EViews 9 Getting Started





TStat S.r.l. Distributore Ufficiale di EViews Via Rettangolo, 12-14 67039 Sulmona - AQ Tel. 0864 210101 | Fax. 0864 206014 tstat@tstat.it | www.tstat.it EViews 9 features a wide range of exciting changes and improvements. The following is an overview of the most important new features in Version 9.

General EViews Interface

- Command capture from the interactive interface ("Command Capture," on page 13).
- Dockable command and capture window ("Window Docking," on page 14)
- Database and workfile object preview ("Database and Workfile Object Preview," on page 15).

The following are mentioned elsewhere in this document, but touch on interface in important ways.

- Graph pan/zoom ("Graph Pan and Zoom" on page 21).
- Multiple graph viewing slideshow ("Multi-graph Slideshow" on page 23).

Data Handling

- Enhanced import and linking of external data ("Enhanced import and linking of data," on page 16).
- A powerful new FRED database interface ("New FRED Database Interface" on page 17).
- Direct read and write access to data stored on cloud drive services ("Cloud Drive Support," on page 18).
- Dated data table template support for saving and importing customized settings ("Dated Data Table Templates," on page 19).
- New frequency conversion methods ("Expanded Frequency Conversion Interpolation Methods" on page 20).

Graphs, Tables, and Spools

- Graph pan/zoom ("Graph Pan and Zoom" on page 21).
- Multiple graph viewing slideshow ("Multi-graph Slideshow" on page 23).
- Improved mixed graph type ("Improved Mixed Graph Type" on page 26).
- Rectangle and ellipse drawing ("Rectangle and Ellipse Drawing" on page 27).

- Data-based anchoring of arrows, rectangles, and ellipses ("Data-based Anchoring" on page 28).
- Tables, graphs, and spools may now be saved in LaTeX format ("LaTeX Output" on page 28).

Econometrics and Statistics

Computation

- Automatic ARIMA forecasting of a series ("Automatic ARIMA Forecasting" on page 30).
- Forecast evaluation and combination testing ("Forecast Evaluation" on page 30).
- Forecast averaging ("Forecast Averaging" on page 31).
- VAR Forecasting ("VAR Forecasting" on page 32).

Estimation

- Autoregressive Distributed Lag regression (ARDL) with automatic lag selection ("Autoregressive Distributed Lag (ARDL) Models" on page 34).
- Maximum Likelihood and GLS ARMA estimation ("ML and GLS ARMA" on page 35)
- Fractional Integration (ARFIMA) model estimation ("Fractional Integration (ARFIMA)" on page 36).
- Pooled mean group estimation of panel data ARDL models ("Panel ARDL and Pooled Mean Group (PMG) Estimation" on page 37).
- Threshold regression ("Threshold Regression" on page 37).
- New optimization engine and associated coefficient covariances ("Optimization Engine and Coefficient Covariances" on page 38).

Testing and Diagnostics

- Unit root tests with a structural break ("Unit Root Tests with a Breakpoint" on page 39).
- Cross-section Dependence Tests ("Panel Cross-section Dependence Tests" on page 41).
- Panel Effects Tests ("Panel Random Effects Tests" on page 42).

Other Features

- New functions for generating series ("Series Generating Functions" on page 44).
- Added matrix language tools ("Matrix Language Tools" on page 45).
- Enhanced table support ("Table Tools," on page 46).

- New general information tools ("General Information Tools," on page 46).
- New object data members ("Object Data Members," on page 47).
- List of new or updated global commands ("Updated Command List" on page 48).
- List of new or updated object commands ("Updated Object List" on page 49).

EViews 9 Compatibility Notes

• Compatibility notes for users of EViews 9 ("EViews 9 Compatibility Notes" on page 53).

General EViews Interface

The general EViews interface has been improved in a number of important ways. The following are some of the highlights.

Command Capture

EViews offers command capture for most object views and procedures, and a large number of interactive operations. With command capture, when you perform an operation using the dialogs or user-interface, EViews will save the equivalent text command for display and export.

You can copy-and-paste the contents of the capture window, or you can save the contents to a file. Right-clicking in the window brings up a menu for copying or clearing the window, saving the contents to a file on disk, or opening a new, untitled program containing the contents of the window.

To display the command capture window or set focus on the window, click on **Window**/ **Display Command Capture Window** from the main EViews menu.

🔏 EViews - 🗆	x
Eile Edit Object View Proc Quick Options Add-ins Window Help	
Workfile: TESTFILE - (c:\eviews\testfile.wf1) View Proc Object Save Freeze Details+/- Show Fetch Store Delete Genr Sample Range: 1954M01 1954M01 1994M12 - 492 obs	Capture
Sample: 1954M01 1994M12 - 492 005 Capture × [®] c [°] c [°] div [°] graph1	
Testfile / New Page /	
Path = c/animum DB = fred WF =	none

Note that not all interactive operations in EViews are capture enabled. Among the notable exceptions are some types of graph creation and customization, object view graph customization, and individual cell editing for tables and spreadsheets.

• For additional discussion, see "Command Capture" on page 3 of the *Command and Programming Reference*.

Window Docking

The EViews 9 command and capture windows are dockable, hideable, and floatable.

Dockable and hideable windows allow you to move frequently used windows out of the way while keeping them close at hand. They offer space saving convenience which is particularly valued when working with smaller screen devices like laptops.



Floatable windows allow you to move them out of the way of your work. You may even go so far as to float a window outside of the EViews frame.

• For discussion, see "Command and Capture Window Docking" on page 6 of the *Command and Programming Reference*.

