.l3	const		
##	-0.0046358761	-0.0039050957	0.0022334480		

##

##

Estimated coefficients for equation Selic.Over:

=====

Call:

Selic.Over = g_BR.l1 + g_BA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_BA.l2 + IPCA.l2 + Selic.Over.l

##

	g_BR.l1	g_BA.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	0.20455721	-0.17373763	2.88623908	0.33764642	-0.05631983

```
##      g_BA.12      IPCA.12 Selic.Over.12      g_BR.13      g_BA.13
##    -0.05740184    1.34132665    0.48291490   -0.13602290    0.11734557
##      IPCA.13 Selic.Over.13      const
##    7.37383783    0.15564712   -0.03735475
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_BA_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_BA_pe
## Chi-squared = 99.602, df = 80, p-value = 0.06807
```

```
arch.test(VAR_BA_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.6152, df = 16, p-value = 1
##
##
## $g_BA
##
## ARCH test (univariate)
##
## data: Residual of g_BA equation
## Chi-squared = 29.587, df = 16, p-value = 0.02027
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 12.945, df = 16, p-value = 0.6768
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 20.81, df = 16, p-value = 0.1859
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_BA_pe
## Chi-squared = 547.4, df = 500, p-value = 0.07013
```

```
#Estimando o VAR Estrutural
```

```
SVAR_BA_pe <- SVAR(VAR_BA_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_BA_pe
```

```
##
```

```
## SVAR Estimation Results:
```

```
## =====
```

```
##
```

```
##
```

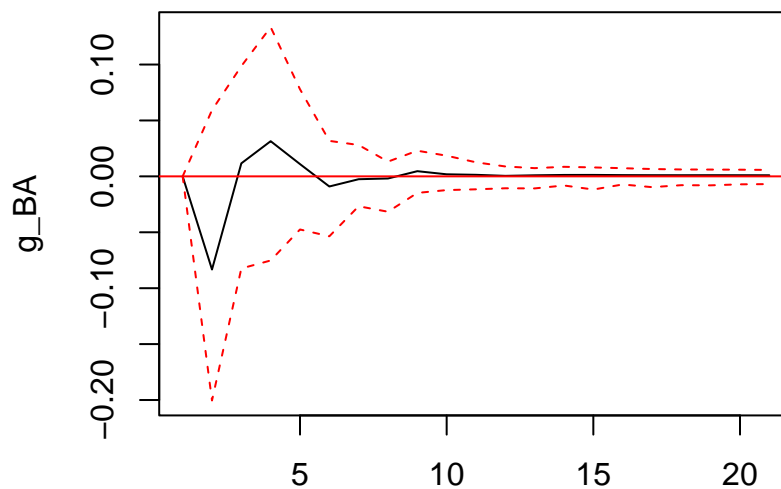
```
## Estimated A matrix:
```

```
##           g_BR    g_BA    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_BA      0.02854  1.00000  0.00000         0
## IPCA      0.08432  0.06657  1.00000         0
## Selic.Over 0.07529  0.01897  0.09202         1
```

```
#Função Impulso-Resposta
```

```
SVAR_BA_irf_pe <- irf(SVAR_BA_pe, impulse = "Selic.Over", response = "g_BA", n.ahead=20, ortho = TRUE)
plot(SVAR_BA_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
```

```
#####CEARÁ#####
```

```
#####
```

```
#Selecionando o Lag
```

```
lagselect_CE <- VARselect(base_SVAR_final[,c(1,4,15,16)],lag.max=10, type="both")
```

```
lagselect_CE$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      5      2      2      5
```

```
lagselect_CE$criteria
```

```
##              1              2              3              4              5
## AIC(n) -2.994918e+01 -3.022596e+01 -3.023991e+01 -3.031836e+01 -3.031956e+01
## HQ(n)  -2.981352e+01 -2.999987e+01 -2.992338e+01 -2.991139e+01 -2.982216e+01
## SC(n)  -2.961208e+01 -2.966413e+01 -2.945336e+01 -2.930708e+01 -2.908355e+01
## FPE(n)  9.845866e-14  7.466303e-14  7.365017e-14  6.812893e-14  6.810276e-14
##              6              7              8              9             10
## AIC(n) -3.027048e+01 -3.020063e+01 -3.023319e+01 -3.021545e+01 -3.015889e+01
## HQ(n)  -2.968264e+01 -2.952235e+01 -2.946447e+01 -2.935630e+01 -2.920930e+01
## SC(n)  -2.880973e+01 -2.851516e+01 -2.832298e+01 -2.808052e+01 -2.779923e+01
## FPE(n)  7.161319e-14  7.691778e-14  7.461136e-14  7.615005e-14  8.084982e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_CE <- VAR(base_SVAR_final[,c(1,4,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_CE
```

```
##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.2859022883  0.1121063886  0.3112156076 -0.0177018400 -0.1162074254
##      g_CE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_CE.l3
## -0.0429613127 -0.0952380590 -0.0005237096  0.0172349628 -0.0459031259
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_CE.l4      IPCA.l4
##  0.1505921972 -0.0075120235  0.0909984342 -0.0394355476 -1.0591002638
## Selic.Over.l4      g_BR.l5      g_CE.l5      IPCA.l5 Selic.Over.l5
##  0.0132778418  0.1089108941 -0.0649632269  0.4857171365  0.0135425191
##      const
##  0.0002923780
##
##
## Estimated coefficients for equation g_CE:
## =====
## Call:
## g_CE = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.135189536  0.052033120  0.387531083 -0.065737493 -0.136889932
##      g_CE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_CE.l3
## -0.183593532  0.940018411  0.028922767  0.064668006 -0.236653861
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_CE.l4      IPCA.l4
##  0.685061367  0.008176148  0.115411082 -0.103765448 -4.402850932
## Selic.Over.l4      g_BR.l5      g_CE.l5      IPCA.l5 Selic.Over.l5
##  0.056572411  0.268087717 -0.182591765  1.818597768 -0.029743571
##      const
```

```

## 0.003402265
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.11 + g_CE.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_CE.12 + IPCA.12 + Selic.Over.12 + g
##
##      g_BR.11      g_CE.11      IPCA.11 Selic.Over.11      g_BR.12
## -0.0192406581  0.0095612573  0.6418854059 -0.0008470346  0.0003457700
##      g_CE.12      IPCA.12 Selic.Over.12      g_BR.13      g_CE.13
## -0.0022522819  0.0475578216  0.0049969653  0.0193665163 -0.0061894010
##      IPCA.13 Selic.Over.13      g_BR.14      g_CE.14      IPCA.14
## -0.0098201797  0.0013042521  0.0027906034  0.0035517655 -0.0181636304
## Selic.Over.14      g_BR.15      g_CE.15      IPCA.15 Selic.Over.15
## 0.0020653445 -0.0123220720  0.0018930546 -0.1421930253 -0.0069757229
##      const
## 0.0019351289
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_CE.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_CE.12 + IPCA.12 + Selic.Over.1
##
##      g_BR.11      g_CE.11      IPCA.11 Selic.Over.11      g_BR.12
## 0.123934138  0.014631888  2.416587723  0.325305511  0.150103417
##      g_CE.12      IPCA.12 Selic.Over.12      g_BR.13      g_CE.13
## -0.049216174  2.977630113  0.566277283  0.380494681 -0.138099419
##      IPCA.13 Selic.Over.13      g_BR.14      g_CE.14      IPCA.14
## 3.532729954  0.386565242  0.344021513 -0.089164673 -0.505034487
## Selic.Over.14      g_BR.15      g_CE.15      IPCA.15 Selic.Over.15
## -0.265390094  0.260938008 -0.102323937  3.494610960 -0.071037115
##      const
## -0.009257879

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```

serial.test(VAR_CE, type="BG")

```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_CE
## Chi-squared = 107.85, df = 80, p-value = 0.0207

```

```

arch.test(VAR_CE, multivariate.only = FALSE)

```

```

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 4.153, df = 16, p-value = 0.9986
##

```

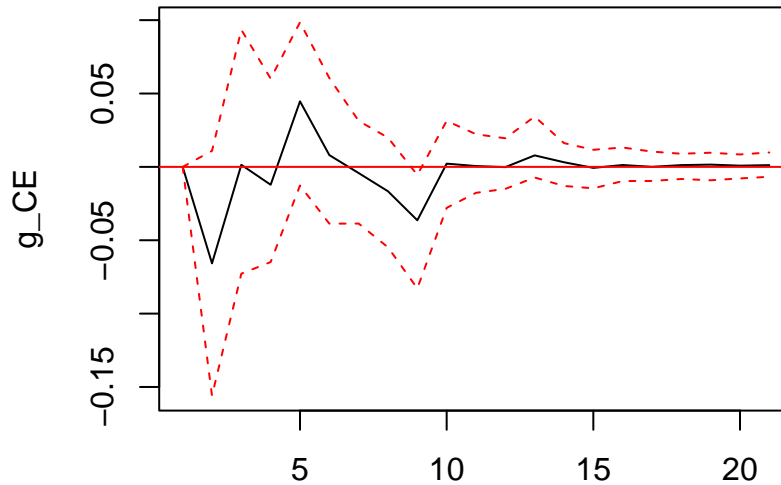
```

##
## $g_CE
##
## ARCH test (univariate)
##
## data: Residual of g_CE equation
## Chi-squared = 17.922, df = 16, p-value = 0.3285
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 13.856, df = 16, p-value = 0.6094
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 51.555, df = 16, p-value = 1.292e-05
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_CE
## Chi-squared = 773, df = 500, p-value = 4.752e-14
##
#Estimando o VAR Estrutural
SVAR_CE <- SVAR(VAR_CE, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod = "scoring")
SVAR_CE

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##      g_BR    g_CE    IPCA Selic.Over
## g_BR      1.000000 0.00000 0.00000      0
## g_CE      0.009694 1.00000 0.00000      0
## IPCA      0.085317 0.06342 1.00000      0
## Selic.Over 0.079761 0.03954 0.09201      1
##
#Função Impulso-Resposta
SVAR_CE_irf <- irf(SVAR_CE, impulse = "Selic.Over", response = "g_CE", n.ahead=20, ortho = TRUE)
plot(SVAR_CE_irf)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

###Comparando Diferentes Períodos###

#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#

#Selecionando o Lag

```
lagselect_CE_pa <- VARselect(base_SVAR_final[1:106,c(1,4,15,16)],lag.max=10, type="both")
lagselect_CE_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      1      1      1      1
```

```
lagselect_CE_pa$criteria
```

```
##           1           2           3           4           5
## AIC(n) -3.127292e+01 -3.126474e+01 -3.124222e+01 -3.121927e+01 -3.119591e+01
## HQ(n)  -3.101378e+01 -3.083284e+01 -3.063756e+01 -3.044185e+01 -3.024574e+01
## SC(n)  -3.063183e+01 -3.019626e+01 -2.974635e+01 -2.929601e+01 -2.884526e+01
## FPE(n)  2.621975e-14  2.649806e-14  2.724686e-14  2.814778e-14  2.925390e-14
##           6           7           8           9          10
## AIC(n) -3.111468e+01 -3.103123e+01 -3.108368e+01 -3.110754e+01 -3.104169e+01
## HQ(n)  -2.999175e+01 -2.973554e+01 -2.961523e+01 -2.946634e+01 -2.922772e+01
## SC(n)  -2.833664e+01 -2.782580e+01 -2.745085e+01 -2.704733e+01 -2.655408e+01
## FPE(n)  3.244446e-14  3.638124e-14  3.599451e-14  3.711989e-14  4.252811e-14
```

#Estimando o VAR Reduzido

```
VAR_CE_pa <- VAR(base_SVAR_final[1:106,c(1,4,15,16)], p = 2, season = NULL, exog = NULL, type = "const")
VAR_CE_pa
```



```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.282962426 -0.010659407  0.127412995 -0.011831128 -0.056417087
##      g_CE.l2      IPCA.l2 Selic.Over.l2      const
## -0.016146604 -0.100065984  0.008175820  0.006059588
##
##
## Estimated coefficients for equation g_CE:
## =====
## Call:
## g_CE = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.275204691 -0.364265706  0.983906090 -0.021919583  0.331305155
##      g_CE.l2      IPCA.l2 Selic.Over.l2      const
## -0.040473562 -0.687650747  0.017297672  0.004263365
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.0017819828 -0.0044004293  0.7637595190  0.0011586232 -0.0159692997
##      g_CE.l2      IPCA.l2 Selic.Over.l2      const
## -0.0021533681 -0.0512537644 -0.0007198308  0.0010545567
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.06561108  0.02747619  3.40914022  0.44590231 -0.23430321
##      g_CE.l2      IPCA.l2 Selic.Over.l2      const
## -0.08706206  6.83240190  0.46879461  0.04117180
##
## Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
##
## serial.test(VAR_CE_pa, type="BG")
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_CE_pa

```

```

## Chi-squared = 101.24, df = 80, p-value = 0.05464
arch.test(VAR_CE_pa, multivariate.only = FALSE)

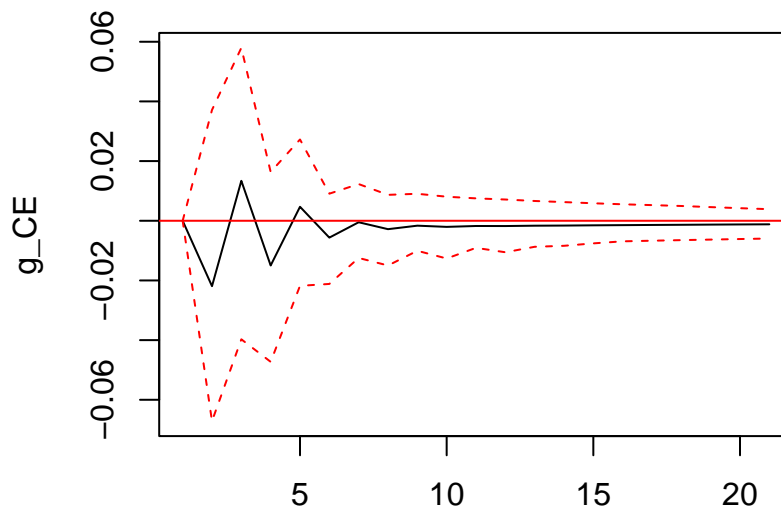
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 15.288, df = 16, p-value = 0.5037
##
##
## $g_CE
##
## ARCH test (univariate)
##
## data: Residual of g_CE equation
## Chi-squared = 16.231, df = 16, p-value = 0.437
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 17.147, df = 16, p-value = 0.3762
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 27.859, df = 16, p-value = 0.03287
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_CE_pa
## Chi-squared = 599.72, df = 500, p-value = 0.001414
#Estimando o VAR Estrutural
SVAR_CE_pa <- SVAR(VAR_CE_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_CE_pa

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_CE    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_CE      0.07743  1.00000  0.00000      0

```

```
## IPCA      0.09402 0.08227 1.00000      0
## Selic.Over 0.09146 0.06217 0.09237      1
#Função Impulso-Resposta
SVAR_CE_irf_pa <- irf(SVAR_CE_pa, impulse = "Selic.Over", response = "g_CE", n.ahead=20, ortho = TRUE)
plot(SVAR_CE_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#
lagselect_CE_pe <- VARselect(base_SVAR_final[107:261,c(1,4,15,16)],lag.max=10, type="both")
lagselect_CE_pe$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4
```

```
lagselect_CE_pe$criteria
```

	1	2	3	4	5
## AIC(n)	-2.964345e+01	-2.999724e+01	-3.003026e+01	-3.007778e+01	-2.998744e+01
## HQ(n)	-2.944325e+01	-2.966357e+01	-2.956313e+01	-2.947717e+01	-2.925337e+01
## SC(n)	-2.915075e+01	-2.917607e+01	-2.888063e+01	-2.859967e+01	-2.818087e+01
## FPE(n)	1.336885e-13	9.391736e-14	9.100608e-14	8.702104e-14	9.565860e-14

	6	7	8	9	10
## AIC(n)	-2.991954e+01	-2.981470e+01	-2.992736e+01	-2.993225e+01	-2.984145e+01
## HQ(n)	-2.905200e+01	-2.881370e+01	-2.879289e+01	-2.866431e+01	-2.844005e+01
## SC(n)	-2.778450e+01	-2.735120e+01	-2.713539e+01	-2.681181e+01	-2.639255e+01
## FPE(n)	1.030190e-13	1.153897e-13	1.042689e-13	1.052778e-13	1.174016e-13

#Estimando o VAR Reduzido

```
VAR_CE_pe <- VAR(base_SVAR_final[107:261,c(1,4,15,16)], p = 4, season = NULL, exog = NULL, type = "cons")
VAR_CE_pe
```

##

VAR Estimation Results:

=====

##

Estimated coefficients for equation g_BR:

=====

Call:

g_BR = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_CE.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.719597796	0.256007493	0.077590008	-0.019483292	-0.470250326
##	g_CE.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_CE.l3
##	0.035415302	0.223103425	0.028804850	-0.206246990	0.020472987
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_CE.l4	IPCA.l4
##	-0.666319902	0.001163805	-0.103439803	0.016223442	-1.353757576
##	Selic.Over.l4	const			
##	-0.016922483	0.010637308			

##

##

Estimated coefficients for equation g_CE:

=====

Call:

g_CE = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_CE.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.34120303	0.29620516	-0.08672733	-0.03519712	-0.46195663
##	g_CE.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_CE.l3
##	-0.18398856	2.15882193	0.08555076	-0.04854917	-0.15135367
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_CE.l4	IPCA.l4
##	-1.49264897	-0.01486056	-0.15384428	-0.04783431	-4.54768750
##	Selic.Over.l4	const			
##	-0.04418325	0.02298797			

##

##

Estimated coefficients for equation IPCA:

=====

Call:

IPCA = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_CE.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.030534920	0.015771465	0.553036193	-0.001035155	0.001975161
##	g_CE.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_CE.l3
##	-0.001187734	0.100785745	0.007822999	0.030027497	-0.010961955
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_CE.l4	IPCA.l4
##	-0.028088310	-0.001467592	0.004038663	0.007176119	-0.121148125
##	Selic.Over.l4	const			
##	-0.005708079	0.002643533			

##

##

Estimated coefficients for equation Selic.Over:

```
## =====
## Call:
## Selic.Over = g_BR.l1 + g_CE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_CE.l2 + IPCA.l2 + Selic.Over.l2
##
##      g_BR.l1      g_CE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      0.24167014 -0.05246778  2.46062127  0.33296602  0.19580686
##      g_CE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_CE.l3
##      -0.11136958  1.74307632  0.63781610 -0.13728628  0.06602679
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_CE.l4      IPCA.l4
##      3.67668125  0.30864235 -0.08714494 -0.01756385  3.28410916
## Selic.Over.l4      const
##      -0.31515088 -0.02700776
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_CE_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_CE_pe
## Chi-squared = 102.29, df = 80, p-value = 0.04722
```

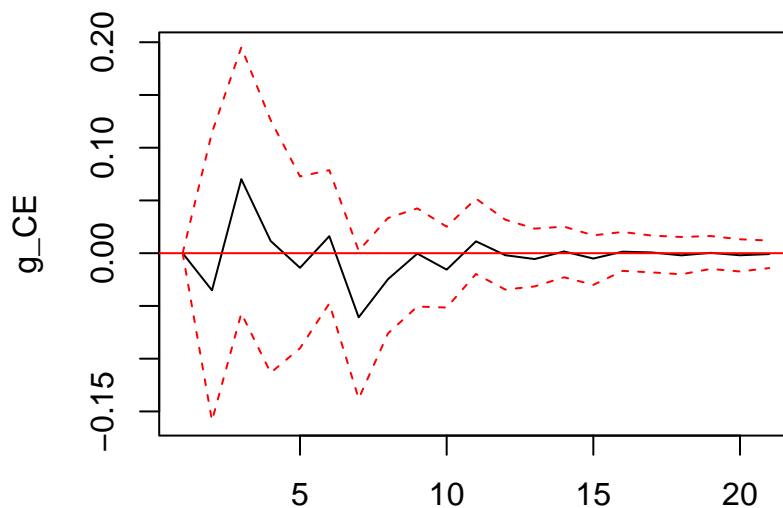
```
arch.test(VAR_CE_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.0575, df = 16, p-value = 1
##
##
## $g_CE
##
## ARCH test (univariate)
##
## data: Residual of g_CE equation
## Chi-squared = 10.631, df = 16, p-value = 0.8317
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 12.198, df = 16, p-value = 0.7303
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 32.7, df = 16, p-value = 0.008096
```

```
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_CE_pe
## Chi-squared = 596.7, df = 500, p-value = 0.001859
#Estimando o VAR Estrutural
SVAR_CE_pe <- SVAR(VAR_CE_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_CE_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##          g_BR      g_CE      IPCA Selic.Over
## g_BR      1.00000 0.000000 0.00000      0
## g_CE     -0.02346 1.000000 0.00000      0
## IPCA       0.08314 0.055465 1.00000      0
## Selic.Over 0.07212 0.004911 0.09088      1
#Função Impulso-Resposta
SVAR_CE_irf_pe <- irf(SVAR_CE_pe, impulse = "Selic.Over", response = "g_CE", n.ahead=20, ortho = TRUE)
plot(SVAR_CE_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####ESPÍRITOSANTO#####
#####
#Selecionando o Lag
lagselect_ES <- VARselect(base_SVAR_final[,c(1,5,15,16)],lag.max=10, type="both")
lagselect_ES$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_ES$criteria

##              1              2              3              4              5
## AIC(n) -2.964388e+01 -2.993813e+01 -2.992947e+01 -3.002109e+01 -3.000298e+01
## HQ(n)  -2.950822e+01 -2.971204e+01 -2.961294e+01 -2.961412e+01 -2.950558e+01
## SC(n)  -2.930678e+01 -2.937631e+01 -2.914291e+01 -2.900980e+01 -2.876697e+01
## FPE(n)  1.336117e-13  9.956512e-14  1.004613e-13  9.171392e-14  9.346616e-14
##              6              7              8              9             10
## AIC(n) -2.996960e+01 -2.988215e+01 -2.983695e+01 -2.981210e+01 -2.975101e+01
## HQ(n)  -2.938176e+01 -2.920388e+01 -2.906823e+01 -2.895295e+01 -2.880142e+01
## SC(n)  -2.850885e+01 -2.819668e+01 -2.792674e+01 -2.767717e+01 -2.739135e+01
## FPE(n)  9.675319e-14  1.057646e-13  1.108890e-13  1.139837e-13  1.215678e-13

#Estimando o VAR Reduzido
VAR_ES <- VAR(base_SVAR_final[,c(1,5,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_ES

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.172266566  0.085975045  0.276466426 -0.015028872 -0.228935821
##      g_ES.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_ES.l3
##  0.036525778 -0.236219109 -0.004978974 -0.174806650  0.023347751
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_ES.l4      IPCA.l4
##  0.218708046 -0.001566812 -0.083049247  0.053000201 -1.392278340
## Selic.Over.l4      g_BR.l5      g_ES.l5      IPCA.l5 Selic.Over.l5
##  0.012449227 -0.068091949  0.013733779  0.508885498  0.010610303
##      const
##  0.002223003
##
##
## Estimated coefficients for equation g_ES:
## =====
## Call:
## g_ES = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.583523517 -0.211325437 -0.119089983  0.008156671  0.231196879
```

```

##      g_ES.12      IPCA.12 Selic.Over.12      g_BR.13      g_ES.13
## -0.044605259 -0.157691476 -0.064606580 -0.169578678 -0.069703775
##      IPCA.13 Selic.Over.13      g_BR.14      g_ES.14      IPCA.14
## 2.470675103 -0.037718199 0.098175210 0.042149672 -2.042487650
## Selic.Over.14      g_BR.15      g_ES.15      IPCA.15 Selic.Over.15
## 0.038611337 -0.152615593 0.101789341 0.341008116 0.057347067
##      const
## -0.005223247
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.11 + g_ES.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_ES.12 + IPCA.12 + Selic.Over.12 + g
##
##      g_BR.11      g_ES.11      IPCA.11 Selic.Over.11      g_BR.12
## -0.0097408489 0.0025622277 0.6554418486 -0.0004131773 -0.0084675670
##      g_ES.12      IPCA.12 Selic.Over.12      g_BR.13      g_ES.13
## 0.0060471932 0.0144614617 0.0041817437 0.0036123201 -0.0007196605
##      IPCA.13 Selic.Over.13      g_BR.14      g_ES.14      IPCA.14
## 0.0063925942 0.0008268368 0.0026250496 -0.0019749559 -0.0015583609
## Selic.Over.14      g_BR.15      g_ES.15      IPCA.15 Selic.Over.15
## 0.0029408172 -0.0105513916 0.0080942595 -0.1785296437 -0.0069735383
##      const
## 0.0020343674
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_ES.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_ES.12 + IPCA.12 + Selic.Over.
##
##      g_BR.11      g_ES.11      IPCA.11 Selic.Over.11      g_BR.12
## 0.15074598 -0.04302102 2.24141890 0.32349682 -0.09569893
##      g_ES.12      IPCA.12 Selic.Over.12      g_BR.13      g_ES.13
## 0.15329368 2.25677927 0.57372010 -0.16601134 0.22123556
##      IPCA.13 Selic.Over.13      g_BR.14      g_ES.14      IPCA.14
## 3.67679173 0.39022028 -0.16116807 0.17509320 -0.44737025
## Selic.Over.14      g_BR.15      g_ES.15      IPCA.15 Selic.Over.15
## -0.27176577 0.01354730 -0.07183605 3.25227982 -0.07282848
##      const
## -0.00499409

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_ES, type="BG")
```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_ES
## Chi-squared = 99.641, df = 80, p-value = 0.06772

```



