

Causality analysis

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PATTERN / WIGNER MTA

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NAP-B PATTERN (2015-2017)
Population Activity Research Unit
MTA WIGNER RESEARCH CENTRE FOR PHYSICS

joint work with

Zsigmond Benkő,



Ádám Zlatniczki,



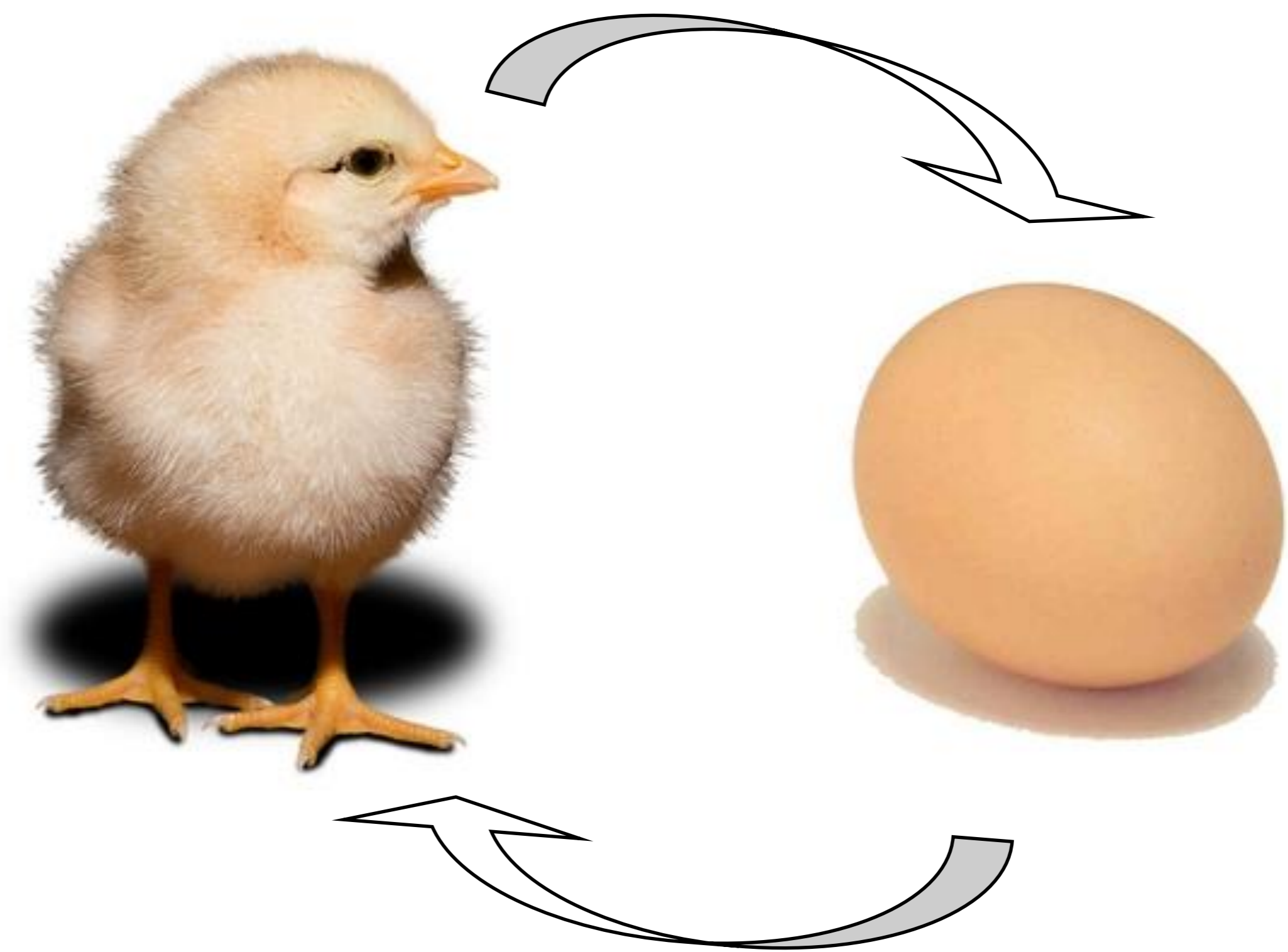
Dániel Fabó,



Zoltán Somogyvári



Which was first?



Puzzle for you:

What on earth to do with GPU about causality?

Find the right spot in the lecture!



Wiener-Granger, predictive causality

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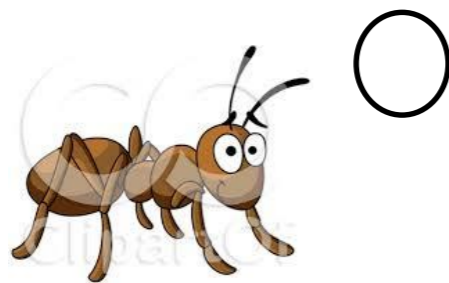
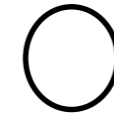
Wiener-Granger causality

1. Axiom – cause precedes caused
2. Axiom – Using the past of the cause improves the forecast of the caused based solely on its own past.

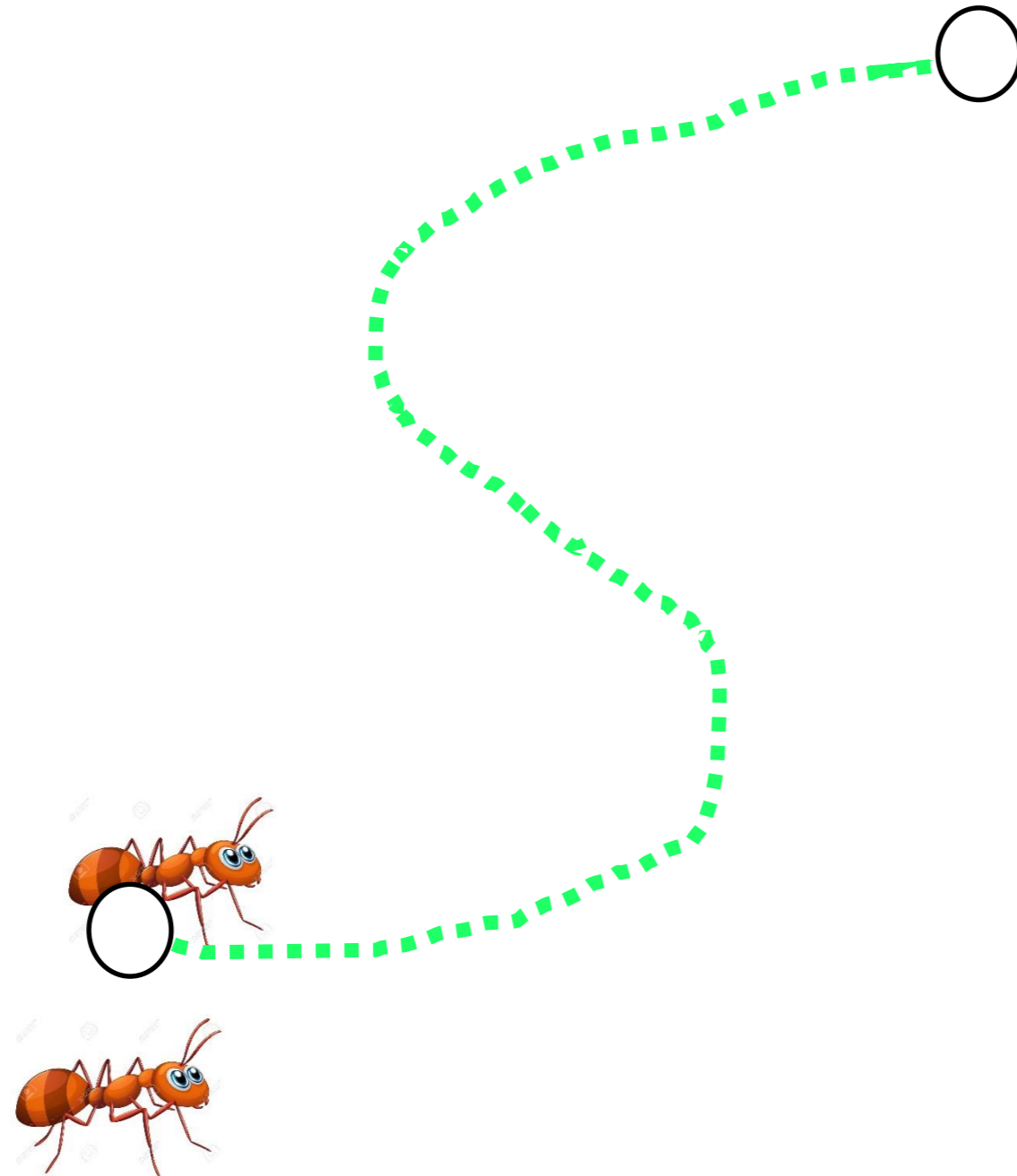
Granger causality

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Life of the ants



Life of the ants



Let $f_t: M \rightarrow M$

$$a_{t+1} = f_t(a_t)$$

the map for a discrete time dynamical system with a strange attractor \mathcal{X} with box counting dimension $d_{\mathcal{X}}$.

$x_t = g(a_t)$ observation

g must be twice-differentiable observation function,
 $m > 2d_{\mathcal{X}}$ then, the delay embedding \mathcal{X}

$X_t = (x_t, x_{t-1}, \dots, x_{t-m+1})$ reconstruct (up to ...) the state space of a

embeds \mathcal{X} into R^m and left $d_{\mathcal{X}}$ invariant.

