

Topic: Bayesian VAR models

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Bayesian Statistics, the 2<sup>nd</sup> Midterm

# What is VAR? Why VAR?

$$y_t = A_1 y_{t-1} + \dots + A_p y_{t-p} + C x_t + \epsilon_t$$

#### where

- $y_t = (y_{1t}, y_{2t}, ..., y_{Kt})$ ' is a k x 1 vector of endogenous variables,
- $x_t = (x_{1t}, x_{2t, ...,} x_{dt})$ ' is a d x 1 vector of exogenous variables,
- $A_1, \ldots, A_p$  are k x k matrices of lag coefficients to be estimated,
- C is a k x d matrix of exogenous variable coefficients to be estimated,
- $e_t = (e_{1t_i}, e_{2t_i, ..., e_{Kt}})$ ' is a k x 1 white noise innovation process, with  $E(e_t) = 0$ ,  $E(e_t, e_t) = \sum_e$ , and  $E(e_t, e_t) = 0$  for  $s \neq t$ .

## Suppose that we have VAR(p) model given by:

$$y_t = a_0 + \sum_{j=1}^p A_j y_{t-j} + \epsilon_t$$

where  $y_t$  for t = 1,...,T is an vector containing observations on m different series and  $e_t$  is an vector of errors where we assume  $e_t$  is i.i.d.  $N(0, \Sigma_e)$ . For compactness we may rewrite the model as:

$$Y = XA + E$$
.

or

$$y = (I_m \otimes X)\theta + e$$

where Y and E are T x m matrices and X =  $(x_1, ..., x_t)$ ' is a T x (mp + 1) matrix for  $x_t = (1, y'_{t-1}, ..., y'_{t-q})$ ,  $I_m$  is the identity matrix of dimention m,  $\theta = \text{vec}(A)$ , and  $e \sim N(0, \Sigma_e \otimes I_T)$ . The likelihood function is:

$$l(\theta, \Sigma_{\epsilon}) \propto \left| \Sigma_{\epsilon} \otimes I_T \right|^{-1/2} \exp \left\{ -\frac{1}{2} (y - (I_m \otimes X)\theta)' (\Sigma_{\epsilon} \otimes I_T)^{-1} (y - (I_m \otimes X)\theta) \right\}$$

To illustrate how to derive the posterior moments, let us assume  $\Sigma_{\epsilon}$  is known and a multivariate normal prior for  $\theta$ :

$$\Pi(\theta) \propto |V_0|^{-1/2} \exp \left\{ -\frac{1}{2} (\theta - \theta_0)' V_0^{-1} (\theta - \theta_0) \right\}$$

where  $\theta_0$  is the prior mean and  $V_0$  is the prior covariance.

## Suppose that we have VAR(p) model given by:

When we combine this prior with the likelihood function given above, the posterior density can be written as:

$$\Pi(\theta | y) = \exp \left\{ -\frac{1}{2} \cdot ((V_0^{-1/2}(\theta - \theta_0))'(V_0^{-1/2}(\theta - \theta_0))'(V_0^{-1/2}(\theta$$

$$+\left.\left\{(\boldsymbol{\Sigma}_{\epsilon}^{-1/2}\otimes\boldsymbol{I}_{T})\boldsymbol{y}-(\boldsymbol{\Sigma}_{\epsilon}^{-1/2}\otimes\boldsymbol{X})\boldsymbol{\theta}\right\}'\left\{(\boldsymbol{\Sigma}_{\epsilon}^{-1/2}\otimes\boldsymbol{I}_{T})\boldsymbol{y}-(\boldsymbol{\Sigma}_{\epsilon}^{-1/2}\otimes\boldsymbol{X})\boldsymbol{\theta}\right\}\right)\right\}$$

which is a multivariate normal pdf. For simplicity, define

$$w \equiv \begin{bmatrix} V_0^{-1/2} \theta_0 \\ (\Sigma_{\epsilon}^{-1/2} \otimes I_T) y \end{bmatrix}$$

$$W \equiv \begin{bmatrix} V_0^{-1/2} \\ (\Sigma_{\epsilon}^{-1/2} \otimes X) \end{bmatrix}$$

Then the exponent in above equation can be written as

$$\Pi(\theta \mid y) \propto \exp \left\{ -\frac{1}{2} (w - W\theta)'(w - W\theta) \right\} \propto$$

$$\exp\left\{-\frac{1}{2}(\theta-\bar{\theta})'W'W(\theta-\bar{\theta})+(w-W\bar{\theta})'(w-W\bar{\theta})\right\}$$

where the posterior mean  $\bar{\theta}$  is

$$\bar{\theta} = (W'W)^{-1}W'w = [V_0^{-1} + (\Sigma_{\epsilon}^{-1} \otimes X'X)]^{-1}[V_0^{-1}\theta_0 + (\Sigma_{\epsilon}^{-1} \otimes X)'y]$$

Since is known, the last part in exponent equation has no randomness about  $\bar{\theta}$ . The posterior therefore may be summarized as:

$$\pi(\theta | y) \propto \exp\left\{-\frac{1}{2}(\theta - \bar{\theta})' W' W(\theta - \bar{\theta})\right\}$$
$$= \exp\left\{-\frac{1}{2}(\theta - \bar{\theta})' \bar{V}^{-1}(\theta - \bar{\theta})\right\}$$

and the posterior covariance  $\overline{V}$  is given as:

$$\overline{V} = [V_0^{-1} + (\Sigma_{\epsilon}^{-1} \otimes X'X)]^{-1}.$$

## **Prior Selection**

- 1. The Litterman/Minnesota prior: A normal prior on  $\beta$  with fixed  $\Sigma$ .
- 2. The normal-flat prior: A normal prior on  $\beta$  that is independent of the distribution for  $\Sigma$ .
- 3. The normal-Wishart prior: A normal prior on  $^{\beta}$  and a Wishart prior on  $^{\Sigma}$ .
- 4. The independent normal-Wishart prior. A normal prior on  $\beta$  and a Wishart prior on where each endogenous equation's coefficients' distributions may be independent from each other.
- 5. The Sims-Zha normal-flat. A structural VAR equivalent of the normal-flat prior.
- 6. The Sims-Zha normal-Wishart prior. A structural VAR equivalent of the normal-Wishart prior.
- 7. The Giannone, Lenza and Primiceri prior. A prior that treats the hyper-parameters as parameters that can be selected through an optimization procedure, etc.

Often a prior is specified that simplifies the posterior analysis. In particular, it is convenient to specify the prior such that the posterior is from a known family of distributions.

#### The Litterman/Minnesota prior:

$$\theta \sim N(\theta_0, V_0)$$

 $\theta_0=0$  (where the hyper-parameter  $\mu_1=0$ , which indicates a zero mean model) and nonzero prior covariance  $V_0\neq 0$ . Note that although the choice of zero mean could lessen the risk of over-fitting, theoretically any value for  $\mu_1$  is possible.

 $V_0$  is assumed to be a diagonal matrix. The diagonal elements corresponding to endogenous variables,  $v_{i,i}$  at lag I are specified by:

$$v_{ij}^{l} = \begin{cases} \left(\frac{\lambda_{1}}{l^{\lambda_{3}}}\right)^{2} & \text{for } (i = j) \\ \left(\frac{\lambda_{1}\lambda_{2}\sigma_{i}}{l^{\lambda_{3}}\sigma_{j}}\right)^{2} & \text{for } (i \neq j) \end{cases}$$