Database and Workfile Object Preview

In earlier versions of EViews, browsing through the contents of a database or workfile could be tedious as it required opening and closing each object that you wished to examine.

EViews 9 offers you the ability to preview objects contained in a database or a workfile. With the preview tool, you may quickly browse through a number of objects. You may use the preview to quickly scroll through a preview of each object, viewing metadata (name, type, description, frequency, last update, source, units, *etc.*) and object type-specific information (for example, series will show observation data and a small graph while equations will show estimation output).



• For additional detail, see "Previewing Objects" on page 103 of User's Guide I.

Data Handling

EViews 9 offers a variety of new features for working with data.

Enhanced import and linking of data

In EViews 9, series objects can be linked to an object in another workfile and to any external data source that can be imported into EViews (*e.g.* – Excel file, Text file, ODBC database, etc). These linked series may be unlinked and re-linked as desired and the data refreshed when the workfile is opened, or on request.

The **Manage Links and Formulae** dialog has been redesigned to help you manage these new data links.

		Manage L	inks and Formula	ae	
ata sources f	or page:	All Pages	~		
Source			Туре		Properties
Yearly			page		Refresh
c: \files \sourc	e.edb		file		D l
c:\files\sourc	e.wf1		file		Break
dsn=mydsn; ed by the fo	dbq=c:\files\te	est1.mdb;driverid=25;f	ii=ms acce: odbc		
dsn=mydsn; ed by the fo Name	lbq=c:\files\te	est1.mdb;driverid=25;f s: Options	iil=ms.acce: odbc	Freq / Merge	Properties
dsn=mydsn; sed by the fo Name Yearly2::X	lbq=c:\files\te llowing objects Spec x	est 1.mdb;driverid = 25; s: Options "Select d	il≔ms acce: odbc	Freq / Merge Average observation	Properties Refresh
dsn=mydsn; ed by the fo Name Yearly2::X Yearly2::Y	lbg=c:\files\te llowing objects Spec X Y	s: Options "Select d	ateid01,x,y from Unti	Freq / Merge Average observatio Average observatio	Properties Refresh Break

You can use the dialog to examine and edit the link properties, refresh the data, and break or restore the links.

• For discussion, see "Breaking links" on page 248 of User's Guide I.

New FRED Database Interface

EViews 9 offers an updated interface to the FRED data service provided by the Federal Reserve Bank of St Louis. The new interface includes a custom browser for navigating the available FRED data and adds support for retrieval of historical releases.

To use the new interface, select **File/Open Database...** from the main EViews menus and open the all new custom FRED database. The browser interface allows you to find data available within FRED by navigating through a set of nested folders. Click on the folder icons to drill down into subtopics:



In addition, the FRED interface offers a number of useful tools for working with your FRED database, including keyword search, popularity and update ordering, filtering, vintage handling, and more.

Available Releases	
Wednesday, August 27, 2014	
Tuesday, February 04, 2014	
Tuesday, February 05, 2013	
Wednesday, August 22, 2012	
Tuesday, January 31, 2012	
Wednesday, February 02, 2011	
C	
Copy Selected Relea	ses

- See the discussion in "FRED" on page 352 of User's Guide I.
- See also dbopen (p. 344) in the Command and Programming Reference.

Cloud Drive Support

You may now access your Box, Dropbox, Google Drive, or OneDrive drive from within EViews. The **File Open...** and **File/Save As...** dialogs allow you to configure access to your accounts, and to read and write files directly from your cloud drives.

	Open		×		
🔄 🄄 👻 🕈 🏭 « EV9 > Chapter 09 - Advanced Wor v 🖒 Search Chapter 09 - Advance 🕫					
Organize 👻 New folde	er		= • 🔟 🔞		
☆ Favorites	Name	Date modified	Туре		
E Desktop	📓 abdata_pan.wf1	3/6/2007 12:12 PM	EViews Workfile		
🚺 Downloads	📓 grunfeld_baltagi_panel.wf1	3/6/2007 12:13 PM	EViews Workfile		
🌉 This PC	🔊 grunfeld_baltagi2.wf1	3/6/2007 12:12 PM	EViews Workfile		
😻 Dropbox	🔊 harrison_panel1.wf1	3/6/2007 12:13 PM	EViews Workfile		
🔚 Recent places	🔊 link_sales.wf1	3/6/2007 12:13 PM	EViews Workfile		
Recycle Bin - Shortc	🖌 restruct.wf1	3/6/2007 12:13 PM	EViews Workfile		
	🛃 statewf.wf1	3/6/2007 12:13 PM	EViews Workfile		
輚 Homegroup	unstack1.wf1	3/6/2007 12:13 PM	EViews Workfile		
_	🔄 wfstack.wf1	3/6/2007 12:13 PM	EViews Workfile		
1 This PC					
辑 Network					
	<		>		
Browse Location: Local Computer V					
File n	ame: New Location	V EViews Workfile	e (*.wf1) ∽		
	Update de	fault directory Open	Cancel		

- See the discussion of "Cloud Drive Support" on page 92 of User's Guide I.
- See also wfopen (p. 513), wfsave (p. 527), pageload (p. 438), and pagesave (p. 440) in the *Command and Programming Reference*.

Dated Data Table Templates

EViews 9 now allows you to store and load all of your dated data table settings to and from a dated data template.