Embedding of single variable

$$(x_t, x_{t-1})$$

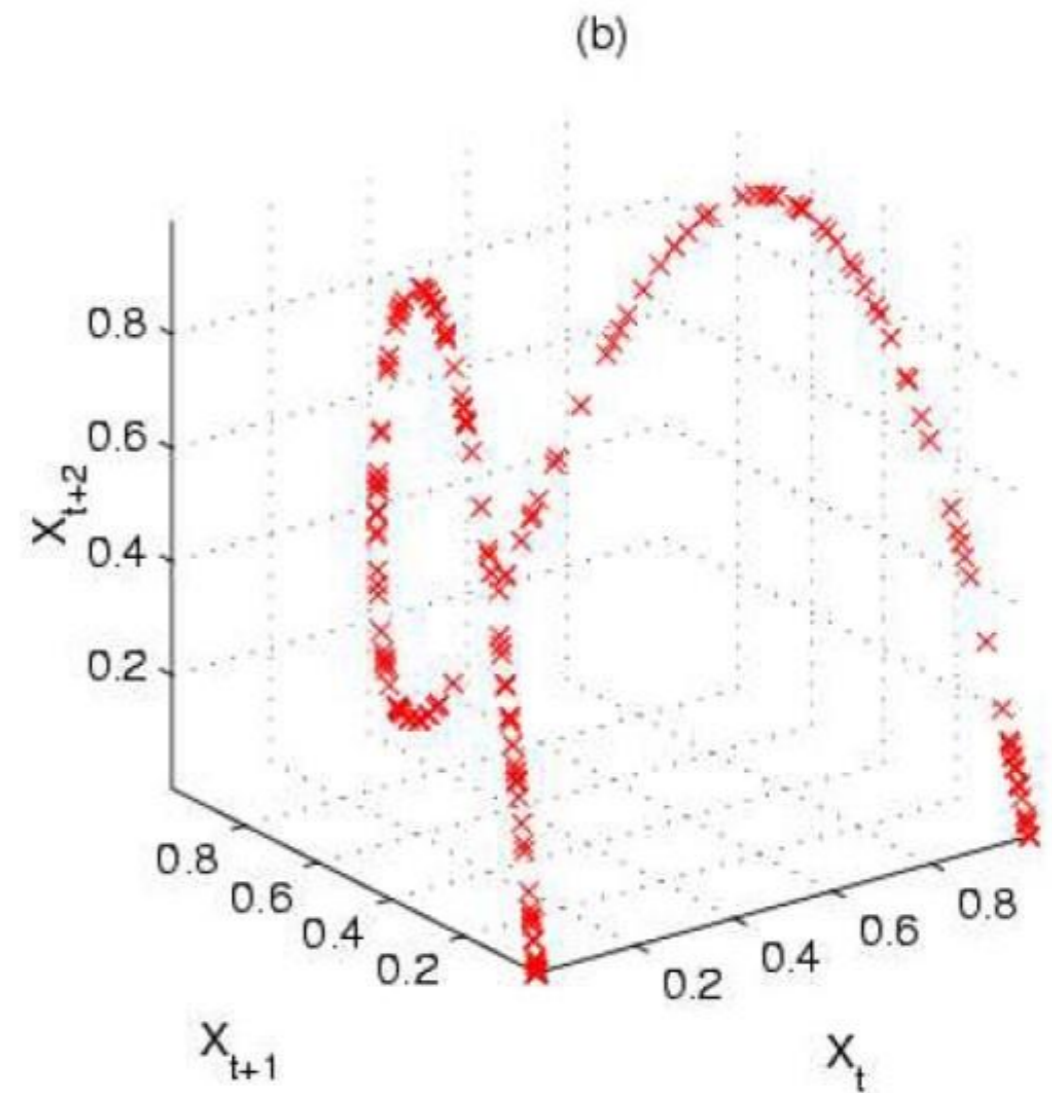
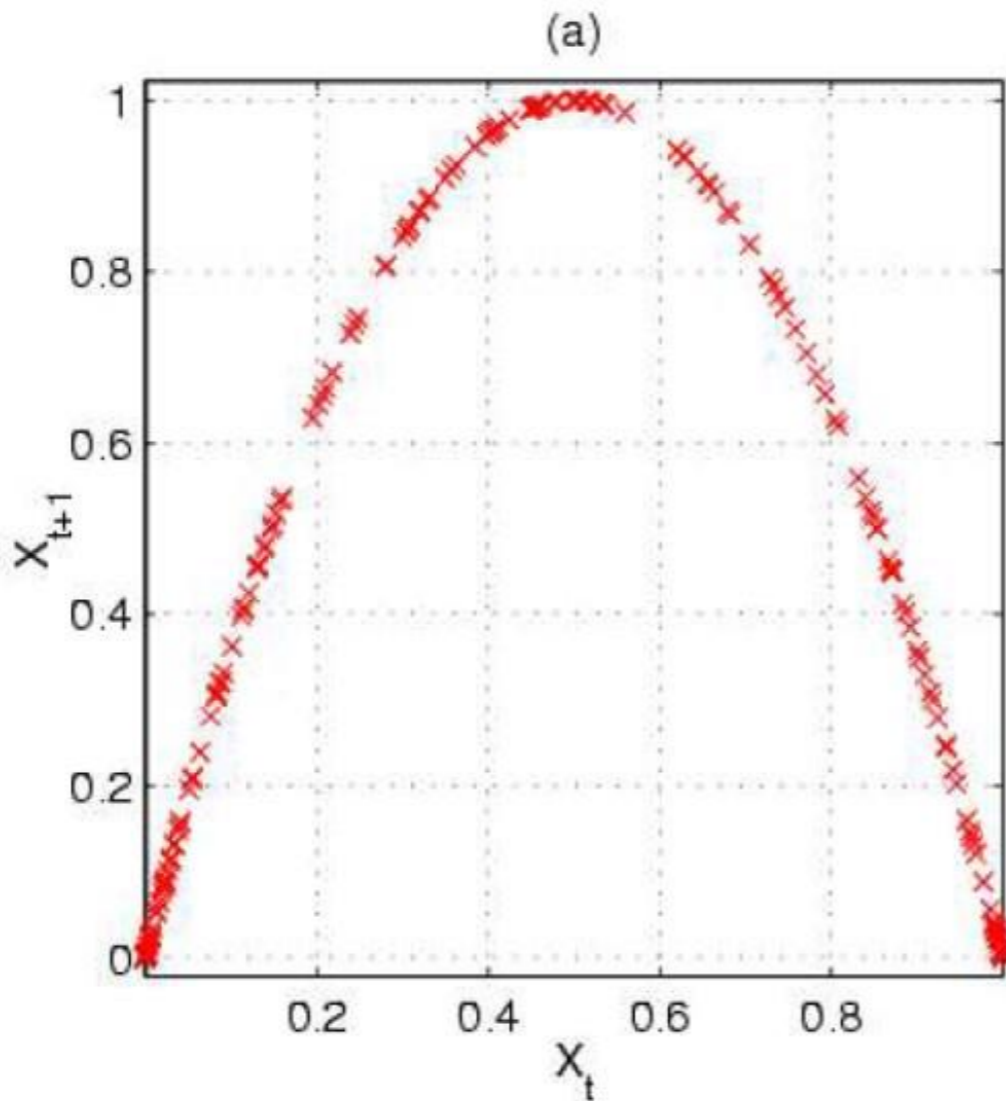
$$m=2$$

$$(x_t, x_{t-1}, x_{t-2})$$

$$m=3$$

Example: logistic map

$$x_{n+1} = r x_n (1 - x_n)$$



Embedded in $D=2,3$, the manifold is still one dimensional.

Embedding of single variables

$$(x_t, x_{t-1}, x_{t-2})$$

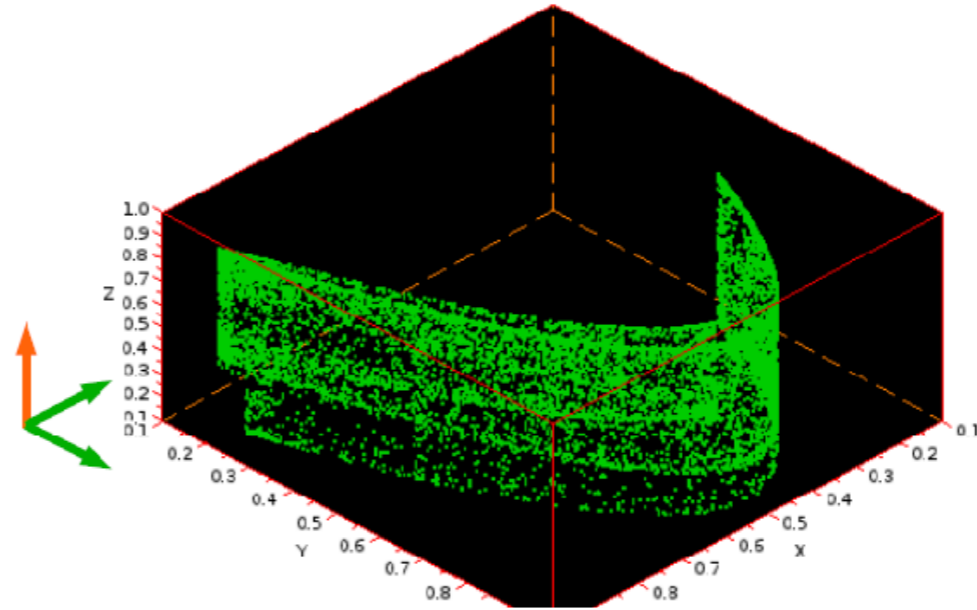
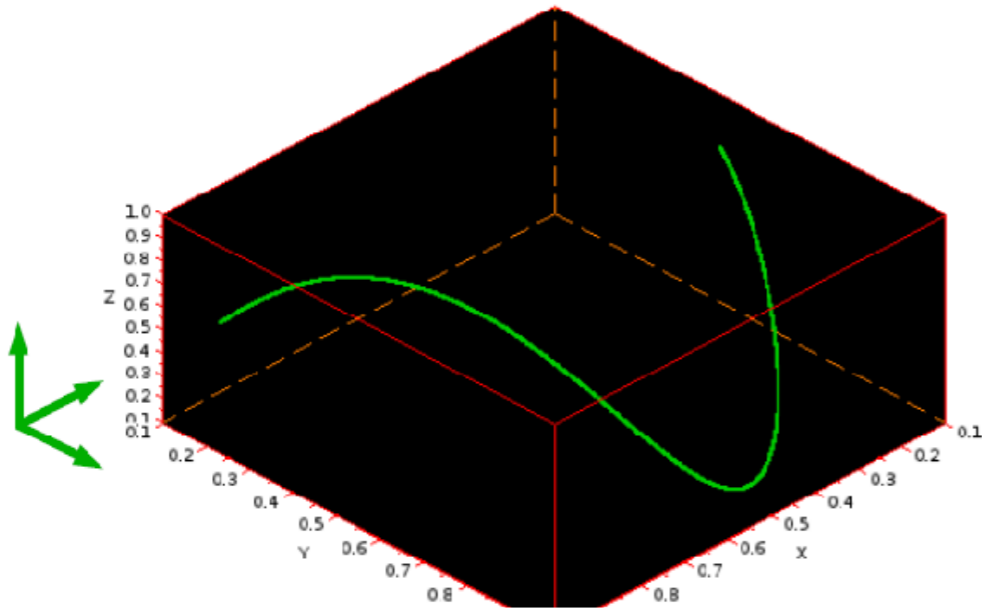
$$(y_t, y_{t-1}, y_{t-2})$$

Joint embedding

$$(x_t, x_{t-1}, y_t)$$

$$(y_{t-2}, y_{t-1}, y_t)$$

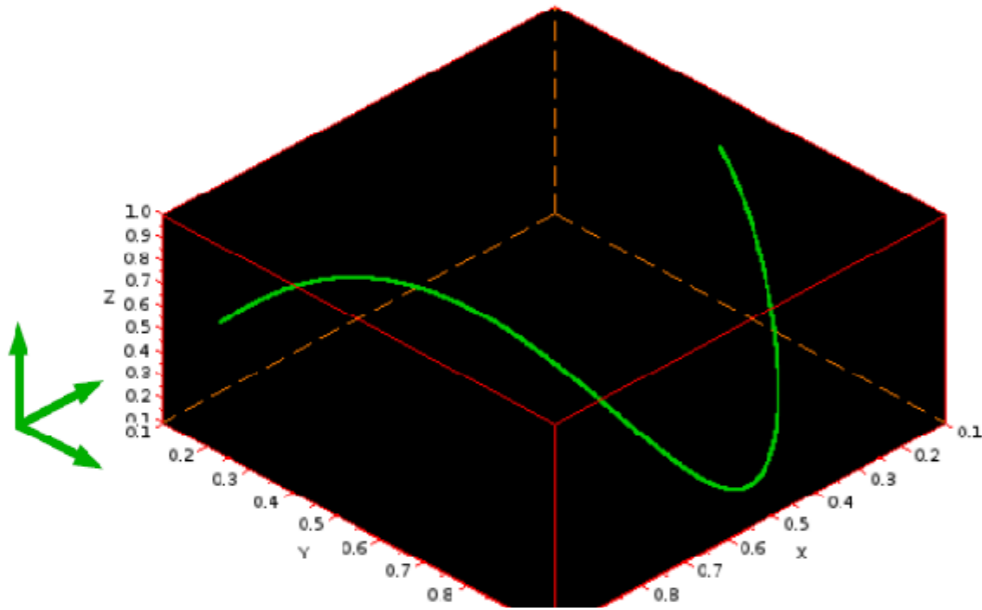
$$(y_{t-2}, y_{t-1}, x_t)$$



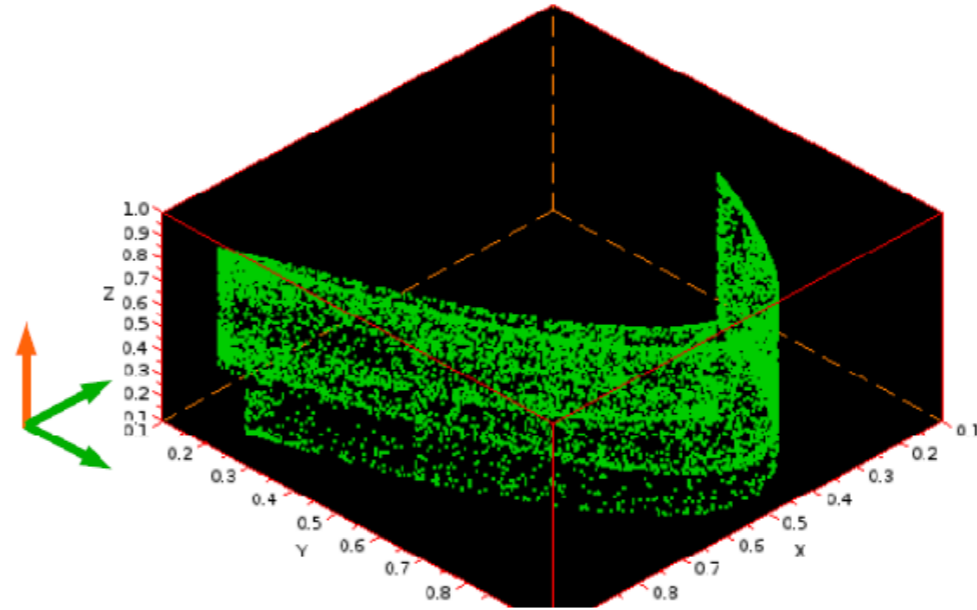
Example 1.