Given this choice of prior, the posterior for  $\theta$  takes the form

$$\theta \sim N(\bar{\theta}, \bar{V})$$

where

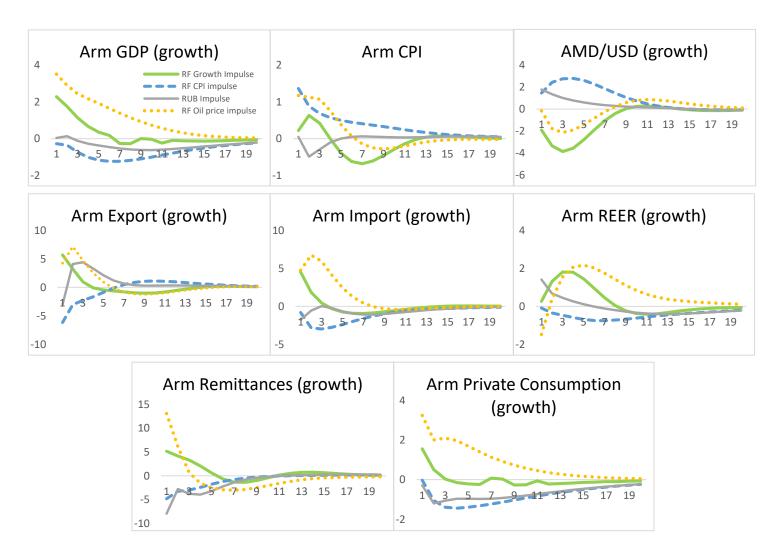
$$\overline{V} = \left[ V_0^{-1} + (\hat{\Sigma}_{\epsilon}^{-1} \otimes XX) \right]^{-1}$$

and

$$\bar{\theta} = \overline{V}[V_0^{-1}\theta_0 + (\hat{\Sigma}_{\epsilon}^{-1} \otimes X)'y]$$

#### A numerical comparison: SVAR vs BVAR

The assessment of the impact of the economic shock in Russia by a SVAR auto-regression model



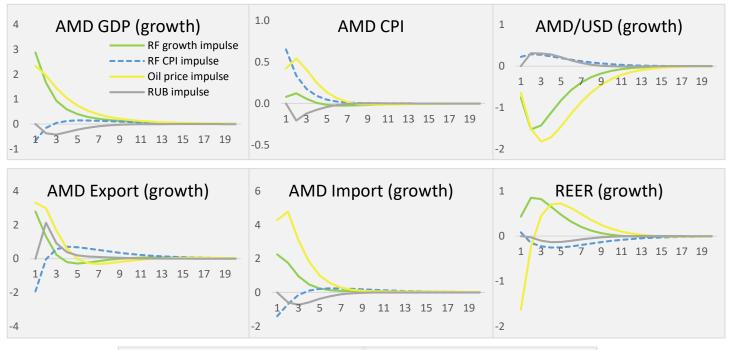
#### Highlight

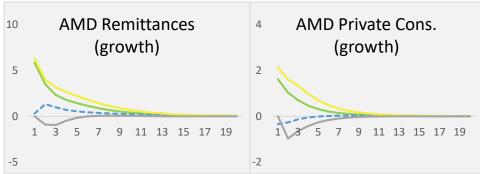
- According to the results of the analysis, the one unit standard deviation change (i.e. 3.1pp change) of the economic growth of the Russian Federation in the short run leads to a change of the RA GDP growth by up to 2.28 percentage points. The impact of the unit sd change (25.8pp) in oil prices (3.5pp) on the GDP growth of the Republic of Armenia is also transmitted by influencing other variables. However, it fades quickly after 1-3 quarters.
- Meanwhile, the effects of the RUB/USD exchange rate and the GDP growth are more stable. Among the mechanisms of transmission of the economic shock in Russia, the impact of the growth rates of remittances, as well as imports and exports is significant. In the case of the former, 1 unit sd shock effect reaches up to 13.1 percentage points. The impact on the growth of private consumption in Armenia is similar to that of GDP growth.
- In a short run (1 quarter) the one unit standard deviation increase (9.8pp) of the RUB/USD exchange rate shocks the CPI of the Republic of Armenia positively (-0.5 percentage points), while in a two year period it leads to a growth of up to 0.06 percentage points.
- The most significant impact on RA CPI has the inflation shock in Russia (up to 1.36pp), while the most significant impact on the AMD / USD exchange rate has the inflation shock in Russia (up to 2.8 percentage points). In a 1-1.5 year period the REER is also significantly affected by the oil price and RF GDP growth, while the impact of RUB / USD exchange rate shock is seen instantly.

Source: Authors' calculations.

#### A numerical comparison: SVAR vs BVAR

The Bayesian VAR model





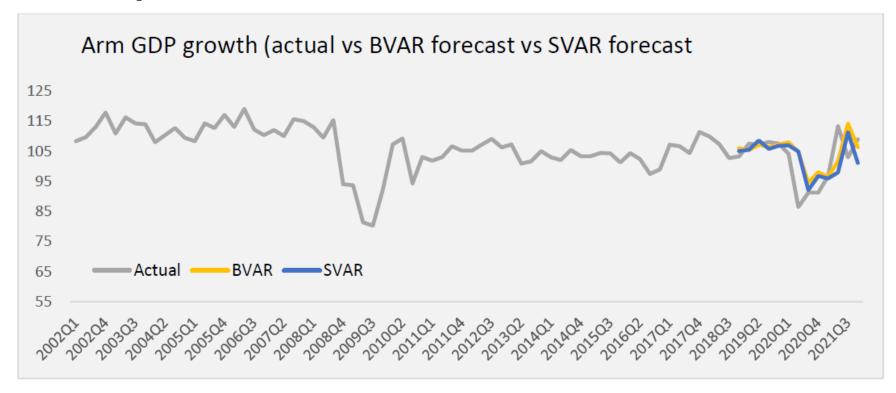
#### Highlight

- An one lag Bayesian VAR model was developed with Minessota/Litterman prior and with the following parameters: Mul (AR(1) coefficient for prior) = 0,  $\lambda 1 = 0.1$ ,  $\lambda 2 = 0.99$ , and  $\lambda 3 = 1$ .
- Initial residual covariance option is selected to be Univariate AR estimate.
- The directions and amplitudes of the effects in the two models are almost the same.
- To compare the performance of the two models, a well-known method in time series analysis of out of sample forecasting is applied.

Source: Authors' calculations.

#### A numerical comparison: SVAR vs BVAR

Out of sample forecasts



RMSE indicator for the BVAR model out of sample forecasts (7.26) is smaller than the RMSE indicator (7.64) for the SVAR model out of sample forecasts

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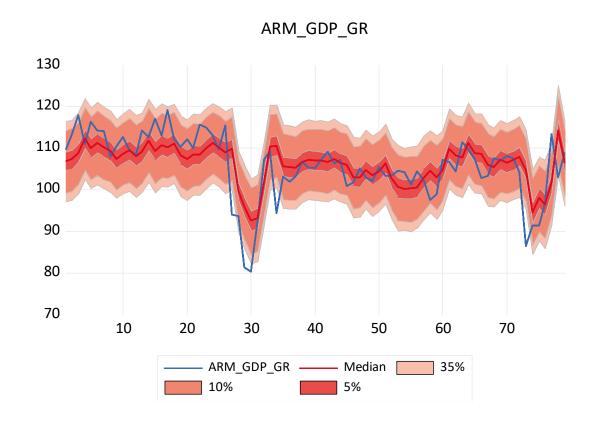
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#### Appendix 1



Forecast Evaluation

Date: 05/22/22 Time: 20:45 Sample: 2002Q1 2021Q4 Included observations: 80

Variable	Inc. obs.	RMSE	MAE	MAPE	Theil
OIL_GR	80	26.68002	20.26156	20.24664	0.116186
USD RUB GR	80	10.81261	7.438419	6.737471	0.050401
RUS CPI INF	80	2.562801	1.949276	82.52901	0.389733
RUS_GDP_GR	80	3.791897	2.217284	401.3456	0.427244
USD_AMD_GR	80	5.833716	4.523339	160.7803	0.293975
REER_GR	80	4.697310	3.873658	3.848860	0.023063
EXPORT_GR	80	24.32010	16.67623	102.7956	0.750270
IMPORT_GR	80	11.54012	8.430824	8.470199	0.054583
REM_GR	80	26.91111	19.72932	108.9811	0.507154
CPI_INF_ARM	80	3.318230	2.792331	202.7482	0.500309
PCONS_GR	80	6.114204	4.295627	4.232168	0.029192
ARM_GDP_GR	80	5.530471	3.907050	3.839054	0.026038

RMSE: Root Mean Square Error

MAE: Mean Absolute Error

MAPE: Mean Absolute Percentage Error

Theil: Theil inequality coefficient

#### Appendix 2

### ADF tests for stationarity

Null Hypothesis: ARM\_GDP\_GR has a unit root

Exogenous: Constant

Lag Length: 0 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	iller test statistic 1% level 5% level 10% level	-3.928795 -3.511262 -2.896779 -2.585626	0.0029