Dated data table options					×	
Table Options	Data Format	Fonts	Labels/Headers	Templates		
Predefined	Predefined templates					
mytempla	ate	Apply	template settings			
			Layout and Trans	formations		
			Appeara	nce		
			Include serie	s settings		
		F	Rename Delete			
Create new	template					
Template N	ame:			Save		
		Г	ОК	Cancel	Apply	

For documentation, see:

- See the discussion of "Dated Data Table Templates" on page 520 of User's Guide I.
- See also Group::dloadtmpl (p. 302), and Group::ddsavetmpl (p. 305) in the *Object Reference*.

Expanded Frequency Conversion Interpolation Methods

EViews 9 supports new low-to-high frequency conversion methods as well as modification to the existing methods.

The new conversion methods are Point, Denton, Chow-Lin and Litterman. Each of these methods supports a number of different match points (*i.e.* how the periods on the source page are connected to those in the destination page), including First, Last, Sum, and Average. The existing Linear and Cubic methods have been extended to allow matching on First as well as Last.

Paste Special				
Paste y as Pattern: 🔭 Name: y	Frequency conversion options High to low frequency method Specified in source			
Paste as O Series (by Value)	Low to high frequency method Chow-Lin			
Merge by Date with frequency conversion General match merge criteria	Match: Last V Rho: Indicator: X			
OK OK to All	Cancel Cancel All			

- For additional detail, see "Frequency Conversion" on page 158 of User's Guide I.
- See also copy (p. 327) in the Command and Programming Reference.

Graphs, Tables, and Spools

Graph Pan and Zoom

Panning and zooming is available on graphs and graph views that feature the observation slider bar (*i.e.*, those graph views that allow for subsample display). You may use these features to focus on particular regions of the graph, much like cropping a photograph.

To zoom, left-click on the graph and drag the cursor to draw a box around the area into which you would like to zoom.



then release the mouse. The display changes to show only the area within the box and the observation slider bar and axis labels will change to reflect the modified display.



Note that there will be no indication that observations from the original graph have been trimmed away.

To pan, right-click on the graph, and when the cursor change to a hand, drag the image and release when the display shows the portion of the graph that is of interest.



Note that zooming and panning does not change the graph, only the display of the portion of the graph in the window. The observation slider bar and axis labels will change to reflect the modified display. Freezing the graph will take a snapshot of the current display.

To restore the original graph display, press ESC.

• For related discussion, see "The Graph Sample" on page 595 of User's Guide I.

Multi-graph Slideshow

EViews 9 makes viewing multi-graphs easier via the introduction of a graph slideshow.

When working with multiple graphs in a single graph frame one sometimes finds that the individual graphs are too small to see any detail. Consider, for example, the graph of impulse responses for a three variable VAR:

EViews 9 allows you to zoom in on the individual graphs and to display the contents of the graph in a slideshow format. Click on the **Zoom** button on the toolbar to enable the slideshow:



Here we see the zoom view. On the left is a gallery of the individual graphs; on the right is a larger display of the selected graph. You can page through the individual graphs using the arrow keys or **Page Up/Page Down** to select the next graph, or the **Home** and **End** keys to move to the first or last graph. Alternately, you can click on a gallery item to jump to that graph.

Click on ESC to exit zoom mode.

If you wish to enable the slideshow for a subset of the graphs, return to the original graph and select the graphs of interest by clicking or drag clicking to select the desired items,



then press the **Zoom** button or right-mouse click and select **View/Selected graphs**. EViews will show the gallery for just the selected items:



• See the related discussion in "Graphing Multiple Series (Groups)" on page 584 of *User's Guide I.*

Improved Mixed Graph Type

EViews 9 offers an entirely new interface where you can flexibly specify the type of each graph in the mixed display.

Previous versions of EViews allowed you to display a mixed graph type where the first series or column of data was depicted using a bar, spike, or area plot, and the remaining data was plotted using line graphs. There was no way to mix more than two graph types, nor was there a way to draw with more than one of the first graph type.

The new EViews interface allows you to specify the type of each graph element. Set the first series to spike, the second to line, and the third to area. Or mix up the assignment. The choice is up to you.

When you select **Mixed** as your **Basic Type**, the left-hand side tree adds an additional page under **Graph Type** for **Mixed settings**. When you click on this node, the right-hand side of the dialog changes to display the settings for each type:

	Graph Op	tions	×
Option Pages			
⊡ Graph Type	Series	Туре	
Mixed settings	nonacadsal	Bar Line	
	pcntnonac	Line	¥
Graph Elements Graph Flements Quick Fonts	pctunemp	Line Stacked Line Bar Stacked Bar	
i - Templates & Objects		Spike Stacked Spike Area	73
		Area Band	
	✓ Draw line types after	er fill types (use group	order otherwise)
Undo Page Edits			OK Cancel

You may, for example, graph your group data with a spike for the first series, a line for the second, and an bar for the remaining two.



- See the discussion in "Mixed" on page 627 of User's Guide I for discussion.
- See also mixed (p. 909) in the Object Reference.

Rectangle and Ellipse Drawing

You may now draw rectangles and ellipses in frozen graph objects.



- For discussion, see "Drawing Rectangles and Ellipses" on page 713 of *User's Guide I* for discussion.
- See also, addellipse (p. 232) and addrect (p. 234) in the Object Reference.

Data-based Anchoring

By default, the positions of arrow and text objects in a graph are specified in virtual inches relative the top left corner of the graph. While useful for some types of drawing, this approach is problematic when one wishes to place objects at particular data points or dates in the graph as in, for example, identifying an outlier or the observation value at a notable date. For one, placing the object is quite difficult programmatically as there is no clear way of translating from data values or observations to virtual inch offset. Moreover, if one changes the axes ranges, sample, or aspect ratio of the graph, the absolute positioning method breaks down.