$$(y_{t-2}, y_{t-1}, y_t)$$

$$(y_{t-2}, y_{t-1}, x_t)$$



y d=1 (in D=3)

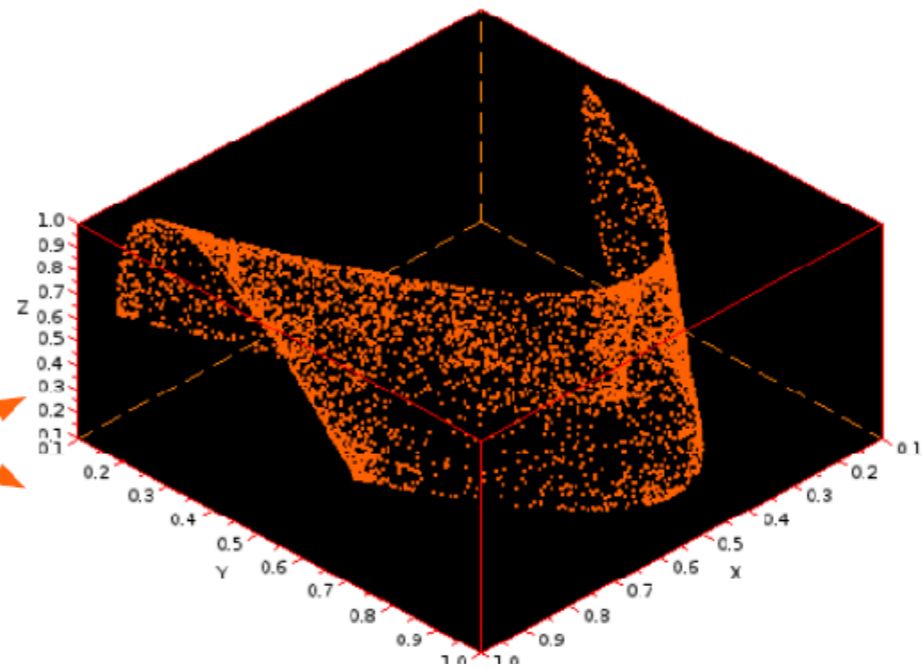
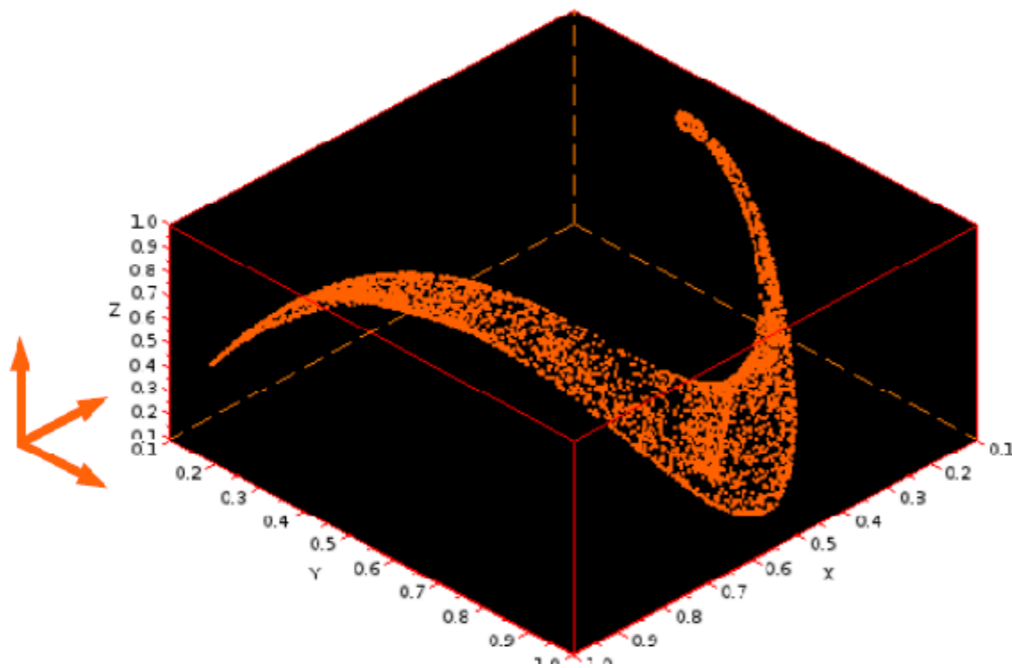


Joint d=2 (in D=3)

Dimension increase indicates independence

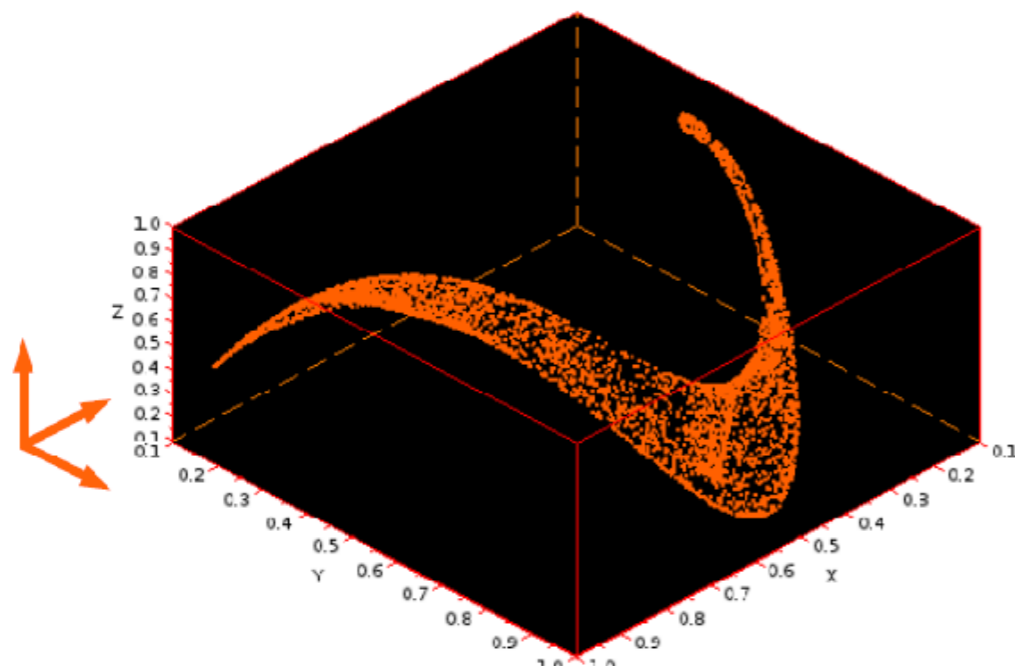
$$(x_{t-2}, x_{t-1}, x_t)$$

$$(x_{t-2}, x_{t-1}, y_t)$$



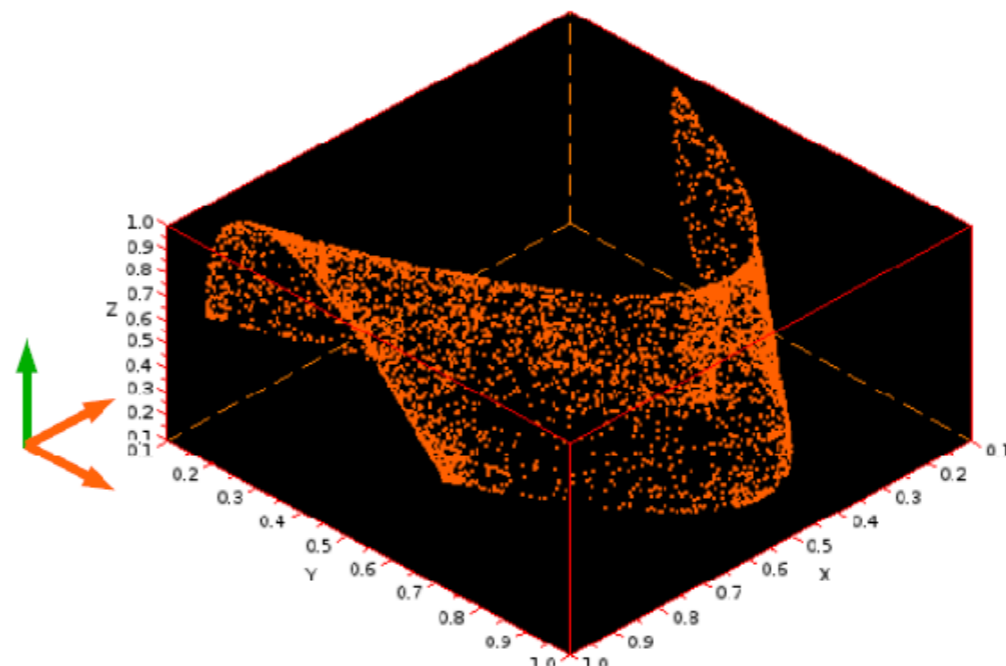
Example 2.