```

arch.test(VAR_ES, multivariate.only = FALSE)

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 4.737, df = 16, p-value = 0.9969
##
##
## $g_ES
##
## ARCH test (univariate)
##
## data: Residual of g_ES equation
## Chi-squared = 12.246, df = 16, p-value = 0.7269
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.939, df = 16, p-value = 0.2718
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 49.844, df = 16, p-value = 2.427e-05
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_ES
## Chi-squared = 649.06, df = 500, p-value = 7.332e-06

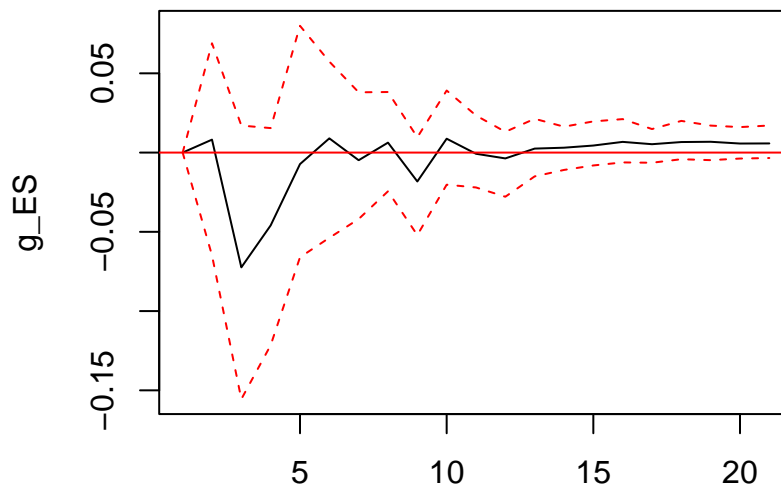
#Estimando o VAR Estrutural
SVAR_ES <- SVAR(VAR_ES, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_ES

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_ES    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_ES      0.05649  1.00000  0.00000         0
## IPCA      0.08834  0.07060  1.00000         0

```

```
## Selic.Over 0.07950 0.04294 0.09171 1
#Função Impulso-Resposta
SVAR_ES_irf <- irf(SVAR_ES, impulse = "Selic.Over", response = "g_ES", n.ahead=20, ortho = TRUE)
plot(SVAR_ES_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
#Selecionando o Lag
lagselect_ES_pa <- VARselect(base_SVAR_final[1:106,c(1,5,15,16)],lag.max=10, type="both")
lagselect_ES_pa$selection
```

```
## AIC(n) HQ(n) SC(n) FPE(n)
##      2      1      1      2
```

```
lagselect_ES_pa$criteria
```

	1	2	3	4	5
## AIC(n)	-3.080396e+01	-3.089463e+01	-3.080690e+01	-3.077216e+01	-3.071952e+01
## HQ(n)	-3.054482e+01	-3.046274e+01	-3.020225e+01	-2.999475e+01	-2.976935e+01
## SC(n)	-3.016287e+01	-2.982616e+01	-2.931103e+01	-2.884890e+01	-2.836887e+01
## FPE(n)	4.190787e-14	3.836617e-14	4.210854e-14	4.401685e-14	4.710610e-14
	6	7	8	9	10
## AIC(n)	-3.062123e+01	-3.054740e+01	-3.057821e+01	-3.053591e+01	-3.044422e+01
## HQ(n)	-2.949830e+01	-2.925171e+01	-2.910976e+01	-2.889470e+01	-2.863026e+01
## SC(n)	-2.784319e+01	-2.734197e+01	-2.694538e+01	-2.647569e+01	-2.595661e+01

```
## FPE(n) 5.314275e-14 5.902029e-14 5.967028e-14 6.574551e-14 7.729515e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_ES_pa <- VAR(base_SVAR_final[1:106,c(1,5,15,16)], p = lagselect_ES_pa$selection[1], season = NULL, c
VAR_ES_pa
```

```
##
```

```
## VAR Estimation Results:
```

```
## =====
```

```
##
```

```
## Estimated coefficients for equation g_BR:
```

```
## =====
```

```
## Call:
```

```
## g_BR = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.091320517 0.112571993 -0.316657617 -0.010154038 -0.177340553
```

```
##      g_ES.l2      IPCA.l2 Selic.Over.l2      const
## 0.083667616 0.161753695 0.008338497 0.004750379
```

```
##
```

```
##
```

```
## Estimated coefficients for equation g_ES:
```

```
## =====
```

```
## Call:
```

```
## g_ES = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.83839974 -0.19030520 -1.73076742 -0.02961845 -0.25444075
```

```
##      g_ES.l2      IPCA.l2 Selic.Over.l2      const
## 0.05028198 1.47869140 0.01888685 0.01788928
```

```
##
```

```
##
```

```
## Estimated coefficients for equation IPCA:
```

```
## =====
```

```
## Call:
```

```
## IPCA = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0154110930 0.0027132668 0.7486442353 0.0016915296 -0.0275470650
```

```
##      g_ES.l2      IPCA.l2 Selic.Over.l2      const
## 0.0117772033 -0.0616259608 -0.0010489137 0.0009474373
```

```
##
```

```
##
```

```
## Estimated coefficients for equation Selic.Over:
```

```
## =====
```

```
## Call:
```

```
## Selic.Over = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l
```

```
##
```

```
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.14494763 -0.17231266 4.17616243 0.45028404 -0.22097904
```

```
##      g_ES.l2      IPCA.l2 Selic.Over.l2      const
## -0.02697219 6.04599635 0.46382644 0.04213364
```

```
#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
```

```

serial.test(VAR_ES_pa, type="BG")

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_ES_pa
## Chi-squared = 100.93, df = 80, p-value = 0.05698
arch.test(VAR_ES_pa, multivariate.only = FALSE)

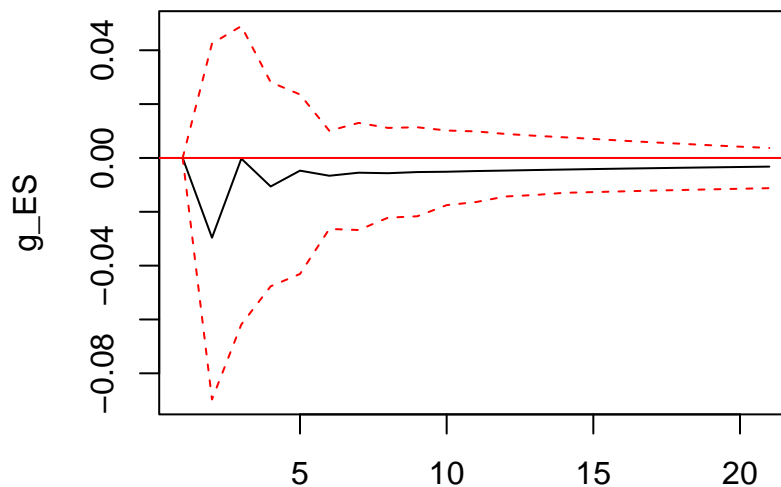
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 12.037, df = 16, p-value = 0.7414
##
##
## $g_ES
##
## ARCH test (univariate)
##
## data: Residual of g_ES equation
## Chi-squared = 19.348, df = 16, p-value = 0.251
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 15.414, df = 16, p-value = 0.4946
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 28.602, df = 16, p-value = 0.02676
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_ES_pa
## Chi-squared = 584.89, df = 500, p-value = 0.005133
#Estimando o VAR Estrutural
SVAR_ES_pa <- SVAR(VAR_ES_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_ES_pa

##
## SVAR Estimation Results:
## =====

```

```
##
##
## Estimated A matrix:
##           g_BR   g_ES   IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000      0
## g_ES      0.07049 1.00000 0.00000      0
## IPCA      0.09303 0.06939 1.00000      0
## Selic.Over 0.08526 0.03968 0.09022      1
#Função Impulso-Resposta
SVAR_ES_irf_pa <- irf(SVAR_ES_pa, impulse = "Selic.Over", response = "g_ES", n.ahead=20, ortho = TRUE)
plot(SVAR_ES_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_ES_pe <- VARselect(base_SVAR_final[107:261,c(1,5,15,16)],lag.max=10, type="both")
lagselect_ES_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      1      4

lagselect_ES_pe$criteria

##           1           2           3           4           5
## AIC(n) -2.922854e+01 -2.953018e+01 -2.952175e+01 -2.959813e+01 -2.955103e+01
## HQ(n)  -2.902834e+01 -2.919651e+01 -2.905461e+01 -2.899753e+01 -2.881695e+01
## SC(n)  -2.873584e+01 -2.870901e+01 -2.837211e+01 -2.812003e+01 -2.774446e+01
## FPE(n)  2.024370e-13 1.498261e-13 1.513270e-13 1.405826e-13 1.479976e-13
```

```

##          6          7          8          9          10
## AIC(n) -2.951878e+01 -2.940903e+01 -2.930894e+01 -2.926485e+01 -2.911262e+01
## HQ(n) -2.865125e+01 -2.840803e+01 -2.817447e+01 -2.799691e+01 -2.771121e+01
## SC(n) -2.738375e+01 -2.694553e+01 -2.651697e+01 -2.614442e+01 -2.566371e+01
## FPE(n) 1.538027e-13 1.731206e-13 1.935223e-13 2.052031e-13 2.433344e-13

#Estimando o VAR Reduzido
VAR_ES_pe <- VAR(base_SVAR_final[107:261,c(1,5,15,16)], p = lagselect_ES_pe$selection[1], season = NULL)
VAR_ES_pe

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.2550566048 0.0403499097 0.4523254039 0.0004745601 -0.2522903231
##      g_ES.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_ES.l3
## -0.0104806690 -0.2482785878 0.0137263692 -0.1792070481 -0.0275968426
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_ES.l4      IPCA.l4
## -0.2735913396 0.0004547716 -0.0851998558 0.0186390775 -1.7912574289
## Selic.Over.l4      const
## -0.0201945596 0.0110975832
##
##
## Estimated coefficients for equation g_ES:
## =====
## Call:
## g_ES = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.501763863 -0.198883767 0.214717457 0.100049811 0.263425051
##      g_ES.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_ES.l3
## -0.082029442 -0.303292766 -0.048817257 -0.239767220 -0.082055456
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_ES.l4      IPCA.l4
## 2.895289054 0.005785548 0.195029446 0.022776863 -4.577981882
## Selic.Over.l4      const
## -0.053414058 0.001681002
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_ES.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_ES.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_ES.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0053386083 0.0009544841 0.5704119364 0.0011069752 -0.0037112209
##      g_ES.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_ES.l3
## 0.0036353905 0.0382721947 0.0068891907 0.0039761051 0.0003554880
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_ES.l4      IPCA.l4
## 0.0237582673 -0.0039894979 0.0074967804 -0.0048726805 -0.1525311604

```



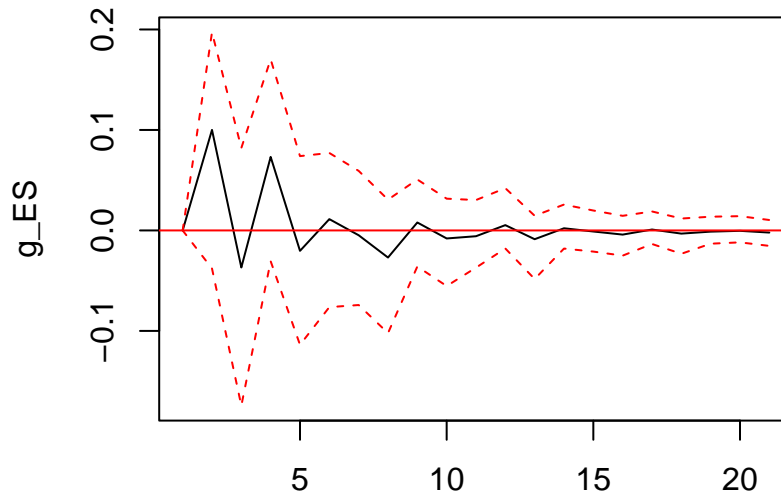
```

##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 25.594, df = 16, p-value = 0.06001
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_ES_pe
## Chi-squared = 587.51, df = 500, p-value = 0.004128
#Estimando o VAR Estrutural
SVAR_ES_pe <- SVAR(VAR_ES_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_ES

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##          g_BR    g_ES    IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_ES      0.05649 1.00000 0.00000         0
## IPCA      0.08834 0.07060 1.00000         0
## Selic.Over 0.07950 0.04294 0.09171         1
#Função Impulso-Resposta
SVAR_ES_irf_pe <- irf(SVAR_ES_pe, impulse = "Selic.Over", response = "g_ES", n.ahead=20, ortho = TRUE)
plot(SVAR_ES_irf_pe)

```


SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####GOIÁS#####
#####
```

#Seleccionando o Lag

```
lagselect_GO <- VARselect(base_SVAR_final[,c(1,6,15,16)],lag.max=10, type="both")
lagselect_GO$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      5      4      2      5
```

```
lagselect_GO$criteria
```

```
##          1          2          3          4          5
## AIC(n) -3.086122e+01 -3.137480e+01 -3.138677e+01 -3.159527e+01 -3.162361e+01
## HQ(n)  -3.072556e+01 -3.114871e+01 -3.107024e+01 -3.118830e+01 -3.112621e+01
## SC(n)  -3.052412e+01 -3.081298e+01 -3.060021e+01 -3.058398e+01 -3.038760e+01
## FPE(n)  3.955119e-14  2.366847e-14  2.339385e-14  1.900105e-14  1.848513e-14
##          6          7          8          9         10
## AIC(n) -3.159714e+01 -3.150360e+01 -3.144382e+01 -3.144409e+01 -3.137666e+01
## HQ(n)  -3.100930e+01 -3.082533e+01 -3.067510e+01 -3.058494e+01 -3.042708e+01
## SC(n)  -3.013640e+01 -2.981813e+01 -2.953361e+01 -2.930915e+01 -2.901700e+01
## FPE(n)  1.900342e-14  2.090032e-14  2.223484e-14  2.228846e-14  2.392253e-14
```

#Estimando o VAR Reduzido

```
VAR_GO <- VAR(base_SVAR_final[,c(1,6,15,16)], p = lagselect_GO$selection[1], season = NULL, exog = NULL)
VAR_GO
```

```
##
```

```

## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.064868049 -0.159321349  0.386537285 -0.006132949 -0.154963343
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## -0.052653497 -0.404281946 -0.005547944 -0.106678050  0.003181092
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
##  0.356941425 -0.013740556 -0.046505909  0.105116595 -1.122271073
## Selic.Over.l4      g_BR.l5      g_GO.l5      IPCA.l5 Selic.Over.l5
##  0.010609798 -0.038534665  0.110301880  0.224990322  0.017041809
##      const
##  0.001108717
##
##
## Estimated coefficients for equation g_GO:
## =====
## Call:
## g_GO = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.019707833 -0.977483907 -0.032789203 -0.002959576  0.012571985
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## -0.917215254  0.086910332 -0.008162934 -0.005035041 -0.624472082
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
##  0.428418067  0.020581978  0.068193878 -0.507966965 -0.440540293
## Selic.Over.l4      g_BR.l5      g_GO.l5      IPCA.l5 Selic.Over.l5
##  0.018861887 -0.074979406 -0.132272080 -0.276850950 -0.018271429
##      const
##  0.001335545
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0023816552 -0.0068438565  0.6535833821 -0.0002769579 -0.0031911377
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
##  0.0040669345  0.0175824614  0.0044467526  0.0042363681  0.0110495917
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
##  0.0096291108  0.0007276907  0.0044892646  0.0056534807 -0.0006459874
## Selic.Over.l4      g_BR.l5      g_GO.l5      IPCA.l5 Selic.Over.l5
##  0.0023812566 -0.0076327646  0.0024791732 -0.1717849933 -0.0067397449
##      const
##  0.0019527143
##
##

```

```
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l3 + g_BR.l4 + g_GO.l4 + IPCA.l4 + Selic.Over.l4 + g_BR.l5 + g_GO.l5 + IPCA.l5 + Selic.Over.l5 + const
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      0.12878939  0.09840873  2.36648241  0.33043559  0.06126779
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
##      -0.04312393  2.78337017  0.56765693  0.11941230 -0.06523925
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
##      3.33340894  0.38498821  0.10776303 -0.04237072 -0.56286039
## Selic.Over.l4      g_BR.l5      g_GO.l5      IPCA.l5 Selic.Over.l5
##      -0.26919710  0.06506364  0.03636922  3.56607519 -0.07083189
##      const
##      -0.00788735
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_GO, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_GO
## Chi-squared = 75.045, df = 80, p-value = 0.6358
```

```
arch.test(VAR_GO, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 5.3154, df = 16, p-value = 0.9939
##
##
## $g_GO
##
## ARCH test (univariate)
##
## data: Residual of g_GO equation
## Chi-squared = 1.6675, df = 16, p-value = 1
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.539, df = 16, p-value = 0.2933
##
##
## $Selic.Over
##
## ARCH test (univariate)
```

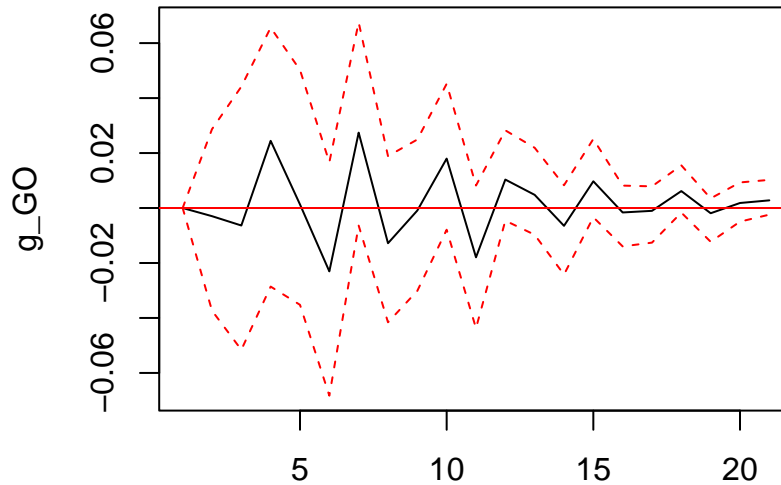
```
##
## data: Residual of Selic.Over equation
## Chi-squared = 56.349, df = 16, p-value = 2.131e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_GO
## Chi-squared = 740.35, df = 500, p-value = 1.365e-11

#Estimando o VAR Estrutural
SVAR_GO <- SVAR(VAR_GO, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_GO

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_GO    IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_GO      0.08269 1.00000 0.00000         0
## IPCA      0.09223 0.09324 1.00000         0
## Selic.Over 0.08159 0.07946 0.09482         1

#Função Impulso-Resposta
SVAR_GO_irf <- irf(SVAR_GO, impulse = "Selic.Over", response = "g_GO", n.ahead=20, ortho = TRUE)
plot(SVAR_GO_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

###Comparando Diferentes Períodos###

#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#

#Selecionando o Lag

```
lagselect_GO_pa <- VARselect(base_SVAR_final[1:106,c(1,6,15,16)],lag.max=10, type="both")
lagselect_GO_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      6      2      2      5
```

```
lagselect_GO_pa$criteria
```

```
##          1          2          3          4          5
## AIC(n) -3.229147e+01 -3.276754e+01 -3.282487e+01 -3.295264e+01 -3.300132e+01
## HQ(n)  -3.203233e+01 -3.233564e+01 -3.222021e+01 -3.217522e+01 -3.205114e+01
## SC(n)  -3.165038e+01 -3.169906e+01 -3.132900e+01 -3.102937e+01 -3.065066e+01
## FPE(n)  9.468413e-15  5.896005e-15  5.597323e-15  4.973377e-15  4.809561e-15
##          6          7          8          9         10
## AIC(n) -3.300201e+01 -3.295748e+01 -3.279237e+01 -3.278556e+01 -3.276919e+01
## HQ(n)  -3.187908e+01 -3.166179e+01 -3.132393e+01 -3.114436e+01 -3.095522e+01
## SC(n)  -3.022396e+01 -2.975205e+01 -2.915955e+01 -2.872534e+01 -2.828158e+01
## FPE(n)  4.914577e-15  5.300518e-15  6.518667e-15  6.931903e-15  7.558444e-15
```

#Estimando o VAR Reduzido

```
VAR_GO_pa <- VAR(base_SVAR_final[1:106,c(1,6,15,16)], p = 4, season = NULL, exog = NULL, type = "const")
VAR_GO_pa
```

```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.251729102 -0.018348867  0.303052471 -0.022540637 -0.048775777
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## 0.084192971 -0.269911293 -0.008402289 -0.012641141  0.143684617
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
## 0.488630599  0.002721962  0.023909033  0.063951716 -0.325185566
## Selic.Over.l4      const
## 0.026044234  0.002245584
##
##
## Estimated coefficients for equation g_GO:
## =====
## Call:
## g_GO = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0334692745 -1.4087086912  0.8266469292 -0.0011589415  0.0872963632
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## -1.5105559365 -0.1295474598 -0.0203346507 -0.0502859920 -1.0121317926
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
## 0.3252224781  0.0006131424  0.1231302293 -0.4668096020 -0.4754423229
## Selic.Over.l4      const
## 0.0181894410  0.0193608172
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0096897240 -0.0119478083  0.7843092727  0.0013871681 -0.0168841885
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## -0.0279679242  0.0200198215 -0.0009372471  0.0217104436 -0.0038041146
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
## -0.1199803151 -0.0019779795  0.0057905612  0.0041339692  0.0255749232
## Selic.Over.l4      const
## 0.0022846680  0.0008035404
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l
##

```

```
##      g_BR.11      g_GO.11      IPCA.11 Selic.Over.11      g_BR.12
##      0.10000188   -0.38680936   4.29283832   0.35820294   -0.25480069
##      g_GO.12      IPCA.12 Selic.Over.12      g_BR.13      g_GO.13
##      -0.83274714   3.23849731   0.49145871   0.40081450   -0.44479921
##      IPCA.13 Selic.Over.13      g_BR.14      g_GO.14      IPCA.14
##      3.02589558   0.33645198   0.25058292   -0.51835246   1.79624845
## Selic.Over.14      const
##      -0.28765192   0.05328355
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_GO_pa, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_GO_pa
## Chi-squared = 108.1, df = 80, p-value = 0.01991
```

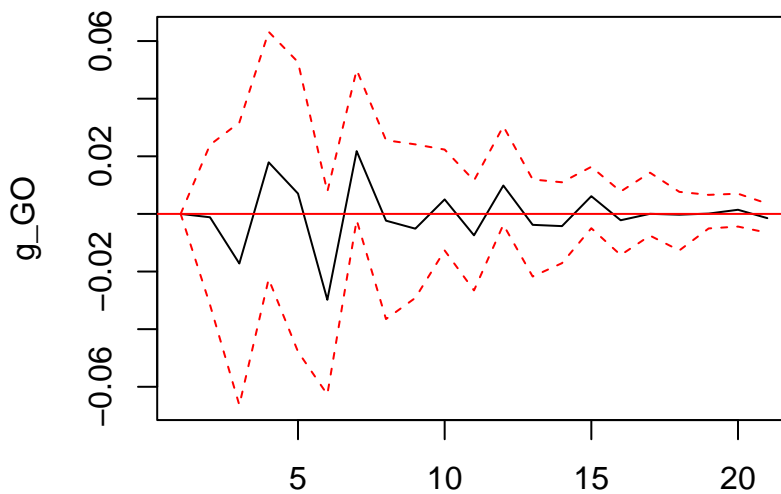
```
arch.test(VAR_GO_pa, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 9.3846, df = 16, p-value = 0.8967
##
##
## $g_GO
##
## ARCH test (univariate)
##
## data: Residual of g_GO equation
## Chi-squared = 15.26, df = 16, p-value = 0.5057
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 5.6264, df = 16, p-value = 0.9917
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 20.76, df = 16, p-value = 0.1879
##
##
## ARCH (multivariate)
```

```
##
## data: Residuals of VAR object VAR_GO_pa
## Chi-squared = 573.66, df = 500, p-value = 0.01239
#Estimando o VAR Estrutural
SVAR_GO_pa <- SVAR(VAR_GO_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_GO_pa

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_GO    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_GO      0.09769  1.00000  0.00000         0
## IPCA      0.09722  0.09785  1.00000         0
## Selic.Over 0.08785  0.09111  0.09374         1
#Função Impulso-Resposta
SVAR_GO_irf_pa <- irf(SVAR_GO_pa, impulse = "Selic.Over", response = "g_GO", n.ahead=20, ortho = TRUE)
plot(SVAR_GO_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#
lagselect_GO_pe <- VARselect(base_SVAR_final[107:261,c(1,6,15,16)],lag.max=10, type="both")
lagselect_GO_pe$selection
```