	Rectangle ×				
Location					
	Anchor center to:	Data (axis un	its) 🗸		
	Bottom axis:	1969m06	(date)		
	Left axis:	4.984	(data units)		
s	ize				
	Width:	111.54437	(observations)		
	Height:	1.4537444	(data units)		
	Angle:	0.00	(degrees)		
-4	ppearance				
	Color Line width Line pattern				
	▶ 1/2 pt →				
QK Cancel					

You can attach arrows and the new rectangle and ellipse drawing objects to fixed data/date points within the graph. You can, for example, choose to place your object at the point defined by the observation for 1990m01 and the maximum value of X.

Similarly, sizes of objects may also be expressed in terms of observations and data units, so that you can draw a rectangle that is "2 years wide" and "10.5" data units tall.

In addition to being easier to specify programmatically, if, you then alter the graph frame or axes, the object will move with the data in the graph.

- See "Data-based Anchoring" on page 715 of User's Guide I for discussion.
- See also addarrow (p. 229), addellipse (p. 232), addrect (p. 234) in the Object Reference.

LaTeX Output

You can now save your table, graph, and spool output in LaTeX format.

This feature is available whenever you save a table, graph, or spool to disk (using commands, the table or graph save to disk proc, or the right-mouse button save from a table or graph view). In all of these case, you should select change the File type combo to "LaTeX file" to see the available options.

The options to save as PDF are included in the standard save dialogs for graphs, tables, and spools. Right-click on the graph, table, or spool object and select **Save graph to disk..., Save table to disk...**, or **Save to Disk**, respectively. The standard file save dialogs will appear. Select PDF from the **File Type** drop-down.

Graphics File Sav	ve
File name/path	
	Browse
File type	Output graph size
LaTeX file (*.tex)	Units: inches V
	inches percent
Options	Width: 5.9 100
✓ Use <u>c</u> olor ● <u>P</u> ortrait	Height: 5 1 100
✓ Make background white ○ Landscape	
<u>P</u> ut input command on clipboard	Dots per Inch: 96
Include document specification	
OK Cancel	Reset

LaTeX also appears as a supported output type for table objects in EViews 9:

Table File Save			
File name/pa	th LaTeX file (*.tex)	✓ Browse	
Options Range: Embed as:	Entire table	Numbers Number format: As displayed V Text representing NA: NA (blank is ok)	
✓ Include o	ocument specification	OK	

For command line documentation

- see Graph::save (p. 262) in the Object Reference
- see Table::save (p. 766) in the Object Reference

• see Spool::save (p. 664) in the Object Reference

Econometrics and Statistics

EViews 9 offers a exciting new additions and improvements to its set of econometric and statistical features. The following is a brief outline of the most important new features, followed by additional discussion and pointers to full documentation.

Automatic ARIMA Forecasting

Automatic ARIMA forecasting is a method of forecasting values for a single series based upon an ARIMA model.

Although EViews provides sophisticated tools for estimating and working with ARIMA models using the familiar equation object, there is considerable value in a quick-and-easy tool for performing this type of forecasting. EViews 9 introduces an automatic ARIMA forecasting series procedure that allows the user to quickly determine an appropriate ARIMAX specification and use it to forecast the series into the future.

You will find this feature using the Proc menu for a series obje	ct
---	----

Automatic A	RIMA Forecasting 🛛 💌
Specification Options	
Transformation Auto (None / Log) None Log Box-Cox Power: 2 Sample specification Estimation sample: 2005M012015M12 Forecast length (optional):	ARIMA Specification Max differencing: 2 v Max. AR: 4 v Max. MA: 4 v Max. SAR: 0 v Max. SMA: 0 v Periodicity: 12 Regressors C
	OK Cancel

- For further documentation, see "Automatic ARIMA Forecasting" on page 449 of *User's Guide II* for discussion.
- See also Series::autoarma (p. 531) in the Object Reference.

Forecast Evaluation

When constructing a forecast of future values of a variable, economic decision makers often have access to different forecasts; perhaps from different models they have created themselves or from forecasts obtained from external sources. When faced with competing forecasts of a single variable, it can be difficult to decide which single or composite forecast is "best". Fortunately, EViews 9 provides tools for evaluating the quality of a forecast which can help you determine which single forecast to use, or whether constructing a composite forecast by averaging would be more appropriate.

Given actual data and one or more sets of forecasts for an evaluation sample, EViews computes four different measures of forecast accuracy; RMSE (Root Mean Squared Error), MAE (Mean Absolute Error), MAPE (Mean Absolute Percentage Error), and the Theil Inequality Coefficient.

In addition, EViews can compute the Combination Test, or Forecast Encompassing Test (Chong and Hendry, 1986; Timmermann, 2006) for evaluating whether averages of forecasts perform better than the individual forecasts.

To perform forecast evaluation in EViews, you must have a series containing the observed values of the variable for which you wish to evaluate forecasts. To begin, open up the series and click on **View/Forecast Evaluation...**, which will open the **Forecast Evaluation** dialog box:

Forecast Evaluation	x
Forecast data objects Enter a list of series, a group, or a name pattern, OR a list of equation objects: eq01	Averaging methods (optional) All Simple mean Trimmed mean Least-squares Mean equivance are power:
Valuation sample	MSE ranks Smooth AIC weights SIC weights Training
ОК	sample: Training forecast type: O Dynamic O Static

- For documentation see "Forecast Evaluation" on page 397 of *User's Guide I* for additional discussion.
- See also Series::forceval (p. 553) in the Object Reference.