$$(x_{t-2}, x_{t-1}, x_t)$$



2d in 3D

$$(x_{t-2}, x_{t-1}, y_t)$$



joint embedding is still 2d

Lack of dimension increase indicates causality,
y causes x

X_t , stationary time series $t=1, \dots, n$

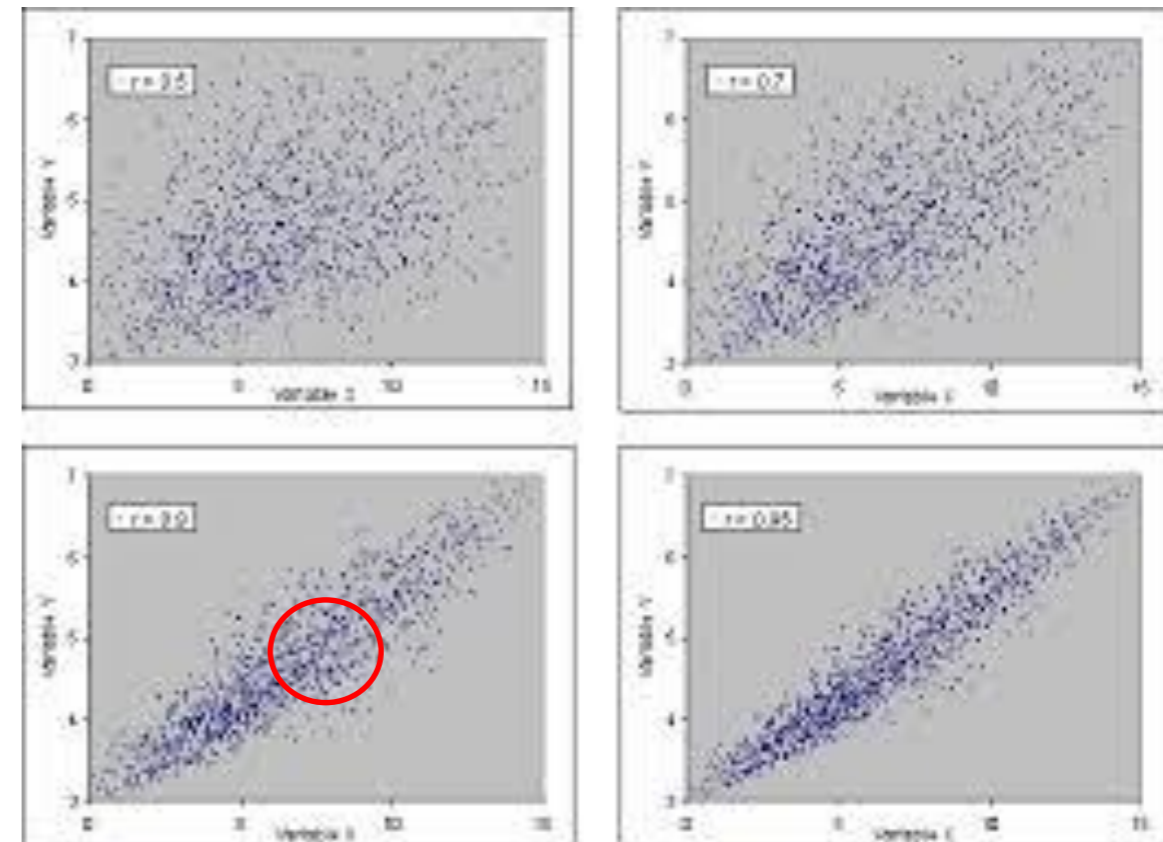
$X_t = (x_t, x_{t-1}, \dots, x_{t-m+1})$ embedded in R^m

$N(X_t, r) = \# \{ s : |X_s - X_t| < r \}$

$N(X_t, r) \approx r^{d(X)}$ local dimensions

d_x average of local dimensions

d_x is the intrinsic dimension of the X -manifold





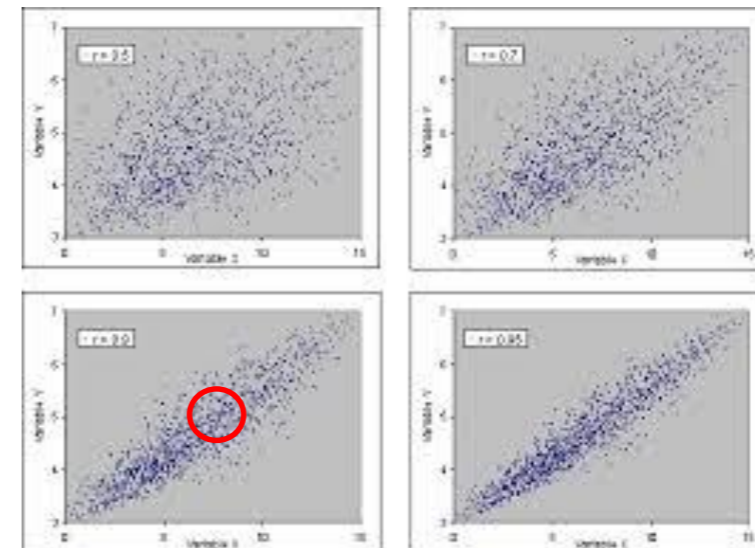
Intrinsic dimension

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intrinsic dimension = information dimension

X_t, Y_t time series

$X_t = (x_t, x_{t-1}, \dots, x_{t-m+1}), Y_t = (y_t, y_{t-1}, \dots, y_{t-m+1})$ embedded in R^m

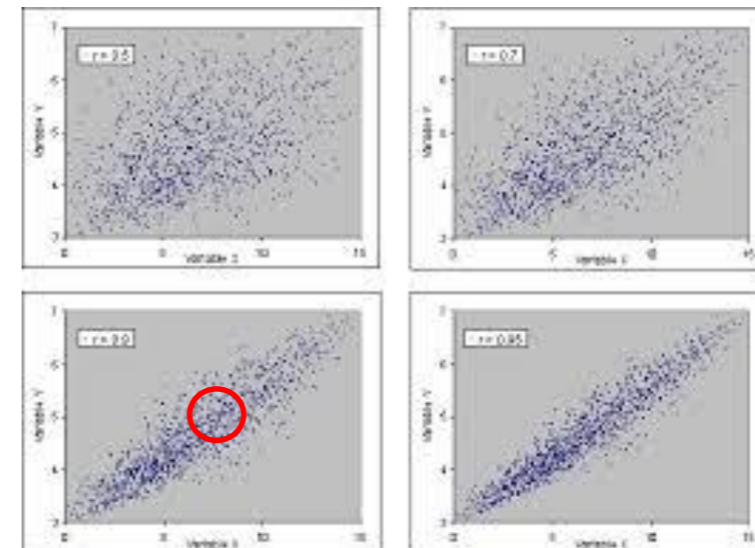


X_t, Y_t time series

$X_t = (x_t, x_{t-1}, \dots, x_{t-m+1}), Y_t = (y_t, y_{t-1}, \dots, y_{t-m+1})$ embedded in R^m

And the joint:

$J_t = (X_t, Y_t) = (x_t, x_{t-1}, \dots, x_{t-m+1}, y_t, y_{t-1}, \dots, y_{t-m+1})$



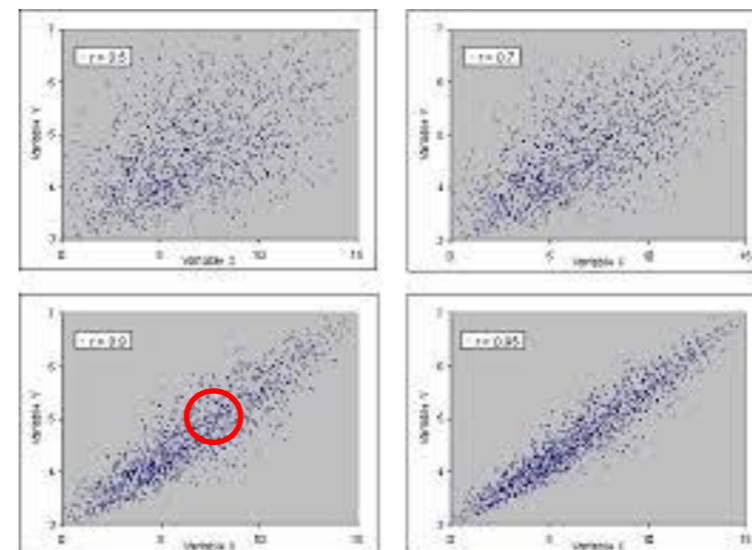
X_t, Y_t time series

$X_t = (x_t, x_{t-1}, \dots, x_{t-m+1}), Y_t = (y_t, y_{t-1}, \dots, y_{t-m+1})$ embedded in R^m

And the joint:

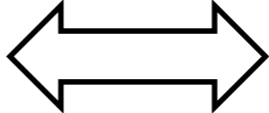
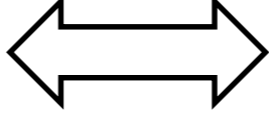
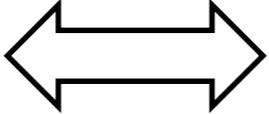
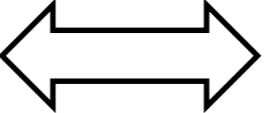
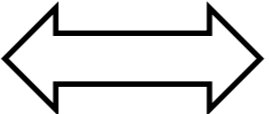
$J_t = (X_t, Y_t) = (x_t, x_{t-1}, \dots, x_{t-m+1}, y_t, y_{t-1}, \dots, y_{t-m+1})$

d_X, d_Y, d_J is the intrinsic dimension of the manifold, X, Y and the joint variable



In general

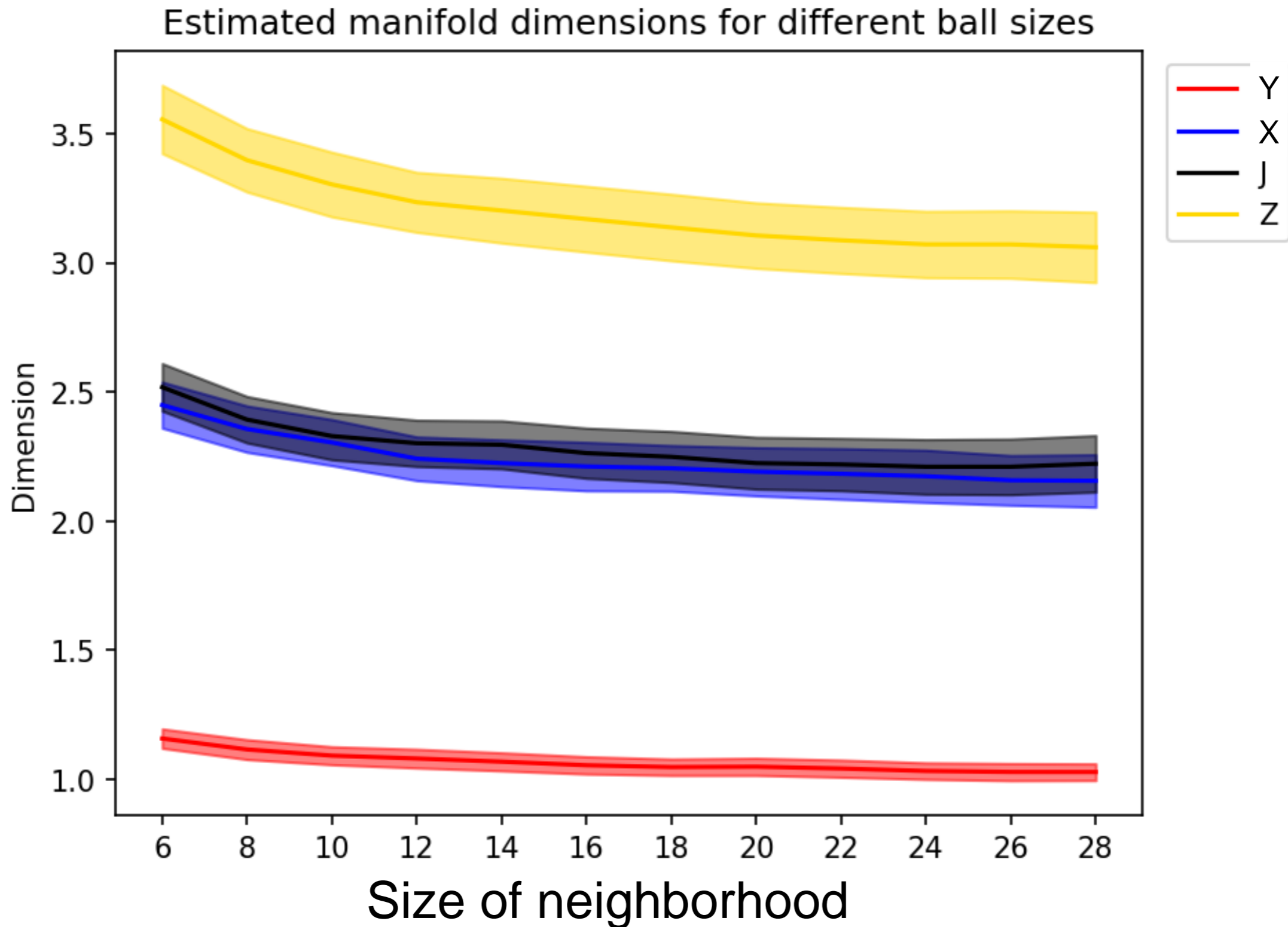
$$\max\{d_X, d_Y\} \leq d_J \leq d_X + d_Y$$

