USING AUTOMETRICS

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The basics of OxMetrics 5:

 \rightarrow Loading, editing and transforming data, creating "special" series (cointegration relations, trends, etc.)

An overview of Autometrics:

 \rightarrow Key concepts and characteristics

Single-equation modelling with Autometrics

 \rightarrow Formulation, Advanced Autometrics settings, fixing variables, example (2007 Econometric Game, Q1)

Multiple-equation modelling with Autometrics

 \rightarrow Formulation, fixing variables, example (2007 Econometric Game, Q2)

OxMetrics basics:

- \bullet Load data: File \rightarrow Open, etc.
- Edit sample/dates: Edit \rightarrow Change Sample
- Missing values (recommendation): Set to "missing" by double-clicking the data cell in question
- \bullet Graph series: Model \to Graphics (or click on the graphics button) \to Actual series or All plot types
- Transform data (algebra feature):

(NOTE: Case sensitivity in variable names!)

 \bullet Create special series (calculator feature): Model \to Calculator (or click on the calculator button) $\to \ldots$

• Example. Edit dates (2007 Econometric Game Case):

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Graphics	3	196403	1.7299	-99.99	169.28	485.5	316.9	554.76		
Code	4	196404	86419	-99.99	170.36	561.06	258.69	542.07		
Text Results	5	196405	81742	322.23	166.66	Change Sam	ple			8
Modules	6	196406	.11888	321.89	174.27	Current Databa	se Sample			
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	26	196602	79059	321.59	168.87				_	
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	28	196604	.28895	323.7	162.72	557.48	287.82	554.08		_
	2.9	196605	. 50448	324.07	163.78	523.04	298.36	566.47		
	30	196606	1.6865	323.75	170.41	564.71	321.53	565.45		
	31	196607	.28445	322.4	167.5	672.49	302.51	583.61		
	32	196608	-2.1055	320.37	170.17	735.03	248.43	570.47		
	32	196609	-1.9345	318.64	169.95	763.91	256.81	581.43		
p 🔺	24	196610	.17696	318.1	169.04	610.38	295.64	574.52		

• Example. Create a differenced series:

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	1966(9)	196609	-1.9345	318.64					
	1966(10)	196610	.17696	318.1					

References

An overview of Autometrics:

• The objective of Autometrics is to automate *General-to-Specific* (Gets) *multiple path* simplification search of a *data coherent*, *General Unrestricted Model* (GUM) in the form of a linear OLS/IV estimable regression (or regressions)

• Default definition of data-coherency: Stable parameters and Gaussian, serially uncorrelated, homoscedastic errors. NOTE: These assumptions can be relaxed through the "Advanced Autometrics settings", and if the GUM fails one or several diagnostic checks Autometrics proceeds anyway

• GUM: A general model (advice: Not too general!) that includes the variables and lags that are believed to possibly have an impact

• Further reading: Doornik and Hendry (2007a, pp. 70-77), Hendry and Krolzig (2001) (Autometrics is an evolution of PcGets) • Key benefits of Gets modelling:

 \rightarrow Estimation and inference while controlling for the impact from other variables: Spurious variables are more likely to be excluded, parameter estimates are "more correct"

 \rightarrow Gets modelling leads to a parsimonious, explanatory model particularly useful for scenario analysis (policy analysis, counterfactual analysis and conditional forecasting)

• Originally, the main disadvantage of Gets modelling was:

 \rightarrow Resource demanding and time consuming if properly implemented

Solution: Automated Gets

 \rightarrow Hoover and Perez (1999), PcGets (Hendry and Krolzig 2001), Autometrics (Doornik 2007, Doornik and Hendry 2007a)

What does multiple path Gets specification search consist in?

• Sequential deletion of insignificant variables (significance level can be user-specified), while checking data-coherency for each variable deletion (variable-deletion that induces data-incoherency, say, serially correlated errors, is not undertaken)

• Multiple path Gets search typically leads to several terminal models; Autometrics either applies Gets on their union or chooses among them by means of information criteria (can be user-specified; keyword: "Tie-breaker")

Single-equation modelling with Autometrics. *Example*: 2007 Econometric Game, Question 1

• My GUM:

$$\Delta COO_t = b_0 + b_1 \Delta COO_{t-1} + b_2 \Delta COO_{t-2} + \sum_{j=1}^{11} c_j d_{j,t} + e_t$$

• The specific model proposed by Autometrics using the default options:

$$\Delta COO_t = b_0 + b_1 \Delta COO_{t-1} + \sum_{j=1}^3 c_j d_{j,t} + \sum_{j=5}^{11} c_j d_{j,t} + e_t$$

• Unfortunately, the missing values cannot be estimated by means of the level representation

$$COO_{t} = b_{0} + (1+b_{1})COO_{t-1} + b_{1}COO_{t-2} + \sum_{j=1}^{3} c_{j}d_{j,t} + \sum_{j=5}^{11} c_{j}d_{j,t} + e_{t}$$

so a different specification is needed with, say, no lags of ΔCOO_t

• Formulate a model: (Model \rightarrow) PcGive \rightarrow Category: "Models for time series data" \rightarrow Model class: "Single-equation dynamic modelling using PcGive" \rightarrow "Formulate"

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Introduction

References

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Use default status	Chear>>	Constant Seasonal Trend CSeasonal	

USEFUL FEATURE: Fixing regressors (that is, preventing Autometrics from deleting them). Select the regressors to fix \rightarrow Right-click mouse \rightarrow A: instrument/fixed. NOTE: This defines instrument if IV is used instead of OLS

• Selected Autometrics options:

 \rightarrow Target size: Significance level

 \rightarrow Outlier detection: Neutralises large residuals in the GUM by means of impulse dummies

 \rightarrow Pre-search lag reduction: Speeds up simplification; GENERAL ADVICE: Turn off!