```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4
```

```
lagselect_GO_pe$criteria
```

```
##              1              2              3              4              5
## AIC(n) -3.032667e+01 -3.077317e+01 -3.077093e+01 -3.099065e+01 -3.097455e+01
## HQ(n)  -3.012647e+01 -3.043950e+01 -3.030379e+01 -3.039004e+01 -3.024048e+01
## SC(n)  -2.983397e+01 -2.995200e+01 -2.962130e+01 -2.951254e+01 -2.916798e+01
## FPE(n)  6.751128e-14  4.322769e-14  4.339135e-14  3.492764e-14  3.564722e-14
##              6              7              8              9             10
## AIC(n) -3.092839e+01 -3.078860e+01 -3.071469e+01 -3.062853e+01 -3.047922e+01
## HQ(n)  -3.006085e+01 -2.978759e+01 -2.958021e+01 -2.936059e+01 -2.907781e+01
## SC(n)  -2.879335e+01 -2.832509e+01 -2.792272e+01 -2.750809e+01 -2.703032e+01
## FPE(n)  3.756471e-14  4.357231e-14  4.744877e-14  5.247428e-14  6.204334e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_GO_pe <- VAR(base_SVAR_final[107:261,c(1,6,15,16)], p = lagselect_GO_pe$selection[1], season = NULL)
VAR_GO_pe
```

```
##
```

```
## VAR Estimation Results:
```

```
## =====
```

```
##
```

```
## Estimated coefficients for equation g_BR:
```

```
## =====
```

```
## Call:
```

```
## g_BR = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
```

```
## -0.169078225 -0.198697246  0.523791440  0.003790553 -0.255137216
```

```
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
```

```
## -0.096960704 -0.571090602  0.015836327 -0.168760524 -0.117921264
```

```
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
```

```
##  0.063006643 -0.011442595 -0.107384870  0.001481406 -1.788384968
```

```
## Selic.Over.l4      const
```

```
## -0.012817937  0.010509524
```

```
##
```

```
##
```

```
## Estimated coefficients for equation g_GO:
```

```
## =====
```

```
## Call:
```

```
## g_GO = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
```

```
## -0.0286084678 -0.8299677405 -0.7366784283 -0.0135500929  0.0012572919
```

```
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
```

```
## -0.7705441196  0.9662421306 -0.0155713265 -0.0211593128 -0.4468775428
```

```
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
```

```
##  0.1382469591  0.0412111612  0.0675041411 -0.4236782516 -0.4513823611
```

```
## Selic.Over.l4      const
```

```
## -0.0055881327  0.0001252806
```

```
##
```

```
##
```

```
## Estimated coefficients for equation IPCA:
```

```

## =====
## Call:
## IPCA = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.001866240 -0.010630088 0.573500008 0.001036627 -0.001539879
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## 0.007552397 0.030519991 0.005909844 0.004843031 0.011444618
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
## 0.048946316 -0.003801963 0.007274436 0.006721589 -0.158490191
## Selic.Over.l4      const
## -0.003601968 0.002711487
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_GO.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_GO.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_GO.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.032502706 0.058263754 2.020012465 0.328206074 -0.004291574
##      g_GO.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_GO.l3
## -0.155087170 1.921665247 0.640933018 -0.020584866 -0.296786581
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_GO.l4      IPCA.l4
## 3.679089392 0.327691625 -0.066991841 -0.106547235 3.237891776
## Selic.Over.l4      const
## -0.331983912 -0.025439666

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```

serial.test(VAR_GO_pe, type="BG")

```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_GO_pe
## Chi-squared = 85.642, df = 80, p-value = 0.3126

```

```

arch.test(VAR_GO_pe, multivariate.only = FALSE)

```

```

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.7815, df = 16, p-value = 1
##
##
## $g_GO
##
## ARCH test (univariate)
##
## data: Residual of g_GO equation
## Chi-squared = 2.507, df = 16, p-value = 0.9999
##

```

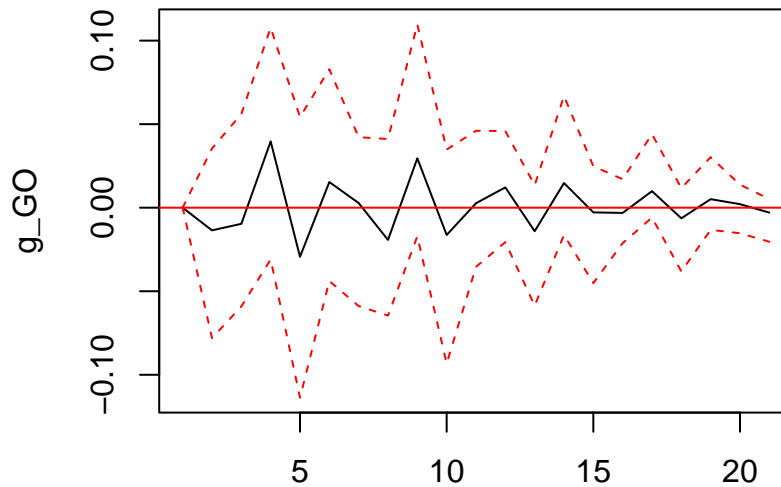
```

##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 14.486, df = 16, p-value = 0.5626
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 33.726, df = 16, p-value = 0.005913
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_GO_pe
## Chi-squared = 616.26, df = 500, p-value = 0.0002849
#Estimando o VAR Estrutural
SVAR_GO_pe <- SVAR(VAR_GO_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_GO

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_GO    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_GO      0.08269  1.00000  0.00000         0
## IPCA      0.09223  0.09324  1.00000         0
## Selic.Over 0.08159  0.07946  0.09482         1
#Função Impulso-Resposta
SVAR_GO_irf_pe <- irf(SVAR_GO_pe, impulse = "Selic.Over", response = "g_GO", n.ahead=20, ortho = TRUE)
plot(SVAR_GO_irf_pe)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####MINAS#####
#####
#Seleccionando o Lag
lagselect_MG <- VARselect(base_SVAR_final[,c(1,7,15,16)],lag.max=10, type="both")
lagselect_MG$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_MG$criteria

##           1           2           3           4           5
## AIC(n) -3.101122e+01 -3.128703e+01 -3.129061e+01 -3.135451e+01 -3.133231e+01
## HQ(n)  -3.087556e+01 -3.106094e+01 -3.097408e+01 -3.094755e+01 -3.083490e+01
## SC(n)  -3.067412e+01 -3.072520e+01 -3.050405e+01 -3.034323e+01 -3.009629e+01
## FPE(n)  3.404209e-14  2.583983e-14  2.575505e-14  2.417336e-14  2.473637e-14
##           6           7           8           9          10
## AIC(n) -3.132470e+01 -3.123231e+01 -3.119482e+01 -3.116514e+01 -3.114031e+01
## HQ(n)  -3.073686e+01 -3.055404e+01 -3.042611e+01 -3.030599e+01 -3.019072e+01
## SC(n)  -2.986395e+01 -2.954684e+01 -2.928461e+01 -2.903020e+01 -2.878064e+01
## FPE(n)  2.495468e-14  2.741401e-14  2.852151e-14  2.945962e-14  3.030087e-14

#Estimando o VAR Reduzido
VAR_MG <- VAR(base_SVAR_final[,c(1,7,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_MG

##
```

```

## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.2697763421  0.1557779200  0.4595298237 -0.0111691570 -0.2811121496
##      g_MG.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_MG.l3
##  0.1137105877 -0.3978557318 -0.0026044945 -0.2610984782  0.1382612520
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_MG.l4      IPCA.l4
##  0.2105418182 -0.0040102733 -0.1705948897  0.1182348843 -1.1919726734
## Selic.Over.l4      g_BR.l5      g_MG.l5      IPCA.l5 Selic.Over.l5
##  0.0101619369 -0.1386490421  0.1297207584  0.2784169667  0.0103802160
##      const
##  0.0009571995
##
##
## Estimated coefficients for equation g_MG:
## =====
## Call:
## g_MG = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.0661776343 -0.1401906220  0.4049535390 -0.0232409129  0.0453902766
##      g_MG.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_MG.l3
## -0.1253587006 -0.4377029105 -0.0181185050 -0.0866095173 -0.0221170935
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_MG.l4      IPCA.l4
##  0.6460575939  0.0068687052 -0.0627141330  0.0211912036 -0.8876958304
## Selic.Over.l4      g_BR.l5      g_MG.l5      IPCA.l5 Selic.Over.l5
##  0.0294928977 -0.0525479653  0.0557063962  0.1764378230  0.0054446675
##      const
##  0.0005577908
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0138633298  0.0090626287  0.6482607515 -0.0001624476 -0.0130749775
##      g_MG.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_MG.l3
##  0.0106967864  0.0231684231  0.0047172239  0.0020648433  0.0045694527
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_MG.l4      IPCA.l4
##  0.0108498630  0.0010287194  0.0069281874 -0.0047414099 -0.0147663114
## Selic.Over.l4      g_BR.l5      g_MG.l5      IPCA.l5 Selic.Over.l5
##  0.0020090823 -0.0087089665  0.0014169445 -0.1708720285 -0.0069802786
##      const
##  0.0019790638
##
##

```

```
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l1
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.141539142 -0.003577998  2.371533691  0.328267886 -0.054177017
##      g_MG.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_MG.l3
## 0.059092452  2.646352805  0.569096773 -0.179800987  0.314727401
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_MG.l4      IPCA.l4
## 3.485999859  0.385991518 -0.040696933  0.127138792 -0.564186231
## Selic.Over.l4      g_BR.l5      g_MG.l5      IPCA.l5 Selic.Over.l5
## -0.264484600 -0.058840195  0.135763983  3.447761854 -0.075023369
##      const
## -0.008236118
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_MG, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_MG
## Chi-squared = 101.78, df = 80, p-value = 0.05067
```

```
arch.test(VAR_MG, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 2.8076, df = 16, p-value = 0.9999
##
##
## $g_MG
##
## ARCH test (univariate)
##
## data: Residual of g_MG equation
## Chi-squared = 28.952, df = 16, p-value = 0.02426
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 22.439, df = 16, p-value = 0.1296
##
##
## $Selic.Over
##
## ARCH test (univariate)
```

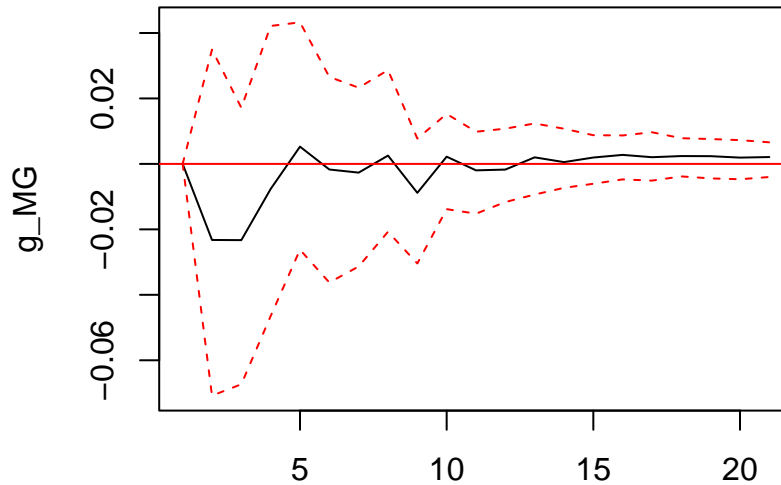
```
##
## data: Residual of Selic.Over equation
## Chi-squared = 56.001, df = 16, p-value = 2.433e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_MG
## Chi-squared = 653.03, df = 500, p-value = 4.519e-06

#Estimando o VAR Estrutural
SVAR_MG <- SVAR(VAR_MG, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_MG

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_MG    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_MG      0.04538  1.00000  0.00000         0
## IPCA      0.08830  0.08432  1.00000         0
## Selic.Over 0.08221  0.08265  0.09532         1

#Função Impulso-Resposta
SVAR_MG_irf <- irf(SVAR_MG, impulse = "Selic.Over", response = "g_MG", n.ahead=20, ortho = TRUE)
plot(SVAR_MG_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
```

```
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
```

```
#Selecionando o Lag
```

```
lagselect_MG_pa <- VARselect(base_SVAR_final[1:106,c(1,7,15,16)],lag.max=10, type="both")
lagselect_MG_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      10      1      1      2
```

```
lagselect_MG_pa$criteria
```

```
##           1           2           3           4           5
## AIC(n) -3.278838e+01 -3.287835e+01 -3.283272e+01 -3.286312e+01 -3.283639e+01
## HQ(n)  -3.252924e+01 -3.244645e+01 -3.222807e+01 -3.208571e+01 -3.188621e+01
## SC(n)  -3.214729e+01 -3.180987e+01 -3.133685e+01 -3.093986e+01 -3.048573e+01
## FPE(n)  5.760638e-15  5.277540e-15  5.553523e-15  5.439088e-15  5.671956e-15
##           6           7           8           9          10
## AIC(n) -3.288552e+01 -3.287745e+01 -3.285132e+01 -3.292473e+01 -3.297563e+01
## HQ(n)  -3.176259e+01 -3.158176e+01 -3.138287e+01 -3.128353e+01 -3.116166e+01
## SC(n)  -3.010747e+01 -2.967202e+01 -2.921849e+01 -2.886451e+01 -2.848802e+01
## FPE(n)  5.521757e-15  5.742143e-15  6.145535e-15  6.031313e-15  6.148628e-15
```

```
#Estimando o VAR Reduzido
```

```
VAR_MG_pa <- VAR(base_SVAR_final[1:106,c(1,7,15,16)], p = 2, season = NULL, exog = NULL, type = "const")
VAR_MG_pa
```



```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.020791460  0.181204527 -0.140715312 -0.006115116 -0.289076578
##      g_MG.l2      IPCA.l2 Selic.Over.l2      const
##  0.141430478  0.022837842  0.002900789  0.006796464
##
##
## Estimated coefficients for equation g_MG:
## =====
## Call:
## g_MG = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.584991986 -0.029139782 -0.521859343 -0.001840282 -0.385315799
##      g_MG.l2      IPCA.l2 Selic.Over.l2      const
##  0.061890142  0.447229977 -0.001707570  0.006359073
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.057209997  0.034882271  0.714400692  0.002144214 -0.048166923
##      g_MG.l2      IPCA.l2 Selic.Over.l2      const
##  0.019275732 -0.024858408 -0.001648398  0.001190188
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.92956404 -0.67740518  3.78548915  0.43242143 -0.53966985
##      g_MG.l2      IPCA.l2 Selic.Over.l2      const
##  0.20422214  6.34723533  0.48439007  0.03862111
##
## Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
##
## serial.test(VAR_MG_pa, type="BG")
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_MG_pa

```

```

## Chi-squared = 106.22, df = 80, p-value = 0.02662
arch.test(VAR_MG_pa, multivariate.only = FALSE)

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 15.15, df = 16, p-value = 0.5137
##
##
## $g_MG
##
## ARCH test (univariate)
##
## data: Residual of g_MG equation
## Chi-squared = 43.392, df = 16, p-value = 0.0002438
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.851, df = 16, p-value = 0.2765
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 28.052, df = 16, p-value = 0.03117
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_MG_pa
## Chi-squared = 637.08, df = 500, p-value = 3.003e-05

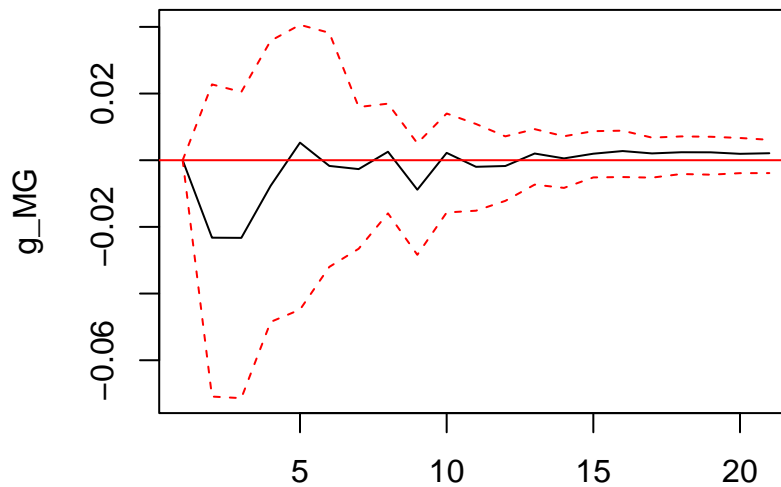
#Estimando o VAR Estrutural
SVAR_MG_pa <- SVAR(VAR_MG, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_MG_pa

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_MG    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_MG      0.04538  1.00000  0.00000      0

```

```
## IPCA      0.08830 0.08432 1.00000      0
## Selic.Over 0.08221 0.08265 0.09532      1
#Função Impulso-Resposta
SVAR_MG_irf_pa <- irf(SVAR_MG_pa, impulse = "Selic.Over", response = "g_MG", n.ahead=20, ortho = TRUE)
plot(SVAR_MG_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_MG_pe <- VARselect(base_SVAR_final[107:261,c(1,7,15,16)],lag.max=10, type="both")
lagselect_MG_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      1      4

lagselect_MG_pe$criteria

##           1           2           3           4           5
## AIC(n) -3.050017e+01 -3.075930e+01 -3.077883e+01 -3.079478e+01 -3.069062e+01
## HQ(n)  -3.029997e+01 -3.042563e+01 -3.031169e+01 -3.019417e+01 -2.995655e+01
## SC(n)  -3.000747e+01 -2.993814e+01 -2.962919e+01 -2.931668e+01 -2.888405e+01
## FPE(n)  5.675802e-14 4.383140e-14 4.304998e-14 4.248485e-14 4.735176e-14
##           6           7           8           9          10
## AIC(n) -3.068574e+01 -3.055348e+01 -3.047593e+01 -3.034331e+01 -3.026518e+01
## HQ(n)  -2.981820e+01 -2.955248e+01 -2.934145e+01 -2.907537e+01 -2.886377e+01
## SC(n)  -2.855070e+01 -2.808998e+01 -2.768395e+01 -2.722287e+01 -2.681627e+01
## FPE(n)  4.788088e-14 5.512122e-14 6.024448e-14 6.979394e-14 7.685151e-14
```

#Estimando o VAR Reduzido

```
VAR_MG_pe <- VAR(base_SVAR_final[107:261,c(1,7,15,16)], p = lagselect_MG_pe$selection[1], season = NULL)
VAR_MG_pe
```

##

VAR Estimation Results:

=====

##

Estimated coefficients for equation g_BR:

=====

Call:

g_BR = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + g

##

g_BR.l1	g_MG.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
---------	---------	---------	---------------	---------

-0.287149983	0.080667929	0.527965856	-0.002603362	-0.233669478
--------------	-------------	-------------	--------------	--------------

g_MG.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_MG.l3
---------	---------	---------------	---------	---------

-0.010127244	-0.307697212	0.022661657	-0.223689129	0.034798945
--------------	--------------	-------------	--------------	-------------

IPCA.l3	Selic.Over.l3	g_BR.l4	g_MG.l4	IPCA.l4
---------	---------------	---------	---------	---------

-0.246401672	-0.003940712	-0.128244415	0.035756131	-1.761851653
--------------	--------------	--------------	-------------	--------------

Selic.Over.l4	const
---------------	-------

-0.021457690	0.010547232
--------------	-------------

##

##

Estimated coefficients for equation g_MG:

=====

Call:

g_MG = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + g

##

g_BR.l1	g_MG.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
---------	---------	---------	---------------	---------

0.032125138	-0.307825292	0.274214792	-0.042055380	0.122431132
-------------	--------------	-------------	--------------	-------------

g_MG.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_MG.l3
---------	---------	---------------	---------	---------

-0.296215092	-0.557636640	0.007068115	-0.009765088	-0.173789023
--------------	--------------	-------------	--------------	--------------

IPCA.l3	Selic.Over.l3	g_BR.l4	g_MG.l4	IPCA.l4
---------	---------------	---------	---------	---------

-0.240055004	0.026484745	-0.077065205	-0.011541343	-0.041412792
--------------	-------------	--------------	--------------	--------------

Selic.Over.l4	const
---------------	-------

-0.001276487	0.008012965
--------------	-------------

##

##

Estimated coefficients for equation IPCA:

=====

Call:

IPCA = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l2 + g

##

g_BR.l1	g_MG.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
---------	---------	---------	---------------	---------

-0.006319149	0.003493220	0.570836946	0.001009775	-0.004437318
--------------	-------------	-------------	-------------	--------------

g_MG.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_MG.l3
---------	---------	---------------	---------	---------

0.006466788	0.043091787	0.006464847	0.006086877	0.001161908
-------------	-------------	-------------	-------------	-------------

IPCA.l3	Selic.Over.l3	g_BR.l4	g_MG.l4	IPCA.l4
---------	---------------	---------	---------	---------

0.023357644	-0.003253647	0.010745221	-0.006789763	-0.155014853
-------------	--------------	-------------	--------------	--------------

Selic.Over.l4	const
---------------	-------

-0.004648354	0.002767264
--------------	-------------

##

##

Estimated coefficients for equation Selic.Over:

```
## =====
## Call:
## Selic.Over = g_BR.l1 + g_MG.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_MG.l2 + IPCA.l2 + Selic.Over.l3
##
##      g_BR.l1      g_MG.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.08406118  0.24186506  2.05726382  0.32562265 -0.08938701
##      g_MG.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_MG.l3
##  0.11325672  2.12179714  0.63678826 -0.33107215  0.32576550
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_MG.l4      IPCA.l4
##  4.06646482  0.32441335 -0.13684405  0.13942334  2.97673341
## Selic.Over.l4      const
## -0.32064756 -0.02863560
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_MG_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_MG_pe
## Chi-squared = 79.788, df = 80, p-value = 0.4857
```

```
arch.test(VAR_MG_pe, multivariate.only = FALSE)
```

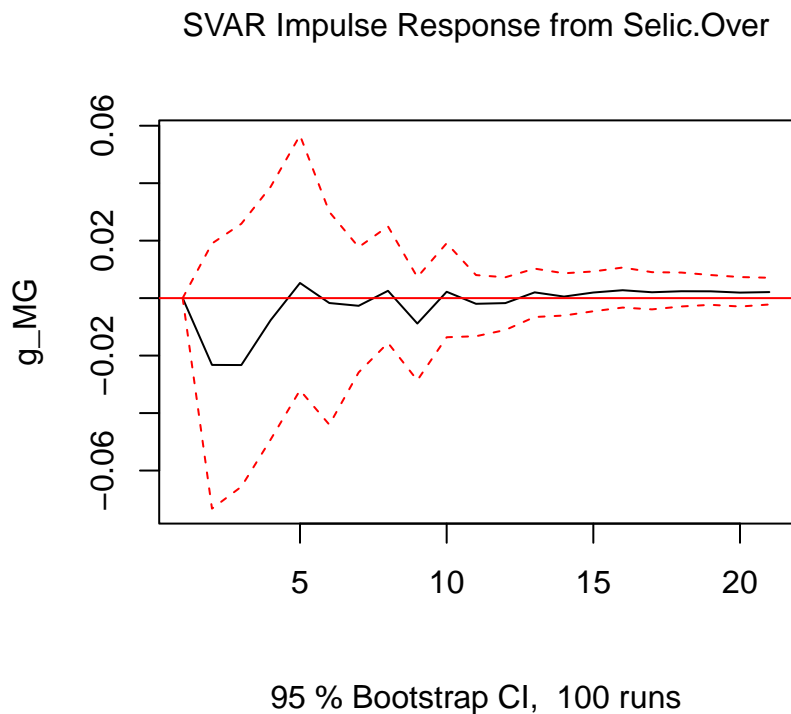
```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.4746, df = 16, p-value = 1
##
##
## $g_MG
##
## ARCH test (univariate)
##
## data: Residual of g_MG equation
## Chi-squared = 12.789, df = 16, p-value = 0.6881
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 15.309, df = 16, p-value = 0.5022
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 33.255, df = 16, p-value = 0.006835
```

```
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_MG_pe
## Chi-squared = 573.04, df = 500, p-value = 0.01298

#Estimando o VAR Estrutural
SVAR_MG_pe <- SVAR(VAR_MG, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_MG_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR   g_MG   IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_MG      0.04538 1.00000 0.00000         0
## IPCA      0.08830 0.08432 1.00000         0
## Selic.Over 0.08221 0.08265 0.09532         1

#Função Impulso-Resposta
SVAR_MG_irf_pe <- irf(SVAR_MG_pe, impulse = "Selic.Over", response = "g_MG", n.ahead=20, ortho = TRUE)
plot(SVAR_MG_irf_pe)
```



```
#####
#####PARÁ#####
#####
#Selecionando o Lag
lagselect_PA <- VARselect(base_SVAR_final[,c(1,8,15,16)],lag.max=10, type="both")
lagselect_PA$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_PA$criteria

##              1              2              3              4              5
## AIC(n) -2.934605e+01 -2.962874e+01 -2.967198e+01 -2.974708e+01 -2.972076e+01
## HQ(n)  -2.921040e+01 -2.940265e+01 -2.935545e+01 -2.934011e+01 -2.922335e+01
## SC(n)  -2.900896e+01 -2.906691e+01 -2.888542e+01 -2.873580e+01 -2.848474e+01
## FPE(n)  1.799646e-13  1.356675e-13  1.299645e-13  1.206244e-13  1.239431e-13
##              6              7              8              9             10
## AIC(n) -2.970480e+01 -2.963086e+01 -2.958638e+01 -2.957475e+01 -2.957826e+01
## HQ(n)  -2.911696e+01 -2.895259e+01 -2.881766e+01 -2.871560e+01 -2.862867e+01
## SC(n)  -2.824405e+01 -2.794539e+01 -2.767617e+01 -2.743982e+01 -2.721860e+01
## FPE(n)  1.260856e-13  1.359795e-13  1.424657e-13  1.445182e-13  1.444918e-13

#Estimando o VAR Reduzido
VAR_PA <- VAR(base_SVAR_final[,c(1,8,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_PA

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -7.089534e-02  5.396737e-02  2.624168e-01 -1.157779e-02 -1.764230e-01
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
## -3.889560e-02 -1.270909e-01 -8.003035e-05 -8.150463e-02  3.567709e-02
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
##  5.213703e-02 -1.088466e-02 -5.560210e-02 -2.498206e-02 -1.002077e+00
## Selic.Over.l4      g_BR.l5      g_PA.l5      IPCA.l5 Selic.Over.l5
##  9.695831e-03  7.419211e-03 -1.727783e-04  2.614368e-01  1.467607e-02
##      const
##  1.344332e-03
##
##
## Estimated coefficients for equation g_PA:
## =====
## Call:
## g_PA = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.2072978317 -0.4812785571 -0.2091583638 -0.0359871891 -0.0083231214
```

```

##      g_PA.12      IPCA.12 Selic.Over.12      g_BR.13      g_PA.13
## -0.3372102122  0.0931379411  0.0015954355  0.2012123938 -0.3393622207
##      IPCA.13 Selic.Over.13      g_BR.14      g_PA.14      IPCA.14
## -0.9650298248  0.0464177156  0.0005224269 -0.2087379263  1.4407332936
## Selic.Over.14      g_BR.15      g_PA.15      IPCA.15 Selic.Over.15
##  0.0289168202 -0.0552544720 -0.1122495322 -0.6046074104 -0.0245019692
##      const
## -0.0062634215
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.11 + g_PA.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_PA.12 + IPCA.12 + Selic.Over.12 + g
##
##      g_BR.11      g_PA.11      IPCA.11 Selic.Over.11      g_BR.12
## -0.0061088117 -0.0028368008  0.6551500583 -0.0003153420 -0.0014156640
##      g_PA.12      IPCA.12 Selic.Over.12      g_BR.13      g_PA.13
##  0.0001155601  0.0122856159  0.0043590514  0.0051455742 -0.0016890722
##      IPCA.13 Selic.Over.13      g_BR.14      g_PA.14      IPCA.14
##  0.0172855250  0.0010382257  0.0039659488  0.0021687949 -0.0129264041
## Selic.Over.14      g_BR.15      g_PA.15      IPCA.15 Selic.Over.15
##  0.0021507979 -0.0086939355  0.0003876094 -0.1619842215 -0.0066514826
##      const
##  0.0019496621
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_PA.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_PA.12 + IPCA.12 + Selic.Over.
##
##      g_BR.11      g_PA.11      IPCA.11 Selic.Over.11      g_BR.12
##  0.120849362  0.028172805  2.549634828  0.331466980  0.033429468
##      g_PA.12      IPCA.12 Selic.Over.12      g_BR.13      g_PA.13
##  0.084276369  2.647851342  0.565391685  0.096923551  0.002512714
##      IPCA.13 Selic.Over.13      g_BR.14      g_PA.14      IPCA.14
##  3.408140353  0.384946680  0.095938764  0.027837992 -0.577738298
## Selic.Over.14      g_BR.15      g_PA.15      IPCA.15 Selic.Over.15
## -0.266579399  0.063723946  0.017960244  3.502202251 -0.073202725
##      const
## -0.007643573

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_PA, type="BG")
```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PA
## Chi-squared = 95.935, df = 80, p-value = 0.1081

```