Dimensions		Causal relation
$d_X < d_Y = d_J$		X drives Y
$d_Y < d_X = d_J$		Y drives X
$d_X = d_Y = d_J$		X circular Y
$\max\{d_X, d_Y\} < d_J = d_X + d_Y$		X and Y are independent
$\max\{d_X, d_Y\} < d_J < d_X + d_Y$		X, Y have a common cause

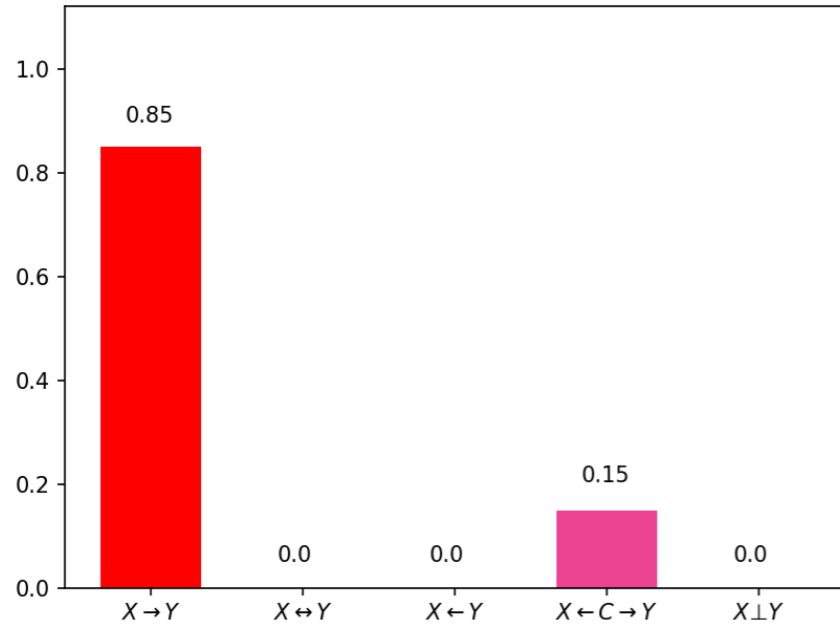


Example and test the logistic map

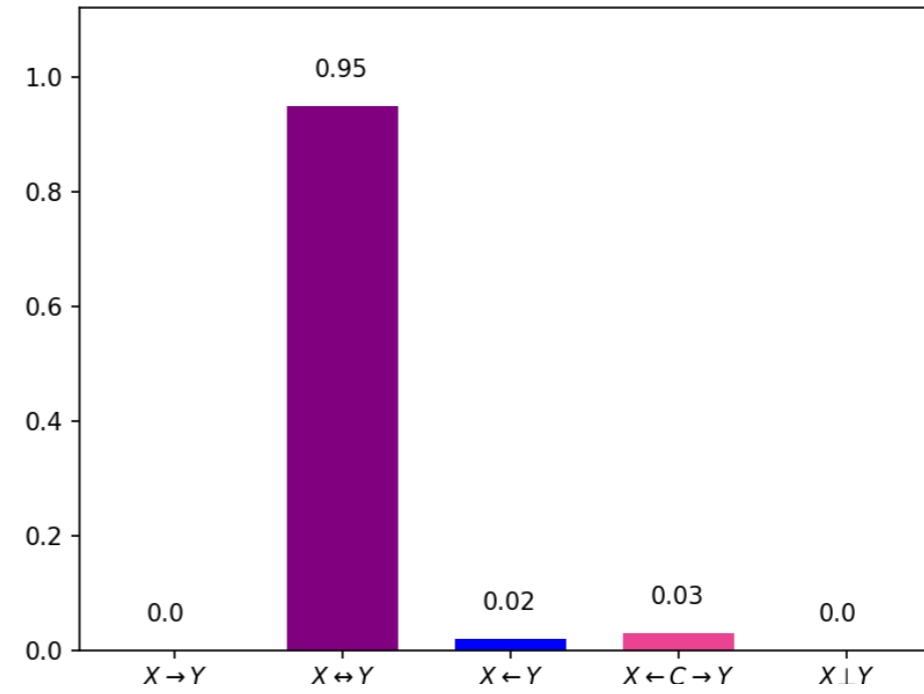
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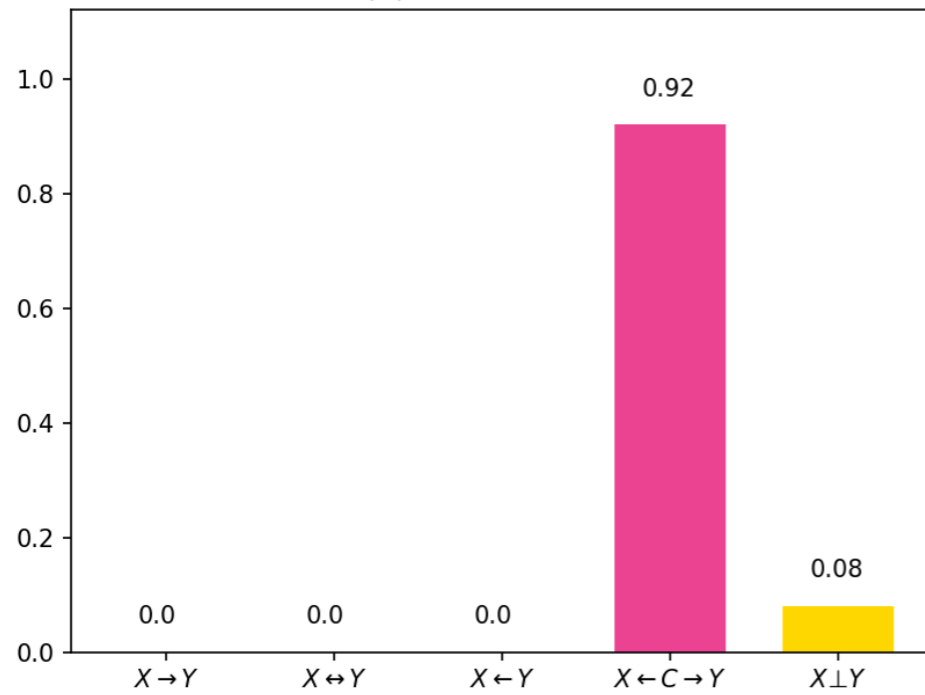
(A) Direct cause



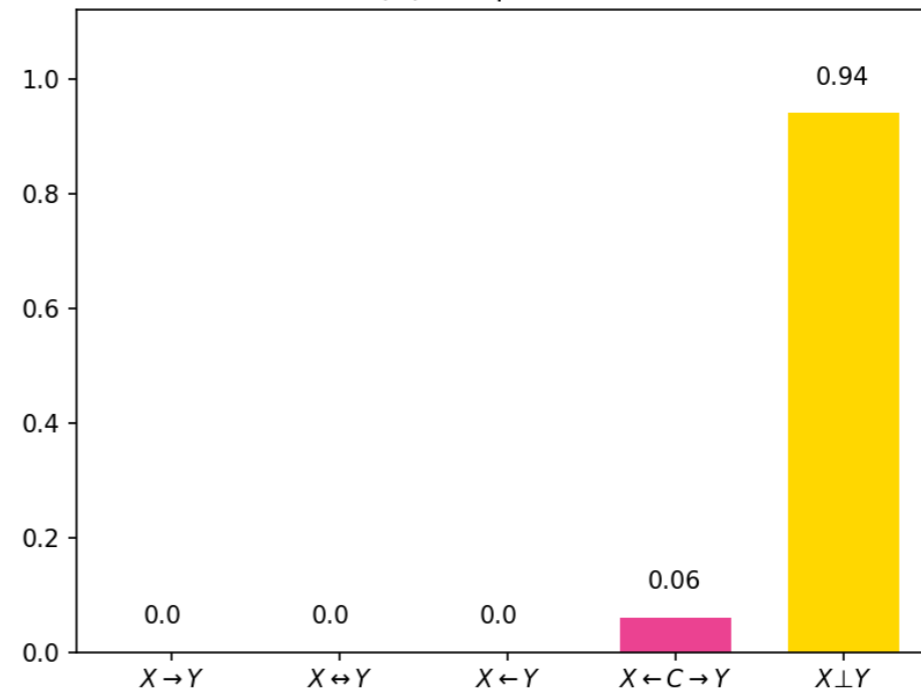
(B) Circular cause



(C) Common cause

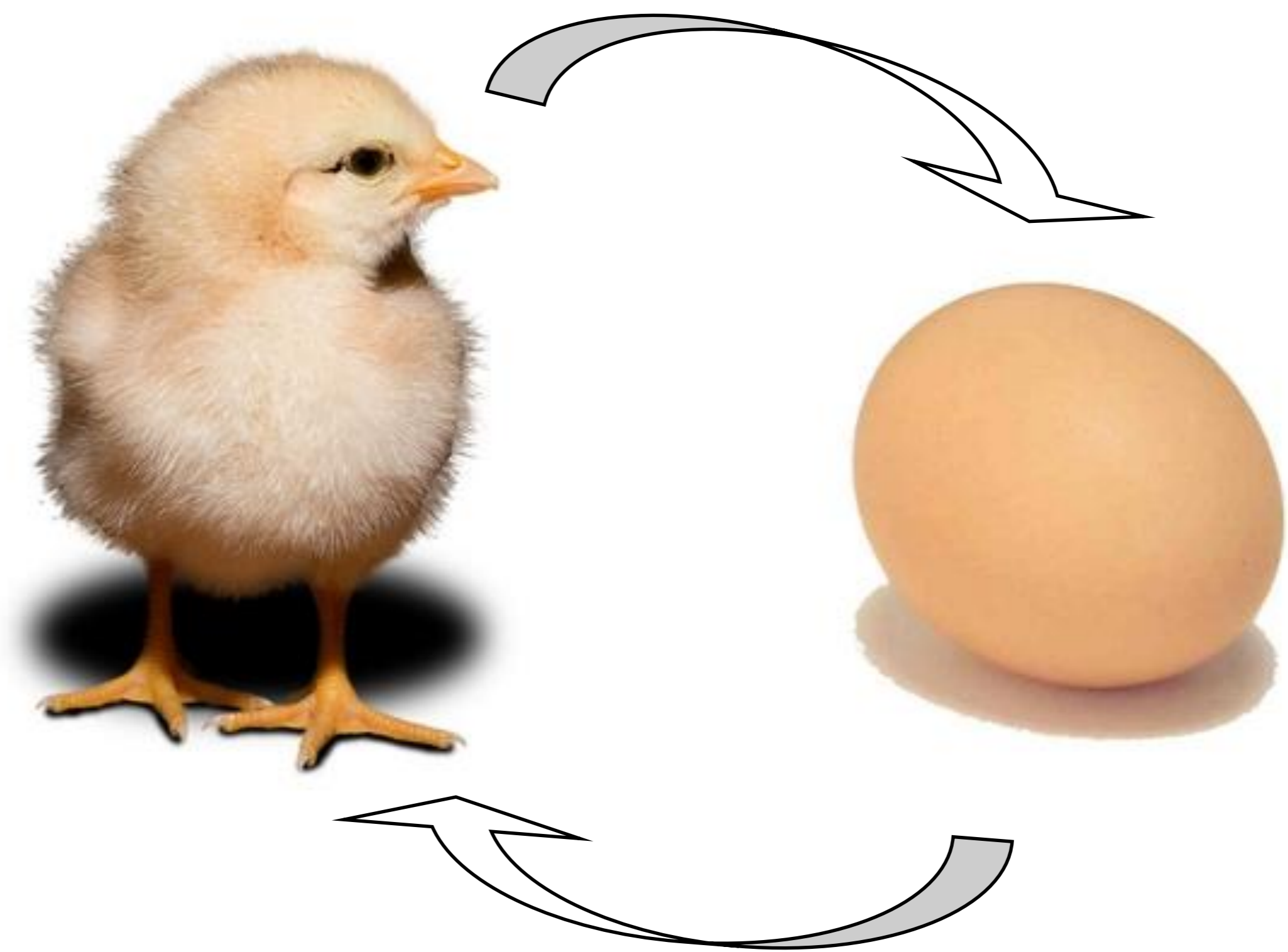


(D) Independence



Which one came first?