Forecast Averaging

Economic forecasters often have a variety of different models and forecasts of the same variable from which to choose. Traditionally the forecasting decision was to pick which single forecast was "best" out of the individual forecasts available. However, a number of studies (Timmermann 2006) have shown that averaging forecasts is more accurate than choosing a single best forecast.

Forecast averaging, or forecast combining, is a methodology for combining multiple forecasts into a single forecast. EViews 9 offers a number of easy-to-use tools for performing forecast averaging using simple mean, least squares, mean square error, mean square error ranks, smoothed AIC, approximate Bayesian model averaging. trimmed mean and simple median methods.

To perform forecast averaging in EViews, you must start with a series representing the data to be forecasted. Some of the forecast weighting techniques require actual data to calculate the weights, and in these cases this series should contain actual values for the variable being forecasted for at least some of the observations for which forecast values are available.

Open the series and click on **Proc/Forecast Averaging...**, which will open the **Forecast Averaging** dialog box:

Forecast Averaging	x
Forecast data objects Enter a list of series, a group, or a name pattern, OR a list of equation objects:	Averaging method Simple mean Percent: 5 Training
e1e2	Training forecast type:
Forecast sample	Forecast name: x1_f
1 100	Weight vector (optional):
ОК	Cancel

The **Forecast data objects** box specifies the forecasts to be used for averaging. Forecasts can be entered either as a collection of series (in which case the names of the series, a series naming pattern, or the name of a group are entered), or as a list of equation objects. If equation objects are entered, EViews will automatically perform a dynamic forecast over the forecast period from each of those equation objects to generate the forecast data.

For documentation see:

- See "Forecast Averaging" on page 458 of User's Guide I for additional discussion.
- Series::forcavg (p. 551) in the Object Reference.

VAR Forecasting

Previously, to forecast from an estimated VAR object, you needed to first make an EViews model object from the VAR and then solve the model. This procedure was cumbersome, especially for those unaccustomed to working with models and scenarios.

In EViews 9, you may produce forecasts directly from an estimated VAR object. Click on the **Forecast** button or select **Proc/Forecast** to display the forecast dialog:

Foreca	ast
Forecast of VAR: VAR1	
Series output Name suffix: F S.E. suffix: SEF (optional): SEF Simulation repetitions: 10000 % Failed reps before halting: 2 Forecast sample	Method Dynamic forecast Static forecast Coef uncertainty in S.E. calc Output Individual graphs Multiple graphs
1959M02 1995M04 ✓ Insert actuals for out-of-sample observ	Forecast evaluation
ОК	Cancel

Fill out the dialog as desired and click on **OK**. EViews performs the forecast and, if appropriate displays output:

(var	Var: VAR	1 Work	cfile: VA	R1::V	ar1\					• 8	3
View Proc Object Print Nan	ne Freeze	Estimate	Forecast	Stats	Impulse	Resids	Zoom				
Evaluation Table	🗏 Evalu	ation Tabl	e								^
TB3_F	Fore Date Sam Inclu	cast Evalu : 03/07/15 ple: 1959 ded obse	iation Time: 1 M02 1995 rvations: 4	2:29 5M04 435							
	Varia	ble	Inc	c. obs.	RN	ISE	MA	E	MAPE	Th	
	IP M1 TB3			435 435 435	7.00 51.1 2.55	4263 5341 6030	5.422 41.22 1.862	211 679 331	6.669999 9.498844 26.74072	0.04 0.04 0.18	
	RMS MAE: MAPE Theil	E: Root M Mean Ab E: Mean A : Theil ing	lean Squ solute Er Absolute F equality c	are En ror ^P ercer oefficio	ror ntage Err ent	or					
	⊟ IP_F										
	200)									¥
										,	.::

In this case, the output consists of a spool containing the forecast evaluation of all of the series in the VAR, along with individual graphs of the forecasts along with the corresponding actuals series.

• For additional information, see "Forecasting" on page 635 in User's Guide II.

- See also Var::fit (p. 829) in the Object Reference.
- See also Var::forecast (p. 830) in the Object Reference.

Autoregressive Distributed Lag (ARDL) Models

EViews 9 offers new tools for estimating and examining the properties of Autoregressive Distributed Lag (ARDL) models. ARDLs are standard least squares regressions which include lags of both the dependent variable and independent variables as regressors.

Although ARDL models have been used in econometrics for decades, they have gained popularity in recent years as a method of examining long-run and cointegrating relationships between variables (Pesaran and Shin, 1999; Pesaran, Shin and Smith, 2001).

EViews offers specialized tools for estimating ARDL models, including built-in lag-length selection methods, cointegrating relationship estimation, and Bounds testing for long-run relationship.

To estimate an ARDL model using the ARDL estimator, open the equation dialog by selecting **Quick/Estimate Equation...**, or by selecting **Object/New Object.../Equation** and then selecting **ARDL** from the **Method** dropdown menu. EViews will then display the ARDL estimation dialog:

		Equation Estin	nation			x
Specificatio	n Options					
-Dynamic Depend	Specification- ent variable fo	llowed by list of dynar	mic regressors			
log(real	cons) log(realg	jdp)				
Autor Fixed	matic Selectior	n Dependent Va Max lags: [8	ariable:	Regressors Max lags:	:: 8 ♥	
Fixed reg Trend sp Constan	gressors ecification It (Level)	List of fixed regre @expand(@qua	ssors rter, @droplast)			
-Estimatio Method: Sample:	ARDL - Auto 1950q1 2000	regressive Distributed	Lag Models		`	
	1			ОК	Cancel	

- See "Autoregressive Distributed Lag (ARDL) Models" on page 283 of *User's Guide II* for discussion.
- See also Equation::ardl (p. 45) in the Object Reference.

