

# THE SENSITIVITY OF GRANGER CAUSALITY

M. Deistler (TU Vienna and IHS Vienna), Brian D.O. Anderson  
(ANU Canberra and Hangzhou University) and Jean Marie  
Dufour (Mc Gill University)

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# Overview

1. **Introduction**
2. Characterisations of Granger Causality
3. Granger Causality and Additive Noise

# Introduction

## Causality

- ▶ often known a priori, in particular in physics. E.g. Ohm's law  $U = RI$ . This formula does not reveal the direction of causation
- ▶ causal inference: the process of drawing a conclusion about a causal connection

The Sveriges Riksbank Prize in Economic Sciences in Memory of Alfred Nobel 2011 has been awarded to T.J. Sargent and C. Sims "for their empirical research on cause and effect in the macroeconomy"

# Granger Causality

Let  $(X(t) \mid t \in \mathbb{Z}), X(t) : \Omega \rightarrow \mathbb{R}^d$  be a weakly stationary, centered and Gaussian vector process with full rank stationary rational spectral density  $f(z)$ .

Write

$$X(t) = \begin{pmatrix} X_A(t) \\ X_B(t) \end{pmatrix}.$$

We say that  $(X_A(t))$  does not **Granger cause** (GC)  $(X_B(t))$  if

$$E(X_B(t) \mid X_A(s), X_B(s), s < t) = E(X_B(s) \mid X_B(s), s < t).$$

## Instantaneous causality

$$E(X_B(t) \mid X_A(t), X_A(s), X_B(s), s < t) \neq E(X_B(t) \mid X_A(s), X_B(s), s < t).$$

Post hoc, ergo propter hoc Fallacy.

## Other concepts:

- ▶ Peter Bühlmann (2018): Invariance, Causality and Robustness. *ETH Zürich*.
- ▶ Jim Heckman (2014): Causal Models, Structural Equations and Identification. *Asia Meeting of The Econometric Society, Taipei*.
- ▶ Judea Pearl (2000): Causality: Models, Reasoning and Inference. *Cambridge University Press*.

# Questions of this Presentation

How sensitive is GC with respect to

- ▶ errors in variables
- ▶ (linear transformations)
- ▶ (subsampling)

Anderson, B.D.O., M. Deistler and J.B. Dufour (2019): On the Sensitivity of Granger Causality to Errors-in-Variables, Linear Transformations and Subsampling. *Journal of Time Series Analysis*, 40, 102-123.

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4. Signal to Noise Ratio and GC



# Characterizations of Granger Causality

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# Characterizations of Granger Causality

Wold representation (based on spectral factorization)

$$f(z) = W(z)QW^*(z) = \begin{pmatrix} f_{AA}(z) & f_{AB}(z) \\ f_{BA}(z) & f_{BB}(z) \end{pmatrix}$$

$W(z)$  square, real rational, stable and (strictly) miniphase transferfunction,  $W(0) = I$ ,  $Q > 0$ .

$$X(t) = W(z)\varepsilon(t); \quad W(z) = \begin{pmatrix} W_{11}(z) & W_{12}(z) \\ W_{21}(z) & W_{22}(z) \end{pmatrix}; \quad Q = \begin{pmatrix} Q_{11} & Q_{12} \\ Q_{21} & Q_{22} \end{pmatrix}$$

$AR(\infty)$  representation of  $(X(t))$ :

$$\pi(z)X(t) = \varepsilon(t); \quad X(t) = \underbrace{a(z)X(t)}_{\sum_{i=1}^{\infty} A_i X(t-i)} + \varepsilon(t)$$
$$a(z) = \begin{pmatrix} a_{11}(z) & a_{12}(z) \\ a_{21}(z) & a_{22}(z) \end{pmatrix}.$$

$(\varepsilon(t))$  are innovations,

# Characterizations of Granger Causality

## **Theorem 1** (Canonical spectral factor characterizaion)

The following conditions are equivalent:

- (i)  $X_A$  does not GC  $X_B$
- (ii)  $W_{21}(z) = 0$ .

The following conditions are also equivalent:

- (i)  $X_A$  neither GC  $X_B$ , nor does it cause  $X_B$  instantaneously
- (ii)  $W_{21}(z) = 0$  and  $Q$  is block diagonal.

## **Theorem 2** (AR representation characterization)

$X_A$  does not GC  $X_B$  if and only if the 21-block of  $a(z)$ , i.e.  $a_{21}(z)$  is zero.

## **Theorem 3** (Spectrum based characterization)

$X_A$  does not GC  $X_B$  if and only if  $f_{AB}(z)f_{BB}^{-1}(z)$  is causal.

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# GC and Additive Noise

Observations:

$$\bar{X}_A = X_A + N_A$$

$$\bar{X}_B = X_B + N_B$$

$N_A, N_B$  stationary noise, with rational spectral densities, uncorrelated with each other and with  $X_A, X_B$ .