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Chickens, Eggs, and Causality, or Which Came First?

1930-1983 egg production and chicken population

Walter N. Thurman and Mark E. Fisher

*

Time-series evidence from the United States indicates unidirectional causality from eggs to chickens.

Key words: causality, chickens, eggs.

Granger's seminal paper entitled "Investigating Causal Relations" has spawned a vast and influential literature. In macroeconomics, for example, the causal relationship between money and income has been investigated time (Sims) and again (Barth and Bennett; Williams, Goodhart, and Gowland; Ciccolo; Feige and Pearce; Hsiao). Some authors have taken exception to Granger's definition of causality *qua* causality (Zellner; Jacobs, Leamer, and Ward; Conway et al.), and even Granger has suggested "a better term might be temporally related" (Granger and Newbold, p. 225). We find ourselves in agreement with the temporal ordering interpretation of Granger causality. In fact, we believe that the most natural application of tests for Granger causality (temporal ordering) has until now been overlooked. We refer, of course, to: "Which came first, the chicken or the egg?" Our purpose in this study is to provide an empirical answer to this venerable question, which theory alone has not resolved.

This measure excludes chickens raised only for meat. Eggs are measured in millions of dozens and include all eggs produced annually in the United States. All are potentially fertilizable.

The notion of Granger causality is simple: If lagged values of X help predict current values of Y in a forecast formed from lagged values of both X and Y , then X is said to Granger cause Y . We implement this notion by regressing eggs on lagged eggs and lagged chickens; if the coefficients on lagged chickens are significant as a group, then chickens cause eggs. A symmetric regression tests the reverse causality.¹ We perform the Granger causality tests using one to four lags. The number of lags in each equation is the same for eggs and chickens.

To conclude that one of the two "came first," we must find unidirectional causality from one to the other. In other words, we must reject the noncausality of the one to the other and at the same time fail to reject the noncausality of the other to the one. If either both cause each other or neither causes the other,

**Mark E Fisher ≠
Ronald Fisher
father of modern statistics*

*, American Journal of Agricultural
Economics (1988)*

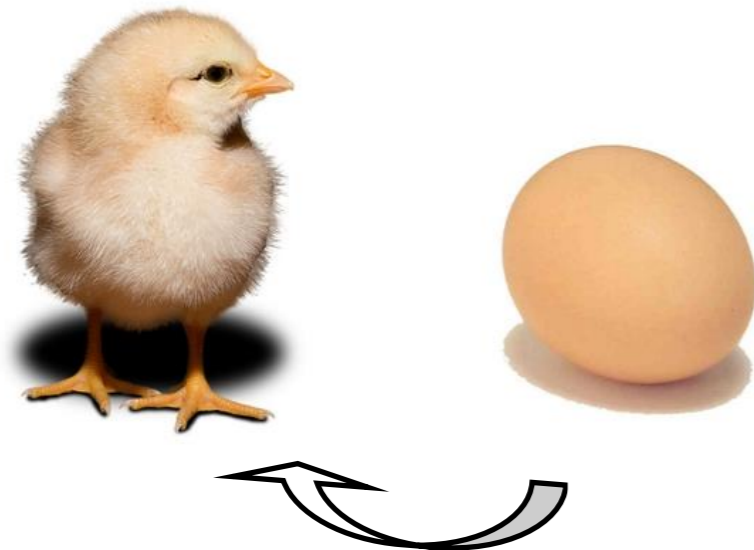


Table 1. Granger Causality Tests

Part 1: Did the Chicken Come First?

The following equation was estimated by OLS:

$$Eggs_t = \mu + \sum_{i=1}^L \alpha_i Eggs_{t-i} + \sum_{i=1}^L \beta_i Chickens_{t-i} + \epsilon_t$$

$H_0: \beta_1 = \dots = \beta_L = 0$ (chickens do not Granger cause eggs).

L = no. of lags	F-statistic	P-value	R ² of the regression
1	.04	.85	.96
2	1.71	.19	.97
3	1.10	.36	.97
4	.79	.54	.97

Part 2: Did the Egg Come First?

The following equation was estimated by OLS:

$$Chickens_t = \mu + \sum_{i=1}^L \alpha_i Chickens_{t-i} + \sum_{i=1}^L \beta_i Eggs_{t-i}$$

$H_0: \beta_1 = \dots = \beta_L = 0$ (eggs do not Granger cause chickens).

L = no. of lags	F-statistic	P-value	R ² of the regression
1	1.23	.27	.73
2	10.36	.0002	.81
3	5.85	.0019	.81
4	4.71	.0032	.82

Data source: U.S. Department of Agriculture, 1983 and others.
 Note: The data are annual, 1930–83.

We perform the Granger causality tests using one to four lags. The number of lags in each equation is the same for eggs and chickens.

To conclude that one of the two “came first,” we must find unidirectional causality from one to the other. In other words, we must reject the noncausality of the one to the other and at the same time fail to reject the noncausality of the other to the one. If either both cause each other or neither causes the other, the question will remain unanswered. The test results are presented in table 1. They indicate a clear rejection of the hypothesis that eggs do not Granger cause chickens. They provide no such rejection of the hypothesis that chickens do not Granger cause eggs. Therefore, we conclude that the egg came first.²

Which one came first?

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interpretation of Granger causality. We believe that the most natural applications for Granger causality (temporal causality) have, as until now been overlooked. We ask the question, to: "Which came first, the egg or the chicken?" Our purpose in this study is to provide an empirical answer to this question, which theory alone has not

Results

We use the annual U.S. time series from 1930 to 2000 for egg production and chicken population. The number of chickens on 1 December 1930 was 11.5 million. The number of chickens on 1 December 2000 was 21.5 million. We assume that all chickens that lay or fertilize eggs are capable of causing eggs.

We perform the Granger causality tests using one to four lags. The number of lags in each equation is the same for eggs and chickens.

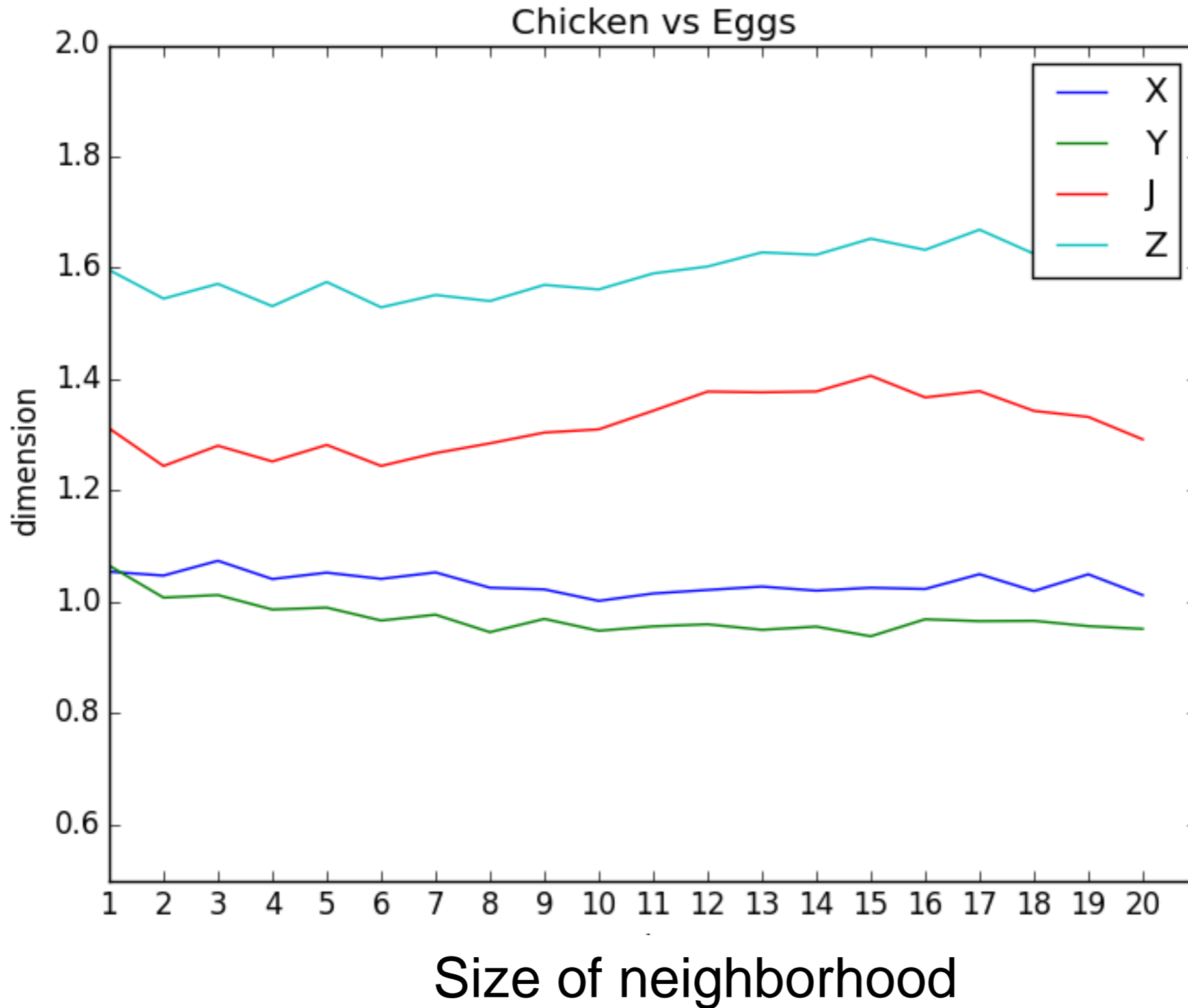
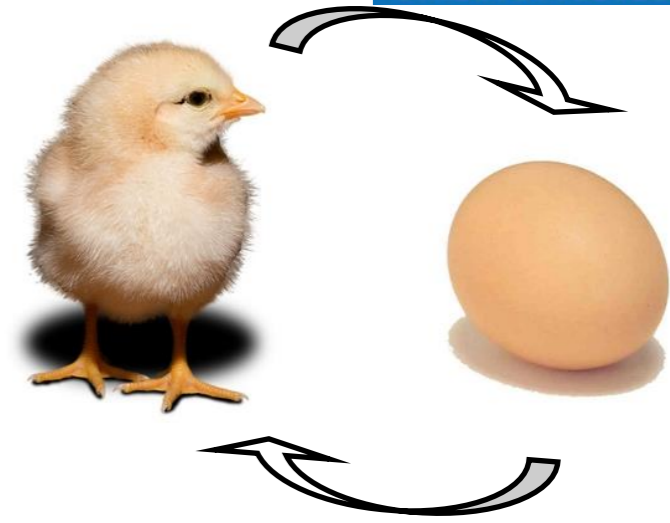
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Therefore, we conclude that eggs came first

¹ Feige and Pearce describe and distinguish among the several Granger causality tests. The validity of our test statistic requires lack of serial correlation, homoskedasticity, and normality of the

Which one came first?