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: IMPORT\_GR has a unit root

Exogenous: Constant

Lag Length: 7 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ıller test statistic 1% level 5% level 10% level	-3.653729 -3.519050 -2.900137 -2.587409	0.0068

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: REER\_GR has a unit root

Exogenous: Constant

Lag Length: 9 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ıller test statistic 1% level 5% level 10% level	-3.430760 -3.521579 -2.901217 -2.587981	0.0129

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: CPI\_INF\_ARM has a unit root

Exogenous: Constant

Lag Length: 4 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ıller test statistic 1% level 5% level 10% level	-2.747058 -3.515536 -2.898623 -2.586605	0.0708

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: OIL GR has a unit root

Exogenous: Constant

Lag Length: 6 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ıller test statistic 1% level 5% level 10% level	-3.292129 -3.517847 -2.899619 -2.587134	0.0186

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: REM GR has a unit root

Exogenous: Constant

Lag Length: 4 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ıller test statistic 1% level 5% level 10% level	-2.332564 -3.515536 -2.898623 -2.586605	0.1645

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: EXPORT\_GR has a unit root

Exogenous: Constant

Lag Length: 5 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	uller test statistic 1% level 5% level 10% level	-2.072311 -3.516676 -2.899115 -2.586866	0.2563

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: PCONS\_GR has a unit root

Exogenous: Constant

Lag Length: 4 (Automatic - based on AIC, maxlag=11)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	uller test statistic 1% level 5% level 10% level	-2.981995 -3.515536 -2.898623 -2.586605	0.0410

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

Null Hypothesis: RUS\_CPI\_INF has a unit root

Exogenous: Constant

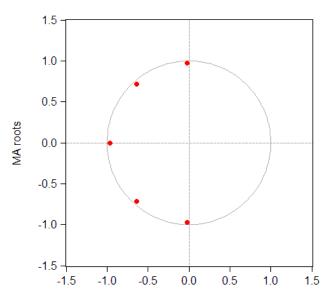
Lag Length: 8 (Automatic - based on AIC, maxlag=11)

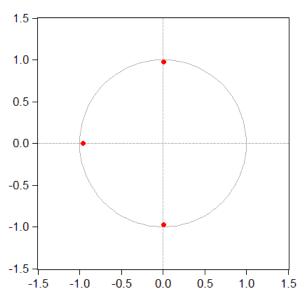
		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ıller test statistic 1% level 5% level 10% level	-2.208425 -3.520307 -2.900670 -2.587691	0.2051

<sup>\*</sup>MacKinnon (1996) one-sided p-values.

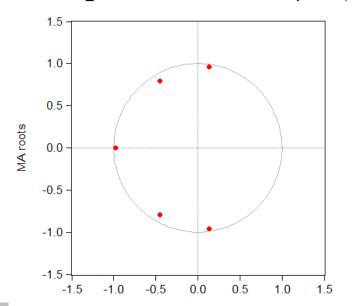
## **Stability conditions: characteristic root test**

RUS\_GDP\_GR: Inverse Roots of AR/MA Polynomial(s) IMPORT\_GR: Inverse Roots of AR/MA Polynomial(s)

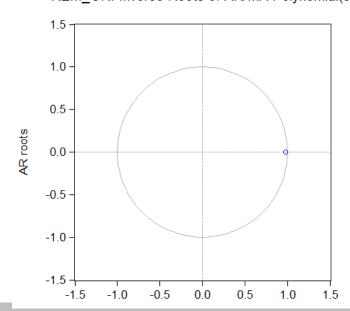


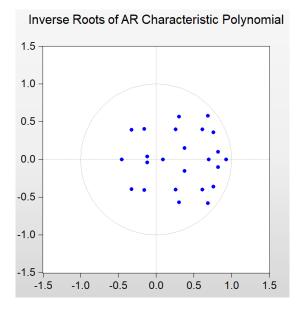


PCONS\_GR: Inverse Roots of AR/MA Polynomial(s)









VAR Lag Order Selection Criteria

Endogenous variables: OIL\_GR USD\_RUB\_GR RUS\_CPI\_INF RUS\_GDP\_GR U...

Exogenous variables: C Date: 03/29/22 Time: 22:16 Sample: 2002Q1 2021Q4 Included observations: 77

\*Note: selection calculation does not impose restricted VAR coefficient restrictions

Lag	LogL**	LR	FPE	AIC	SC	HQ
0 1 2	-2708.768	NA 713.4435 230.6612 243.3121*	1.89e+19	78.84535 78.14981	83.59383* 87.28152	80.74470* 81.80242

\* indicates lag order selected by the criterion LR: sequential modified LR test statistic (each test at 5% level)

FPE: Final prediction error AIC: Akaike information criterion SC: Schwarz information criterion HQ: Hannan-Quinn information criterion

	OIL_GR	USD_RUB	. RUS_CPI_INF	RUS_GDP	USD_AMD	REER_GR	EXPORT_GR	IMPORT_GR	REM_GR	CPI_INF_ARM	PCONS_GR	ARM_GDP
OIL_GR(-1)	0.862695 (0.12705) [ 6.79022]	-0.094086 (0.05534) [-1.70009]	0.009626 (0.01280) [ 0.75217]	0.040879 (0.02045) [ 1.99894]	0.019439 (0.02911) [ 0.66769]	-0.001769 (0.02205) [-0.08025]	0.257135 (0.11344) [ 2.26672]	0.189621 (0.06401) [ 2.96230]	-0.039455 (0.14303) [-0.27585]	-0.021164 (0.01709) [-1.23809]	0.030845 (0.03462) [ 0.89086]	0.025363 (0.03108) [ 0.81601]
USD_RUB_GR(-1)	0.074828 (0.23209) [ 0.32241]	0.626706 (0.10110) [ 6.19908]	0.007424 (0.02338) [ 0.31753]	0.059872 (0.03736) [ 1.60268]	-0.061327 (0.05959) [-1.02917]	-0.047560 (0.04537) [-1.04834]	0.725994 (0.23436) [ 3.09775]	Inverse	Roots of A	R Characte	ristic Polyr	nomia 4791 3194) 7736]
RUS_CPI_INF(-1)	-1.543173 (0.80140) [-1.92559]	0.741801 (0.34909) [ 2.12498]	0.806875 (0.08073) [ 9.99491]	-0.297732 (0.12900) [-2.30808]	0.497768 (0.20264) [ 2.45645]	-0.062755 (0.15417) [-0.40704]	0.183926 (0.79607) [ 0.23104]	1.0 -				2812  133)  1454]
RUS_GDP_GR(-1)	-0.973906 (0.62625) [-1.55513]	0.104122 (0.27279) [ 0.38169]	0.049642 (0.06309) [ 0.78690]	0.591557 (0.10080) [ 5.86843]	-0.549293 (0.18390) [-2.98685]	0.292996 (0.14071) [ 2.08232]	0.218614 (0.72945) [ 0.29970]	0.5 -		_	•	2427 3625) )363]
USD_AMD_GR(-1)	0.000000	0.000000	0.000000	0.000000	0.680461 (0.07016) [ 9.69923]	0.097158 (0.05453) [ 1.78182]	-0.141098 (0.28577) [-0.49376]	0.0				5877 3461) 3096]
REER_GR(-1)	0.000000	0.000000	0.000000	0.000000	0.023191 (0.09355) [ 0.24789]	0.871843 (0.07271) [ 11.9907]	0.072911 (0.38106) [ 0.19134]	0.5 –				6970 3615) 3126]
EXPORT_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.049283 (0.02491) [-1.97813]	0.016018 (0.01936) [ 0.82723]	0.467444 (0.10148) [ 4.60623]	1.0 -				2595 2294) 2064]
IMPORT_GR(-1)	0.000000	0.000000	0.000000	0.000000	0.121662 (0.05721) [ 2.12647]	-0.028456 (0.04447) [-0.63991]	-0.119674 (0.23305) [-0.51352]	1.5 <del> </del> -1.5 [ 3.63553]	-1.0 -0.5 [-0.90877]	0.0 0	[2.57364]	1.5 7339 5269) [ 0.70866]
REM_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.027448 (0.01827) [-1.50242]	0.026756 (0.01420) [ 1.88430]	0.180934 (0.07442) [ 2.43140]	0.045609 (( [ 1	0.349152 NA NA	0.008964	0.014587	-0.021519 (0.01682) [-1.27902]
CPI_INF_ARM(-1)	0.000000	0.000000	0.000000	0.000000	-0.536171 (0.13909) [-3.85484]	0.229451 (0.10810) [ 2.12249]	0.211374 (0.56655) [ 0.37309]	0 NA (0 NA (1 NA (1 NA	NA NA NA NA NA NA NA NA	0 0 0 0 0 0 NA NA	0 0 0 NA	-0.107020 (0.12809) [-0.83549]
PCONS_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.119523 (0.10696) [-1.11746]	-0.102328 (0.08313) [-1.23092]	1.423155 (0.43567) [ 3.26657]	-0 NA NA NA NA	NA NA NA NA NA NA	NA NA NA NA NA NA	NA NA NA	0.085839 (0.09850) [ 0.87144]
ARM_GDP_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.077829 (0.11034) [-0.70534]	0.155314 (0.08576) [ 1.81101]	-1.343934 (0.44946) [-2.99014]	0.470004 (0.23864) [ 2.00335]	0.12/115 (0.51023) [ 0.24913]	0.06500 (0.06388) [1.38715]	0.230250 (0.13023) [ 1.81413]	0.610738 (0.10162) [ 6.01012]
С	7.151472 (35.9386) [ 0.19899]	50.45731 (15.6546) [ 3.22316]	-1.898176 (3.62023) [-0.52432]	-11.06179 (5.78474) [-1.91224]	10.26415 (15.1999) [ 0.67528]	15.58141 (11.7254) [ 1.32886]	-106.1520 (61.1363) [-1.73632]	-24.36057 (33.0345) [-0.73743]	67.61116 (71.6013) [ 0.94427]	18.66452 (8.83630) [ 2.11226]	23.86645 (17.9795) [ 1.32742]	20.15908 (14.6877) [ 1.37252]
R-squared Sum sq. resids Mean dependent S.D. dependent	0.562692 43872.11 111.9310 35.86355	0.659894 8324.359 105.9113 17.71417	0.656453 445.1839 -0.697553 4.075954	0.540163 1136.671 -0.006524 5.629475	0.815904 1991.301 0.027317 11.77604	0.765264 1132.198 101.6277 7.863646	0.483588 29790.13 -0.496203 27.19509	0.572080 9674.696 104.3785 17.02512	0.511285 48918.90 -0.058780 35.82310	0.566918 689.4776 0.231481 4.517809	0.438782 2822.923 104.4136 8.030388	0.497126 2360.734 105.8241 7.757944
Determinant resid covaria Log likelihood Akaike information criterion Schwarz criterion Number of coefficients		7.66E+18 -2977.504 78.51909 82.23822 124										

