

Improved Tests for Granger Non-Causality in Panel Data

Jiaqi Xiao¹, Arturas Juodis², Yiannis Karavias¹, Vasilis Sarafidis³, Jan Ditzen⁴

¹University of Birmingham, Birmingham, UK

²University Amsterdam, Amsterdam, The Netherlands, UK

³BI Norwegian Business School, Oslo, Norway

⁴Free University of Bozen-Bolzano, Bozen, Italy

Swiss Stata Conference 2022
November 18, 2022

Motivation

- “Granger Causality”: a statistical concept of causality (Granger, 1969)
- A variable x “Granger-causes” a variable y , if the variable y can be better predicted using past values of both x and y , than using solely past values of y
- Adopted in economics, medicine, chemistry, physics, biology, engineering, and beyond
- Widely used in panel data analysis

- Holtz-Eakin, Newey and Rosen (1988)
 - Homogeneous panels, large N , small T
 - `pvargranger` by Abrigo and Love (2016)
- Dumitrescu and Hurlin (2012)
 - Heterogeneous panels, large T , small N
 - `xtgcause` by Lopez and Weber (2017)
- Juodis, Karavias and Sarafidis (2021)
 - large N , small-moderate T with heterogeneous coefficients
- `xtgranger` introduces the JKS test in Stata.

The JKS test

- We consider the following linear dynamic panel data model:

$$y_{i,t} = \phi_{0,i} + \sum_{p=1}^P \phi_{p,i} y_{i,t-p} + \sum_{p=1}^P \beta_{p,i} x_{i,t-p} + \varepsilon_{i,t},$$

- $\beta_{p,i}$ are the heterogeneous feedback coefficients, or Granger causality parameters.
- The hypothesis of Granger non-causality is:

$$H_0 : \beta_{p,i} = 0, \text{ for all } i \text{ and } p.$$

The alternative hypothesis is:

$$H_1 : \beta_{p,i} \neq 0 \text{ for some } i \text{ and } p.$$

The JKS test: key estimation ideas

- Idea 1: under the null hypothesis: $\beta_{p,i} = 0$, for all i and p , and thus we can use a more efficient (\sqrt{NT}) pooled estimator

$$\hat{\beta} = \left(\sum_{i=1}^N \mathbf{X}'_i \mathbf{M}_{\mathbf{Z}_i} \mathbf{X}_i \right)^{-1} \left(\sum_{i=1}^N \mathbf{X}'_i \mathbf{M}_{\mathbf{Z}_i} \mathbf{y}_i \right).$$

distributed as

$$\sqrt{NT} (\hat{\beta} - \beta_0) \rightarrow \mathbf{J}^{-1} N(-\kappa \mathbf{B}, \mathbf{V}),$$

- Idea 2: Remove the bias using the half panel jackknife estimator (Dhaene and Jochmans, 2015):

$$\tilde{\beta} = \hat{\beta} + \left(\hat{\beta} - \frac{1}{2} (\hat{\beta}_{1/2} + \hat{\beta}_{2/1}) \right) = \hat{\beta} + T^{-1} \hat{\mathbf{B}}.$$

The JKS test: key estimation ideas

- As $N, T \rightarrow \infty$ with $N/T \rightarrow \kappa^2 \in [0, \infty)$, we have:

$$\hat{W}_{HPJ} = NT\tilde{\beta}' (\hat{\mathbf{J}}^{-1}\hat{\mathbf{V}}\hat{\mathbf{J}}^{-1})^{-1} \tilde{\beta} \rightarrow \chi^2(P),$$

where $\hat{\mathbf{J}} = (NT)^{-1} \sum_{i=1}^N \mathbf{X}_i' \mathbf{M}_{\mathbf{Z}_i} \mathbf{X}_i$.

- Power against both homogeneous as well as heterogeneous alternatives
- Cross-section heteroskedasticity
- Weak Cross-section dependence
- Multivariate $x_{i,t}$
- Fast computation

The xtgranger command

- The command syntax is:

```
xtgranger depvar [indepvars] [if] [in] [, options]
```

Data must be `xtset` before using `xtgranger`. The panel must be balanced.

- `lags(#)` specifies the number of lags of dependent and independent variables to be added to the regression. If `lags(#)` is not specified, the default is `lags(1)`.
- `maxlags(#)` specifies the upper bound of lags. The BIC criterion is used to select the number of lags that provides the best model fit.
- `het` allows for cross-sectional heteroskedasticity.
- `bootstrap(reps, seed(seed))` employs a bootstrap variance estimator in the HPJ Wald statistic with the custom set `seed` and `reps` repetitions.