- See also Equation::icgraph (p. 107) in the Object Reference.
- See also Equation::ictable (p. 108) in the Object Reference.
- See also Equation::cointrep (p. 76) in the Object Reference.
- See also Equation::cointgraph (p. 67) in the Object Reference.
- See also Equation::boundstest (p. 51) in the Object Reference.
- See also Equation: :makecoint (p. 121) in the Object Reference.

ML and GLS ARMA

EViews now allows you to estimate ARMA models specified by list using ML or GLS (in addition to the previously existing CLS-based estimator). Estimation of these models features the use of the Kalman filter to evaluation the exact likelihood (Hamilton 1994).

In the equation object dialog, click on the **Options** tab. The **ARMA** section in the upper right-hand corner of the dialog presents new options for estimating your ARMA model.

Equation	1 Estimation
Specification Options	
Coefficient covariance Covariance method: Information matrix: d.f. Adjustment	ARMA Method: ML V Starting ARMA coefficient values: Automatic V
Estimation algorithm Optimization method: BFGS Step method: Marquardt Maximum iterations: 500 Convergence tolerance: 0.0001 Display settings in output	Coefficient name
	OK Cancel

In addition to a choice between ML, GLS, or CLS estimation of ARMA methods, you will have a choice of methods for determining ARMA starting values, optimization method, and the approach to computing the estimate of the coefficient covariance.

- For additional discussion, see Chapter 22. "Time Series Regression," on page 87 of *User's Guide II*.
- See also Equation::ls (p. 114) in the Object Reference.

Fractional Integration (ARFIMA)

Stationary processes are said to have *long memory* when autocorrelations are persistent, decaying more slowly than the rate associated with ARMA models. Modeling long term dependence is difficult for standard ARMA specifications as it requires non-parsimonious, large-order ARMA representations that are generally accompanied by undesirable short-run dynamics (Sowell, 1992).

One popular approach to modeling long memory processes is to employ the notion of *fractional integration* (Granger and Joyeux, 1980; Hosking, 1981). A fractionally integrated series is one with long-memory that is not I(1). By combining fractional integration with an ARMA specification we obtain a fractional ARIMA (ARFIMA) model. The ARFIMA model includes a fractional integration parameter d that controls the degree of long-run dependence, allowing for the separate modeling of the short and the long-run dynamics

EViews supports exact maximum likelihood estimation of ARFIMA models via ML or GLS using efficient algorithms as described in Sowell (1992) and Doornik and Ooms (2003). Among the supported features are automatic initialization of the integration parameter d estimates using the Geweke and Porter-Hundlak (1983) log-periodogram regression, and concentration of the likelihood with respect to regression coefficients and scale.

	Equation Estimation	×
Specification	Options	
Equation [a	specification >ependent variable followed by list of regressors including ARMA ind PDL terms, OR an explicit equation like Y=c(1)+c(2)®X.	
dlog(gnp) ar(1 to 3) ma(1 to 2) c d	
Estimatio	n settings	
Method:	LS - Least Squares (NLS and ARMA)	~
Sample:	1947q1 1989q4	
	ОК	Cancel

To specify an ARFIMA model you should specify your ARMA model by list, then include "D" as a regressor to instruct EViews to estimate an integration term.

- For additional discussion, see Chapter 22. "Time Series Regression," on page 87 of *User's Guide II*.
- See also Equation::1s (p. 114) in the Object Reference.

Panel ARDL and Pooled Mean Group (PMG) Estimation

EViews 9 supports estimation of the Pooled Mean Group (PMG) estimator of Pesaran, Shin and Smith (PSS, 1999) for ARDL models with individual effects. This model is particularly popular in panel settings where the number of periods is large, since alternative GMM estimators may not be appropriate in those settings.

The PMG takes the cointegration form of the simple ARDL model and adapts it for a panel setting by allowing the intercepts, short-run coefficients and cointegrating terms to differ across cross-sections.

	E	quation Estimation	×
Specificatio	on Options		
Dynamic Depend	Specification	ed by list of dynamic regresso	prs
 Auto Fixed 	matic Selection	Dependent Variable: Max lags: 4 v	Regressors: Max lags: 4 V
Fixed re Trend sp Constar	gressors becification ht (Level) V	List of fixed regressors	
-Estimation Method: Sample:	PMG/ARDL - Poo 1990 2000	oled Mean Group / AR Distribu	ited Lag Models 🗸
	1		OK Cancel

- See "Pooled Mean Group Estimation" on page 855 in User's Guide II for discussion.
- See also Equation::ardl (p. 45) in the Object Reference.

Threshold Regression

EViews 9 estimates threshold regression models (TR) which are linear regression models where the coefficients on explanatory variables may change as a threshold variable crosses boundary values. This class of model includes the popular threshold autogressive models (TAR).