oil\_gr usd\_rub\_gr d\_rus\_cpi\_inf d\_rus\_gdp\_gr d\_usd\_amd\_gr reer\_gr d\_export\_gr import\_gr d\_rem\_gr

d\_cpi\_inf\_arm pcons\_gr arm\_gdp\_gr

0	0	0	0	0	0	1
0	0	0	0	0	0	
0	0	0	0	0	0	
0	0	0	0	0	0	
NA	NA	NA	NA	NA	NA	
NA	NA	NA	NA	NA	NA	
NA	NA	NA	NA	NA	NA	
NA	NA	NA	NA	NA	NA	
<						>

VAR Lag Order Selection Criteria

Endogenous variables: OIL\_GR USD\_RUB\_GR D\_RUS\_CPI\_INF D\_RUS\_GDP...

Exogenous variables: C Date: 03/23/22 Time: 23:53 Sample: 2001Q1 2021Q4 Included observations: 78

\*Note: selection calculation does not impose restricted VAR coefficient restrictions

Lag	LogL**	LR	FPE	AIC	SC	HQ
0	-3077.016	NA	4.06e+19	79.20553	79.56810	79.35067
1	-2712.508	607.5134	1.48e+17	73.55148	78.26489*	75.43834
2	-2551.128	219.3110	1.21e+17	73.10584	82.17010	76.73443
3	-2360.926	199.9558	7.52e+16	71.92118	85.33629	77.29149
4	-2079.391	209.3466*	1.16e+16	68.39464	86.16060	75.50668
5	-1719.711	156.7836	2.05e+15*	62.86438*	84.98118	71.71814*

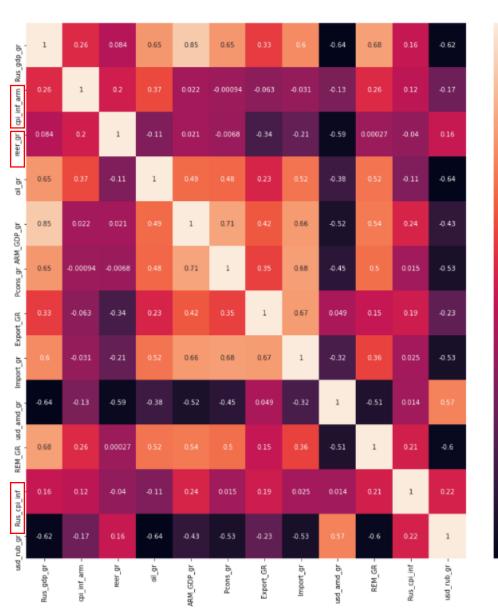
\* indicates lag order selected by the criterion

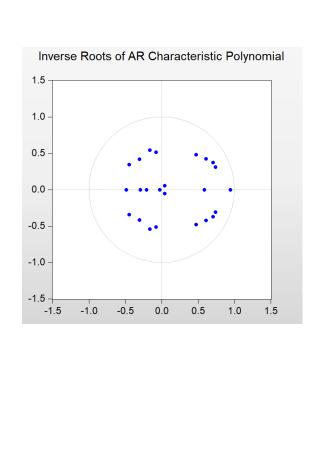
LR: sequential modified LR test statistic (each test at 5% level)

