stata.com

Title

varbasic — Fit a simple VAR and graph IRFs or FEVDs

Description	Quick start	Menu	Syntax
Options	Remarks and examples	Stored results	Methods and formulas
References	Also see		

## Description

varbasic fits a basic vector autoregressive (VAR) model and graphs the impulse-response functions (IRFs), the orthogonalized impulse-response functions (OIRFs), or the forecast-error variance decompositions (FEVDs).

# Quick start

- Fit vector autoregressive model for dependent variables y1, y2, and y3 and their first and second lags, and graph orthogonalized impulse responses using tsset data varbasic y1 y2 y3
- Same as above, but include second and third lags instead of first and second lags var y1 y2 y3, lags(2 3)
- Same as above, but produce forecast-error variance decompositions instead of impulse responses var y1 y2 y3, lags(2 3) fevd
- Same as above, but set the forecast horizon for the forecast-error variance decompositions to be 12 periods

var y1 y2 y3, lags(2 3) fevd step(12)

### Menu

Statistics > Multivariate time series > Basic VAR

### Syntax

```
varbasic depvarlist [if] [in] [, options]
```

options Description

Main

<u>lags(numlist)</u>	use lags <i>numlist</i> in the model; default is lags(1 2)
<u>i</u> rf	produce matrix graph of IRFs
<u>f</u> evd	produce matrix graph of FEVDs
nograph	do not produce a graph
<u>s</u> tep(#)	set forecast horizon # for estimating the OIRFs, IRFs, and FEVDs; default is step(8)

You must tsset your data before using varbasic; see [TS] tsset.

depvarlist may contain time-series operators; see [U] 11.4.4 Time-series varlists.

collect, rolling, statsby, and xi are allowed; see [U] 11.1.10 Prefix commands.

See [U] 20 Estimation and postestimation commands for more capabilities of estimation commands.

# Options

Main

- lags (numlist) specifies the lags to be included in the model. The default is lags (1 2). This option takes a numlist and not simply an integer for the maximum lag. For instance, lags (2) would include only the second lag in the model, whereas lags (1/2) would include both the first and second lags in the model. See [U] 11.1.8 numlist and [U] 11.4.4 Time-series varlists for more discussion of numlists and lags.
- irf causes varbasic to produce a matrix graph of the IRFs instead of a matrix graph of the OIRFs, which is produced by default.
- fevd causes varbasic to produce a matrix graph of the FEVDs instead of a matrix graph of the OIRFs, which is produced by default.
- nograph specifies that no graph be produced. The IRFs, OIRFs, and FEVDs are still estimated and saved in the IRF file \_varbasic.irf.
- step(#) specifies the forecast horizon for estimating the IRFs, OIRFs, and FEVDs. The default is eight
  periods.

### **Remarks and examples**

varbasic simplifies fitting simple VAR models and graphing the IRFs, the OIRFs, or the FEVDs. See [TS] var and [TS] var svar for fitting more advanced VAR models and structural vector autoregressive (SVAR) models. All the postestimation commands discussed in [TS] var postestimation work after varbasic.

This entry does not discuss the methods for fitting a VAR model or the methods surrounding the IRFs, OIRFs, and FEVDs. See [TS] var and [TS] irf create for more on these methods. This entry illustrates how to use varbasic to easily obtain results. It also illustrates how varbasic serves as an entry point to further analysis.

Example 1

We fit a three-variable VAR model with two lags to the German macro data used by Lütkepohl (2005). The three variables are the first difference of natural log of investment, dln\_inv; the first difference of the natural log of income, dln\_inc; and the first difference of the natural log of consumption, dln\_consump. In addition to fitting the VAR model, we want to see the OIRFs. Below we use varbasic to fit a VAR(2) model on the data from the second quarter of 1961 through the fourth quarter of 1978. By default, varbasic produces graphs of the OIRFs.