```

arch.test(VAR_PA, multivariate.only = FALSE)

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 5.1579, df = 16, p-value = 0.9949
##
##
## $g_PA
##
## ARCH test (univariate)
##
## data: Residual of g_PA equation
## Chi-squared = 35.234, df = 16, p-value = 0.003692
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.509, df = 16, p-value = 0.295
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 56.449, df = 16, p-value = 2.051e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PA
## Chi-squared = 733.47, df = 500, p-value = 4.238e-11

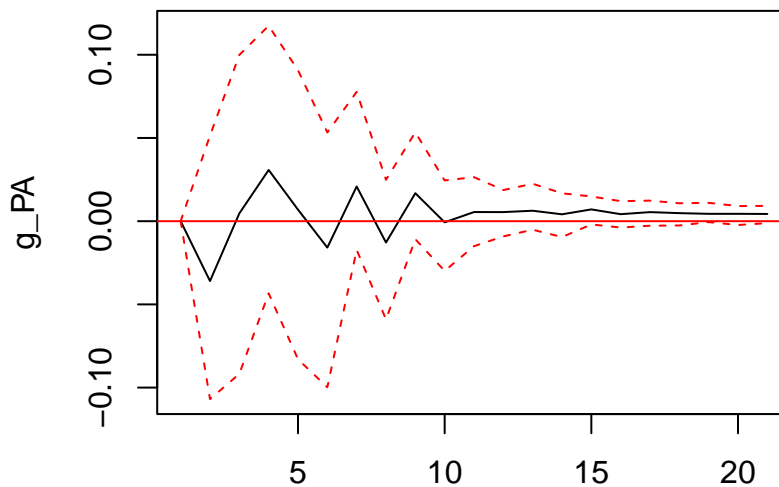
#Estimando o VAR Estrutural
SVAR_PA <- SVAR(VAR_PA, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_PA

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PA    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_PA      0.07636  1.00000  0.00000      0
## IPCA      0.08969  0.07076  1.00000      0

```

```
## Selic.Over 0.07907 0.06113 0.09357 1
#Função Impulso-Resposta
SVAR_PA_irf <- irf(SVAR_PA, impulse = "Selic.Over", response = "g_PA", n.ahead=20, ortho = TRUE)
plot(SVAR_PA_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
#Selecionando o Lag
lagselect_PA_pa <- VARselect(base_SVAR_final[1:106,c(1,8,15,16)],lag.max=10, type="both")
lagselect_PA_pa$selection
```

```
## AIC(n) HQ(n) SC(n) FPE(n)
##      5      1      1      5
```

```
lagselect_PA_pa$criteria
```

	1	2	3	4	5
## AIC(n)	-3.148402e+01	-3.148906e+01	-3.143577e+01	-3.145904e+01	-3.154018e+01
## HQ(n)	-3.122489e+01	-3.105716e+01	-3.083112e+01	-3.068163e+01	-3.059000e+01
## SC(n)	-3.084294e+01	-3.042058e+01	-2.993990e+01	-2.953578e+01	-2.918952e+01
## FPE(n)	2.122979e-14	2.117357e-14	2.245207e-14	2.214675e-14	2.073337e-14
	6	7	8	9	10
## AIC(n)	-3.144627e+01	-3.145068e+01	-3.132617e+01	-3.122753e+01	-3.126373e+01
## HQ(n)	-3.032334e+01	-3.015499e+01	-2.985772e+01	-2.958632e+01	-2.944977e+01
## SC(n)	-2.866823e+01	-2.824524e+01	-2.769334e+01	-2.716731e+01	-2.677612e+01

```
## FPE(n) 2.328801e-14 2.391738e-14 2.824378e-14 3.292287e-14 3.406002e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_PA_pa <- VAR(base_SVAR_final[1:106,c(1,8,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_PA_pa
```

```
##
```

```
## VAR Estimation Results:
```

```
## =====
```

```
##
```

```
## Estimated coefficients for equation g_BR:
```

```
## =====
```

```
## Call:
```

```
## g_BR = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
```

```
## 0.2393770513 0.0486380187 0.0399869732 -0.0255721558 -0.0624886900
```

```
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
```

```
## -0.0908283755 -0.0244877880 -0.0090514094 0.0463271238 -0.0677994607
```

```
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
```

```
## 0.5266543852 -0.0004905052 0.0743393146 -0.0774858442 -0.1410536932
```

```
## Selic.Over.l4      g_BR.l5      g_PA.l5      IPCA.l5 Selic.Over.l5
```

```
## 0.0207753716 -0.0341941703 -0.0701340163 0.1396104647 0.0112644439
```

```
##      const
```

```
## 0.0033516995
```

```
##
```

```
##
```

```
## Estimated coefficients for equation g_PA:
```

```
## =====
```

```
## Call:
```

```
## g_PA = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
```

```
## 0.366398002 -0.491849628 0.182355811 -0.043888771 -0.091232795
```

```
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
```

```
## -0.277454605 1.881583252 -0.021061325 0.008859302 -0.297923122
```

```
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
```

```
## -1.910993312 0.068475609 0.522920827 -0.145887480 1.968505757
```

```
## Selic.Over.l4      g_BR.l5      g_PA.l5      IPCA.l5 Selic.Over.l5
```

```
## 0.013369438 0.503513336 -0.067343372 -0.574878389 -0.008976589
```

```
##      const
```

```
## -0.010251707
```

```
##
```

```
##
```

```
## Estimated coefficients for equation IPCA:
```

```
## =====
```

```
## Call:
```

```
## IPCA = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
```

```
##
```

```
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
```

```
## 0.0114465580 -0.0021262237 0.8225949804 -0.0014083083 -0.0071232394
```

```
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
```

```
## -0.0316741984 0.0168763688 0.0033663995 0.0284674523 -0.0065637041
```

```
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
```

```
## -0.1621640582 0.0022547414 -0.0042347741 -0.0002989588 0.2321827123
```

```
## Selic.Over.14      g_BR.15      g_PA.15      IPCA.15 Selic.Over.15
## 0.0052068377 -0.0002803578 -0.0024115391 -0.3228875435 -0.0079083880
##      const
## 0.0005973522
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_PA.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_PA.12 + IPCA.12 + Selic.Over.13
##
##      g_BR.11      g_PA.11      IPCA.11 Selic.Over.11      g_BR.12
## 0.31487374 -0.36523298 3.38061254 0.34715761 -0.10380490
##      g_PA.12      IPCA.12 Selic.Over.12      g_BR.13      g_PA.13
## 0.03360377 4.97128747 0.47348033 0.22532586 -0.11866024
##      IPCA.13 Selic.Over.13      g_BR.14      g_PA.14      IPCA.14
## 3.20186074 0.38409204 0.20835814 0.30127133 -4.70444136
## Selic.Over.14      g_BR.15      g_PA.15      IPCA.15 Selic.Over.15
## -0.26595928 0.53093809 -0.01634056 7.56209971 -0.05061457
##      const
## 0.04707996
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_PA_pa, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PA_pa
## Chi-squared = 102.79, df = 80, p-value = 0.04399
```

```
arch.test(VAR_PA_pa, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 12.042, df = 16, p-value = 0.7411
##
##
## $g_PA
##
## ARCH test (univariate)
##
## data: Residual of g_PA equation
## Chi-squared = 19.89, df = 16, p-value = 0.2252
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
```

```
## Chi-squared = 11.829, df = 16, p-value = 0.7556
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 20.644, df = 16, p-value = 0.1926
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PA_pa
## Chi-squared = 565.67, df = 500, p-value = 0.02205
```

```
#Estimando o VAR Estrutural
```

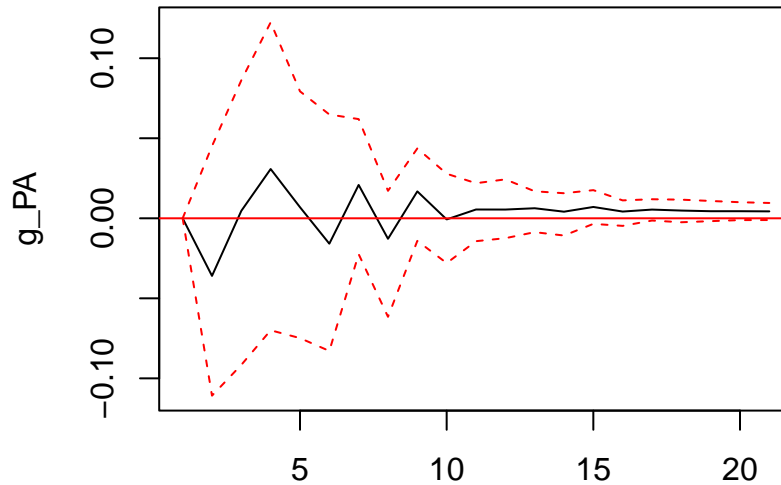
```
SVAR_PA_pa <- SVAR(VAR_PA, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_PA_pa
```

```
##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PA    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_PA      0.07636  1.00000  0.00000         0
## IPCA      0.08969  0.07076  1.00000         0
## Selic.Over 0.07907  0.06113  0.09357         1
```

```
#Função Impulso-Resposta
```

```
SVAR_PA_irf_pa <- irf(SVAR_PA_pa, impulse = "Selic.Over", response = "g_PA", n.ahead=20, ortho = TRUE)
plot(SVAR_PA_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#
```

```
lagselect_PA_pe <- VARselect(base_SVAR_final[107:261,c(1,8,15,16)],lag.max=10, type="both")
lagselect_PA_pe$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      1      4
```

```
lagselect_PA_pe$criteria
```

```
##           1           2           3           4           5
## AIC(n) -2.859921e+01 -2.885297e+01 -2.890058e+01 -2.893791e+01 -2.884384e+01
## HQ(n)  -2.839901e+01 -2.851930e+01 -2.843345e+01 -2.833731e+01 -2.810977e+01
## SC(n)  -2.810651e+01 -2.803180e+01 -2.775095e+01 -2.745981e+01 -2.703727e+01
## FPE(n)  3.798434e-13  2.949146e-13  2.816332e-13  2.720584e-13  3.001806e-13
##           6           7           8           9          10
## AIC(n) -2.881700e+01 -2.872291e+01 -2.860125e+01 -2.854960e+01 -2.850612e+01
## HQ(n)  -2.794947e+01 -2.772191e+01 -2.746677e+01 -2.728166e+01 -2.710472e+01
## SC(n)  -2.668197e+01 -2.625941e+01 -2.580927e+01 -2.542916e+01 -2.505722e+01
## FPE(n)  3.102723e-13  3.438155e-13  3.927178e-13  4.195814e-13  4.462732e-13
```

```
#Estimando o VAR Reduzido
```

```
VAR_PA_pe <- VAR(base_SVAR_final[107:261,c(1,8,15,16)], p = lagselect_PA_pe$selection[1], season = NULL)
VAR_PA_pe
```

```
##
## VAR Estimation Results:
## =====
```

```

##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.199378982    0.058091214    0.282444184    0.000331353   -0.273633027
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
## -0.027828922    0.048426894    0.018921616   -0.173161032    0.043233710
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
## -0.431312232   -0.004851350   -0.121294856   -0.019595052   -1.716840661
## Selic.Over.l4      const
## -0.020685174    0.011225429
##
##
## Estimated coefficients for equation g_PA:
## =====
## Call:
## g_PA = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.234671951   -0.481793157   -0.784450708   -0.048011419   -0.016395421
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
## -0.335719077   -0.738462795    0.001705441    0.237383797   -0.315717789
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
## -0.606165035    0.016019952   -0.152007220   -0.161542319    1.607946550
## Selic.Over.l4      const
##  0.052428572   -0.008706611
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0059914836 -0.0031537589    0.5810208754    0.0005436530    0.0016042036
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3
##  0.0029637707  0.0208859065    0.0065861048    0.0044609536   -0.0007227287
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PA.l4      IPCA.l4
##  0.0398457242 -0.0031598982    0.0069232821    0.0026802199   -0.1499600213
## Selic.Over.l4      const
## -0.0044386624    0.0027466405
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PA.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PA.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_PA.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.03533968    0.06002390    2.25693249    0.32976136   -0.04939389
##      g_PA.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PA.l3

```

```
##      0.08107839      2.19321625      0.63507412      -0.08638775      0.01852346
##      IPCA.13 Selic.Over.13      g_BR.14      g_PA.14      IPCA.14
##      3.59618570      0.32423880      -0.05321357      -0.01128235      3.06340821
## Selic.Over.14      const
##      -0.32633269      -0.02609435
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_PA_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PA_pe
## Chi-squared = 91.142, df = 80, p-value = 0.1854
```

```
arch.test(VAR_PA_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.4872, df = 16, p-value = 1
##
##
## $g_PA
##
## ARCH test (univariate)
##
## data: Residual of g_PA equation
## Chi-squared = 24.108, df = 16, p-value = 0.08718
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 13.558, df = 16, p-value = 0.6316
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 34.256, df = 16, p-value = 0.005017
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PA_pe
## Chi-squared = 623.37, df = 500, p-value = 0.0001359
```



```
#Estimando o VAR Estrutural
```

```
SVAR_PA_pe <- SVAR(VAR_PA, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_PA
```

```
##
```

```
## SVAR Estimation Results:
```

```
## =====
```

```
##
```

```
##
```

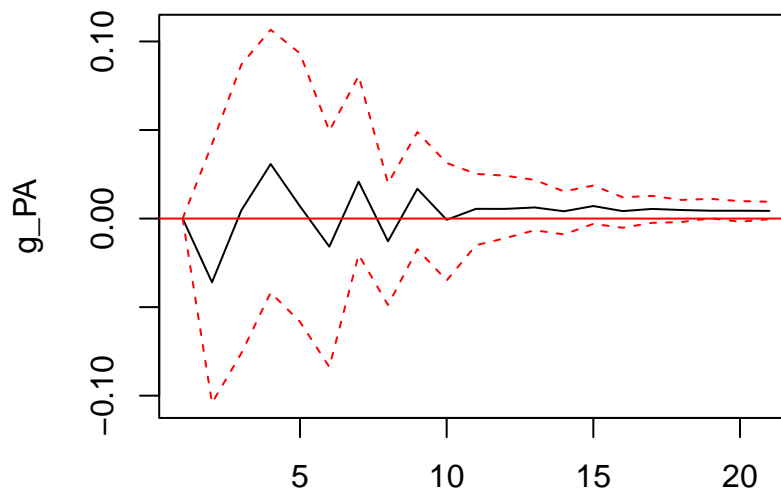
```
## Estimated A matrix:
```

```
##           g_BR    g_PA    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_PA      0.07636  1.00000  0.00000         0
## IPCA      0.08969  0.07076  1.00000         0
## Selic.Over 0.07907  0.06113  0.09357         1
```

```
#Função Impulso-Resposta
```

```
SVAR_PA_irf_pe <- irf(SVAR_PA, impulse = "Selic.Over", response = "g_PA", n.ahead=20, ortho = TRUE)
plot(SVAR_PA_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
```

```
#####PERNAMBUCO#####
```

```
#####
```

```
#Selecionando o Lag
```

```
lagselect_PE <- VARselect(base_SVAR_final[,c(1,9,15,16)],lag.max=10, type="both")
```

```
lagselect_PE$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4
```

```
lagselect_PE$criteria
```

```
##              1              2              3              4              5
## AIC(n) -3.018659e+01 -3.045924e+01 -3.046920e+01 -3.055885e+01 -3.055653e+01
## HQ(n)  -3.005093e+01 -3.023315e+01 -3.015267e+01 -3.015188e+01 -3.005912e+01
## SC(n)  -2.984949e+01 -2.989742e+01 -2.968264e+01 -2.954756e+01 -2.932051e+01
## FPE(n)  7.765126e-14  5.912787e-14  5.855944e-14  5.356600e-14  5.373440e-14
##              6              7              8              9             10
## AIC(n) -3.053017e+01 -3.046770e+01 -3.041906e+01 -3.040510e+01 -3.035254e+01
## HQ(n)  -2.994233e+01 -2.978942e+01 -2.965034e+01 -2.954595e+01 -2.940295e+01
## SC(n)  -2.906942e+01 -2.878222e+01 -2.850885e+01 -2.827017e+01 -2.799288e+01
## FPE(n)  5.523472e-14  5.888998e-14  6.195583e-14  6.299497e-14  6.661581e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_PE <- VAR(base_SVAR_final[,c(1,9,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_PE
```

```
##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.1284131056  0.0334205059  0.4299914229 -0.0097303919 -0.1845177257
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
##  0.0527283107 -0.3856028669 -0.0060278857 -0.0815784215 -0.0301516818
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PE.l4      IPCA.l4
##  0.2703650784 -0.0057965489 -0.0832182280  0.0456910786 -1.3396436314
## Selic.Over.l4      g_BR.l5      g_PE.l5      IPCA.l5 Selic.Over.l5
##  0.0063724683 -0.0167175440  0.0131323545  0.5509092948  0.0171135102
##      const
##  0.0009013208
##
##
## Estimated coefficients for equation g_PE:
## =====
## Call:
## g_PE = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.116034289 -0.152501690 -0.362670884 -0.031618602  0.030709296
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## -0.151628612  1.947688788  0.021630843  0.048010276 -0.203759177
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PE.l4      IPCA.l4
## -0.518056017  0.015795040  0.010142202 -0.115919110 -2.741457011
## Selic.Over.l4      g_BR.l5      g_PE.l5      IPCA.l5 Selic.Over.l5
## -0.016944413 -0.079587697  0.013716535  0.923391665  0.012446728
##      const
```

```

## 0.003778441
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0014212831 -0.0042831614  0.6504590907 -0.0006207946 -0.0031120882
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## -0.0003228020  0.0129069280  0.0042832874  0.0131044229 -0.0067717044
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PE.l4      IPCA.l4
##  0.0341046066  0.0014270014  0.0098841627 -0.0081667156 -0.0119338214
## Selic.Over.l4      g_BR.l5      g_PE.l5      IPCA.l5 Selic.Over.l5
##  0.0023995004 -0.0024416595 -0.0074876093 -0.1681946327 -0.0069444002
##      const
##  0.0019600817
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.262363877 -0.066712991  2.488702231  0.327765044  0.076270937
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## -0.039760966  2.465957531  0.558518612  0.345010574 -0.265396873
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PE.l4      IPCA.l4
##  4.126120255  0.394707908  0.146330342 -0.064366327 -1.032018603
## Selic.Over.l4      g_BR.l5      g_PE.l5      IPCA.l5 Selic.Over.l5
## -0.271178840  0.288698274 -0.249028829  3.838635452 -0.067554514
##      const
## -0.008653178

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```

serial.test(VAR_PE, type="BG")

```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PE
## Chi-squared = 101.69, df = 80, p-value = 0.05136

```

```

arch.test(VAR_PE, multivariate.only = FALSE)

```

```

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 4.4084, df = 16, p-value = 0.998
##

```

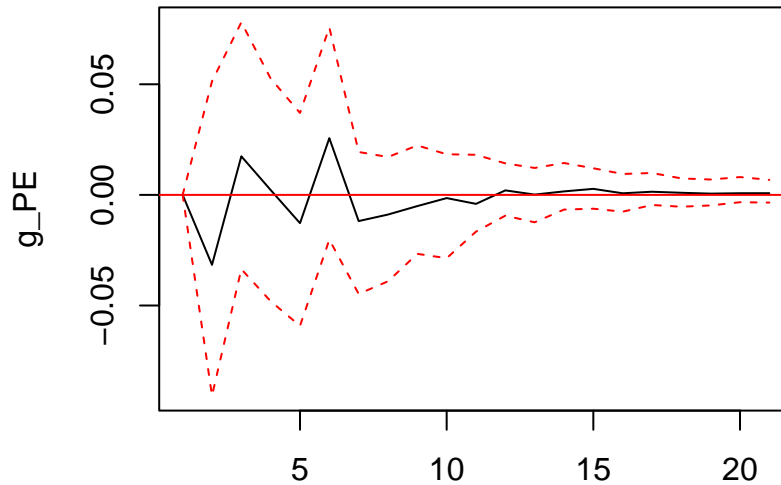
```

##
## $g_PE
##
## ARCH test (univariate)
##
## data: Residual of g_PE equation
## Chi-squared = 18.425, df = 16, p-value = 0.2996
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 22.714, df = 16, p-value = 0.1216
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 58.422, df = 16, p-value = 9.631e-07
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PE
## Chi-squared = 621.57, df = 500, p-value = 0.0001645
##
#Estimando o VAR Estrutural
SVAR_PE <- SVAR(VAR_PE, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_PE

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PE    IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_PE      0.05066 1.00000 0.00000         0
## IPCA      0.08814 0.07816 1.00000         0
## Selic.Over 0.07911 0.04977 0.09272         1
##
#Função Impulso-Resposta
SVAR_PE_irf <- irf(SVAR_PE, impulse = "Selic.Over", response = "g_PE", n.ahead=20, ortho = TRUE)
plot(SVAR_PE_irf)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

###Comparando Diferentes Períodos###

#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#

#Selecionando o Lag

```
lagselect_PE_pa <- VARselect(base_SVAR_final[1:106,c(1,9,15,16)],lag.max=10, type="both")
lagselect_PE_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      9      1      1      5
```

```
lagselect_PE_pa$criteria
```

```
##          1          2          3          4          5
## AIC(n) -3.159450e+01 -3.174551e+01 -3.170849e+01 -3.174826e+01 -3.182212e+01
## HQ(n)  -3.133536e+01 -3.131361e+01 -3.110384e+01 -3.097085e+01 -3.087195e+01
## SC(n)  -3.095341e+01 -3.067703e+01 -3.021262e+01 -2.982500e+01 -2.947147e+01
## FPE(n)  1.900935e-14  1.638395e-14  1.709289e-14  1.658455e-14  1.563956e-14
##          6          7          8          9         10
## AIC(n) -3.182748e+01 -3.179637e+01 -3.176336e+01 -3.189849e+01 -3.185799e+01
## HQ(n)  -3.070455e+01 -3.050068e+01 -3.029491e+01 -3.025728e+01 -3.004403e+01
## SC(n)  -2.904944e+01 -2.859093e+01 -2.813053e+01 -2.783827e+01 -2.737038e+01
## FPE(n)  1.590654e-14  1.692712e-14  1.824123e-14  1.683080e-14  1.880014e-14
```

#Estimando o VAR Reduzido