FPE: Final prediction error

AIC: Akaike information criterion SC: Schwarz information criterion

HQ: Hannan-Quinn information criterion





- 0.6

- 0.4

- 0.2

0.0

- -0.2

- -0.4



VAR Residual Cross-Correlations

Ordered by variables Date: 03/29/22 Time: 22:38

Sample: 2002Q1 2021Q4 Included observations: 79

	OIL_GR	USD_RUB_GR	RUS_CPI_INF	RUS_GDP_GR	USD_AMD_GR	REER_GR	EXPORT_GR	IMPORT_GR	REM_GR	CPI_INF_ARM	PCONS_GR	ARM_GDP_GR
OIL_GR	1.000000	-0.479793	0.137369	0.598651	-0.030391	-0.356869	0.198438	0.392353	0.480466	0.363486	0.495655	0.587781
OIL_GR(-1)	0.217935	-0.172486	-0.078645	0.250844	-0.030880	-0.153771	0.029636	0.025654	0.082006	0.153598	0.012220	0.161625
USD_RUB_GR	-0.479793	1.000000	0.278993	-0.304882	0.296556	0.470214	-0.219383	-0.316501	-0.487401	-0.161256	-0.276340	-0.277017
USD_RUB_GR(-1)	-0.101114	0.118989	0.327436	-0.328996	0.245852	0.134079	-0.223676	-0.145989	-0.179335	0.103159	-0.097290	-0.250354
RUS_CPI_INF	0.137369	0.278993	1.000000	-0.176775	0.362283	0.065354	-0.291816	-0.065522	-0.210130	0.439202	0.047471	0.040911
RUS_CPI_INF(-1)	0.231223	-0.210055	0.362899	0.066517	0.060028	-0.134329	-0.022060	0.045390	0.054006	0.377522	0.115548	0.092435
RUS_GDP_GR	0.598651	-0.304882	-0.176775	1.000000	-0.355931	-0.167430	0.402680	0.536381	0.484870	0.152054	0.477786	0.651638
RUS_GDP_GR(-1)	0.021352	-0.038510	-0.196015	0.205803	-0.050801	-0.031893	0.034025	0.010998	-0.024594	-0.080096	-0.079228	0.024030
USD_AMD_GR	-0.030391	0.296556	0.362283	-0.355931	1.000000	-0.378141	-0.237142	-0.153639	-0.189160	0.058586	-0.105244	-0.211342
USD_AMD_GR(-1)	0.106349	-0.042767	0.207722	-0.134880	0.263729	-0.092475	-0.169939	-0.014267	-0.004098	0.118179	0.051650	0.061165
REER_GR	-0.356869	0.470214	0.065354	-0.167430	-0.378141	1.000000	-0.271650	-0.339641	-0.377276	0.105223	-0.297921	-0.238932
REER_GR(-1)	-0.225647	0.207827	0.171387	-0.179364	-0.037754	0.248146	-0.178528	-0.271387	-0.147921	0.044754	-0.215431	-0.243315
EXPORT_GR	0.198438	-0.219383	-0.291816	0.402680	-0.237142	-0.271650	1.000000	0.641697	0.292425	-0.163731	0.401555	0.359465
EXPORT_GR(-1)	0.132075	0.018872	-0.089094	0.143374	0.081258	0.004510	-0.017969	-0.012394	0.026342	0.043777	-0.002745	0.102097
IMPORT_GR	0.392353	-0.316501	-0.065522	0.536381	-0.153639	-0.339641	0.641697	1.000000	0.333404	-0.066453	0.628255	0.618181
IMPORT_GR(-1)	0.072753	-0.020146	-0.161621	0.239662	0.033699	-0.105441	0.050088	0.017357	0.051020	-0.026550	0.033892	0.058650
REM_GR	0.480466	-0.487401	-0.210130	0.484870	-0.189160	-0.377276	0.292425	0.333404	1.000000	0.064573	0.421201	0.321781
REM_GR(-1)	-0.056495	-0.079646	-0.206215	0.157750	-0.054597	-0.069741	0.136492	0.094253	0.035571	-0.027226	-0.112918	0.053567
CPI_INF_ARM	0.363486	-0.161256	0.439202	0.152054	0.058586	0.105223	-0.163731	-0.066453	0.064573	1.000000	0.024028	0.151288
CPI_INF_ARM(-1)	0.113121	-0.216396	0.191170	0.191454	-0.027693	-0.160262	0.045857	0.121516	0.046279	0.200005	0.048767	0.194736
PCONS_GR	0.495655	-0.276340	0.047471	0.477786	-0.105244	-0.297921	0.401555	0.628255	0.421201	0.024028	1.000000	0.661120
PCONS_GR(-1)	0.037235	-0.093856	-0.129430	0.172175	-0.028649	-0.118731	0.045128	-0.001637	0.109936	-0.041803	0.038400	0.043338
ARM_GDP_GR	0.587781	-0.277017	0.040911	0.651638	-0.211342	-0.238932	0.359465	0.618181	0.321781	0.151288	0.661120	1.000000
ARM_GDP_GR(-1)	0.025108	-0.123701	-0.158350	0.160239	0.040141	-0.166430	-0.009339	-0.032356	0.058025	-0.031353	-0.009956	-0.077853

Bayesian VAR Estimates Date: 05/20/22 Time: 00:32

Sample (adjusted): 2002Q3 2021Q4 Included observations: 78 after adjustments

Prior type: Litterman / Minnesota

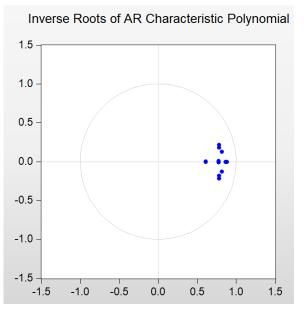
Initial residual covariance: Univariate AR

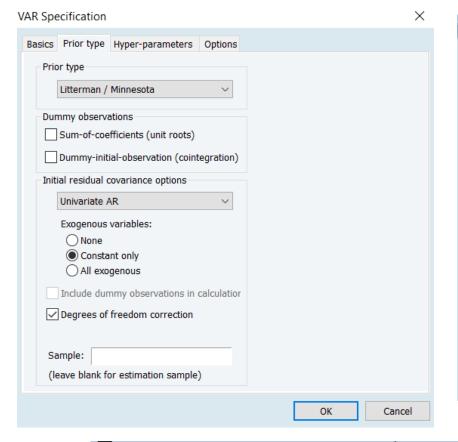
Constant included in covariance calculation Hyper-parameters: Mu1: 1, L1: 0.1, L2: 0.99, L3: 1, L4: inf

Standard errors in ( )

	OIL_GR	USD_RUB	RUS_CPI_INF	RUS_GDP	USD_AMD	REER_GR	EXPORT_GR	IMPORT_GR	REM_GR	CPI_INF_ARM	PCONS_GR	ARM_GDP
OIL_GR(-1)	0.936620	-0.007390	0.002791	0.005174	-0.001906	0.009294	0.058745	0.038547	-0.029685	0.000591	0.005023	0.008947
	(0.07874)	(0.03591)	(0.00685)	(0.01250)	(0.01782)	(0.01442)	(0.07750)	(0.04148)	(0.09883)	(0.01048)	(0.02229)	(0.01888)
OIL_GR(-2)	-0.039718	0.012890	0.000138	-0.004446	0.002454	0.003522	-0.001223	-0.002856	-0.005107	-0.002538	0.004628	-0.002053
	(0.04683)	(0.02130)	(0.00406)	(0.00742)	(0.01057)	(0.00855)	(0.04598)	(0.02460)	(0.05863)	(0.00622)	(0.01322)	(0.01120)
USD_RUB_GR(-1)	0.022612	0.894129	0.004358	-0.012398	-0.015910	-0.016862	0.073412	0.003060	0.042647	-0.002802	-0.028539	-0.018628
	(0.16886)	(0.07794)	(0.01478)	(0.02697)	(0.03845)	(0.03110)	(0.16721)	(0.08948)	(0.21323)	(0.02261)	(0.04810)	(0.04073)
USD_RUB_GR(-2)	0.019985	-0.031435	-0.006602	0.008636	-0.021629	-0.004991	0.042476	0.026855	0.026868	-0.001312	0.008052	0.010914
	(0.10015)	(0.04635)	(0.00877)	(0.01600)	(0.02281)	(0.01845)	(0.09920)	(0.05309)	(0.12650)	(0.01342)	(0.02853)	(0.02416)
RUS_CPI_INF(-1)	-0.554366	0.072975	0.903007	-0.033450	0.147493	-0.062516	0.377410	-0.160792	-0.497631	-0.010866	-0.155649	-0.034310
	(0.75976)	(0.34860)	(0.06685)	(0.12139)	(0.17304)	(0.14000)	(0.75253)	(0.40272)	(0.95963)	(0.10179)	(0.21645)	(0.18330)
RUS_CPI_INF(-2)	-0.445465	0.182147	-0.083504	-0.027615	0.004102	0.031725	-0.212295	-0.071755	-0.298627	-0.062083	-0.044919	-0.066794
	(0.50611)	(0.23221)	(0.04468)	(0.08087)	(0.11528)	(0.09325)	(0.50136)	(0.26829)	(0.63930)	(0.06780)	(0.14421)	(0.12212)
RUS_GDP_GR(-1)	-0.405576	-0.023860	0.025069	0.912341	-0.206965	0.145225	0.032104	-0.028653	0.045616	0.031134	-0.032570	-0.000799
	(0.49778)	(0.22833)	(0.04357)	(0.08005)	(0.11337)	(0.09170)	(0.49304)	(0.26384)	(0.62873)	(0.06668)	(0.14180)	(0.12009)
RUS_GDP_GR(-2)	-0.041547	0.021610	0.012649	-0.031855	-0.005230	0.017056	-0.037345	-0.031827	0.085172	0.008863	0.026149	0.001920
	(0.29216)	(0.13404)	(0.02558)	(0.04711)	(0.06654)	(0.05383)	(0.28940)	(0.15487)	(0.36903)	(0.03914)	(0.08324)	(0.07049)
USD_AMD_GR(-1)	-0.136621	-0.013881	-0.010832	0.036045	0.884908	0.028786	0.125198	0.017138	0.155917	0.043452	-0.017871	0.004414
	(0.29603)	(0.13583)	(0.02592)	(0.04732)	(0.06780)	(0.05456)	(0.29325)	(0.15693)	(0.37398)	(0.03966)	(0.08435)	(0.07143)
USD_AMD_GR(-2)	-0.082333	-0.009766	-0.012238	0.012230	-0.104829	0.028583	0.114594	0.027258	0.079999	0.000358	0.013652	-0.010402
	(0.19147)	(0.08785)	(0.01676)	(0.03060)	(0.04397)	(0.03528)	(0.18967)	(0.10150)	(0.24186)	(0.02565)	(0.05455)	(0.04620)
REER_GR(-1)	-0.250336	0.095507	0.029439	-0.074515	0.102748	0.923052	0.034164	-0.161287	-0.203717	-0.066041	0.001778	-0.046584
	(0.38156)	(0.17506)	(0.03341)	(0.06097)	(0.08693)	(0.07069)	(0.37796)	(0.20227)	(0.48197)	(0.05112)	(0.10871)	(0.09206)
REER_GR(-2)	0.076237	-0.019178	0.000405	0.001341	0.057785	-0.052241	0.042570	0.059366	-0.143301	-0.002808	0.007930	0.021286
	(0.24569)	(0.11272)	(0.02151)	(0.03926)	(0.05596)	(0.04566)	(0.24338)	(0.13024)	(0.31035)	(0.03292)	(0.07000)	(0.05928)
EXPORT_GR(-1)	0.011898	-0.010602	-0.002381	-0.002277	-0.010318	0.002674	0.846303	-0.004464	-0.129643	0.003654	-0.005337	-0.002176
	(0.08148)	(0.03738)	(0.00713)	(0.01302)	(0.01856)	(0.01501)	(0.08125)	(0.04320)	(0.10292)	(0.01092)	(0.02321)	(0.01966)
EXPORT_GR(-2)	-0.013318	-0.006918	-0.001849	-0.001687	-0.001364	-0.001597	-0.014005	-0.007118	0.031143	-0.000972	-0.004699	-0.002852
	(0.04749)	(0.02179)	(0.00416)	(0.00759)	(0.01082)	(0.00875)	(0.04747)	(0.02517)	(0.05999)	(0.00636)	(0.01353)	(0.01146)
IMPORT_GR(-1)	-0.089623	0.022182	-0.004181	-0.010486	0.011214	0.006933	-0.137542	0.862383	-0.118968	0.005262	-0.011035	-0.016042
	(0.15714)	(0.07209)	(0.01376)	(0.02511)	(0.03579)	(0.02895)	(0.15568)	(0.08390)	(0.19849)	(0.02105)	(0.04478)	(0.03792)
IMPORT_GR(-2)	-0.041151	0.014024	0.000620	-0.011645	0.007511	0.002984	-0.047172	-0.033896	0.000643	-0.000644	-0.013419	-0.008675
	(0.08783)	(0.04030)	(0.00769)	(0.01403)	(0.02000)	(0.01618)	(0.08700)	(0.04698)	(0.11095)	(0.01177)	(0.02503)	(0.02119)
REM GR(-1)	-0 002002	-0 004290	0.001835	-0.001083	-0 017923	0.012122	0 075939	0 023207	0.818195	0 001824	-0 002295	-0 001387