To estimate a threshold equation, select **Threshold Regression** from the equation estimation method combo.

cification Op	otions	
Equation spec	rification	
Dependent va	ariable followed by list of thresh	old regressors:
lvnx transf	c lvnx_transf(-1 to -11)	-
ist of non-th	reshold regressors:	
Threshold var	iable specification	
[hreshold var	iable specification	r range pairs. Integers or range pairs
Threshold var Enter a series may be used t	iable specification 5, group, list of series, integer, f for self-exciting models. Multiple	r range pairs. Integers or range pairs variables indicate model selection.
[hreshold var Enter a series nay be used 15	iable specification , group, list of series, integer , for self-exciting models. Multiple	vr range pairs. Integers or range pairs : variables indicate model selection. Ex: "3", "3 6" (SETAR), or "21 22" (TAR)
Threshold var Enter a series nay be used 1 5	iable specification , group, list of series, integer, , for self-exciting models. Multiple	or range pairs. Integers or range pairs : variables indicate model selection. Ex: "3", "3 6" (SETAR), or "21 Z2" (TAR)
Threshold var Enter a series may be used 1 5 Estimation set	iable specification , group, list of series, integer, , for self-exciting models. Multiple ttings	or range pairs. Integers or range pairs variables indicate model selection. Ex: "3", "3 6" (SETAR), or "Z1 Z2" (TAR)
Threshold var Enter a series may be used 1 5 Estimation set Method: THR	iable specification 5, group, list of series, integer, , for self-exciting models. Multiple ttings ttings	or range pairs. Integers or range pairs : variables indicate model selection. Ex: "3", "3 6" (SETAR), or "Z1 Z2" (TAR) n v
Inreshold var Enter a series may be used 1 5 Estimation set Method: THR Sample: @al	iable specification 5, group, list of series, integer, , for self-exciting models. Multiple ttings ESHOLD - Threshold Regressic	or range pairs. Integers or range pairs variables indicate model selection. Ex: "3", "3 6" (SETAR), or "Z1 Z2" (TAR) n v

- See "Threshold Regression" on page 427 in User's Guide II for discussion.
- See also Equation::threshold (p. 161) in the Object Reference.

Optimization Engine and Coefficient Covariances

In EViews 9, we have integrated an all new estimation engine into many familiar EViews estimators.

The new engine, which implements a sophisticated trust region approach (More and Sorensen, 1983), supports versions of the Broyden-Fletcher-Goldfarb-Shanno (BFGS), Gauss-Newton/BHHH, Newton-Raphson, and Fisher Scoring algorithms.

Equation Es	timation
Specification Options	
Coefficient covariance	Weights
Covariance Ordinary V	Type: None V
Information OPG V	Weight
matrix:	Scaling: EViews default V
Gauss-Newton V Step method: Marquardt V	_
step method: Marguardt	
Maximum iterations: 500	Use numeric only
Convergence tolerance: 1e-8	Coefficient name
Display settings in output	C
	OK Cancel

One important benefit of the use of the new optimization engine is the availability of numeric second derivative estimates of the Hessian in cases where they were not previously available. Having second derivatives allows us additional choice in the estimators of the coefficient covariance, both through alternative estimators of the information matrix and through an expanded ability to compute Huber-White type sandwich estimators.

The following existing EViews estimators have been updated to support the new estimation engine: single equation nonlinear least squares and ARMA, binary, count, ordered, censored, ARCH (single equation and system), switching regression, GLM, Heckman selection, system FIML, state space, user-defined likelihood.

- See "Optimization Method" on page 1006 in the User's Guide II for discussion.
- See also the command entries for the various estimators, for example, Equation::ls (p. 114) in the *Object Reference*.

Unit Root Tests with a Breakpoint

EViews 9 supports the computation of modified Dickey-Fuller tests which allow for levels and trends that differ across a single break date. The testing framework follows the work of Perron (1989), Perron and Vogelsang (1992), Vogelsang and Perron (1998), Banerjee, *et al.* (1992).

EViews offers unit root tests under a variety of scenarios for the type and timing of the break and different data trend specifications. To compute a unit root test with breakpoint select **View/Breakpoint Unit Root Test...** from the series menu.

Breakpoint U	nit Root Test
Test for unit root in Level 1st difference 2nd difference	Break type Innovation Outlier Additive Outlier
Trend specification Basic: Intercept Breaking: Intercept	Breakpoint selection Dickey-Fuller min-t V
Lag length Method: Schwarz criterion V Max. lags: 10	Additional output Jisplay test and selection graphs Results matrix:
OK	Cancel

Fill out the dialog as desired, and click on **OK**. EViews will display the test results, and will, by default, display a spool object that contains the test results table,

	Series: R Workfile: TESTFILE::Testfile\	x
View Proc Object Properties	Print Name Freeze Sample Genr Sheet Graph Stats Ident	
 Unit root test results Test statistics graph AR coefficients graph 	□ Unit root test results	^
	Unit Root with Break Test on R	
	Null Hypothesis: R has a unit root Trend Specification: Intercept only Break Type: Innovational outlier Lag Length: 6 (Automatic - based on Schwarz information criterion, maxlag=17)	
	Break Selection: Minimize Dickey-Fuller t-statistic Break Date: 1967M04	
	t-Statistic Prob.*	
	Augmented Dickey-Fuller test statistic -2.252724 0.9548 Test critical values: 1% level -4.949133 5% level -4.443649 10% level -4.193627	
	*Vogelsang (1993) asymptotic one-sided p-values.	
	Auamented Dickev-Fuller Test Equation	>

and graphs of the breakpoint selection results,



- For discussion, see "Unit Root Tests with a Breakpoint" on page 539 in the User's *Guide II*.
- See also Series::buroot (p. 537) in the Object Reference.

Panel Cross-section Dependence Tests

EViews 9 performs tests for cross-section dependence (CD) in panel data. You may perform the Breusch-Pagan LM (1980), Pesaran (2004) scaled LM and CD, and the Baltagi, Feng, and Kao (2012) bias-corrected scaled LM tests in panel equation and panel series settings.