Determinants of Bank Profitability

- We examine temporal relation between profitability, cost efficiency and asset quality in the U.S. banking industry.
- Random sample of 450 U.S. bank holding companies (BHC), each one observed over 56 time periods, namely 2006:Q1-2019:Q4.
- We estimate

$$ROA_{i,t} = \phi_{0,i} + \sum_{p=1}^P \phi_{p,i} ROA_{i,t-p} + \sum_{p=1}^P \beta_{1,p,i} INEFFICIENCY_{i,t-p} \\ + \sum_{p=1}^P \beta_{2,p,i} QUALITY_{i,t-p} + \varepsilon_{i,t},$$

Determinants of Bank Profitability

```
. xtgranger roa inefficiency quality, maxlags(4) het
Juodis, Karavias and Sarafidis (2021) Granger non-causality Test
-----
Number of units= 450          Obs. per unit (T) = 55
Number of lags = 1            BIC      = -34257.34
-----
JKS non-causality test
H0: Selected covariates do not Granger-cause roa.
H1: H0 is violated.
HPJ Wald test : 30.2387
p-value       : 0.0000
-----
BIC selection:
lags = 1, BIC = -34257.336*
lags = 2, BIC = -33371.195
lags = 3, BIC = -32727.595
lags = 4, BIC = -32715.923
-----
Results for the Half-Panel Jackknife estimator
Cross-sectional heteroskedasticity-robust variance estimation

```

	Coefficient	Std. err.	z	P> z	[95% conf. interval]
inefficiency	.2562039	.0572807	4.47	0.000	.1439358 .368472
L1.					
quality	-.0162294	.0444754	-0.36	0.715	-.1033996 .0709409
L1.					

Determinants of Bank Profitability

```
. xtgranger roa inefficiency, maxlags(4) het
```

Juodis, Karavias and Sarafidis (2021) Granger non-causality Test

Number of units= 450

Obs. per unit (T) = 55

Number of lags = 1

BIC = -33295.8

JKS non-causality test

H0: inefficiency does not Granger-cause roa.

H1: inefficiency does Granger-cause roa for at least one panelvar.

HPJ Wald test : 24.3174

p-value : 0.0000

BIC selection:

lags = 1, BIC = -33295.799*

lags = 2, BIC = -32170.227

lags = 3, BIC = -31112.604

lags = 4, BIC = -30724.676

Results for the Half-Panel Jackknife estimator

Cross-sectional heteroskedasticity-robust variance estimation

	Coefficient	Std. err.	z	P> z	[95% conf. interval]
inefficiency	.2549723	.0517052	4.93	0.000	.1536319 .3563127
L1.					

Determinants of Bank Profitability

```
. xtgranger roa quality, maxlags(4) het  
Juodis, Karavias and Sarafidis (2021) Granger non-causality Test
```

Number of units= 450	Obs. per unit (T) = 55
Number of lags = 1	BIC = -33816.06

JKS non-causality test

H0: quality does not Granger-cause roa.

H1: quality does Granger-cause roa for at least one panelvar.

HPJ Wald test : 0.2090

p-value : 0.6476

BIC selection:

lags = 1, BIC = -33816.061*

lags = 2, BIC = -32649.24

lags = 3, BIC = -31479.433

lags = 4, BIC = -30607.217

Results for the Half-Panel Jackknife estimator

Cross-sectional heteroskedasticity-robust variance estimation

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
quality	-.0201426	.0440637	-0.46	0.648	-.1065059	.0662207
L1.						

Determinants of Bank Profitability

```
. xtgranger roa inefficiency quality, maxlags(4) het sum, if cluster==2 & time>20  
Juodis, Karavias and Sarafidis (2021) Granger non-causality Test
```

Number of units= 183	Obs. per unit (T) = 34
Number of lags = 2	BIC = -10307.93

JKS non-causality test

H0: Selected covariates do not Granger-cause roa.
H1: H0 is violated.

HPJ Wald test : 36.3572
p-value : 0.0000

BIC selection:

lags = 1, BIC = -10249.44
lags = 2, BIC = -10307.934*
lags = 3, BIC = -9788.8299
lags = 4, BIC = -9963.9685

Sum of Half-Panel Jackknife coefficients across lags (lags>1)
Cross-sectional heteroskedasticity-robust variance estimation

	Coefficient	Std. err.	z	P> z	[95% conf. interval]
inefficiency	.4906756	.2405474	2.04	0.041	.0192113 .9621398
quality	-.1765458	.1235961	-1.43	0.153	-.4187897 .0656981

Conclusions

- Introduced new community contributed package called `xtgranger`
- Tests for Granger non-causality when N is large and T is moderate/small
- The command allows for cross-section heteroskedasticity, weak cross-section dependence, multivariate systems
- The command is available on the SSC Archive
- `ssc install xtgranger`
- More information in the paper [Improved Tests for Granger Non-Causality in Panel Data](#)