```
VAR_PE_pa <- VAR(base_SVAR_final[1:106,c(1,9,15,16)], p = 3, season = NULL, exog = NULL, type = "const")
VAR_PE_pa
```

```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.251721886 0.033233885 0.170761219 -0.022620519 -0.027826867
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## -0.013677120 -0.232730498 0.002772174 -0.015667082 -0.044270653
##      IPCA.l3 Selic.Over.l3      const
## 0.249219519 0.016118540 0.005358368
##
##
## Estimated coefficients for equation g_PE:
## =====
## Call:
## g_PE = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.232207223 -0.195653952 -0.068804387 -0.025705662 0.617437004
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## -0.270868965 1.455937839 0.009343999 0.014332835 -0.052602565
##      IPCA.l3 Selic.Over.l3      const
## -2.025069034 0.018601008 0.003054977
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.0049776908 -0.0131282926 0.7614830769 0.0022039261 -0.0307961823
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## 0.0018699060 -0.0081621664 -0.0009878786 0.0268661186 0.0058998520
##      IPCA.l3 Selic.Over.l3      const
## -0.0732344039 -0.0005177777 0.0008620950
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.09757821 0.04907219 3.80727938 0.30906862 -0.17528118
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
## -0.47521754 4.53875371 0.39823725 0.44044065 -0.02839344
##      IPCA.l3 Selic.Over.l3      const
## 5.19124399 0.19714347 0.03267045

```

```
#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
```

```
serial.test(VAR_PE_pa, type="BG")
```

```
##  
## Breusch-Godfrey LM test  
##  
## data: Residuals of VAR object VAR_PE_pa  
## Chi-squared = 115.06, df = 80, p-value = 0.006253
```

```
arch.test(VAR_PE_pa, multivariate.only = FALSE)
```

```
## $g_BR  
##  
## ARCH test (univariate)  
##  
## data: Residual of g_BR equation  
## Chi-squared = 11.923, df = 16, p-value = 0.7493  
##
```

```
##  
## $g_PE  
##  
## ARCH test (univariate)  
##  
## data: Residual of g_PE equation  
## Chi-squared = 18.756, df = 16, p-value = 0.2815  
##
```

```
##  
## $IPCA  
##  
## ARCH test (univariate)  
##  
## data: Residual of IPCA equation  
## Chi-squared = 14.405, df = 16, p-value = 0.5686  
##
```

```
##  
## $Selic.Over  
##  
## ARCH test (univariate)  
##  
## data: Residual of Selic.Over equation  
## Chi-squared = 23.464, df = 16, p-value = 0.1019  
##
```

```
##  
##  
## ARCH (multivariate)  
##  
## data: Residuals of VAR object VAR_PE_pa  
## Chi-squared = 545.09, df = 500, p-value = 0.07985
```

```
#Estimando o VAR Estrutural
```

```
SVAR_PE_pa <- SVAR(VAR_PE_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")  
SVAR_PE_pa
```

```
##
```

```
## SVAR Estimation Results:
```

```
## =====
```

```
##
```

```
##
```

```
## Estimated A matrix:
```

```
##           g_BR    g_PE    IPCA Selic.Over
```

```
## g_BR      1.00000 0.00000 0.00000      0
```

```
## g_PE      0.08597 1.00000 0.00000      0
```

```
## IPCA      0.09520 0.09004 1.00000      0
```

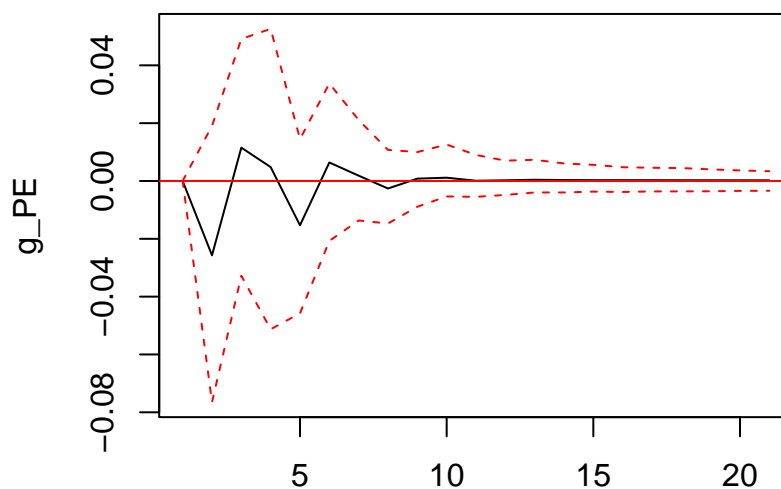
```
## Selic.Over 0.09530 0.04662 0.09136      1
```

```
#Função Impulso-Resposta
```

```
SVAR_PE_irf_pa <- irf(SVAR_PE_pa, impulse = "Selic.Over", response = "g_PE", n.ahead=20, ortho = TRUE)
```

```
plot(SVAR_PE_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#
```

```
lagselect_PE_pe <- VARselect(base_SVAR_final[107:261,c(1,9,15,16)],lag.max=10, type="both")
```

```
lagselect_PE_pe$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
```

```
##      4      2      1      4
```

```
lagselect_PE_pe$criteria
```

```
##           1           2           3           4           5
```

```
## AIC(n) -2.966712e+01 -2.988362e+01 -2.993014e+01 -3.000555e+01 -2.992359e+01
```

```
## HQ(n)  -2.946692e+01 -2.954996e+01 -2.946300e+01 -2.940495e+01 -2.918952e+01
```



```

## SC(n) -2.917442e+01 -2.906246e+01 -2.878050e+01 -2.852745e+01 -2.811702e+01
## FPE(n) 1.305618e-13 1.052175e-13 1.005901e-13 9.353853e-14 1.019656e-13
## 6 7 8 9 10
## AIC(n) -2.993758e+01 -2.989319e+01 -2.984894e+01 -2.975866e+01 -2.968033e+01
## HQ(n) -2.907005e+01 -2.889219e+01 -2.871447e+01 -2.849072e+01 -2.827892e+01
## SC(n) -2.780255e+01 -2.742969e+01 -2.705697e+01 -2.663823e+01 -2.623142e+01
## FPE(n) 1.011768e-13 1.066792e-13 1.127752e-13 1.252347e-13 1.379271e-13

#Estimando o VAR Reduzido
VAR_PE_pe <- VAR(base_SVAR_final[107:261,c(1,9,15,16)], p = lagselect_PE_pe$selection[1], season = NULL)
VAR_PE_pe

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
## g_BR.l1 g_PE.l1 IPCA.l1 Selic.Over.l1 g_BR.l2
## -0.303807485 0.075709840 0.446751206 0.006135358 -0.324342904
## g_PE.l2 IPCA.l2 Selic.Over.l2 g_BR.l3 g_PE.l3
## 0.088027462 -0.365984215 0.012207724 -0.186682588 -0.010399024
## IPCA.l3 Selic.Over.l3 g_BR.l4 g_PE.l4 IPCA.l4
## -0.258725266 0.002957653 -0.187508422 0.092879252 -1.978411832
## Selic.Over.l4 const
## -0.026480430 0.011932607
##
##
## Estimated coefficients for equation g_PE:
## =====
## Call:
## g_PE = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
## g_BR.l1 g_PE.l1 IPCA.l1 Selic.Over.l1 g_BR.l2
## -0.238471200 -0.103326838 -0.646914284 -0.010583233 -0.220370068
## g_PE.l2 IPCA.l2 Selic.Over.l2 g_BR.l3 g_PE.l3
## -0.050518852 2.262827346 0.010070768 -0.017990069 -0.235365143
## IPCA.l3 Selic.Over.l3 g_BR.l4 g_PE.l4 IPCA.l4
## 0.086944494 0.003401182 -0.126589608 -0.047730551 -3.626541707
## Selic.Over.l4 const
## -0.009763885 0.013324243
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l2 + g
##
## g_BR.l1 g_PE.l1 IPCA.l1 Selic.Over.l1 g_BR.l2
## -0.004577187 0.001994150 0.564886128 0.001603556 -0.002923916
## g_PE.l2 IPCA.l2 Selic.Over.l2 g_BR.l3 g_PE.l3
## 0.005820527 0.031972289 0.005719808 0.016960262 -0.010844246

```

```
##      IPCA.13 Selic.Over.13      g_BR.14      g_PE.14      IPCA.14
##      0.056359539 -0.003005818  0.007860075 -0.002139233 -0.174075261
## Selic.Over.14      const
##      -0.004722259  0.002774125
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PE.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PE.l2 + IPCA.l2 + Selic.Over.l3
##
##      g_BR.l1      g_PE.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      0.22292688 -0.18296588  2.41435666  0.35740147 -0.11498468
##      g_PE.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PE.l3
##      0.11415618  1.44856427  0.60621204  0.04604560 -0.14661668
##      IPCA.13 Selic.Over.13      g_BR.14      g_PE.14      IPCA.14
##      4.70433165  0.32236610 -0.04083411 -0.02734451  2.36604055
## Selic.Over.14      const
##      -0.32178095 -0.02590087
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_PE_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PE_pe
## Chi-squared = 101.2, df = 80, p-value = 0.05487
```

```
arch.test(VAR_PE_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.1251, df = 16, p-value = 1
##
##
## $g_PE
##
## ARCH test (univariate)
##
## data: Residual of g_PE equation
## Chi-squared = 8.1258, df = 16, p-value = 0.945
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 13.971, df = 16, p-value = 0.6009
##
```

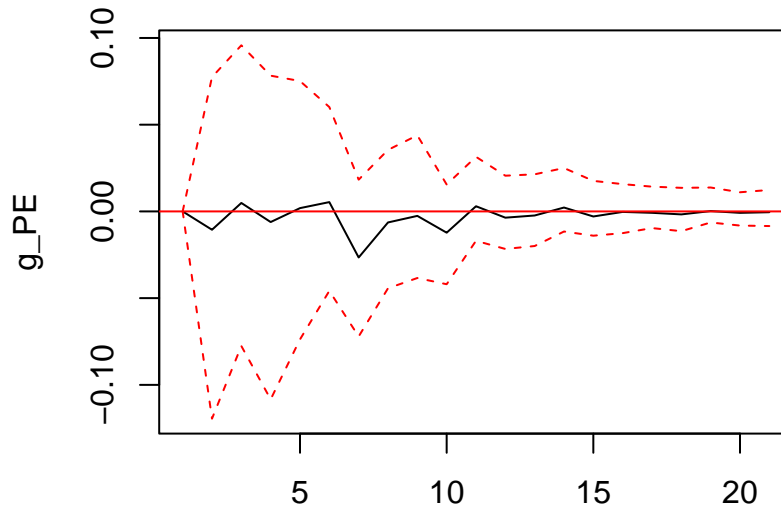
```
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 31.864, df = 16, p-value = 0.01041
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PE_pe
## Chi-squared = 586.68, df = 500, p-value = 0.004425

#Estimando o VAR Estrutural
SVAR_PE_pe <- SVAR(VAR_PE_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_PE_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PE    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_PE      0.03319  1.00000  0.00000         0
## IPCA      0.08536  0.07220  1.00000         0
## Selic.Over 0.07600  0.03782  0.09375         1

#Função Impulso-Resposta
SVAR_PE_irf_pe <- irf(SVAR_PE_pe, impulse = "Selic.Over", response = "g_PE", n.ahead=20, ortho = TRUE)
plot(SVAR_PE_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####PARANÁ#####
#####
```

#Seleccionando o Lag

```
lagselect_PR <- VARselect(base_SVAR_final[,c(1,10,15,16)],lag.max=10, type="both")
lagselect_PR$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4
```

```
lagselect_PR$criteria
```

```
##          1          2          3          4          5
## AIC(n) -2.993834e+01 -3.025245e+01 -3.028591e+01 -3.035540e+01 -3.034793e+01
## HQ(n)  -2.980268e+01 -3.002636e+01 -2.996938e+01 -2.994843e+01 -2.985053e+01
## SC(n)  -2.960124e+01 -2.969062e+01 -2.949936e+01 -2.934411e+01 -2.911192e+01
## FPE(n)  9.953151e-14  7.271120e-14  7.033899e-14  6.565185e-14  6.619769e-14
##          6          7          8          9         10
## AIC(n) -3.030269e+01 -3.023454e+01 -3.018146e+01 -3.016588e+01 -3.014876e+01
## HQ(n)  -2.971485e+01 -2.955626e+01 -2.941275e+01 -2.930673e+01 -2.919918e+01
## SC(n)  -2.884195e+01 -2.854906e+01 -2.827126e+01 -2.803095e+01 -2.778910e+01
## FPE(n)  6.934308e-14  7.435360e-14  7.857218e-14  8.001981e-14  8.167277e-14
```

#Estimando o VAR Reduzido

```
VAR_PR <- VAR(base_SVAR_final[,c(1,10,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_PR
```

```
##
```

```

## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.071886310 -0.028902364   0.386307991 -0.007379030 -0.082075768
##      g_PR.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PR.l3
## -0.047616917 -0.392023931 -0.004074030 -0.132100207   0.007043595
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PR.l4      IPCA.l4
##   0.287241254 -0.012250825 -0.105813389   0.049468032 -0.979937676
## Selic.Over.l4      g_BR.l5      g_PR.l5      IPCA.l5 Selic.Over.l5
##   0.006824974 -0.043344669   0.037500425   0.109817181   0.019216131
##      const
##   0.001179077
##
##
## Estimated coefficients for equation g_PR:
## =====
## Call:
## g_PR = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.022186066 -0.400376679   0.209597804 -0.001762904   0.051168279
##      g_PR.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PR.l3
## -0.278284432   2.774930070 -0.050350549   0.242197922 -0.200571728
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PR.l4      IPCA.l4
## -0.572624463   0.026223741   0.027599625 -0.093884638 -3.756331775
## Selic.Over.l4      g_BR.l5      g_PR.l5      IPCA.l5 Selic.Over.l5
##   0.031236602   0.067802587 -0.020801236   1.541294882 -0.008513584
##      const
##   0.005498779
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0066608569  0.0009384339   0.6522096218 -0.0005623772 -0.0106820538
##      g_PR.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PR.l3
##   0.0051335453  0.0269593124   0.0044224580   0.0086093987   0.0006159340
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PR.l4      IPCA.l4
##   0.0010837506  0.0011142527   0.0027274793   0.0006882098 -0.0206994336
## Selic.Over.l4      g_BR.l5      g_PR.l5      IPCA.l5 Selic.Over.l5
##   0.0025485507 -0.0149262223   0.0042607020 -0.1506092737 -0.0069464725
##      const
##   0.0019352108
##
##

```

```
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l1
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.012749394 0.093861551 2.606354747 0.326977355 0.038208689
##      g_PR.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PR.l3
## 0.033471185 2.775746664 0.563955889 0.175673150 -0.024165234
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PR.l4      IPCA.l4
## 2.886266583 0.388714482 0.064094426 0.005539290 -0.444125471
## Selic.Over.l4      g_BR.l5      g_PR.l5      IPCA.l5 Selic.Over.l5
## -0.264478304 -0.038343169 0.078898249 3.847654794 -0.072508572
##      const
## -0.008817606
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_PR, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PR
## Chi-squared = 105.96, df = 80, p-value = 0.02765
```

```
arch.test(VAR_PR, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 4.8741, df = 16, p-value = 0.9963
##
##
## $g_PR
##
## ARCH test (univariate)
##
## data: Residual of g_PR equation
## Chi-squared = 6.6008, df = 16, p-value = 0.9802
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 19.184, df = 16, p-value = 0.2592
##
##
## $Selic.Over
##
## ARCH test (univariate)
```

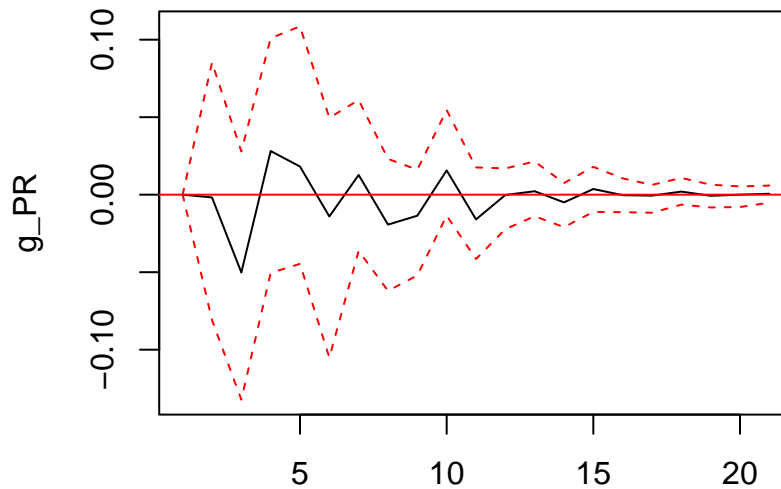
```
##
## data: Residual of Selic.Over equation
## Chi-squared = 60.547, df = 16, p-value = 4.232e-07
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PR
## Chi-squared = 664.91, df = 500, p-value = 1.006e-06

#Estimando o VAR Estrutural
SVAR_PR <- SVAR(VAR_PR, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_PR

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PR    IPCA Selic.Over
## g_BR      1.000000 0.00000 0.00000         0
## g_PR      0.009055 1.00000 0.00000         0
## IPCA      0.084790 0.06688 1.00000         0
## Selic.Over 0.075781 0.03214 0.09107         1

#Função Impulso-Resposta
SVAR_PR_irf <- irf(SVAR_PR, impulse = "Selic.Over", response = "g_PR", n.ahead=20, ortho = TRUE)
plot(SVAR_PR_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
```

```
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
```

```
#Selecionando o Lag
```

```
lagselect_PR_pa <- VARselect(base_SVAR_final[1:106,c(1,10,15,16)],lag.max=10, type="both")
lagselect_PR_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      10      1      1      2
```

```
lagselect_PR_pa$criteria
```

```
##           1           2           3           4           5
## AIC(n) -3.089257e+01 -3.095871e+01 -3.082245e+01 -3.077642e+01 -3.082884e+01
## HQ(n)  -3.063344e+01 -3.052682e+01 -3.021779e+01 -2.999901e+01 -2.987867e+01
## SC(n)  -3.025149e+01 -2.989023e+01 -2.932658e+01 -2.885316e+01 -2.847819e+01
## FPE(n)  3.835385e-14  3.598482e-14  4.145908e-14  4.382986e-14  4.222778e-14
##           6           7           8           9          10
## AIC(n) -3.076554e+01 -3.071027e+01 -3.073840e+01 -3.086437e+01 -3.096780e+01
## HQ(n)  -2.964261e+01 -2.941458e+01 -2.926995e+01 -2.922316e+01 -2.915383e+01
## SC(n)  -2.798750e+01 -2.750484e+01 -2.710557e+01 -2.680415e+01 -2.648019e+01
## FPE(n)  4.600150e-14  5.014974e-14  5.083820e-14  4.733887e-14  4.578970e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_PR_pa <- VAR(base_SVAR_final[1:106,c(1,10,15,16)], p = 2, season = NULL, exog = NULL, type = "const")
VAR_PR_pa
```



```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.432496845 -0.104198948  0.156523800 -0.014298104  0.029906798
##      g_PR.l2      IPCA.l2 Selic.Over.l2      const
## -0.071930386  0.052165657  0.009629236  0.006314384
##
##
## Estimated coefficients for equation g_PR:
## =====
## Call:
## g_PR = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  1.093159069 -0.630023086  1.275340876 -0.002973937  0.209265517
##      g_PR.l2      IPCA.l2 Selic.Over.l2      const
## -0.267796456  1.036330329 -0.019167464  0.018175710
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0060114503  0.0032862511  0.7536798487  0.0011973083 -0.0222889446
##      g_PR.l2      IPCA.l2 Selic.Over.l2      const
##  0.0033036414 -0.0535624885 -0.0006835395  0.0010311369
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.22934553 -0.19794320  3.61120376  0.44253518 -0.03784828
##      g_PR.l2      IPCA.l2 Selic.Over.l2      const
## -0.17695638  6.97525271  0.47027099  0.04163249
##
## #Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
##
## serial.test(VAR_PR_pa, type="BG")
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PR_pa

```

```

## Chi-squared = 104.63, df = 80, p-value = 0.03372
arch.test(VAR_PR_pa, multivariate.only = FALSE)

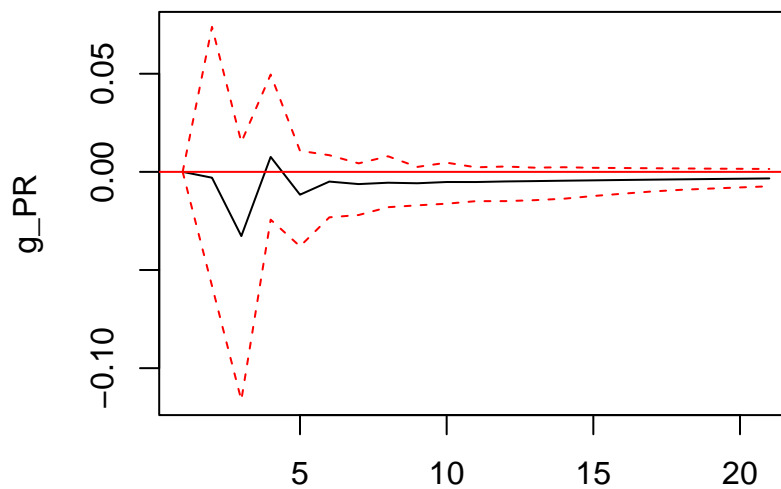
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 15.051, df = 16, p-value = 0.5209
##
##
## $g_PR
##
## ARCH test (univariate)
##
## data: Residual of g_PR equation
## Chi-squared = 17.256, df = 16, p-value = 0.3692
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 15.908, df = 16, p-value = 0.4594
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 26.828, df = 16, p-value = 0.04343
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PR_pa
## Chi-squared = 594.95, df = 500, p-value = 0.002174
#Estimando o VAR Estrutural
SVAR_PR_pa <- SVAR(VAR_PR_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod = "scoring")
SVAR_PR_pa

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PR    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_PR      0.06295  1.00000  0.00000      0

```

```
## IPCA      0.09235 0.07282 1.00000      0
## Selic.Over 0.09833 0.09924 0.09591      1
#Função Impulso-Resposta
SVAR_PR_irf_pa <- irf(SVAR_PR_pa, impulse = "Selic.Over", response = "g_PR", n.ahead=20, ortho = TRUE)
plot(SVAR_PR_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_PR_pe <- VARselect(base_SVAR_final[107:261,c(1,10,15,16)],lag.max=10, type="both")
lagselect_PR_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_PR_pe$criteria

##           1           2           3           4           5
## AIC(n) -2.964696e+01 -2.997794e+01 -3.003389e+01 -3.008811e+01 -3.006045e+01
## HQ(n)  -2.944676e+01 -2.964427e+01 -2.956676e+01 -2.948751e+01 -2.932637e+01
## SC(n)  -2.915426e+01 -2.915677e+01 -2.888426e+01 -2.861001e+01 -2.825388e+01
## FPE(n)  1.332202e-13  9.574750e-14  9.067656e-14  8.612610e-14  8.892346e-14
##           6           7           8           9          10
## AIC(n) -3.002672e+01 -2.993224e+01 -2.981272e+01 -2.973877e+01 -2.967898e+01
## HQ(n)  -2.915919e+01 -2.893124e+01 -2.867825e+01 -2.847083e+01 -2.827757e+01
## SC(n)  -2.789169e+01 -2.746874e+01 -2.702075e+01 -2.661834e+01 -2.623007e+01
## FPE(n)  9.254814e-14  1.025939e-13  1.169343e-13  1.277506e-13  1.381134e-13
```

#Estimando o VAR Reduzido

```
VAR_PR_pe <- VAR(base_SVAR_final[107:261,c(1,10,15,16)], p = lagselect_PR_pe$selection[1], season = NULL)
VAR_PR_pe
```

##

VAR Estimation Results:

=====

##

Estimated coefficients for equation g_BR:

=====

Call:

g_BR = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_PR.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.267809007	0.022152252	0.600368358	-0.002771924	-0.180008899
##	g_PR.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_PR.l3
##	-0.032448718	-0.377342252	0.020991873	-0.177184814	-0.024700715
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_PR.l4	IPCA.l4
##	-0.450300646	-0.001190937	-0.201832124	0.051660324	-1.499664124
##	Selic.Over.l4	const			
##	-0.023188639	0.010747681			

##

##

Estimated coefficients for equation g_PR:

=====

Call:

g_PR = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_PR.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.51981488	-0.21014812	-0.84887384	-0.01494466	-0.05277651
##	g_PR.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_PR.l3
##	-0.26058419	3.53103319	-0.01312118	0.06270835	-0.16504621
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_PR.l4	IPCA.l4
##	-1.12827647	0.06991644	-0.15503154	-0.04613851	-3.79370462
##	Selic.Over.l4	const			
##	-0.05185294	0.01819748			

##

##

Estimated coefficients for equation IPCA:

=====

Call:

IPCA = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_PR.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.0040666839	-0.0003980389	0.5700942031	0.0010567747	-0.0061020347
##	g_PR.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_PR.l3
##	0.0041693047	0.0436607182	0.0064031704	0.0152010031	-0.0045564469
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_PR.l4	IPCA.l4
##	0.0196212802	-0.0031480508	0.0059076346	-0.0012126268	-0.1574348589
##	Selic.Over.l4	const			
##	-0.0047509308	0.0028059698			

##

##

Estimated coefficients for equation Selic.Over:

```
## =====
## Call:
## Selic.Over = g_BR.l1 + g_PR.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_PR.l2 + IPCA.l2 + Selic.Over.l3
##
##      g_BR.l1      g_PR.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.219679132  0.187199400  2.308387067  0.321321952 -0.001815965
##      g_PR.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_PR.l3
##  0.041346026  2.455401188  0.650485880  0.005950243 -0.051997876
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_PR.l4      IPCA.l4
##  2.684986829  0.339261717  0.043147518 -0.112578212  3.320746663
## Selic.Over.l4      const
## -0.346648946 -0.025587639
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_PR_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_PR_pe
## Chi-squared = 90.727, df = 80, p-value = 0.1935
```

```
arch.test(VAR_PR_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.5059, df = 16, p-value = 1
##
##
## $g_PR
##
## ARCH test (univariate)
##
## data: Residual of g_PR equation
## Chi-squared = 3.8446, df = 16, p-value = 0.9991
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 14.428, df = 16, p-value = 0.5669
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 31.6, df = 16, p-value = 0.01127
```

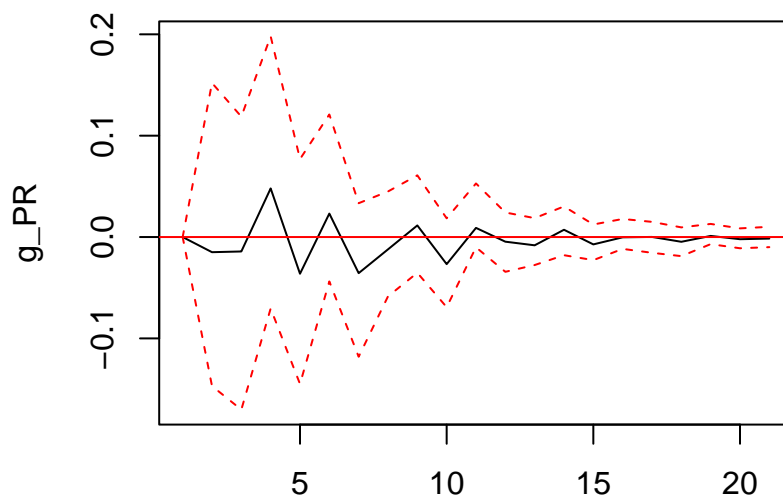
```
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_PR_pe
## Chi-squared = 637.34, df = 500, p-value = 2.914e-05

#Estimando o VAR Estrutural
SVAR_PR_pe <- SVAR(VAR_PR_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_PR_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_PR    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_PR     -0.01500  1.00000  0.00000         0
## IPCA       0.08165  0.06432  1.00000         0
## Selic.Over 0.07345  0.01138  0.09157         1

#Função Impulso-Resposta
SVAR_PR_irf_pe <- irf(SVAR_PR_pe, impulse = "Selic.Over", response = "g_PR", n.ahead=20, ortho = TRUE)
plot(SVAR_PR_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####RIO DE JANEIRO#####
#####
#Selecionando o Lag
lagselect_RJ <- VARselect(base_SVAR_final[,c(1,11,15,16)],lag.max=10, type="both")
lagselect_RJ$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_RJ$criteria

##              1              2              3              4              5
## AIC(n) -3.065990e+01 -3.100127e+01 -3.101475e+01 -3.109707e+01 -3.109650e+01
## HQ(n)  -3.052424e+01 -3.077518e+01 -3.069822e+01 -3.069010e+01 -3.059910e+01
## SC(n)  -3.032280e+01 -3.043944e+01 -3.022819e+01 -3.008578e+01 -2.986049e+01
## FPE(n)  4.837177e-14  3.438694e-14  3.393647e-14  3.127119e-14  3.131436e-14
##              6              7              8              9             10
## AIC(n) -3.105920e+01 -3.095309e+01 -3.092725e+01 -3.092652e+01 -3.089069e+01
## HQ(n)  -3.047136e+01 -3.027481e+01 -3.015853e+01 -3.006737e+01 -2.994111e+01
## SC(n)  -2.959846e+01 -2.926762e+01 -2.901704e+01 -2.879159e+01 -2.853103e+01
## FPE(n)  3.254276e-14  3.624413e-14  3.727164e-14  3.739870e-14  3.889209e-14

#Estimando o VAR Reduzido
VAR_RJ <- VAR(base_SVAR_final[,c(1,11,15,16)], p = lagselect_RJ$selection[1], season = NULL, exog = NULL)
VAR_RJ

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.048795365 -0.080643964  0.371992632 -0.013517071 -0.160997528
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RJ.l3
##  0.008681440 -0.263997407  0.001209008 -0.079513252 -0.020191278
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RJ.l4      IPCA.l4
##  0.105297308  0.001091089 -0.072737867  0.035916650 -0.963546832
## Selic.Over.l4      const
##  0.013658220  0.002003502
##
##
## Estimated coefficients for equation g_RJ:
## =====
## Call:
## g_RJ = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.221302265 -0.494147119  1.104251897 -0.015957611  0.203674671
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RJ.l3
## -0.362094795 -1.372975660  0.014065984  0.080895829 -0.267568817
```

```

##      IPCA.13 Selic.Over.13      g_BR.14      g_RJ.14      IPCA.14
##      0.058184258      0.008320017      -0.069987736      -0.090264994      -0.074496256
## Selic.Over.14      const
##      -0.008067513      0.004871617
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.11 + g_RJ.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_RJ.12 + IPCA.12 + Selic.Over.12 + g
##
##      g_BR.11      g_RJ.11      IPCA.11 Selic.Over.11      g_BR.12
##      -0.0022005770 -0.0019897769      0.6523669672      0.0013254952 -0.0031867952
##      g_RJ.12      IPCA.12 Selic.Over.12      g_BR.13      g_RJ.13
##      0.0007173778      0.0311573751      0.0018065230      0.0122661272 -0.0078485355
##      IPCA.13 Selic.Over.13      g_BR.14      g_RJ.14      IPCA.14
##      -0.0069188688 -0.0024552154      0.0128135228 -0.0103180696 -0.0659767145
## Selic.Over.14      const
##      -0.0002879923      0.0016095139
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_RJ.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_RJ.12 + IPCA.12 + Selic.Over.
##
##      g_BR.11      g_RJ.11      IPCA.11 Selic.Over.11      g_BR.12
##      0.035963524      0.101916104      2.548214898      0.364318850 -0.087064025
##      g_RJ.12      IPCA.12 Selic.Over.12      g_BR.13      g_RJ.13
##      0.125016710      2.472151812      0.548835889      -0.019752681      0.077396815
##      IPCA.13 Selic.Over.13      g_BR.14      g_RJ.14      IPCA.14
##      3.319400366      0.337903657      -0.021367518      0.091692019      1.962946320
## Selic.Over.14      const
##      -0.302863423      -0.007280986

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```

serial.test(VAR_RJ, type="BG")

```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_RJ
## Chi-squared = 99.694, df = 80, p-value = 0.06724

```

```

arch.test(VAR_RJ, multivariate.only = FALSE)

```

```

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 5.0267, df = 16, p-value = 0.9956
##
##

```



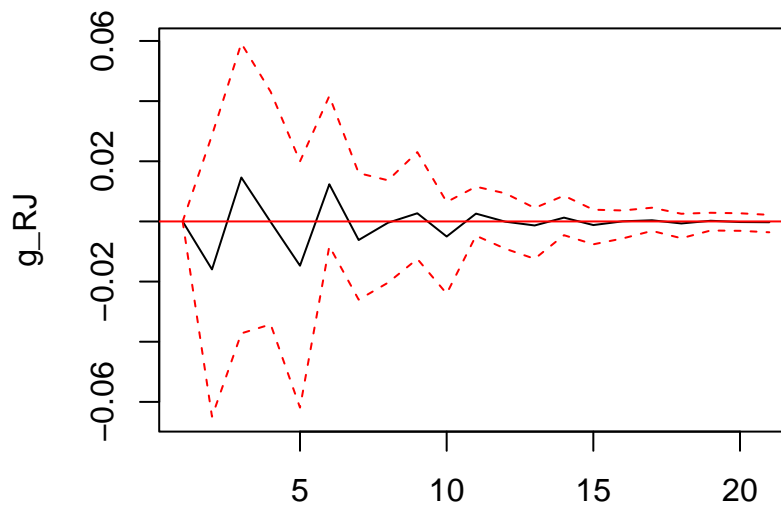
```

## $g_RJ
##
## ARCH test (univariate)
##
## data: Residual of g_RJ equation
## Chi-squared = 6.6659, df = 16, p-value = 0.9792
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 23.861, df = 16, p-value = 0.09259
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 58.229, df = 16, p-value = 1.038e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_RJ
## Chi-squared = 658.78, df = 500, p-value = 2.206e-06
#Estimando o VAR Estrutural
SVAR_RJ <- SVAR(VAR_RJ, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod = "scoring")
SVAR_RJ

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##      g_BR   g_RJ   IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000      0
## g_RJ      0.05764 1.00000 0.00000      0
## IPCA      0.08933 0.08589 1.00000      0
## Selic.Over 0.08105 0.07711 0.09488      1
#Função Impulso-Resposta
SVAR_RJ_irf <- irf(SVAR_RJ, impulse = "Selic.Over", response = "g_RJ", n.ahead=20, ortho = TRUE)
plot(SVAR_RJ_irf)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

###Comparando Diferentes Períodos###

#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#

#Selecionando o Lag

```
lagselect_RJ_pa <- VARselect(base_SVAR_final[1:106,c(1,11,15,16)],lag.max=10, type="both")
lagselect_RJ_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      2      1      1      2
```

```
lagselect_RJ_pa$criteria
```

```
##          1          2          3          4          5
## AIC(n) -3.227876e+01 -3.237662e+01 -3.234297e+01 -3.228254e+01 -3.227934e+01
## HQ(n)  -3.201963e+01 -3.194472e+01 -3.173832e+01 -3.150512e+01 -3.132917e+01
## SC(n)  -3.163768e+01 -3.130814e+01 -3.084710e+01 -3.035927e+01 -2.992869e+01
## FPE(n)  9.589466e-15  8.716301e-15  9.062822e-15  9.720133e-15  9.900423e-15
##          6          7          8          9         10
## AIC(n) -3.222423e+01 -3.206132e+01 -3.206205e+01 -3.208454e+01 -3.198482e+01
## HQ(n)  -3.110130e+01 -3.076563e+01 -3.059360e+01 -3.044334e+01 -3.017085e+01
## SC(n)  -2.944619e+01 -2.885589e+01 -2.842922e+01 -2.802433e+01 -2.749721e+01
## FPE(n)  1.069720e-14  1.298716e-14  1.353113e-14  1.397338e-14  1.656077e-14
```

#Estimando o VAR Reduzido

```
VAR_RJ_pa <- VAR(base_SVAR_final[1:106,c(1,11,15,16)], p = lagselect_RJ_pa$selection[1], season = NULL,
VAR_RJ_pa
```

```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.341150061 -0.075440883  0.102514019 -0.009329155 -0.093374561
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      const
##  0.023375819 -0.066943916  0.005507342  0.006242760
##
##
## Estimated coefficients for equation g_RJ:
## =====
## Call:
## g_RJ = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.422036628 -0.475338147  2.178709361 -0.022500951  0.332358529
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      const
## -0.172201905 -1.703261068  0.014818691  0.008449818
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.0088011222 -0.0187592906  0.7404114085  0.0013290889  0.0007457115
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      const
## -0.0222933106 -0.0165775892 -0.0010157771  0.0011644575
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.08041482 -0.25311944  3.22869098  0.44861079  0.02348132
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      const
## -0.40634014  7.17489058  0.46430357  0.04275935
##
## Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
##
## serial.test(VAR_RJ_pa, type="BG")
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_RJ_pa

```

```

## Chi-squared = 92.157, df = 80, p-value = 0.1664
arch.test(VAR_RJ_pa, multivariate.only = FALSE)

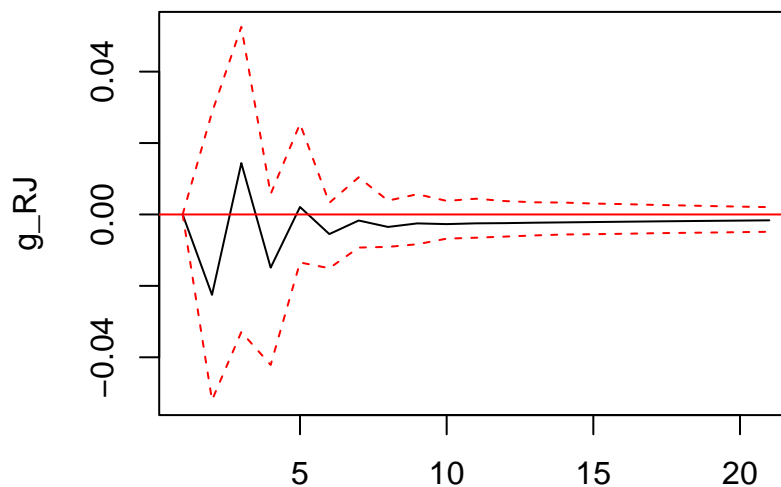
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 16.113, df = 16, p-value = 0.4451
##
##
## $g_RJ
##
## ARCH test (univariate)
##
## data: Residual of g_RJ equation
## Chi-squared = 9.4483, df = 16, p-value = 0.8938
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 16.259, df = 16, p-value = 0.435
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 26.704, df = 16, p-value = 0.04489
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_RJ_pa
## Chi-squared = 577.82, df = 500, p-value = 0.00903
#Estimando o VAR Estrutural
SVAR_RJ_pa <- SVAR(VAR_RJ_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_RJ_pa

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_RJ    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_RJ      0.08011  1.00000  0.00000      0

```

```
## IPCA      0.09512 0.09259 1.00000      0
## Selic.Over 0.09080 0.05349 0.09208      1
#Função Impulso-Resposta
SVAR_RJ_irf_pa <- irf(SVAR_RJ_pa, impulse = "Selic.Over", response = "g_RJ", n.ahead=20, ortho = TRUE)
plot(SVAR_RJ_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_RJ_pe <- VARselect(base_SVAR_final[107:261,c(1,11,15,16)],lag.max=10, type="both")
lagselect_RJ_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      1      4

lagselect_RJ_pe$criteria

##           1           2           3           4           5
## AIC(n) -3.007134e+01 -3.037582e+01 -3.038693e+01 -3.042702e+01 -3.038072e+01
## HQ(n)  -2.987114e+01 -3.004215e+01 -2.991979e+01 -2.982642e+01 -2.964665e+01
## SC(n)  -2.957864e+01 -2.955465e+01 -2.923729e+01 -2.894892e+01 -2.857415e+01
## FPE(n)  8.714958e-14  6.431747e-14  6.370505e-14  6.136903e-14  6.455404e-14
##           6           7           8           9          10
## AIC(n) -3.034318e+01 -3.020009e+01 -3.011981e+01 -3.007875e+01 -2.994259e+01
## HQ(n)  -2.947565e+01 -2.919908e+01 -2.898534e+01 -2.881081e+01 -2.854118e+01
## SC(n)  -2.820815e+01 -2.773658e+01 -2.732784e+01 -2.695831e+01 -2.649368e+01
## FPE(n)  6.744209e-14  7.848688e-14  8.601519e-14  9.093134e-14  1.061089e-13
```

#Estimando o VAR Reduzido

```
VAR_RJ_pe <- VAR(base_SVAR_final[107:261,c(1,11,15,16)], p = lagselect_RJ_pe$selection[1], season = NULL)
VAR_RJ_pe
```

##

VAR Estimation Results:

=====

##

Estimated coefficients for equation g_BR:

=====

Call:

g_BR = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_RJ.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.166120562	-0.096332347	0.568629587	-0.001742462	-0.252041137
##	g_RJ.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_RJ.l3
##	-0.001667966	-0.303343451	0.020840341	-0.140548170	-0.065860843
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_RJ.l4	IPCA.l4
##	-0.390712712	-0.005461564	-0.120780838	0.017050595	-1.704406186
##	Selic.Over.l4	const			
##	-0.019178825	0.011018467			

##

##

Estimated coefficients for equation g_RJ:

=====

Call:

g_RJ = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_RJ.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	0.11635176	-0.47229875	0.61893385	0.00356481	0.11050611
##	g_RJ.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_RJ.l3
##	-0.38416729	-0.57026083	0.01305670	0.03081446	-0.27181133
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_RJ.l4	IPCA.l4
##	-0.96070742	0.00886043	-0.09255216	-0.12203364	-0.47309606
##	Selic.Over.l4	const			
##	-0.03009796	0.01067200			

##

##

Estimated coefficients for equation IPCA:

=====

Call:

IPCA = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_RJ.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.0032200532	0.0014909703	0.5684189663	0.0006678351	-0.0069478192
##	g_RJ.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_RJ.l3
##	0.0100474324	0.0649508869	0.0070883488	0.0104060260	-0.0082277863
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_RJ.l4	IPCA.l4
##	0.0086151287	-0.0033249990	0.0138494881	-0.0122546237	-0.1511089227
##	Selic.Over.l4	const			
##	-0.0048986588	0.0027552933			

##

##

Estimated coefficients for equation Selic.Over:

```
## =====
## Call:
## Selic.Over = g_BR.l1 + g_RJ.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RJ.l2 + IPCA.l2 + Selic.Over.l3
##
##      g_BR.l1      g_RJ.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      0.05163991  0.06520218  1.97956746  0.32865731 -0.08668100
##      g_RJ.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RJ.l3
##      0.12361276  2.14079773  0.63130206 -0.10862526  0.04562561
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RJ.l4      IPCA.l4
##      3.75958583  0.32518533 -0.08048161  0.04440764  3.21118557
## Selic.Over.l4      const
##      -0.32106447 -0.02687604
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_RJ_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_RJ_pe
## Chi-squared = 90.821, df = 80, p-value = 0.1916
```

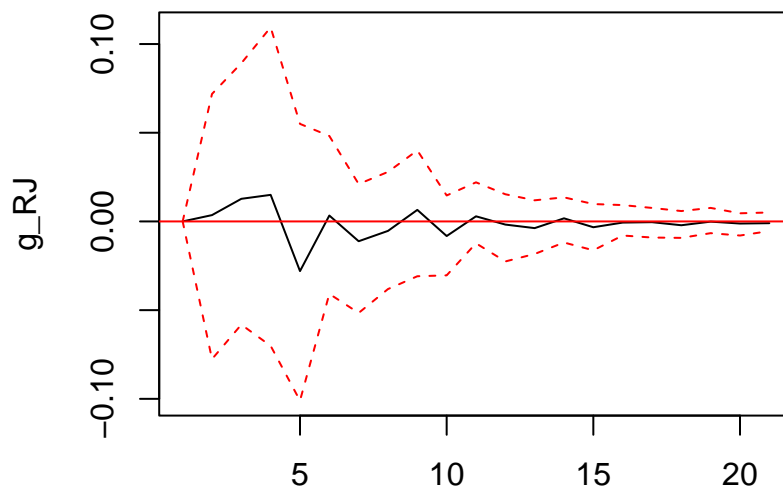
```
arch.test(VAR_RJ_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 1.4871, df = 16, p-value = 1
##
##
## $g_RJ
##
## ARCH test (univariate)
##
## data: Residual of g_RJ equation
## Chi-squared = 4.9124, df = 16, p-value = 0.9962
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.079, df = 16, p-value = 0.3193
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 34.773, df = 16, p-value = 0.004268
```

```
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_RJ_pe
## Chi-squared = 544.37, df = 500, p-value = 0.0831
#Estimando o VAR Estrutural
SVAR_RJ_pe <- SVAR(VAR_RJ_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_RJ_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR   g_RJ   IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_RJ      0.04878 1.00000 0.00000         0
## IPCA      0.08674 0.08207 1.00000         0
## Selic.Over 0.07872 0.08455 0.09749         1
#Função Impulso-Resposta
SVAR_RJ_irf_pe <- irf(SVAR_RJ_pe, impulse = "Selic.Over", response = "g_RJ", n.ahead=20, ortho = TRUE)
plot(SVAR_RJ_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs


```
#####
#####RIO GRANDE DO SUL#####
#####
#Selecionando o Lag
lagselect_RS <- VARselect(base_SVAR_final[,c(1,12,15,16)],lag.max=10, type="both")
lagselect_RS$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      5      2      2      5

lagselect_RS$criteria

##              1              2              3              4              5
## AIC(n) -3.056292e+01 -3.088148e+01 -3.091514e+01 -3.096974e+01 -3.097587e+01
## HQ(n)  -3.042726e+01 -3.065539e+01 -3.059861e+01 -3.056277e+01 -3.047846e+01
## SC(n)  -3.022582e+01 -3.031965e+01 -3.012858e+01 -2.995845e+01 -2.973985e+01
## FPE(n)  5.329771e-14  3.876304e-14  3.749110e-14  3.551744e-14  3.532937e-14
##              6              7              8              9             10
## AIC(n) -3.095673e+01 -3.087240e+01 -3.084142e+01 -3.082122e+01 -3.080797e+01
## HQ(n)  -3.036889e+01 -3.019412e+01 -3.007271e+01 -2.996207e+01 -2.985838e+01
## SC(n)  -2.949599e+01 -2.918692e+01 -2.893121e+01 -2.868629e+01 -2.844830e+01
## FPE(n)  3.605426e-14  3.929003e-14  4.061179e-14  4.155154e-14  4.224634e-14

#Estimando o VAR Reduzido
VAR_RS <- VAR(base_SVAR_final[,c(1,12,15,16)], p = lagselect_RS$selection[1], season = NULL, exog = NULL)
VAR_RS

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.284475880  0.168966337  0.427954356 -0.017679471 -0.136970184
##      g_RS.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RS.l3
##  0.009768174 -0.386074199  0.007997155 -0.127400691  0.019690108
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RS.l4      IPCA.l4
## -0.103289092 -0.003325950 -0.034353049  0.009867779 -0.775486884
## Selic.Over.l4      g_BR.l5      g_RS.l5      IPCA.l5 Selic.Over.l5
##  0.007902115  0.056708891 -0.048960942  0.223134312  0.007810717
##      const
##  0.001024255
##
##
## Estimated coefficients for equation g_RS:
## =====
## Call:
## g_RS = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.043033560 -0.134612904  0.498992473 -0.079863547  0.138058522
```

```

##      g_RS.12      IPCA.12 Selic.Over.12      g_BR.13      g_RS.13
## -0.317949333    1.787358734    0.011199642 -0.033453473 -0.089357391
##      IPCA.13 Selic.Over.13      g_BR.14      g_RS.14      IPCA.14
## -1.653889389    0.028295213 -0.039404237 -0.087534233 -0.238392236
## Selic.Over.14      g_BR.15      g_RS.15      IPCA.15 Selic.Over.15
## 0.069789432    0.041775908 -0.100913928 -0.128804087 -0.037469581
##      const
## 0.005786048
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.11 + g_RS.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_RS.12 + IPCA.12 + Selic.Over.12 + g
##
##      g_BR.11      g_RS.11      IPCA.11 Selic.Over.11      g_BR.12
## -0.0089876677 0.0022690477 0.6542409948 -0.0001007373 -0.0044769228
##      g_RS.12      IPCA.12 Selic.Over.12      g_BR.13      g_RS.13
## 0.0027682014 0.0277132211 0.0039851504 0.0099010619 -0.0027643740
##      IPCA.13 Selic.Over.13      g_BR.14      g_RS.14      IPCA.14
## -0.0085827700 0.0011983147 -0.0040662941 0.0034428470 -0.0011480382
## Selic.Over.14      g_BR.15      g_RS.15      IPCA.15 Selic.Over.15
## 0.0023656279 -0.0195910634 0.0114414942 -0.1646207167 -0.0068044248
##      const
## 0.0019032655
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_RS.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_RS.12 + IPCA.12 + Selic.Over.
##
##      g_BR.11      g_RS.11      IPCA.11 Selic.Over.11      g_BR.12
## 0.195162753 -0.045008613 2.111621207 0.344952299 -0.136368752
##      g_RS.12      IPCA.12 Selic.Over.12      g_BR.13      g_RS.13
## 0.092950067 3.206780070 0.558339431 0.013697027 0.164627355
##      IPCA.13 Selic.Over.13      g_BR.14      g_RS.14      IPCA.14
## 3.217839948 0.370693076 0.278372843 -0.183817743 -0.929949062
## Selic.Over.14      g_BR.15      g_RS.15      IPCA.15 Selic.Over.15
## -0.255203827 -0.093920326 0.132824450 3.667962330 -0.074278543
##      const
## -0.008090261

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_RS, type="BG")
```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_RS
## Chi-squared = 100.52, df = 80, p-value = 0.06019

```

```

arch.test(VAR_RS, multivariate.only = FALSE)