Questions:

- ▶ Suppose that  $X_A$  does not GC  $X_B$ , does  $\bar{X}_A$  not GC  $\bar{X}_B$  ?
- ▶ Suppose that  $\bar{X}_A$  does not GC  $\bar{X}_B$ , can one conclude that  $X_A$  does not GC  $X_B$  ?

Note:

$$\begin{aligned}f_{\bar{A}\bar{A}} &= f_{AA} + f_{N_A N_A}; & f_{\bar{B}\bar{B}} &= f_{BB} + f_{N_B N_B} \\f_{AB} &= f_{\bar{A}\bar{B}}.\end{aligned}$$

#### Theorem 4:

1. If  $N_B = 0$  then  $X_A$  does not GC  $X_B$  if and only if  $\bar{X}_A$  does not GC  $\bar{X}_B$ .
2. If  $N_B \neq 0$  and not all unstable zeros of  $f_{BB} + f_{N_B N_B}$  cancel the (unstable) zeros of  $f_{AB}$  (this "generically" is the case), then if  $X_A$  does not GC  $X_B$ , we have that  $\bar{X}_A$  GC  $\bar{X}_B$ .
3. It generically holds, that if  $\bar{X}_A$  does not GC  $\bar{X}_B$ , then  $X_A$  GC  $X_B$ .

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# Signal to Noise Ratio and GC

**Lemma** (Continuity Result)

Consider a complex matrix function  $M(z)$ , analytic in  $\rho < |z| < \rho^{-1}$ ;  $0 < \rho < 1$  with  $M(z) = M^T(z^{-1})$  and positive definite on  $|z| = 1$ . Suppose

$$M(z) = \sum_{i=-\infty}^{\infty} m_i z^i, \quad m_i = m_{-i}^T \in \mathbb{R}^{d \times d}$$

and define the causal and anticausal parts by

$$M_+(z) = \frac{1}{2}m_0 + \sum_{i=1}^{\infty} m_i z^i, \quad M_-(z) = \frac{1}{2}m_0 + \sum_{i=-\infty}^{-1} m_i z^i$$

Then the matrix function  $L(z) = I + \varepsilon M$  is analytic in  $\rho < |z| < \rho^{-1}$  with  $L(z) = L^T(z^{-1})$  and positive definite on  $|z| = 1$ .

Further, to the first order in  $\varepsilon > 0$  there holds

$$L = I + \varepsilon M \approx (I + \varepsilon M_+)(I + \varepsilon M_-)$$

with  $I + \varepsilon M_+$  stable and miniphase.



# Signal to Noise Ratio and GC

Terminology "to first order in  $\epsilon$ " is shorthand for saying that the  $L_2$  norm of the error  $\Delta z$ , defined via  $\text{tr}(2\pi)^{-1} \int \Delta(e^{-\iota\lambda}) \Delta^*(e^{-\iota\lambda}) d\lambda$  is of order  $\epsilon$ :  $O(\epsilon)$ .

Idea: Small perturbations in a spectrum give small perturbations in a spectral factor: If  $X_A$  does not GC  $X_B$  and if the noise  $N_B$  is small, then the "violation of GC is small too".

**Theorem 5** Define  $\bar{X}_B = X_B + \varepsilon^{\frac{1}{2}} N_B$ , so that  $f_{\bar{X}\bar{X}} = f_{XX} + \varepsilon f_{NN}$  where the 11, 12, 21 blocks of  $f_{NN}$  are zero and the 22 block in  $f_{N_B N_B}$ . Let  $W(z)Q$  with  $W(z)$  upper block triangular and  $Q$  block diagonal and  $\bar{W}(z), \bar{Q}$  define the canonical spectral factorizations  $f_{XX}$  and  $f_{\bar{X}\bar{X}}$  respectively. Then

- (i)  $\bar{W}(z) - W(z)$  is  $O(\varepsilon)$  on  $|z| = 1$
- (ii)  $\bar{Q} - Q$  is  $O(\varepsilon)$
- (iii)  $f_{A\bar{B}} f_{\bar{B}\bar{B}}^{-1} - f_{AB} f_{BB}^{-1}$  is  $O(\varepsilon)$  on  $|z| = 1$  and the anticausal part of  $f_{A\bar{B}} f_{\bar{B}\bar{B}}^{-1}$  is  $O(\varepsilon)$  on  $|z| = 1$
- (iv) For suitably small  $\varepsilon$ ,  $\bar{W}_{22}(z)$  is miniphase

So  $X$  is "close to" a process in which  $X_A$  does not cause  $\bar{X}_B$ .

# Conclusion

- ▶ Noise on  $X_A$  only does not change the "Granger causality relation of the  $A$  part to the  $B$  part".
- ▶ Noise on  $X_B$  generically changes the "Granger causality relation of the  $A$  part to the  $B$  part".
- ▶ Small noise on  $X_B$  leads to "weak" violations of Granger non causality relations

**THANK YOU**