To perform a panel series test, open a panel series and select **View/Cross-section Dependence Test**. To perform a test on the residuals from an estimated panel equation, select **View/Residual Diagnostics/Cross-Section Dependence Test**:

Equation: EQ01 Workfile: GF	RUNFELD_BALTAG 🖃 💷 🗾
View Proc Object Print Name Freeze	Estimate Forecast Stats Resids
Representations	
Estimation Output	
Fixed/Random Effects	+
Actual, Fitted, Residual	•
Gradients and Derivatives	•
ARMA Structure	
Covariance Matrix	Std. Error t-Statistic Prob.
Coefficient Diagnostics	0.005836 19.80259 0.0000 511676 -4 490730 0.0000
Fixed/Random Effects Testing	 Redundant Fixed Effects - Likelihood Ratio
Residual Diagnostics	Omitted Random Effects - Lagrange Multiplier
Label	Correlated Raksom Effects - Hausman Test
Sum squared resid 1755850.	Schwarz criterion 11.99750
Log likelihood -1191.802	Hannan-Quinn criter. 11.96805
F-statistic 426.5757	Durbin-Watson stat 0.219599
Prob(F-statistic) 0.000000	

EViews will compute the relevant statistics, and display the results in tabular form:

Equation: GA	AS W	orkfil	e: GAS	OLINE::0	Gasoline		
View Proc Object	Print	Name	Freeze	Estimate	Forecast	Stats	Resids
Residual Cross-S Null hypothesis: N Equation: GAS Periods included: Cross-sections in Total panel obsen Cross-section effe	Section I lo cross 19 icluded: vations: ects we	Deper s-sect 18 342 re rem	ndence 1 ion depe	Fest endence (i	correlation	n) in re	esiduals
				ing source	- and a state of the		
Te	est			Statisti	c	d.f.	Prob.
Te Breusch-Pagan LI	est M			Statisti	c 35	d.f. 153	Prob.
Te Breusch-Pagan LI Scaled Breusch-P	est M Pagan L	M		Statisti 1027.1: 48.942	c 35 00	d.f. 153	Prob. 0.0000 0.0000
Te Breusch-Pagan LI Scaled Breusch-P Bias-corrected sca	est M Pagan L aled LM	M		Statisti 1027.1: 48.942 48.442	c 35 00 00	d.f. 153	Prob. 0.0000 0.0000 0.0000

- For discussion, see "Panel Cross-section Dependence Test" on page 872 of *User's Guide II*.
- See also Equation::cdtest (p. 55) in the Object Reference.

Panel Random Effects Tests

EViews 9 allows you to test for individual and time unobserved random effects in a panel or pool equation. EViews computes the Breush-Pagan LM (1980), Baltagi and Li (199), Honda

(1985), King and Wu (1997), Gourieroux, Holly, and Monfort (1982), Moulton and Randolph Standardized LM (1989) tests.

To compute the test statistics, select **View/Fixed-Random Effects Testing/Omitted Random Effects - Lagrange Multiplier** from an estimated panel equation:

Equation: EQ01 View Proc Object Print	Workfile: GF	RUNFELD_BALTAG	ts Resids	
Representations	1 1			
Estimation Output				
Fixed/Random Effec	ts	•		
Actual, Fitted, Residu	al	•		
Gradients and Deriva	atives	•		
ARMA Structure				
Covariance Matrix		Std. Error t-Stati	stic Prob.	
Coefficient Diagnos	tics	0.005836 19.80	259 0.0000 730 0.0000	
Fixed/Random Effec	ts Testing	 Redundant Fixe 	ed Effects - Likelihoo	od Ratio
Residual Diagnostics	;	Omitted Rando	om Effects - Lagrang	e Multiplier
Label		Correlated Ray	or Effects - Hausi	man Test
Sum equared reaid	1755950	arke mio criterion	11.94802	
Log likelihood	-1191 802	Hannan-Quinn criter	11 96805	
E-statistic	426 5757	Durbin-Watson stat	0 219599	
Prob(F-statistic)	0.000000	22	2.2.10000	

The results for all of the tests are displayed in the object window:

(all others) alternati	ves	cn-Pagan) and	one-sided
	T Cross-section	est Hypothesis Time	Both
Breusch-Pagan	798.1615	6.453882	804.6154
	(0.0000)	(0.0111)	(0.0000)
Honda	28.25175	-2.540449	18. 1 8064
	(0.0000)		(0.0000)
King-Wu	28.25175	-2.540449	21.83221
	(0.0000)		(0.0000)
Standardized Honda	32.66605	-2.432565	16.29814
	(0.0000)		(0.0000)
Standardized King-Wu	32.66605	-2.432565	20.96591
	(0.0000)		(0.0000)
Gourierioux, et al.*		-	798.1615 (< 0.01)

- For discussion, see "LM Tests for Random Effects" on page 865 of User's Guide II.
- See also Equation::rcomptest (p. 145) in the Object Reference.

Other Features

There are a number of other features and improvements that are not mentioned above.

Series Generating Functions

These functions return cumulative values for each observation:

Function	Description
<pre>@cummedian(arg1[, s])</pre>	Cumulative median of <i>arg1</i> .
<pre>@cummedianb(arg1[, s])</pre>	Backward cumulative median of <i>arg1</i> .
<pre>@cumquantile(arg1, arg2[,s])</pre>	Cumulative <i>arg2</i> quantile of <i>arg1</i> .
<pre>@cumquantileb(arg1, arg2[,s])</pre>	Backward cumulative <i>arg2</i> quantile of <i>arg1</i> .