 \rightarrow Advanced Autometrics settings: In general, use only if default settings and options are unsatisfactory

 \rightarrow Recursive graphics: TURN ON! Slows down computation (slightly), but enables some very useful stability diagnostics

Introduction

OxMetrics basics Autometrics

• Advanced Autometrics settings:

tometrics Settings - Single-equation Dynam Search settings	iic Modelling
Search settings Outlier detection	None
Pre-search lag reduction	
Pre-search lag reduction Pre-search variable reduction	
Search effort	1
Backtesting	GUM 0
Tie-breaker	SC
Print level	Default output
Target size	Default: 0.05
User determined p-value	.05
Diagnostic test p-value	.01
Standard errors	Default
GIVE: first do reduced form	
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RESET test (using squares)	
Error autocorrelation test	
Portmanteau statistic	
ARCH test	
Diagnostic test arguments	
Use default	V
Chow-test sample split (%)	70
Error autocorrelation to lag	2
Portmanteau lag length	
ARCH test to lag	2
ОК	Cancel

• Specific model proposed by Autometrics:

EQ(2) Modelling DCOO by OLS

The dataset is: C:\Documents and Settings\sucarrat\Mis documentos\files\teaching The estimation sample is: 1964(8) - 2000(12)

	Coefficient				Part.R^2
DC00_1	-0.213999	0.04512	-4.74	0.0000	0.0503
Seasonal			-3.40		0.0265
Seasonal_1	-0.507123	0.05851	-8.67	0.0000	0.1502
Seasonal_2	-0.451573	0.06034	-7.48	0.0000	0.1165
Seasonal_4	-0.707806	0.05885	-12.0	0.0000	0.2540
Seasonal_5	-1.94870	0.06413	-30.4	0.0000	0.6848
Seasonal_6	-3.02784	0.09527	-31.8	0.0000	0.7038
Seasonal_7	-3.77271	0.1275	-29.6	0.0000	0.6733
Seasonal_8 Seasonal_9	-3.69728	0.1505	-24.6	0.0000	0.5868
Seasonal_9	-1.89487	0.1424	-13.3	0.0000	0.2941
Seasonal_10	-0.176490	0.07692	-2.29	0.0222	0.0122
Constant U	1.50663	0.06042	24.9	0.0000	0.5940
sigma	0.286776	RSS		34.95208	321
R^2	0.946555	F(11,425)	= 684.3	[0.000]	**
log-likelihood	-68.1549	DW		2.	05
no. of observatio	ns 437	no. of par	ameters		12
mean (DCOO)	0.112334	var (DCOO)		1.496	554
AR 1-7 test:	F(7,418) =	1.0154 [0	.4196]		
ARCH 1-7 test:	F(7,411) =	0.78778 [0	. 5979]		
Normality test:	Chi^2(2) =	2.5460 [0	.2800]		
Hetero test:	F(12, 412) =	0.85756 [0	.5908]		
Hetero-X test:	F(22, 402) =	1.2915 [0	.1715]		
RESET test:	F(1,424) =	0.00067988 [0.9792]		

Further diagnostic tests:

- \bullet Residuals graphs: Model \rightarrow Test \rightarrow Graphical analysis $\rightarrow \dots$
- \bullet User specified residuals tests: Model \rightarrow Test \rightarrow Test $\rightarrow \ldots$

 \bullet Recursive graphics (VERY useful!): Model \rightarrow Test \rightarrow Recursive graphics $\rightarrow \ldots$

Multiple-equation modelling with Autometrics: Two possibilities

• Seemingly Unrelated Regression (SUR) using OLS/IV, that is, single-equation modelling of each equation separately (requires stationarity of regressors)

• Simultaneous variable deletion (or non-deletion) across equations using vector diagnostic tests but estimation still by OLS (does not require stationarity of regressors), see Doornik and Hendry (2007b, pp. 29-31). (NOTE: IV estimation not available with this strategy)

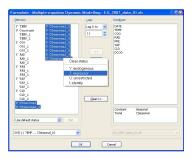
 \rightarrow Model type: "Unrestricted system" (system of URFs), see Doornik and Hendry (2007b, chapter 3)

Formulate a system: (Model \rightarrow) PcGive \rightarrow Category: "Models for time series data" \rightarrow Model class: "Multiple-equation dynamic modelling using PcGive" \rightarrow "Formulate"

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	Formulate> Estimate> Test	
	Options Close	

Multiple-equation modelling with Autometrics using second possibility. *Example*: 2007 Econometric Game, Question 2

• My GUM: A six-dimensional VAR(2) of $y_t = (TEMP_t, COO_t, RAD_t, PRE_t, VAP_t, CLD_t)$, with a constant and 11 centered seasonals in each of the six equations:



NOTE: Fixing variables (that is, not allowing Autometrics to delete them) now differs. Select the variables to delete \rightarrow Right-click mouse \rightarrow U: Unrestricted. (Unfixing: Z: regressor)

Results:

• NOTE: Autometrics simplifies even though the GUM does not pass all diagnostic checks

 \bullet Four variables are removed from all of the equations: The second lag of TEMP, VAP and CLD, and CSeasonal 10

Other type of analysis:

• Cointegration analysis (applied on the Unrestricted system, not on the simplified model): Model \rightarrow Test \rightarrow Dynamic Analysis and Cointegration Tests $\rightarrow \ldots$

See Doornik and Hendry (2007b, chapter 4)

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