. use https://www.stata-press.com/data/r18/lutkepohl2 (Quarterly SA West German macro data, Bil DM, from Lutkepohl 1993 Table E.1)

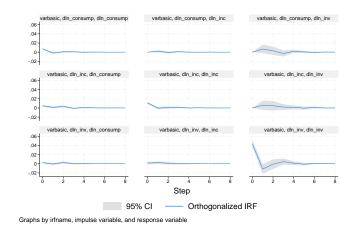
#### stata.com

. varbasic dln\_inv dln\_inc dln\_consump if qtr<=tq(1978q4)

Vector autoregression

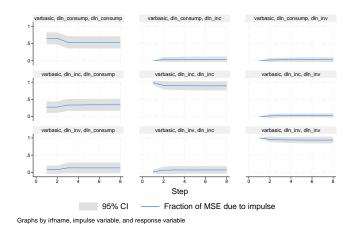
Sample: 1960q4 Log likelihood FPE Det(Sigma_ml)	u 1978q4 606.307 2.18e-11 1.23e-11			Number of AIC HQIC SBIC	f obs	= = =	73 -16.03581 -15.77323 -15.37691
Equation	Parms	RMSE	R-sq	chi2	P>chi2		
dln_inv dln_inc dln_consump	7 7 7	.046148 .011719 .009445	0.1286 0.1142 0.2513	10.76961 9.410683 24.50031	0.0958 0.1518 0.0004		

	Coefficient	Std. err.	Z	P> z	[95% conf.	interval]
dln_inv						
dln_inv						
L1.	3196318	.1192898	-2.68	0.007	5534355	0858282
L2.	1605508	.118767	-1.35	0.176	39333	.0722283
dln_inc						
L1.	.1459851	.5188451	0.28	0.778	8709326	1.162903
L2.	.1146009	.508295	0.23	0.822	881639	1.110841
dln_consump						
L1.	.9612288	.6316557	1.52	0.128	2767936	2.199251
L2.	.9344001	.6324034	1.48	0.140	3050877	2.173888
_cons	0167221	.0163796	-1.02	0.307	0488257	.0153814
dln_inc						
dln_inv						
L1.	.0439309	.0302933	1.45	0.147	0154427	.1033046
L2.	.0500302	.0301605	1.66	0.097	0090833	.1091437
dln_inc						
L1.	1527311	.131759	-1.16	0.246	4109741	.1055118
L2.	.0191634	.1290799	0.15	0.882	2338285	.2721552
dln_consump						
L1.	.2884992	.1604069	1.80	0.072	0258926	.6028909
L2.	0102	.1605968	-0.06	0.949	3249639	.3045639
_cons	.0157672	.0041596	3.79	0.000	.0076146	.0239198
dln_consump						
dln_inv						
L1.	002423	.0244142	-0.10	0.921	050274	.045428
L2.	.0338806	.0243072	1.39	0.163	0137607	.0815219
dln_inc						
L1.	.2248134	.1061884	2.12	0.034	.0166879	.4329389
L2.	.3549135	.1040292	3.41	0.001	.1510199	.558807
dln_consump						
L1.	2639695	.1292766	-2.04	0.041	517347	010592
L2.	0222264	.1294296	-0.17	0.864	2759039	.231451
_cons	.0129258	.0033523	3.86	0.000	.0063554	.0194962



Because we are also interested in looking at the FEVDs, we can use irf graph to obtain the graphs. Although the details are available in [TS] irf and [TS] irf graph, the command below produces what we want after the call to varbasic.

. irf graph fevd, lstep(1)



4

### Technical note

Stata stores the estimated IRFs, OIRFs, and FEVDs in a IRF file called \_varbasic.irf in the current working directory. varbasic replaces any \_varbasic.irf that already exists. Finally, varbasic makes \_varbasic.irf the active IRF file. This means that the graph and table commands irf graph, irf cgraph, irf ograph, irf table, and irf ctable will all display results that correspond to the VAR model fit by varbasic.

### Stored results

See Stored results in [TS] var.

### Methods and formulas

varbasic uses var and irf graph to obtain its results. See [TS] var and [TS] irf graph for a discussion of how those commands obtain their results.

### References

### Also see

- [TS] varbasic postestimation Postestimation tools for varbasic
- [TS] tsset Declare data to be time-series data
- [TS] **var** Vector autoregressive models<sup>+</sup>
- [TS] var intro Introduction to vector autoregressive models
- [TS] var ivsvar Instrumental-variables structural vector autoregressive models<sup>+</sup>
- [TS] var svar Structural vector autoregressive models
- [U] 20 Estimation and postestimation commands

Stata, Stata Press, and Mata are registered trademarks of StataCorp LLC. Stata and Stata Press are registered trademarks with the World Intellectual Property Organization of the United Nations. StataNow and NetCourseNow are trademarks of StataCorp LLC. Other brand and product names are registered trademarks or trademarks of their respective companies. Copyright © 1985–2023 StataCorp LLC, College Station, TX, USA. All rights reserved.



For suggested citations, see the FAQ on citing Stata documentation.