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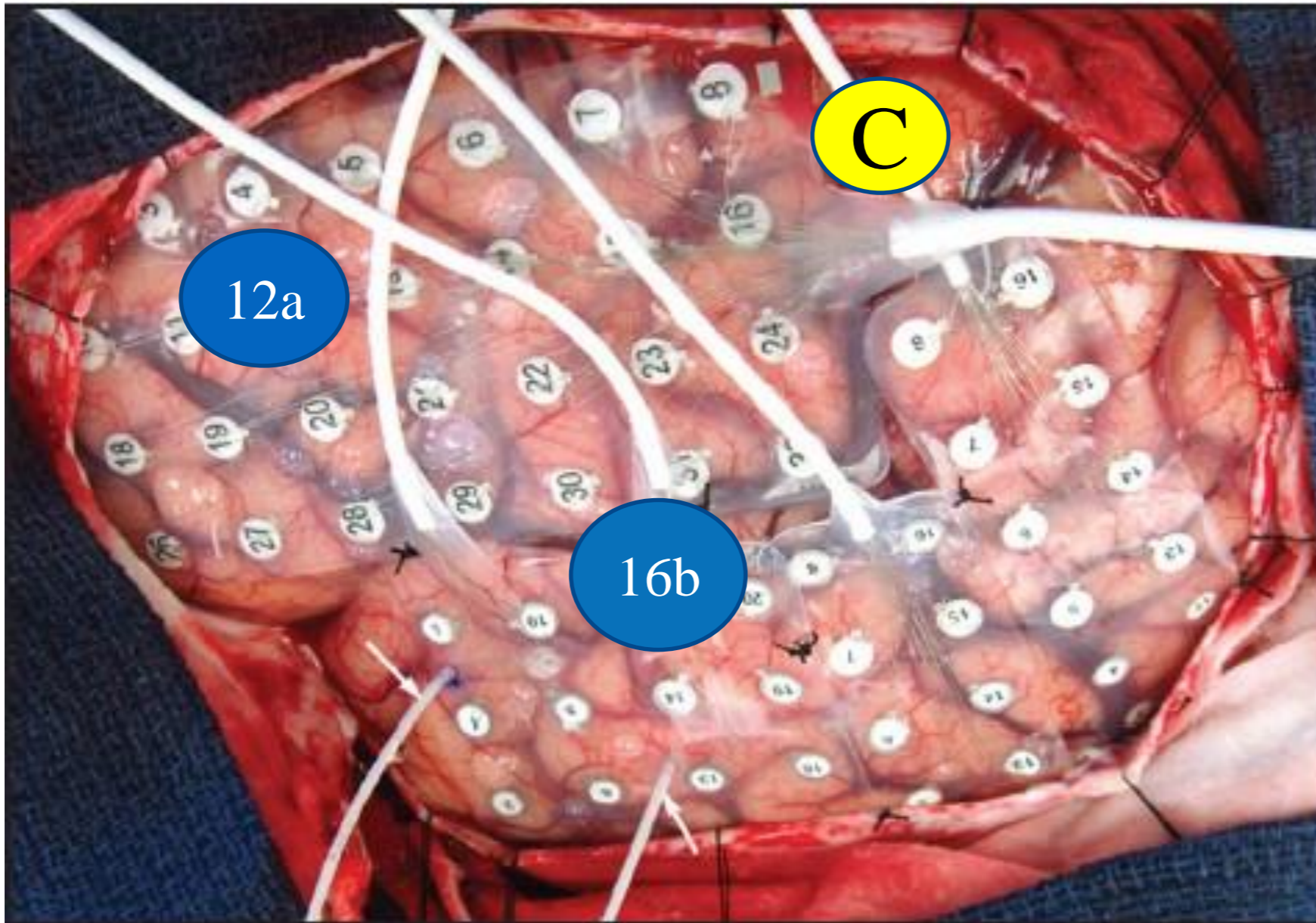


Application to epilepsy focus detection

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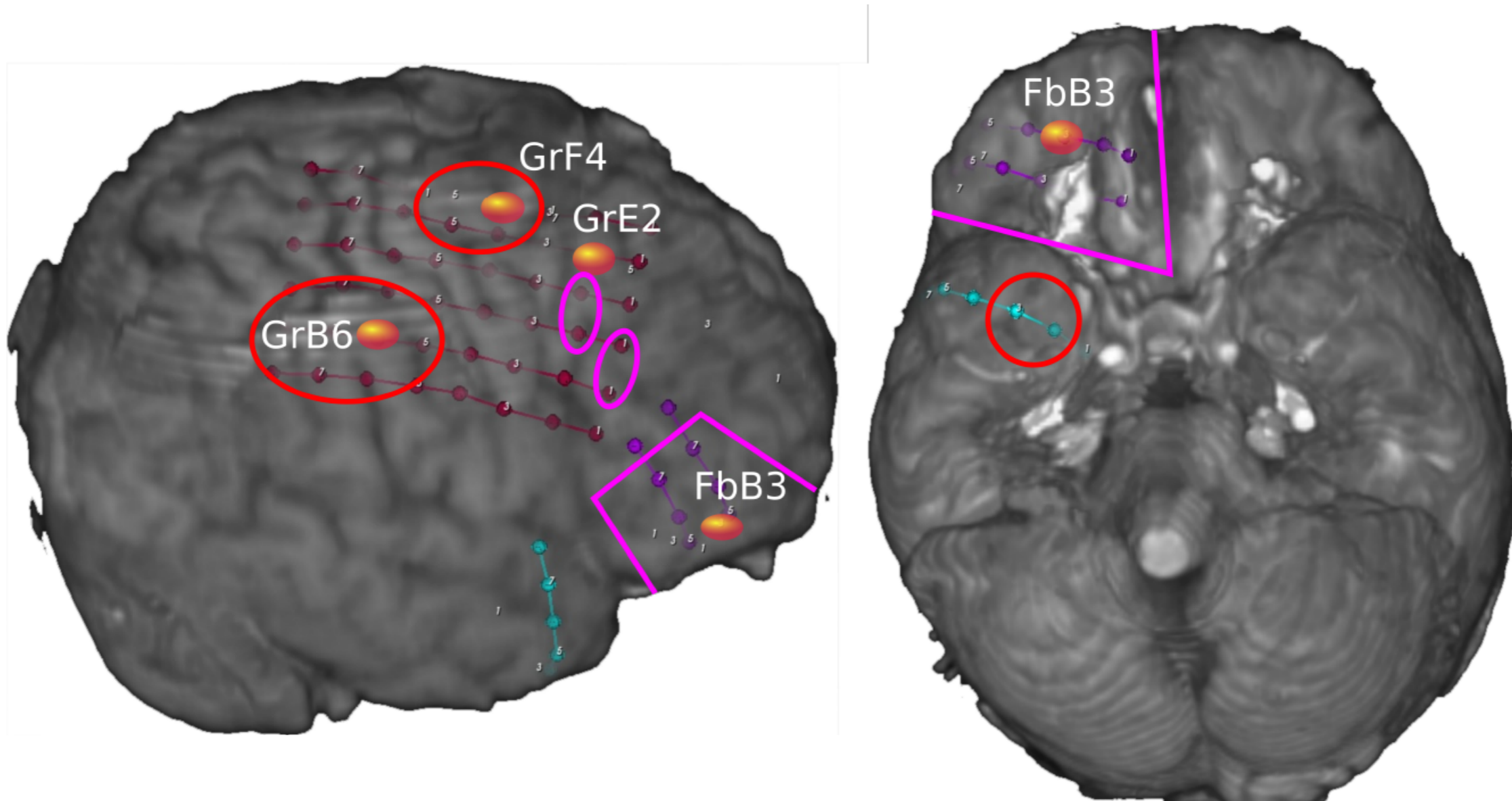


Which region is the source of the epileptic seizure?

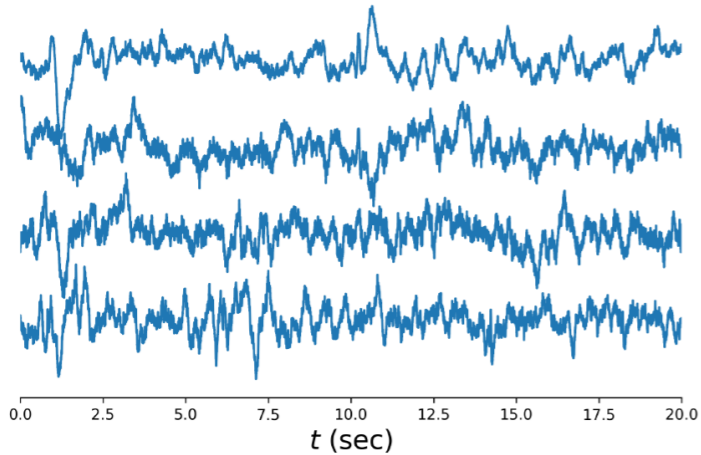


Shah XK, Mittal S. Invasive electroencephalography monitoring: Indications and presurgical planning. Xnn Indian Xcad Neurol 2014;17, Suppl S1:89-94

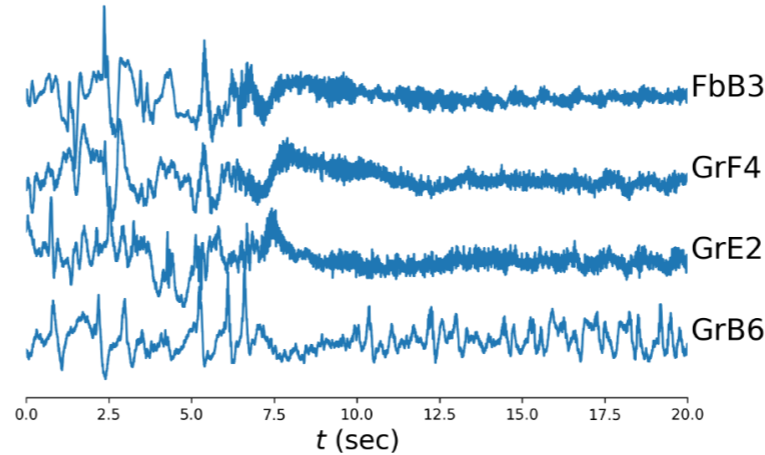
Areas to be analysed



Asymptomatic



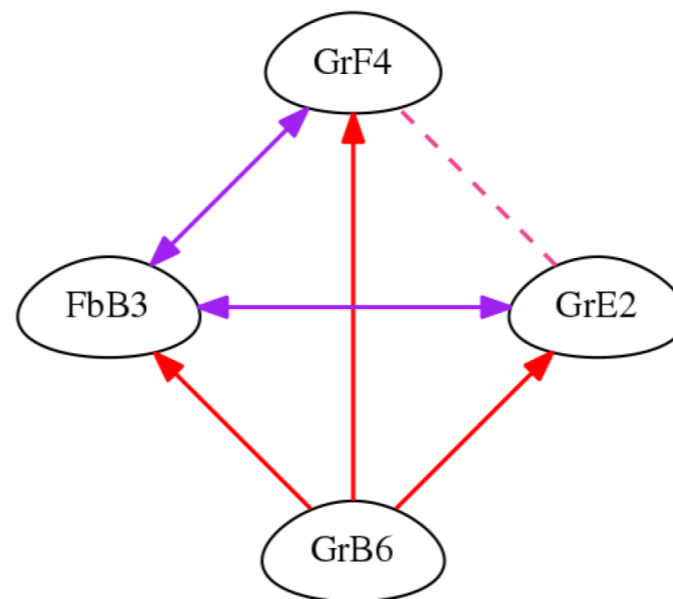
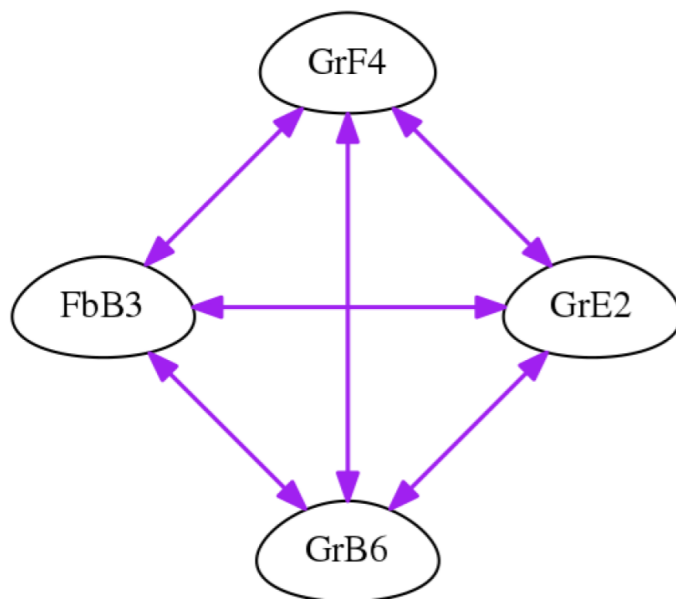
Epileptic seizure