## With AR (myu) = 1







				•									
View Proc	Object Print	Name Freeze	Estimate Forecas	t Stats Impulse	Resids								
	VAR Posterior Residual Covariance Matrix												
	OIL_GR	USD_RUB	. RUS_CPI_INF	RUS_GDP	USD_AMD	REER_GR	EXPORT_GR	IMPORT_GR	REM_GR	CPI_INF_ARM	PCONS_GR	ARM_GDP	Γ
OIL_GR	776.4536	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
USD	0.000000	163.4391	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
RUS	0.000000	0.000000	5.950329	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
RUS	0.000000	0.000000	0.000000	19.82564	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
USD	0.000000	0.000000	0.000000	0.000000	40.27831	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
REER	0.000000	0.000000	0.000000	0.000000	0.000000	26.35700	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
EXPO	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	761.8983	0.000000	0.000000	0.000000	0.000000	0.000000	
IMPO	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	218.1912	0.000000	0.000000	0.000000	0.000000	
REM_GR	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1238.897	0.000000	0.000000	0.000000	
CPI_IN	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	13.93528	0.000000	0.000000	
PCON	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	63.03501	0.000000	
ARM	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	45.20268	

#### Out of sample – 15%: BVAR

Bayesian VAR Estimates Date: 05/22/22 Time: 20:43

Sample (adjusted): 2002Q2 2018Q4 Included observations: 67 after adjustments

Prior type: Litterman / Minnesota Initial residual covariance: Univariate AR Constant included in covariance calculation

Hyper-parameters: Mu1: 0, L1: 0.1, L2: 0.99, L3: 1, L4: inf

	OIL_GR	USD_RUB	RUS_CPI_INF	RUS_GDP	USD_AMD	REER_GR	EXPORT_GR	IMPORT_GR	REM_GR	CPI_INF_ARM	PCONS_GR	ARM_GDP.
OIL_GR(-1)	0.293524	-0.091134	0.010843	0.019188	0.003210	0.004986	0.004689	0.058012	0.010464	0.003113	0.019023	0.019969
	(0.07770)	(0.04001)	(0.00911)	(0.01154)	(0.02614)	(0.01760)	(0.06912)	(0.03827)	(0.09735)	(0.01212)	(0.01839)	(0.01813)
USD_RUB_GR(-1)	-0.315466	0.290465	0.021763	-0.000197	0.022209	-0.001670	0.090004	-0.052254	-0.046148	-0.014309	-0.081712	-0.028996
	(0.14811)	(0.07711)	(0.01745)	(0.02212)	(0.05011)	(0.03374)	(0.13248)	(0.07335)	(0.18661)	(0.02323)	(0.03525)	(0.03475)
RUS_CPI_INF(-1)	-0.361335	0.614903	0.397810	-0.158084	0.274279	0.194118	-0.005298	-0.419528	-1.344309	0.050623	-0.230836	0.037112
	(0.59459)	(0.30790)	(0.07044)	(0.08883)	(0.20125)	(0.13552)	(0.53207)	(0.29460)	(0.74949)	(0.09330)	(0.14155)	(0.13956)
RUS_GDP_GR(-1)	0.146589	0.002253	-0.019551	0.370173	-0.537032	0.153811	0.317181	0.264157	1.188631	0.102985	-0.011769	0.183256
	(0.50067)	(0.25921)	(0.05902)	(0.07522)	(0.16948)	(0.11408)	(0.44803)	(0.24807)	(0.63110)	(0.07856)	(0.11919)	(0.11752)
USD_AMD_GR(-1)	-0.154104	0.124176	-0.001361	-0.003530	0.324649	-0.057724	-0.068362	-0.028820	-0.071965	0.028856	-0.006923	-0.015163
	(0.21280)	(0.11020)	(0.02509)	(0.03180)	(0.07241)	(0.04852)	(0.19045)	(0.10544)	(0.26827)	(0.03339)	(0.05066)	(0.04995)
REER_GR(-1)	-0.254568	0.203189	0.088948	-0.093913	-0.062291	0.424107	-0.122358	-0.296431	-0.445514	-0.047563	0.020682	-0.029814
	(0.30451)	(0.15767)	(0.03591)	(0.04549)	(0.10312)	(0.06973)	(0.27251)	(0.15089)	(0.38383)	(0.04778)	(0.07250)	(0.07147)
EXPORT_GR(-1)	0.018059	-0.003606	0.001090	0.014456	-0.029975	-0.007442	0.121440	0.043489	-0.164157	0.001521	-0.013500	-0.011553
	(0.08990)	(0.04655)	(0.01060)	(0.01343)	(0.03043)	(0.02049)	(0.08098)	(0.04455)	(0.11332)	(0.01411)	(0.02140)	(0.02110)
IMPORT_GR(-1)	0.028157	-0.030197	-0.007401	0.009415	0.000580	-0.022917	-0.027546	0.180664	-0.150208	0.020562	0.044737	0.030006
	(0.16032)	(0.08302)	(0.01890)	(0.02395)	(0.05427)	(0.03654)	(0.14349)	(0.07995)	(0.20209)	(0.02516)	(0.03817)	(0.03764)
REM_GR(-1)	-0.005394	-0.030339	-0.001197	0.005667	-0.045570	0.015293	0.047123	0.018198	0.195442	0.010152	0.007310	-0.004916
	(0.06204)	(0.03213)	(0.00731)	(0.00927)	(0.02100)	(0.01414)	(0.05552)	(0.03074)	(0.07870)	(0.00974)	(0.01477)	(0.01456)
CPI_INF_ARM(-1)	-0.260478	-0.163260	0.048840	0.005061	-0.281118	0.174095	-0.168691	-0.002092	0.921662	0.295914	-0.065726	-0.048682
	(0.47284)	(0.24484)	(0.05575)	(0.07065)	(0.16005)	(0.10776)	(0.42317)	(0.23429)	(0.59607)	(0.07461)	(0.11257)	(0.11099)
PCONS_GR(-1)	0.348757	-0.191161	0.038380	-0.026808	0.025176	0.004718	0.021125	0.131490	-0.144272	-0.012431	0.134722	0.123081
	(0.33863)	(0.17536)	(0.03992)	(0.05059)	(0.11461)	(0.07717)	(0.30303)	(0.16780)	(0.42682)	(0.05313)	(0.08115)	(0.07950)
ARM_GDP_GR(-1)	-0.013610	-0.035543	0.020483	0.014642	-0.042601	0.070306	-0.467232	0.192707	-0.010348	0.025133	0.102919	0.257854
	(0.32687)	(0.16924)	(0.03853)	(0.04884)	(0.11063)	(0.07449)	(0.29252)	(0.16198)	(0.41200)	(0.05129)	(0.07784)	(0.07718)
С	100.1233	92.44797	-18.57556	7.454350	5.389637	52.98164	51.85818	80.91392	77.58084	2.560156	79.33779	66.80055
	(58.0293)	(30.0816)	(6.84187)	(8.66960)	(19.6430)	(13.2405)	(51.9297)	(28.7573)	(73.1423)	(9.10495)	(13.8338)	(13.6336)
R-squared	0.523672	0.636889	0.609220	0.633067	0.763426	0.659882	0.205802	0.594259	0.495951	0.452737	0.577332	0.522331
di. R-squared	0.417821	0.556198	0.522380	0.551526	0.710854	0.584301	0.029314	0.504094	0.383940	0.331123	0.483406	0.416182
dum sq. resids	33406.17	8606.423	463.6288	692.2432	2483.024	1548.296	23231.31	6167.545	40905.95	784.6563	1278.437	1765.977
decension	24.87233	12.62451	2.930140	3.580408	6.780997	5.354639	20.74149	10.68709	27.52304	3.811912	4.865672	5.718678
-statistic	4.947272	7.892914	7.015419	7.763818	14.52156	8.730717	1.166094	6.590815	4.427704	3.722732	6.146664	4.920749
Mean dependent	112.3512	105.9168	-1.049787	-0.122579	-0.149283	101.6753	-1.294030	105.6015	-2.575302	-0.006210	105.0340	106.4836
dependent	32.59780	18.95046	4.239815	5.346431	12.61059	8.305017	21.05235	15.17608	35.06590	4.660898	6.769682	7.484400