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 6.087, df = 16, p-value = 0.9871
##
##
## $g_RS
##
## ARCH test (univariate)
##
## data: Residual of g_RS equation
## Chi-squared = 34.021, df = 16, p-value = 0.005398
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 16.01, df = 16, p-value = 0.4522
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 48.732, df = 16, p-value = 3.643e-05
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_RS
## Chi-squared = 637.48, df = 500, p-value = 2.867e-05

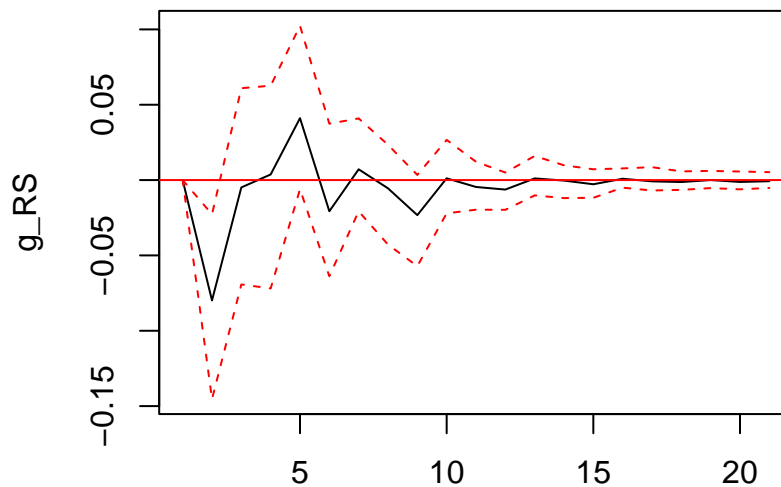
#Estimando o VAR Estrutural
SVAR_RS <- SVAR(VAR_RS, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_RS

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_RS    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_RS      0.03212  1.00000  0.00000      0
## IPCA      0.08687  0.07796  1.00000      0

```

```
## Selic.Over 0.07492 0.04775 0.09187 1
#Função Impulso-Resposta
SVAR_RS_irf <- irf(SVAR_RS, impulse = "Selic.Over", response = "g_RS", n.ahead=20, ortho = TRUE)
plot(SVAR_RS_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
#Selecionando o Lag
lagselect_RS_pa <- VARselect(base_SVAR_final[1:106,c(1,12,15,16)],lag.max=10, type="both")
lagselect_RS_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      2      1      1      2
```

```
lagselect_RS_pa$criteria
```

	1	2	3	4	5
## AIC(n)	-3.152054e+01	-3.160842e+01	-3.154491e+01	-3.153730e+01	-3.157317e+01
## HQ(n)	-3.126140e+01	-3.117652e+01	-3.094026e+01	-3.075989e+01	-3.062300e+01
## SC(n)	-3.087945e+01	-3.053994e+01	-3.004904e+01	-2.961404e+01	-2.922252e+01
## FPE(n)	2.046850e-14	1.879130e-14	2.013065e-14	2.047968e-14	2.006044e-14
	6	7	8	9	10
## AIC(n)	-3.153055e+01	-3.151127e+01	-3.143147e+01	-3.153137e+01	-3.153960e+01
## HQ(n)	-3.040762e+01	-3.021558e+01	-2.996303e+01	-2.989016e+01	-2.972564e+01
## SC(n)	-2.875251e+01	-2.830584e+01	-2.779865e+01	-2.747115e+01	-2.705199e+01

```
## FPE(n) 2.140576e-14 2.251122e-14 2.542084e-14 2.429648e-14 2.584854e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_RS_pa <- VAR(base_SVAR_final[1:106,c(1,12,15,16)], p = lagselect_RS_pa$selection[1], season = NULL,
VAR_RS_pa
```

```
##
```

```
## VAR Estimation Results:
```

```
## =====
```

```
##
```

```
## Estimated coefficients for equation g_BR:
```

```
## =====
```

```
## Call:
```

```
## g_BR = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.213370611 0.050109064 0.085637241 -0.011251502 -0.063473188
```

```
##      g_RS.l2      IPCA.l2 Selic.Over.l2      const
## 0.013396858 -0.105049870 0.007905146 0.006080328
```

```
##
```

```
##
```

```
## Estimated coefficients for equation g_RS:
```

```
## =====
```

```
## Call:
```

```
## g_RS = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.17389764 -0.15626045 -0.39534532 -0.03588422 0.22333805
```

```
##      g_RS.l2      IPCA.l2 Selic.Over.l2      const
## -0.25800879 1.32457075 0.02400079 0.00935722
```

```
##
```

```
##
```

```
## Estimated coefficients for equation IPCA:
```

```
## =====
```

```
## Call:
```

```
## IPCA = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.0026498081 -0.0053610145 0.7637135210 0.0011016254 -0.0271396235
```

```
##      g_RS.l2      IPCA.l2 Selic.Over.l2      const
## 0.0081314871 -0.0569399942 -0.0006273406 0.0010565173
```

```
##
```

```
##
```

```
## Estimated coefficients for equation Selic.Over:
```

```
## =====
```

```
## Call:
```

```
## Selic.Over = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + c
```

```
##
```

```
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.16822497 -0.15834120 3.53809104 0.44522470 -0.25060419
```

```
##      g_RS.l2      IPCA.l2 Selic.Over.l2      const
## -0.11205825 6.71102320 0.46947310 0.04050309
```

```
#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
```

```

serial.test(VAR_RS_pa, type="BG")

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_RS_pa
## Chi-squared = 100.59, df = 80, p-value = 0.05966
arch.test(VAR_RS_pa, multivariate.only = FALSE)

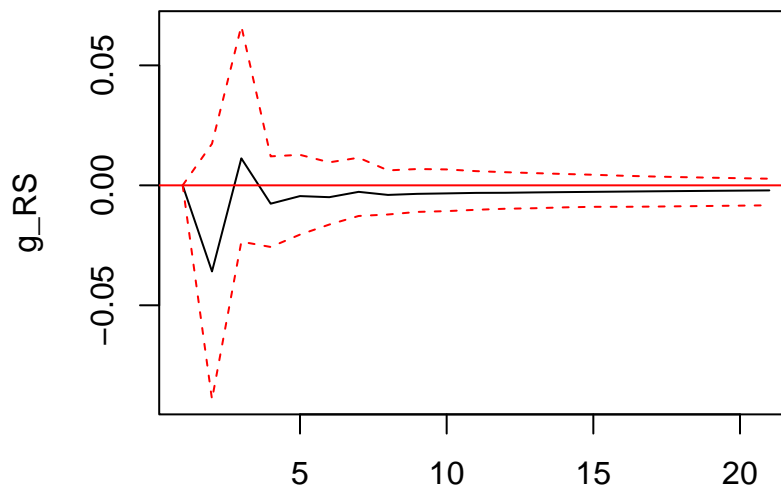
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 13.571, df = 16, p-value = 0.6307
##
##
## $g_RS
##
## ARCH test (univariate)
##
## data: Residual of g_RS equation
## Chi-squared = 16.006, df = 16, p-value = 0.4526
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 19.711, df = 16, p-value = 0.2335
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 28.234, df = 16, p-value = 0.02964
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_RS_pa
## Chi-squared = 579.59, df = 500, p-value = 0.00786
#Estimando o VAR Estrutural
SVAR_RS_pa <- SVAR(VAR_RS_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_RS_pa

##
## SVAR Estimation Results:
## =====

```

```
##
##
## Estimated A matrix:
##           g_BR   g_RS   IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000      0
## g_RS      0.06920 1.00000 0.00000      0
## IPCA       0.09343 0.08373 1.00000      0
## Selic.Over 0.09480 0.12323 0.09772      1
#Função Impulso-Resposta
SVAR_RS_irf_pa <- irf(SVAR_RS_pa, impulse = "Selic.Over", response = "g_RS", n.ahead=20, ortho = TRUE)
plot(SVAR_RS_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_RS_pe <- VARselect(base_SVAR_final[107:261,c(1,12,15,16)],lag.max=10, type="both")
lagselect_RS_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      1      3

lagselect_RS_pe$criteria

##           1           2           3           4           5
## AIC(n) -3.026722e+01 -3.054349e+01 -3.059804e+01 -3.059927e+01 -3.050594e+01
## HQ(n)  -3.006702e+01 -3.020983e+01 -3.013091e+01 -2.999867e+01 -2.977187e+01
## SC(n)  -2.977452e+01 -2.972233e+01 -2.944841e+01 -2.912117e+01 -2.869937e+01
## FPE(n)  7.164676e-14  5.438884e-14  5.158064e-14  5.165848e-14  5.695591e-14
```

```

##          6          7          8          9          10
## AIC(n) -3.046730e+01 -3.036414e+01 -3.026329e+01 -3.016689e+01 -3.006454e+01
## HQ(n) -2.959976e+01 -2.936313e+01 -2.912882e+01 -2.889895e+01 -2.866313e+01
## SC(n) -2.833227e+01 -2.790064e+01 -2.747132e+01 -2.704646e+01 -2.661563e+01
## FPE(n) 5.957000e-14 6.661171e-14 7.451841e-14 8.325935e-14 9.392706e-14

#Estimando o VAR Reduzido
VAR_RS_pe <- VAR(base_SVAR_final[107:261,c(1,12,15,16)], p = lagselect_RS_pe$selection[1], season = NULL)
VAR_RS_pe

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.563918643  0.296753295  0.810966792 -0.039196051 -0.293535272
##      g_RS.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RS.l3
##  0.063407113 -0.371270296  0.040462636 -0.303510383  0.075776317
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RS.l4      IPCA.l4
## -0.815279258  0.018047655 -0.232168400  0.104409437 -1.087544269
## Selic.Over.l4      const
## -0.022495302  0.006982952
##
##
## Estimated coefficients for equation g_RS:
## =====
## Call:
## g_RS = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.24711576 -0.04204130  0.97297693 -0.07255862 -0.06247468
##      g_RS.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RS.l3
## -0.23166175  1.22158071  0.03583901 -0.27517362  0.04800560
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RS.l4      IPCA.l4
## -2.10251463  0.01054517 -0.23318539  0.03619521 -0.65475074
## Selic.Over.l4      const
##  0.01082583  0.01255742
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0012855072 -0.0006826639  0.5597613381  0.0009790117  0.0142365637
##      g_RS.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RS.l3
## -0.0069755104  0.0391331255  0.0063720729  0.0263406170 -0.0163307808
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RS.l4      IPCA.l4
##  0.0176250324 -0.0025888955  0.0102426083 -0.0038726259 -0.1233713961

```



```
## Selic.Over.l4          const
## -0.0053773219  0.0028766133
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_RS.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_RS.l2 + IPCA.l2 + Selic.Over.l3 + Selic.Over.l4
##
##      g_BR.l1      g_RS.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      0.17008321  -0.10268460   2.14991060   0.33279228  -0.25228155
##      g_RS.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_RS.l3
##      0.10033765   2.08774451   0.63058318  -0.47374285   0.31315197
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_RS.l4      IPCA.l4
##      4.12004795   0.30756838  -0.14642000   0.10096486   2.45368720
## Selic.Over.l4          const
##      -0.30445256  -0.02746139
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_RS_pe, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_RS_pe
## Chi-squared = 86.86, df = 80, p-value = 0.281
```

```
arch.test(VAR_RS_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 2.4726, df = 16, p-value = 1
##
##
## $g_RS
##
## ARCH test (univariate)
##
## data: Residual of g_RS equation
## Chi-squared = 19.138, df = 16, p-value = 0.2616
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 11.824, df = 16, p-value = 0.756
##
##
## $Selic.Over
```

```

##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 33.834, df = 16, p-value = 0.00572
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_RS_pe
## Chi-squared = 543.88, df = 500, p-value = 0.08535

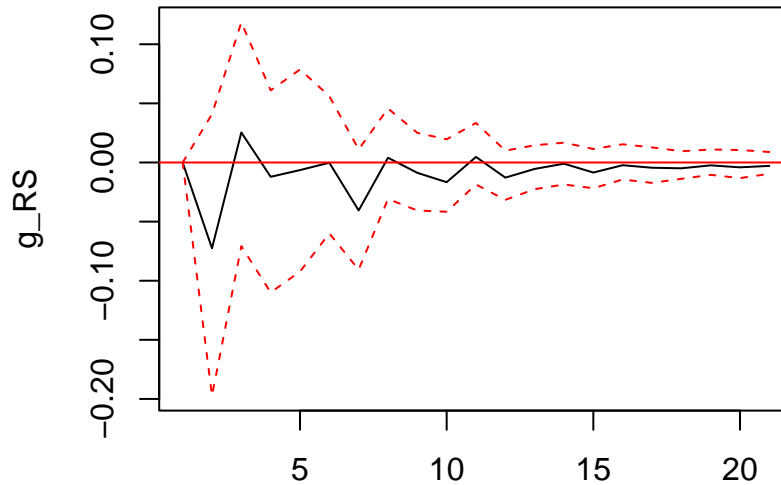
#Estimando o VAR Estrutural
SVAR_RS_pe <- SVAR(VAR_RS_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_RS_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_RS    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_RS      0.01315  1.00000  0.00000         0
## IPCA      0.08441  0.07338  1.00000         0
## Selic.Over 0.07130  0.01482  0.09069         1

#Função Impulso-Resposta
SVAR_RS_irf_pe <- irf(SVAR_RS_pe, impulse = "Selic.Over", response = "g_RS", n.ahead=20, ortho = TRUE)
plot(SVAR_RS_irf_pe)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#####
#####SANTA CATARINA#####
#####
#Selecionando o Lag
lagselect_SC <- VARselect(base_SVAR_final[,c(1,13,15,16)],lag.max=10, type="both")
lagselect_SC$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_SC$criteria

##           1           2           3           4           5
## AIC(n) -3.104193e+01 -3.135177e+01 -3.141203e+01 -3.148175e+01 -3.146194e+01
## HQ(n)  -3.090628e+01 -3.112568e+01 -3.109550e+01 -3.107478e+01 -3.096454e+01
## SC(n)  -3.070484e+01 -3.078995e+01 -3.062548e+01 -3.047046e+01 -3.022592e+01
## FPE(n)  3.301231e-14  2.421989e-14  2.281022e-14  2.128529e-14  2.172884e-14
##           6           7           8           9          10
## AIC(n) -3.141820e+01 -3.132352e+01 -3.125103e+01 -3.122266e+01 -3.117679e+01
## HQ(n)  -3.083036e+01 -3.064524e+01 -3.048231e+01 -3.036351e+01 -3.022720e+01
## SC(n)  -2.995745e+01 -2.963804e+01 -2.934082e+01 -2.908773e+01 -2.881712e+01
## FPE(n)  2.272717e-14  2.502438e-14  2.696259e-14  2.781283e-14  2.921544e-14

#Estimando o VAR Reduzido
VAR_SC <- VAR(base_SVAR_final[,c(1,13,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_SC

##
```

```

## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.3479949044  0.2462474455  0.4717865679 -0.0139739747 -0.2245854221
##      g_SC.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SC.l3
##  0.0872762367 -0.5103067543 -0.0010289027 -0.1400569037  0.0020124668
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SC.l4      IPCA.l4
##  0.0176678708  0.0001398154 -0.0492930774 -0.0093325104 -0.6770980104
## Selic.Over.l4      g_BR.l5      g_SC.l5      IPCA.l5 Selic.Over.l5
##  0.0091940317 -0.0152735797  0.0019661974  0.3057460001  0.0080165750
##      const
##  0.0002448392
##
##
## Estimated coefficients for equation g_SC:
## =====
## Call:
## g_SC = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0276067702 -0.1183750153  1.3152961214 -0.0299564281  0.0200569550
##      g_SC.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SC.l3
## -0.0047115836  0.3484153029 -0.0284320738  0.0910263781 -0.1448535732
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SC.l4      IPCA.l4
## -1.8109129507  0.0133327289  0.1296313599 -0.1858955170 -0.2529177009
## Selic.Over.l4      g_BR.l5      g_SC.l5      IPCA.l5 Selic.Over.l5
##  0.0427018740 -0.0093730190 -0.1057312491  0.0687176699 -0.0005051599
##      const
##  0.0041509970
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0161208545  0.0068349752  0.6628589318 -0.0007649006 -0.0205420840
##      g_SC.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SC.l3
##  0.0201214323  0.0108689299  0.0046005272  0.0073510145 -0.0021646562
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SC.l4      IPCA.l4
## -0.0170520014  0.0016818498 -0.0011800376  0.0007588086  0.0083566810
## Selic.Over.l4      g_BR.l5      g_SC.l5      IPCA.l5 Selic.Over.l5
##  0.0025244694 -0.0117360265  0.0048090105 -0.1446713906 -0.0074307413
##      const
##  0.0018711548
##
##

```

```
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.286427378 -0.046619101  2.338494874  0.327213670  0.386125488
##      g_SC.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SC.l3
## -0.333176514  2.973107165  0.562527322  0.342529708 -0.160970879
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SC.l4      IPCA.l4
##  3.995440707  0.372668056  0.513768531 -0.287132744 -0.748296295
## Selic.Over.l4      g_BR.l5      g_SC.l5      IPCA.l5 Selic.Over.l5
## -0.264016444  0.435332190 -0.354033706  3.382968221 -0.060432479
##      const
## -0.006025034
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_SC, type="BG")
```

```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_SC
## Chi-squared = 83.856, df = 80, p-value = 0.3622
```

```
arch.test(VAR_SC, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 7.6777, df = 16, p-value = 0.9579
##
##
## $g_SC
##
## ARCH test (univariate)
##
## data: Residual of g_SC equation
## Chi-squared = 56.96, df = 16, p-value = 1.688e-06
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 13.207, df = 16, p-value = 0.6576
##
##
## $Selic.Over
##
## ARCH test (univariate)
```

```

##
## data: Residual of Selic.Over equation
## Chi-squared = 52.94, df = 16, p-value = 7.715e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_SC
## Chi-squared = 849.43, df = 500, p-value < 2.2e-16

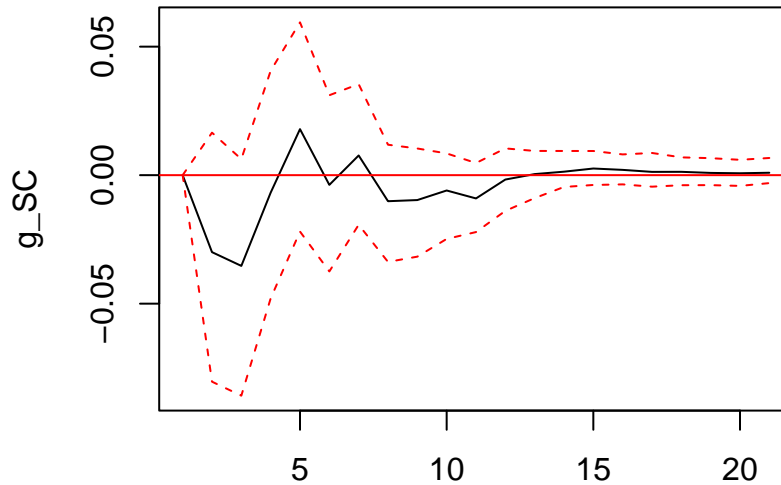
#Estimando o VAR Estrutural
SVAR_SC <- SVAR(VAR_SC, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_SC

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_SC    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000         0
## g_SC      0.04179  1.00000  0.00000         0
## IPCA      0.08804  0.08374  1.00000         0
## Selic.Over 0.07922  0.06795  0.09345         1

#Função Impulso-Resposta
SVAR_SC_irf <- irf(SVAR_SC, impulse = "Selic.Over", response = "g_SC", n.ahead=20, ortho = TRUE)
plot(SVAR_SC_irf)

```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
```

```
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
```

```
#Selecionando o Lag
```

```
lagselect_SC_pa <- VARselect(base_SVAR_final[1:106,c(1,13,15,16)],lag.max=10, type="both")
lagselect_SC_pa$selection
```

```
## AIC(n)  HQ(n)  SC(n) FPE(n)
##      2      1      1      2
```

```
lagselect_SC_pa$criteria
```

```
##          1          2          3          4          5
## AIC(n) -3.228388e+01 -3.243170e+01 -3.235858e+01 -3.239537e+01 -3.237809e+01
## HQ(n)  -3.202474e+01 -3.199980e+01 -3.175393e+01 -3.161795e+01 -3.142792e+01
## SC(n)  -3.164280e+01 -3.136322e+01 -3.086271e+01 -3.047211e+01 -3.002743e+01
## FPE(n)  9.540508e-15  8.249176e-15  8.922463e-15  8.683009e-15  8.969534e-15
##          6          7          8          9         10
## AIC(n) -3.226877e+01 -3.227481e+01 -3.226710e+01 -3.241417e+01 -3.228720e+01
## HQ(n)  -3.114584e+01 -3.097912e+01 -3.079865e+01 -3.077296e+01 -3.047324e+01
## SC(n)  -2.949073e+01 -2.906938e+01 -2.863427e+01 -2.835395e+01 -2.779959e+01
## FPE(n)  1.023121e-14  1.049051e-14  1.102258e-14  1.004958e-14  1.223929e-14
```

```
#Estimando o VAR Reduzido
```

```
VAR_SC_pa <- VAR(base_SVAR_final[1:106,c(1,13,15,16)], p = 2 , season = NULL, exog = NULL, type = "cons")
VAR_SC_pa
```

```

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.286274971 -0.009380465 0.059768391 -0.010998600 -0.174306615
##      g_SC.l2      IPCA.l2 Selic.Over.l2      const
## 0.118391828 -0.088036817 0.007873487 0.005900407
##
##
## Estimated coefficients for equation g_SC:
## =====
## Call:
## g_SC = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.015937342 -0.052707407 0.608840882 -0.024235384 0.309593704
##      g_SC.l2      IPCA.l2 Selic.Over.l2      const
## 0.055876875 -1.223797389 0.023622149 0.003745998
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0089597006 0.0077640114 0.7526264897 0.0010810511 -0.0138433991
##      g_SC.l2      IPCA.l2 Selic.Over.l2      const
## -0.0020847860 -0.0436362904 -0.0005965415 0.0010239729
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 0.05071203 -0.12189749 3.93409092 0.44260203 0.35232705
##      g_SC.l2      IPCA.l2 Selic.Over.l2      const
## -0.76835818 6.40385289 0.46981051 0.04208205
##
## #Testando para presença de autocorrelação e heterocedasticidade nos resíduos:
##
## serial.test(VAR_SC_pa, type="BG")
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_SC_pa

```



```

## Chi-squared = 97.081, df = 80, p-value = 0.09399
arch.test(VAR_SC_pa, multivariate.only = FALSE)

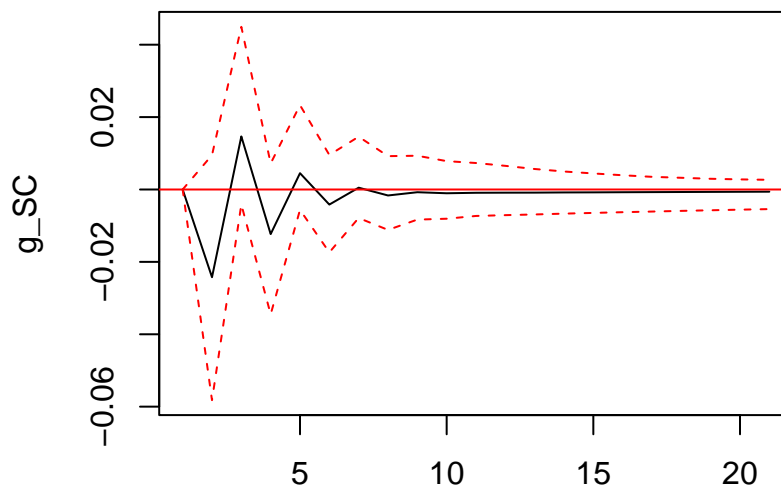
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 14.012, df = 16, p-value = 0.5978
##
##
## $g_SC
##
## ARCH test (univariate)
##
## data: Residual of g_SC equation
## Chi-squared = 21.884, df = 16, p-value = 0.147
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.704, df = 16, p-value = 0.2843
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 27.136, df = 16, p-value = 0.04
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_SC_pa
## Chi-squared = 576.6, df = 500, p-value = 0.009922
#Estimando o VAR Estrutural
SVAR_SC_pa <- SVAR(VAR_SC_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod = "scoring")
SVAR_SC_pa

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_SC    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_SC      0.07601  1.00000  0.00000      0

```

```
## IPCA      0.09457 0.09067 1.00000      0
## Selic.Over 0.08925 0.08726 0.09432      1
#Função Impulso-Resposta
SVAR_SC_irf_pa <- irf(SVAR_SC_pa, impulse = "Selic.Over", response = "g_SC", n.ahead=20, ortho = TRUE)
plot(SVAR_SC_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_SC_pe <- VARselect(base_SVAR_final[107:261,c(1,13,15,16)],lag.max=10, type="both")
lagselect_SC_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      2      4

lagselect_SC_pe$criteria

##           1           2           3           4           5
## AIC(n) -3.069346e+01 -3.102274e+01 -3.115411e+01 -3.124688e+01 -3.112947e+01
## HQ(n)  -3.049325e+01 -3.068908e+01 -3.068697e+01 -3.064628e+01 -3.039540e+01
## SC(n)  -3.020075e+01 -3.020158e+01 -3.000447e+01 -2.976878e+01 -2.932290e+01
## FPE(n)  4.678253e-14  3.368015e-14  2.957959e-14  2.703262e-14  3.053131e-14
##           6           7           8           9          10
## AIC(n) -3.106735e+01 -3.093888e+01 -3.082316e+01 -3.070447e+01 -3.064266e+01
## HQ(n)  -3.019981e+01 -2.993788e+01 -2.968869e+01 -2.943654e+01 -2.924126e+01
## SC(n)  -2.893231e+01 -2.847538e+01 -2.803119e+01 -2.758404e+01 -2.719376e+01
## FPE(n)  3.269121e-14  3.749236e-14  4.257121e-14  4.863673e-14  5.268818e-14
```