2 – Data preprocessing
 Band-pass filtering (1-30 Hz)
 Normalization

3 – Dimension-causality analysis

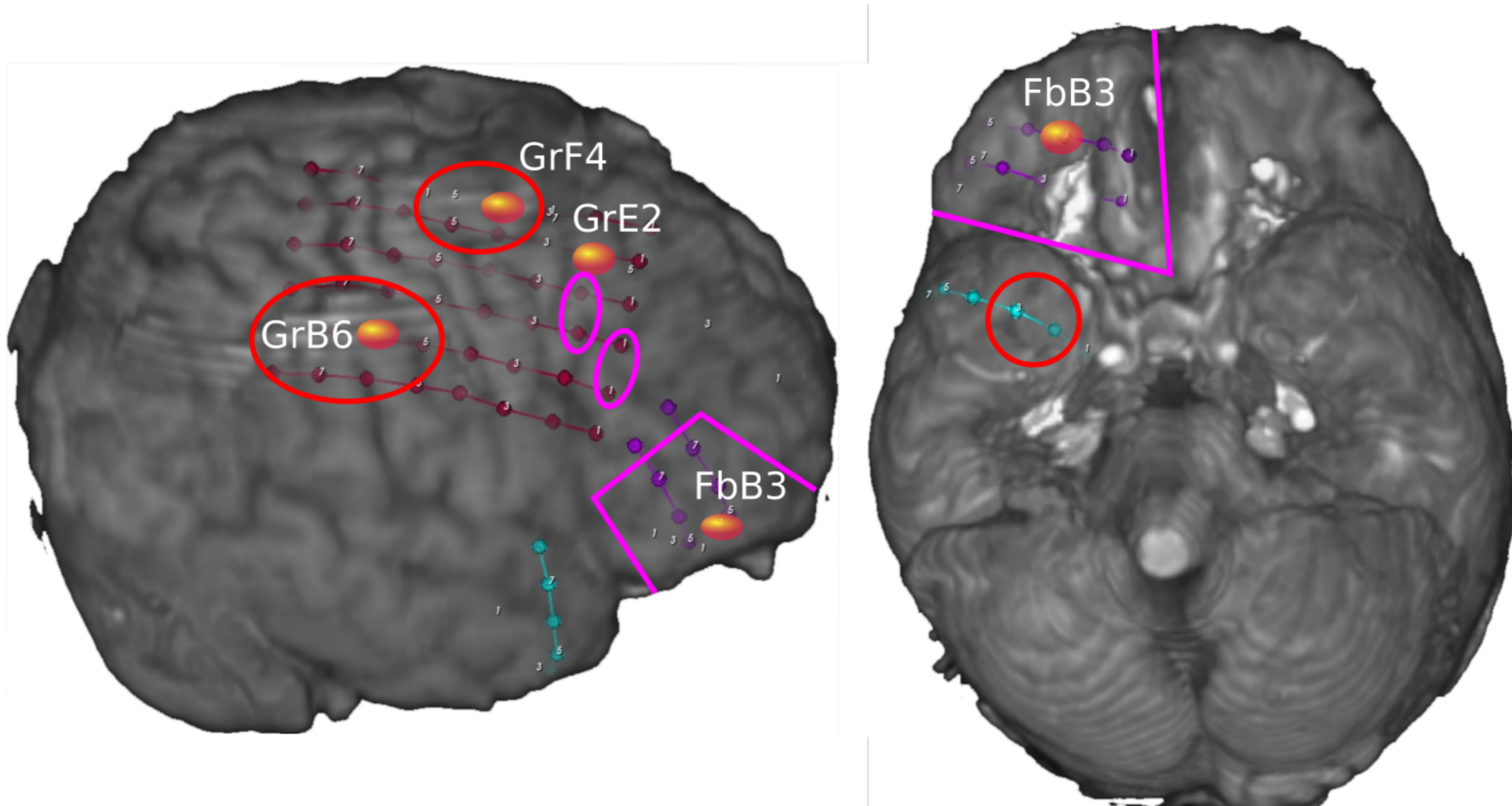
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 embedding delay: 11 step

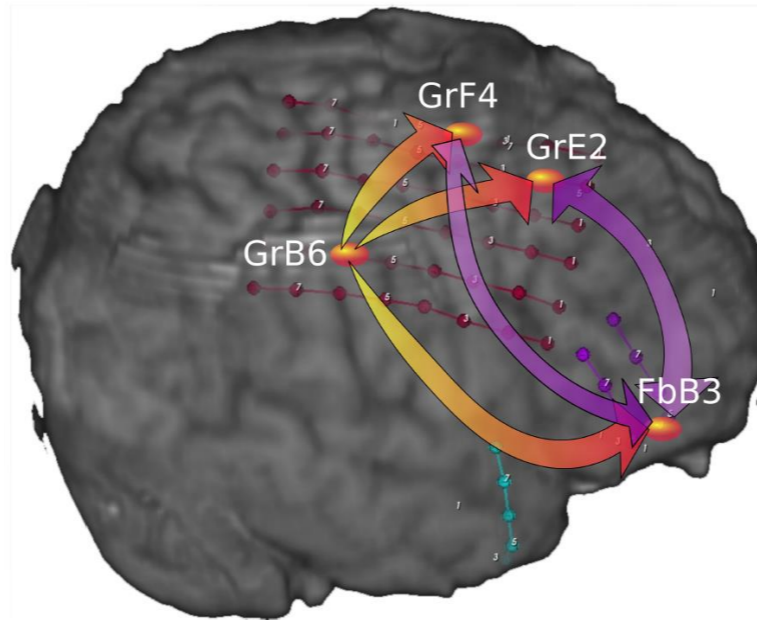
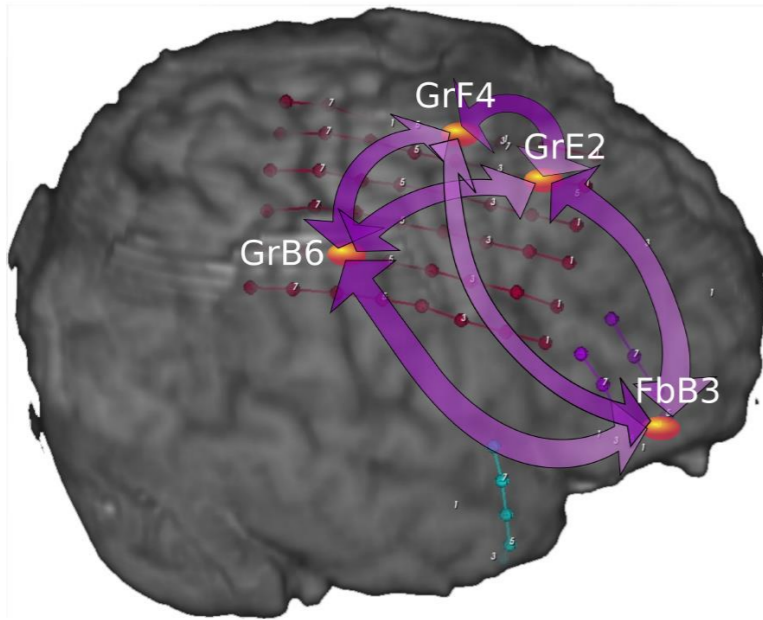
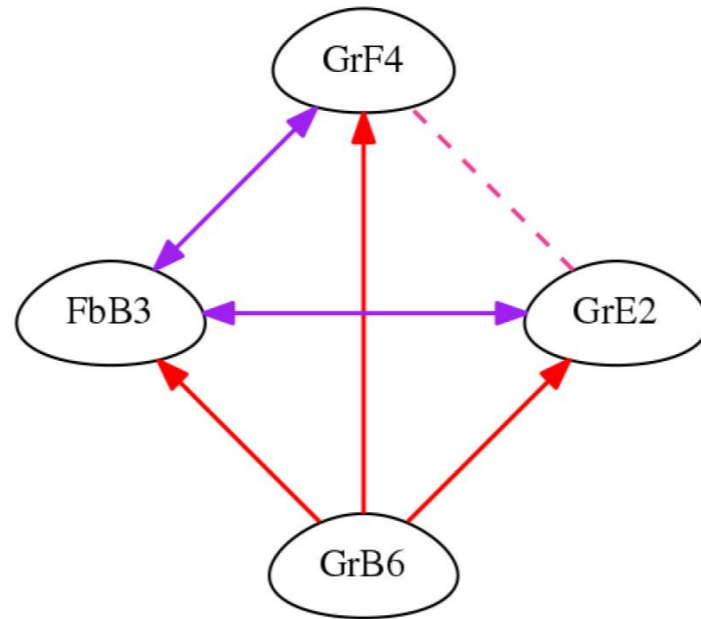
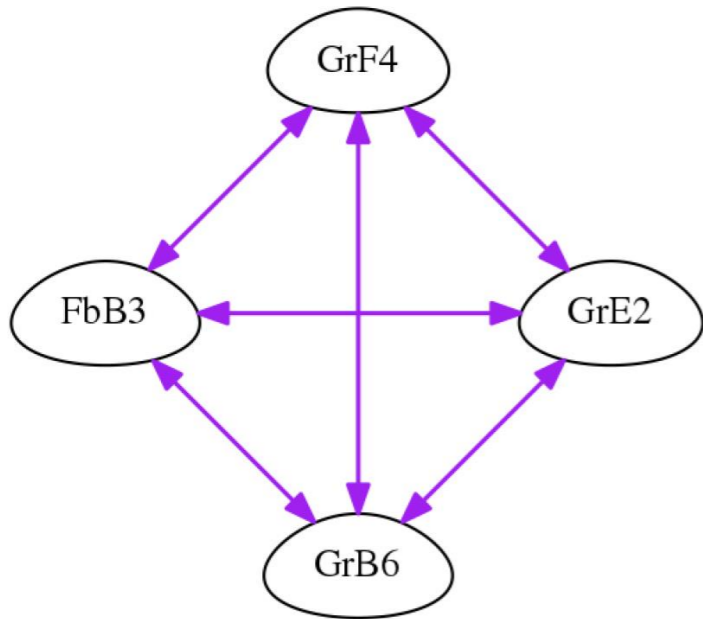


4 – Result

Our causality analysis showed that all the 4 area in question were mutually interconnected during normal, interictal activity, but the infero-temporal (GrB6) area became the dominant cause during seizure.

Magenta areas have been removed





Discussion

These results can be interpreted that, although, the resection of the large part of a highly interconnected epileptic network significantly reduced the seizure activity for a while, the untouched primary cause transformed the remained tissue towards epilepsy and the seizures were restored.

Our method

- Detects and distinguish all causality relations
- Assigns probability to causality relations



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Any suggestion for test data?



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Thanks for the attention