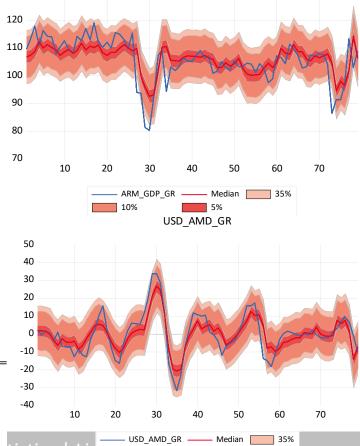
Forecast Evaluation Date: 05/22/22 Time: 20:45 Sample: 2002Q1 2021Q4 Included observations: 80

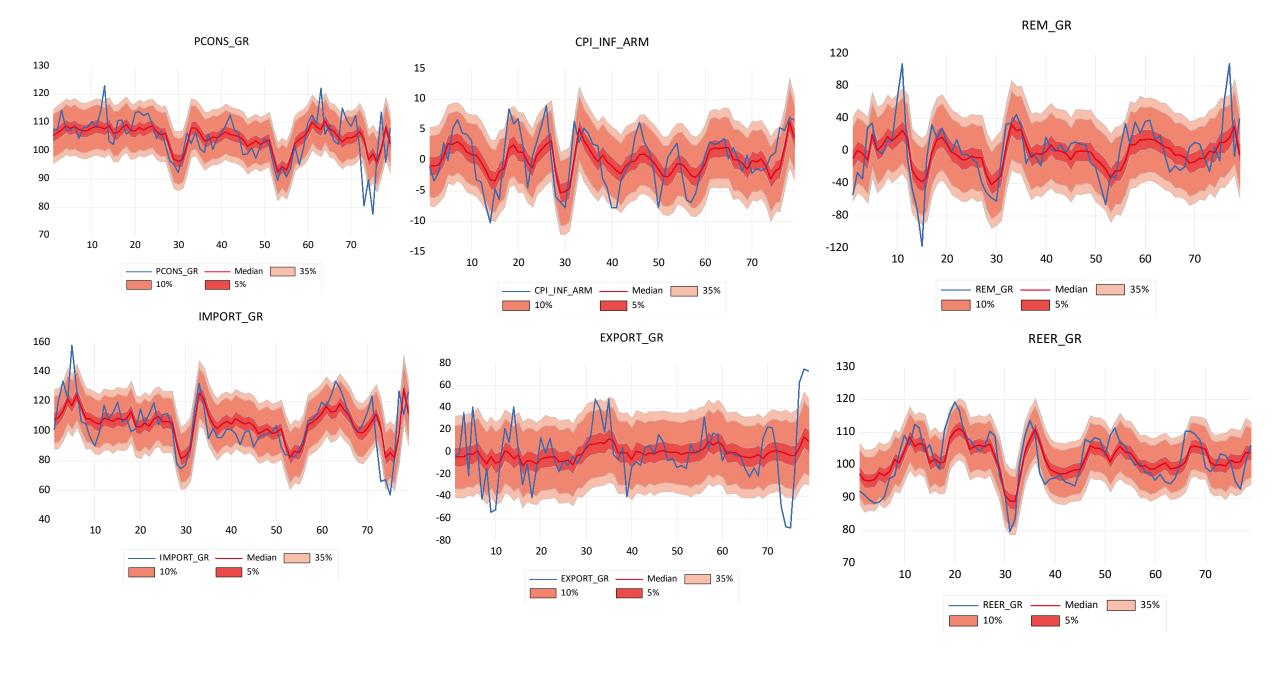
Variable	Inc. obs.	RMSE	MAE	MAPE	Theil
OIL GR	80	26.68002	20.26156	20.24664	0.116186
USD RUB GR	80	10.81261	7.438419	6.737471	0.050401
RUS CPI INF	80	2.562801	1.949276	82.52901	0.389733
RUS GDP GR	80	3.791897	2.217284	401.3456	0.427244
USD_AMD_GR	80	5.833716	4.523339	160.7803	0.293975
REER GR	80	4.697310	3.873658	3.848860	0.023063
EXPORT GR	80	24.32010	16.67623	102.7956	0.750270
IMPORT GR	80	11.54012	8.430824	8.470199	0.054583
REM GR	80	26.91111	19.72932	108.9811	0.507154
CPI INF ARM	80	3.318230	2.792331	202.7482	0.500309
PCONS GR	80	6.114204	4.295627	4.232168	0.029192
ARM_GDP_GR	80	5.530471	3.907050	3.839054	0.026038

RMSE: Root Mean Square Error MAE: Mean Absolute Error MAPE: Mean Absolute Percentage Error

130

ARM GDP GR Theil: Theil inequality coefficient





Vector Autoregression Estimates (with restrictions)
Date: 05/22/22 Time: 21:34

Out of sample - 15%: SVAR

Sample (adjusted): 2002Q2 2018Q4

Ineluded observations: 67 after adjustments

Iterated GLS convergence achieved after 15 iterations

Standard errors in ( ) & t-statistics in [ ]