#Estimando o VAR Reduzido

```
VAR_SC_pe <- VAR(base_SVAR_final[107:261,c(1,13,15,16)], p = lagselect_SC_pe$selection[1], season = NULL)
VAR_SC_pe
```

##

VAR Estimation Results:

=====

##

Estimated coefficients for equation g_BR:

=====

Call:

g_BR = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_SC.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.705248083	0.421393537	0.680817284	-0.006056272	-0.487338206
##	g_SC.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_SC.l3
##	0.216398358	-0.570474822	0.014406602	-0.313761395	0.059466874
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_SC.l4	IPCA.l4
##	-0.946375426	0.014840215	-0.166243672	0.003445850	-0.831007890
##	Selic.Over.l4	const			
##	-0.027973621	0.008863371			

##

##

Estimated coefficients for equation g_SC:

=====

Call:

g_SC = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_SC.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-0.223941637	-0.008272447	1.369260521	-0.010948807	-0.222579964
##	g_SC.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_SC.l3
##	0.122277352	1.029441538	-0.034985170	-0.059899327	-0.041709048
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_SC.l4	IPCA.l4
##	-3.777133982	0.043998964	-0.001679000	-0.126192556	0.466351655
##	Selic.Over.l4	const			
##	-0.006616140	0.009586812			

##

##

Estimated coefficients for equation IPCA:

=====

Call:

IPCA = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l2 + g

##

	g_BR.l1	g_SC.l1	IPCA.l1	Selic.Over.l1	g_BR.l2
##	-1.622994e-02	2.235993e-03	6.069065e-01	1.457236e-05	-2.068837e-02
##	g_SC.l2	IPCA.l2	Selic.Over.l2	g_BR.l3	g_SC.l3
##	2.528126e-02	4.727959e-02	6.420931e-03	2.085523e-02	-1.361750e-02
##	IPCA.l3	Selic.Over.l3	g_BR.l4	g_SC.l4	IPCA.l4
##	-3.139402e-02	-3.639288e-03	1.515997e-02	-1.328254e-02	-1.095648e-01
##	Selic.Over.l4	const			
##	-3.288497e-03	2.653838e-03			

##

##

Estimated coefficients for equation Selic.Over:

```
## =====
## Call:
## Selic.Over = g_BR.l1 + g_SC.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SC.l2 + IPCA.l2 + Selic.Over.l3
##
##      g_BR.l1      g_SC.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##      0.11450064  0.12242987  2.57476947  0.31195005  0.55619266
##      g_SC.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SC.l3
##      -0.41680404  1.28787305  0.63545976  0.37896152 -0.43210378
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SC.l4      IPCA.l4
##      3.93026829  0.34175408  0.08786925 -0.04529452  4.10152211
## Selic.Over.l4      const
##      -0.32762904  -0.02792780
```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_SC_pe, type="BG")
```

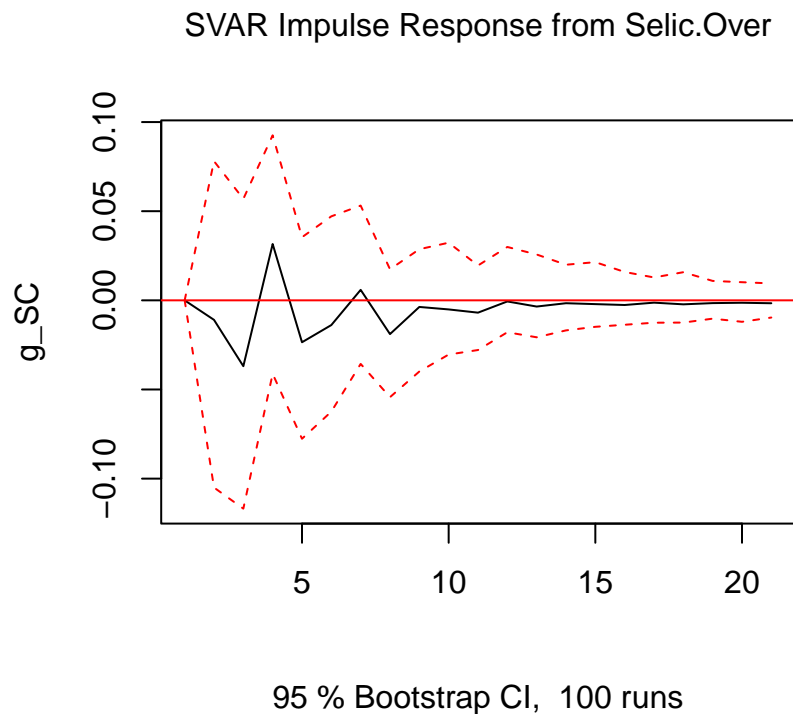
```
##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_SC_pe
## Chi-squared = 86.235, df = 80, p-value = 0.297
```

```
arch.test(VAR_SC_pe, multivariate.only = FALSE)
```

```
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 4.4746, df = 16, p-value = 0.9978
##
##
## $g_SC
##
## ARCH test (univariate)
##
## data: Residual of g_SC equation
## Chi-squared = 29.563, df = 16, p-value = 0.02041
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 7.6948, df = 16, p-value = 0.9574
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 35.07, df = 16, p-value = 0.003887
```

```
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_SC_pe
## Chi-squared = 670.8, df = 500, p-value = 4.643e-07
#Estimando o VAR Estrutural
SVAR_SC_pe <- SVAR(VAR_SC_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod ="scoring")
SVAR_SC_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR   g_SC   IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_SC      0.02139 1.00000 0.00000         0
## IPCA      0.08527 0.07898 1.00000         0
## Selic.Over 0.07915 0.06655 0.09555         1
#Função Impulso-Resposta
SVAR_SC_irf_pe <- irf(SVAR_SC_pe, impulse = "Selic.Over", response = "g_SC", n.ahead=20, ortho = TRUE)
plot(SVAR_SC_irf_pe)
```



```
#####
#####SÃO PAULO#####
#####
#Selecionando o Lag
lagselect_SP <- VARselect(base_SVAR_final[,c(1,14,15,16)],lag.max=10, type="both")
lagselect_SP$selection

## AIC(n)   HQ(n)   SC(n) FPE(n)
##      4      2      2      4

lagselect_SP$criteria

##              1              2              3              4              5
## AIC(n) -3.265692e+01 -3.298857e+01 -3.300615e+01 -3.307523e+01 -3.305144e+01
## HQ(n)  -3.252126e+01 -3.276248e+01 -3.268962e+01 -3.266827e+01 -3.255404e+01
## SC(n)  -3.231982e+01 -3.242674e+01 -3.221959e+01 -3.206395e+01 -3.181543e+01
## FPE(n)  6.565946e-15  4.713248e-15  4.632472e-15  4.325515e-15  4.433267e-15
##              6              7              8              9             10
## AIC(n) -3.302295e+01 -3.296458e+01 -3.291869e+01 -3.289324e+01 -3.284460e+01
## HQ(n)  -3.243511e+01 -3.228630e+01 -3.214997e+01 -3.203409e+01 -3.189501e+01
## SC(n)  -3.156221e+01 -3.127910e+01 -3.100848e+01 -3.075830e+01 -3.048493e+01
## FPE(n)  4.566781e-15  4.849087e-15  5.087515e-15  5.232673e-15  5.511778e-15

#Estimando o VAR Reduzido
VAR_SP <- VAR(base_SVAR_final[,c(1,14,15,16)], p = 5, season = NULL, exog = NULL, type = "const")
VAR_SP

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.1641917417 -0.2693404347  0.4309912934 -0.0101227184  0.3931597618
##      g_SP.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SP.l3
## -0.5010791495 -0.3175112173 -0.0052268505 -0.0328929741 -0.0627480758
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SP.l4      IPCA.l4
##  0.1114813271 -0.0134668734 -0.1551012280  0.0991004039 -1.1421677920
## Selic.Over.l4      g_BR.l5      g_SP.l5      IPCA.l5 Selic.Over.l5
##  0.0076783685 -0.1118474825  0.1309231442  0.4476129632  0.0234606066
##      const
##  0.0005297324
##
##
## Estimated coefficients for equation g_SP:
## =====
## Call:
## g_SP = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  8.042698e-01 -8.550969e-01  3.177976e-01 -1.641193e-02  6.231901e-01
```

```

##      g_SP.12      IPCA.12 Selic.Over.12      g_BR.13      g_SP.13
## -7.749229e-01 -4.014894e-01 -1.973775e-02  2.188098e-01 -2.526399e-01
##      IPCA.13 Selic.Over.13      g_BR.14      g_SP.14      IPCA.14
##  2.668389e-01  4.708160e-03 -5.357747e-03  1.517544e-02 -1.132165e+00
## Selic.Over.14      g_BR.15      g_SP.15      IPCA.15 Selic.Over.15
##  2.211410e-02 -7.385674e-02  1.095574e-01  3.645930e-01  1.267903e-02
##      const
## -1.659096e-05
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.11 + g_SP.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_SP.12 + IPCA.12 + Selic.Over.12 + g
##
##      g_BR.11      g_SP.11      IPCA.11 Selic.Over.11      g_BR.12
## -0.0082758973  0.0028857976  0.6501383181 -0.0004412149 -0.0033106878
##      g_SP.12      IPCA.12 Selic.Over.12      g_BR.13      g_SP.13
##  0.0001177041  0.0200584269  0.0045217211  0.0094338686 -0.0031809984
##      IPCA.13 Selic.Over.13      g_BR.14      g_SP.14      IPCA.14
##  0.0089282942  0.0011782639  0.0041371707 -0.0005433259 -0.0052711645
## Selic.Over.14      g_BR.15      g_SP.15      IPCA.15 Selic.Over.15
##  0.0020770078  0.0001228855 -0.0076451751 -0.1675938317 -0.0067655059
##      const
##  0.0019709367
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.11 + g_SP.11 + IPCA.11 + Selic.Over.11 + g_BR.12 + g_SP.12 + IPCA.12 + Selic.Over.1
##
##      g_BR.11      g_SP.11      IPCA.11 Selic.Over.11      g_BR.12
##  0.322505710 -0.156288112  2.416825705  0.329459263  0.078437240
##      g_SP.12      IPCA.12 Selic.Over.12      g_BR.13      g_SP.13
## -0.054167187  2.723602263  0.563924448  0.469804557 -0.383359019
##      IPCA.13 Selic.Over.13      g_BR.14      g_SP.14      IPCA.14
##  3.384170711  0.380720244  0.613234353 -0.471147091 -0.522013269
## Selic.Over.14      g_BR.15      g_SP.15      IPCA.15 Selic.Over.15
## -0.261645580  0.496797910 -0.402359023  3.454677720 -0.069035276
##      const
## -0.008225199

```

#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```
serial.test(VAR_SP, type="BG")
```

```

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_SP
## Chi-squared = 91.117, df = 80, p-value = 0.1858

```

```

arch.test(VAR_SP, multivariate.only = FALSE)

## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 3.865, df = 16, p-value = 0.9991
##
##
## $g_SP
##
## ARCH test (univariate)
##
## data: Residual of g_SP equation
## Chi-squared = 4.43, df = 16, p-value = 0.9979
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 18.128, df = 16, p-value = 0.3165
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 53.497, df = 16, p-value = 6.263e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_SP
## Chi-squared = 716.28, df = 500, p-value = 6.556e-10

#Estimando o VAR Estrutural
SVAR_SP <- SVAR(VAR_SP, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_SP

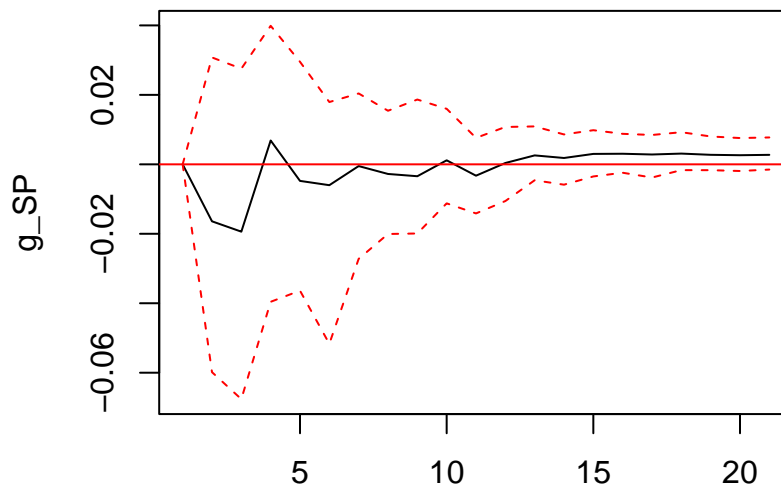
##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##           g_BR    g_SP    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_SP      0.03679  1.00000  0.00000      0
## IPCA      0.08752  0.08612  1.00000      0

```



```
## Selic.Over 0.07778 0.07830 0.09451 1
#Função Impulso-Resposta
SVAR_SP_irf <- irf(SVAR_SP, impulse = "Selic.Over", response = "g_SP", n.ahead=20, ortho = TRUE)
plot(SVAR_SP_irf)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
###Comparando Diferentes Períodos###
#-----Período dos Autores - 01-01-2002 até 01-11-2010-----#
#Selecionando o Lag
lagselect_SP_pa <- VARselect(base_SVAR_final[1:106,c(1,14,15,16)],lag.max=10, type="both")
lagselect_SP_pa$selection
```

```
## AIC(n) HQ(n) SC(n) FPE(n)
##      2      1      1      2
```

```
lagselect_SP_pa$criteria
```

	1	2	3	4	5
## AIC(n)	-3.430362e+01	-3.446953e+01	-3.441367e+01	-3.438187e+01	-3.439724e+01
## HQ(n)	-3.404449e+01	-3.403763e+01	-3.380901e+01	-3.360445e+01	-3.344707e+01
## SC(n)	-3.366254e+01	-3.340105e+01	-3.291780e+01	-3.245860e+01	-3.204659e+01
## FPE(n)	1.265927e-15	1.074957e-15	1.142808e-15	1.191091e-15	1.190862e-15
	6	7	8	9	10
## AIC(n)	-3.430976e+01	-3.426668e+01	-3.423992e+01	-3.425207e+01	-3.418295e+01
## HQ(n)	-3.318683e+01	-3.297099e+01	-3.277147e+01	-3.261087e+01	-3.236899e+01
## SC(n)	-3.153172e+01	-3.106124e+01	-3.060709e+01	-3.019186e+01	-2.969534e+01

```

## FPE(n) 1.329032e-15 1.431334e-15 1.532844e-15 1.599392e-15 1.838410e-15
#Estimando o VAR Reduzido
VAR_SP_pa <- VAR(base_SVAR_final[1:106,c(1,14,15,16)], p = 2, season = NULL, exog = NULL, type = "const",
VAR_SP_pa

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 1.003732462 -0.798307356 0.199993512 -0.013113753 0.321023468
##      g_SP.l2      IPCA.l2 Selic.Over.l2      const
## -0.365810047 -0.404298933 0.011099187 0.005763138
##
##
## Estimated coefficients for equation g_SP:
## =====
## Call:
## g_SP = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 1.400091944 -1.273660500 0.124964003 -0.005209773 0.652821055
##      g_SP.l2      IPCA.l2 Selic.Over.l2      const
## -0.581880384 -0.699340322 0.005638509 0.005256551
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.0516303315 0.0415347324 0.7796514122 0.0009374246 0.0119720855
##      g_SP.l2      IPCA.l2 Selic.Over.l2      const
## -0.0258376845 -0.0612463252 -0.0005648315 0.0011137397
##
##
## Estimated coefficients for equation Selic.Over:
## =====
## Call:
## Selic.Over = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + c
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## 1.2598425 -1.6022588 4.0524225 0.4412432 1.1491980
##      g_SP.l2      IPCA.l2 Selic.Over.l2      const
## -1.3504564 5.6864158 0.4767510 0.0412250
#Testando para presença de autocorrelação e heterocedasticidade nos resíduos:

```

```

serial.test(VAR_SP_pa, type="BG")

##
## Breusch-Godfrey LM test
##
## data: Residuals of VAR object VAR_SP_pa
## Chi-squared = 103.28, df = 80, p-value = 0.04104
arch.test(VAR_SP_pa, multivariate.only = FALSE)

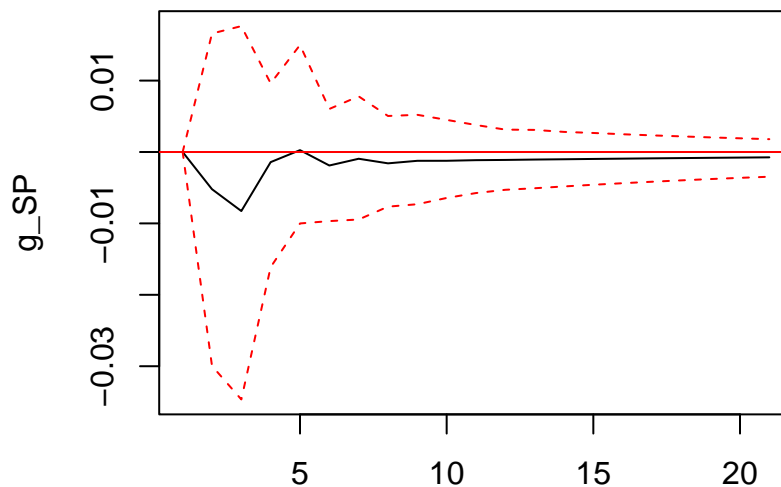
## $g_BR
##
## ARCH test (univariate)
##
## data: Residual of g_BR equation
## Chi-squared = 34.114, df = 16, p-value = 0.005245
##
##
## $g_SP
##
## ARCH test (univariate)
##
## data: Residual of g_SP equation
## Chi-squared = 5.3501, df = 16, p-value = 0.9937
##
##
## $IPCA
##
## ARCH test (univariate)
##
## data: Residual of IPCA equation
## Chi-squared = 12.1, df = 16, p-value = 0.7371
##
##
## $Selic.Over
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 24.219, df = 16, p-value = 0.08482
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_SP_pa
## Chi-squared = 657.4, df = 500, p-value = 2.624e-06
#Estimando o VAR Estrutural
SVAR_SP_pa <- SVAR(VAR_SP_pa, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_SP_pa

##
## SVAR Estimation Results:
## =====

```

```
##
##
## Estimated A matrix:
##           g_BR    g_SP    IPCA Selic.Over
## g_BR      1.00000  0.00000  0.00000      0
## g_SP      0.08142  1.00000  0.00000      0
## IPCA      0.09558  0.09524  1.00000      0
## Selic.Over 0.10572  0.10306  0.09676      1
#Função Impulso-Resposta
SVAR_SP_irf_pa <- irf(SVAR_SP_pa, impulse = "Selic.Over", response = "g_SP", n.ahead=20, ortho = TRUE)
plot(SVAR_SP_irf_pa)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs

```
#-----Período Estendido - 01-12-2010 até 01-10-2023-----#

lagselect_SP_pe <- VARselect(base_SVAR_final[107:261,c(1,14,15,16)],lag.max=10, type="both")
lagselect_SP_pe$selection

## AIC(n)  HQ(n)  SC(n) FPE(n)
##      4      2      1      4

lagselect_SP_pe$criteria

##           1           2           3           4           5
## AIC(n) -3.214089e+01 -3.245577e+01 -3.246674e+01 -3.253224e+01 -3.243688e+01
## HQ(n)  -3.194069e+01 -3.212211e+01 -3.199961e+01 -3.193164e+01 -3.170281e+01
## SC(n)  -3.164819e+01 -3.163461e+01 -3.131711e+01 -3.105414e+01 -3.063031e+01
## FPE(n)  1.100201e-14  8.035577e-15  7.960151e-15  7.475884e-15  8.259329e-15
```

```

##              6              7              8              9              10
## AIC(n) -3.241605e+01 -3.238769e+01 -3.229559e+01 -3.223639e+01 -3.209839e+01
## HQ(n) -3.154851e+01 -3.138669e+01 -3.116112e+01 -3.096846e+01 -3.069698e+01
## SC(n) -3.028101e+01 -2.992419e+01 -2.950362e+01 -2.911596e+01 -2.864948e+01
## FPE(n) 8.485873e-15 8.805065e-15 9.764426e-15 1.051139e-14 1.228854e-14

#Estimando o VAR Reduzido
VAR_SP_pe <- VAR(base_SVAR_final[107:261,c(1,14,15,16)], p = lagselect_SP_pe$selection[1], season = NULL)
VAR_SP_pe

##
## VAR Estimation Results:
## =====
##
## Estimated coefficients for equation g_BR:
## =====
## Call:
## g_BR = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
## -0.219206090 -0.004810144  0.373982993 -0.005712785  0.146423166
##      g_SP.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SP.l3
## -0.364140126 -0.082551376  0.021213794 -0.138122056 -0.040239744
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SP.l4      IPCA.l4
## -0.394465179  0.003151522 -0.175745658  0.030446776 -1.693332522
## Selic.Over.l4      const
## -0.024827873  0.011064278
##
##
## Estimated coefficients for equation g_SP:
## =====
## Call:
## g_SP = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.450544278 -0.600018291  0.265531869 -0.014144862  0.367873304
##      g_SP.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SP.l3
## -0.645553962 -0.093545130 -0.002219787  0.055782886 -0.180664788
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SP.l4      IPCA.l4
## -0.251699568  0.021631578 -0.032747654 -0.051101843 -1.545435153
## Selic.Over.l4      const
## -0.012063806  0.010287539
##
##
## Estimated coefficients for equation IPCA:
## =====
## Call:
## IPCA = g_BR.l1 + g_SP.l1 + IPCA.l1 + Selic.Over.l1 + g_BR.l2 + g_SP.l2 + IPCA.l2 + Selic.Over.l2 + g
##
##      g_BR.l1      g_SP.l1      IPCA.l1 Selic.Over.l1      g_BR.l2
##  0.0129722638 -0.0151915026  0.5678243117  0.0006559568  0.0177498332
##      g_SP.l2      IPCA.l2 Selic.Over.l2      g_BR.l3      g_SP.l3
## -0.0150107305  0.0363456534  0.0063300194  0.0088823861 -0.0030698312
##      IPCA.l3 Selic.Over.l3      g_BR.l4      g_SP.l4      IPCA.l4
##  0.0301811718 -0.0033168994 -0.0024187804  0.0083152446 -0.1443844992

```



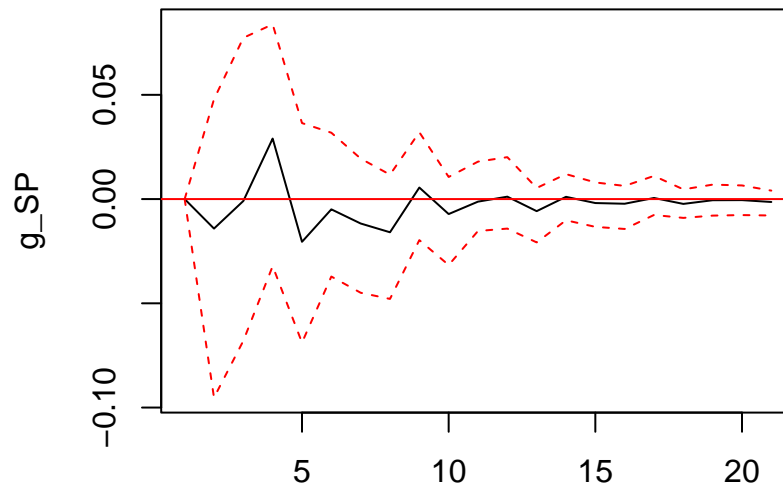
```
##
## ARCH test (univariate)
##
## data: Residual of Selic.Over equation
## Chi-squared = 53.497, df = 16, p-value = 6.263e-06
##
##
## ARCH (multivariate)
##
## data: Residuals of VAR object VAR_SP
## Chi-squared = 716.28, df = 500, p-value = 6.556e-10

#Estimando o VAR Estrutural
SVAR_SP_pe <- SVAR(VAR_SP_pe, Amat=amat, Bmat = NULL, hessian = TRUE, estmethod="scoring")
SVAR_SP_pe

##
## SVAR Estimation Results:
## =====
##
##
## Estimated A matrix:
##          g_BR    g_SP    IPCA Selic.Over
## g_BR      1.00000 0.00000 0.00000         0
## g_SP      0.01196 1.00000 0.00000         0
## IPCA      0.08378 0.08105 1.00000         0
## Selic.Over 0.07179 0.07132 0.09542         1

#Função Impulso-Resposta
SVAR_SP_irf_pe <- irf(SVAR_SP_pe, impulse = "Selic.Over", response = "g_SP", n.ahead=20, ortho = TRUE)
plot(SVAR_SP_irf_pe)
```

SVAR Impulse Response from Selic.Over



95 % Bootstrap CI, 100 runs