Determinant resid covariance

Akaike information criterion

Log likelihood

Schwarz criterion Number of coefficients

Number of restrictions

3.11E+18

-2480.551

77.74780 81.82813

124

32

Standard errors in ( ) & t-s	statistics in [ ]											
	OIL_GR	USD_RUB	RUS_CPI_INF	RUS_GDP	USD_AMD	REER_GR	EXPORT_GR	IMPORT_GR	REM_GR	CPI_INF_ARM	PCONS_GR	ARM_GDP
OIL_GR(-1)	0.774918 (0.13173) [ 5.88274]	-0.117896 (0.06909) [-1.70639]	0.015143 (0.01607) [ 0.94203]	0.051098 (0.01906) [ 2.68065]	-0.009485 (0.03585) [-0.26462]	0.007996 (0.02638) [ 0.30306]	0.144514 (0.12319) [ 1.17308]	0.204375 (0.06232) [ 3.27971]	-0.013151 (0.15260) [-0.08618]	-0.016720 (0.02139) [-0.78165]	0.002293 (0.02855) [ 0.08034]	0.004432 (0.03339) [ 0.13271]
USD_RUB_GR(-1)	-0.038525 (0.22961) [-0.16778]	0.587677 (0.12043) [ 4.87971]	0.011785 (0.02802) [ 0.42059]	0.083348 (0.03323) [ 2.50848]	-0.144695 (0.07001) [-2.06691]	-0.023031 (0.05219) [-0.44134]	0.592824 (0.24465) [ 2.42314]	0.260929 (0.12509) [ 2.08588]	0.141160 (0.30073) [ 0.46940]	-0.058565 (0.04178) [-1.40178]	-0.178629 (0.05709) [-3.12894]	-0.034404 (0.06502) [-0.52911]
RUS_CPI_INF(-1)	-1.817179 (0.73871) [-2.45995]	0.976427 (0.38745) [ 2.52013]	0.805138 (0.09014) [ 8.93159]	-0.379311 (0.10689) [-3.54845]	0.524480 (0.22594) [ 2.32131]	-0.000122 (0.16848) [-0.00073]	0.127515 (0.78993) [ 0.16142]	-1.153052 (0.40401) [-2.85404]	-2.191142 (0.97081) [-2.25702]	0.057658 (0.13484) [ 0.42760]	-0.401978 (0.18436) [-2.18037]	-0.006999 (0.20985) [-0.03335]
RUS_GDP_GR(-1)	-0.809443 (0.57763) [-1.40132]	0.155033 (0.30297) [ 0.51172]	0.067443 (0.07049) [ 0.95679]	0.669729 (0.08359) [ 8.01246]	-0.546135 (0.20136) [-2.71224]	0.300770 (0.15182) [ 1.98104]	0.382893 (0.71427) [ 0.53606]	-0.501044 (0.36856) [-1.35945]	2.322788 (0.87207) [ 2.66352]	0.156703 (0.12018) [ 1.30389]	-0.087451 (0.16768) [-0.52154]	0.273044 (0.18651) [ 1.46398]
USD_AMD_GR(-1)	0.000000	0.000000	0.000000	0.000000	0.659923 (0.07206) [ 9.15793]	0.120687 (0.05629) [ 2.14394]	-0.407547 (0.26756) [-1.52321]	-0.121015 (0.14161) [-0.85456]	0.163455 (0.32019) [ 0.51049]	0.135621 (0.04302) [ 3.15244]	0.110642 (0.06388) [1.73212]	-0.009596 (0.06613) [-0.14509]
REER_GR(-1)	0.000000	0.000000	0.000000	0.000000	0.027259 (0.09579) [ 0.28456]	0.879111 (0.07483) [ 11.7477]	-0.078399 (0.35568) [-0.22042]	-0.263998 (0.18825) [-1.40236]	-0.343901 (0.42565) [-0.80794]	-0.043323 (0.05719) [-0.75753]	0.201215 (0.08492) [ 2.36960]	-0.004686 (0.08792) [-0.05330]
EXPORT_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.057730 (0.03133) [-1.84246]	0.021916 (0.02448) [ 0.89537]	0.289074 (0.11634) [ 2.48475]	0.086398 (0.06158) [ 1.40313]	-0.544768 (0.13923) [-3.91285]	-0.000260 (0.01871) [-0.01392]	-0.046908 (0.02777) [-1.68888]	-0.051778 (0.02876) [-1.80059]
IMPORT_GR(-1)	0.000000	0.000000	0.000000	0.000000	0.127140 (0.05879) [ 2.16255]	-0.032928 (0.04593) [-0.71696]	-0.062417 (0.21829) [-0.28593]	0.419875 (0.11554) [ 3.63416]	-0.228293 (0.26123) [-0.87390]	0.062662 (0.03510) [ 1.78528]	0.123123 (0.05211) [ 2.36253]	0.011478 (0.05396) [ 0.21273]
REM_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.058060 (0.01995) [-2.91031]	0.044156 (0.01558) [ 2.83336]	0.115878 (0.07407) [ 1.56437]	0.056173 (0.03920) [ 1.43279]	0.395683 (0.08864) [ 4.46368]	0.013587 (0.01191) [ 1.14082]	0.017463 (0.01768) [ 0.98747]	-0.014106 (0.01831) [-0.77042]
CPI_INF_ARM(-1)	0.000000	0.000000	0.000000	0.000000	-0.579157 (0.14443) [-4.01003]	0.255088 (0.11282) [ 2.26094]	-0.283543 (0.53625) [-0.52875]	0.119186 (0.28382) [ 0.41993]	1.239597 (0.64175) [1.93160]	0.598722 (0.08622) [ 6.94373]	-0.247201 (0.12803) [-1.93088]	-0.170505 (0.13255) [-1.28635]
PCONS_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.236649 (0.12654) [-1.87021]	-0.071646 (0.09885) [-0.72481]	1.406533 (0.46983) [ 2.99373]	0.065462 (0.24867) [ 0.26325]	-0.278824 (0.56225) [-0.49591]	-0.190060 (0.07554) [-2.51590]	0.006299 (0.11217) [ 0.05616]	0.082545 (0.11613) [ 0.71080]
ARM_GDP_GR(-1)	0.000000	0.000000	0.000000	0.000000	-0.059810 (0.11456) [-0.52206]	0.155206 (0.08950) [ 1.73422]	-1.517890 (0.42538) [-3.56834]	0.384004 (0.22514) [ 1.70562]	-0.075017 (0.50906) [-0.14736]	0.099443 (0.06840) [ 1.45391]	0.216428 (0.10155) [ 2.13115]	0.593799 (0.10514) [ 5.64753]
С	27.48560 (37.1448) [ 0.73996]	58.05475 (19.4825) [ 2.97985]	-3.066765 (4.53281) [-0.67657]	-14.98040 (5.37505) [-2.78703]	31.59518 (19.2356) [ 1.64254]	8.729737 (14.7918) [ 0.59017]	-50.87744 (69.9946) [-0.72688]	-11.47690 (36.6490) [-0.31316]	78.82251 (84.4977) [ 0.93284]	15.37648 (11.4825) [1.33913]	66.04982 (16.5908) [ 3.98111]	36.87282 (17.7277) [ 2.07995]
R-squared Sum sq. resids Mean dependent S.D. dependent	0.602896 27849.95 112.3512 32.59780	0.676755 7661.531 105.9168 18.95046	0.650437 414.7283 -1.049787 4.239815	0.690885 583.1667 -0.122579 5.346431	0.829382 1790.763 -0.149283 12.61059	0.790968 951.5636 101.6753 8.305017	0.295219 20615.77 -1.294030 21.05235	0.658948 5184.220 105.6015 15.17608	0.605489 32016.46 -2.575302 35.06590	0.555291 637.6153 -0.006210 4.660898	0.638112 1094.597 105.0340 6.769682	0.577702 1561.267 106.4836 7.484400
D-4iii-		0.445.40										

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ARM_GDP_GR = 0.00443192037393*OIL_GR(-1) -
0.0344044378544*USD_RUB_GR(-1) -
0.00699862447512*RUS_CPI_INF(-1) +
0.273044178892*RUS_GDP_GR(-1) -
0.00959552751209*USD_AMD_GR(-1) -
0.00468634031582*REER_GR(-1) -
0.0517783172753*EXPORT_GR(-1) +
0.0114783716861*IMPORT_GR(-1) - 0.0141057391752*REM_GR(-1) - 0.170505486638*CPI_INF_ARM(-1) +
0.0825447424991*PCONS_GR(-1) +
```

0.593798504952\*ARM\_GDP\_GR(-1) + 36.8728209358