

Welcome to **instats**

The Session Will Begin Shortly
(At the top of the hour, Eastern USA time)

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Nonlinear Time Series Analysis, Part I: Detecting Nonlinearity

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Seminar Overview

- Day 1
 - Session 1: Introduction to Nonlinear Time Series (NTLS)
 - Session 2: Behaviors and State Spaces
- Day 2
 - Session 3: State Spaces (continued)
 - Session 4: Recurrences
- Day 3
 - Session 5: Tests
 - Session 6: Singular Spectrum Analysis and Noise
- Day 4
 - Session 7: Surrogate Data
 - **Session 8: Convergent Cross Mapping**

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Linear and Nonlinear

- Detecting Changes
 - Linear: Changepoint detection
 - Nonlinear: Singular spectrum transformation
- Multivariate Connections
 - Linear: Correlation
 - Nonlinear: Mutual Information
- Model Selection
 - Linear: AIC
- (Pseudo-) Causality
 - Linear: Granger
 - Nonlinear: Causality without correlation (and nonlinear Granger)
- Convergent Cross Mapping
 - Not separable: Multivariate connections, model selection, and (pseudo-)causal intertwined

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Changepoint Detection (Linear)

- Changepoints
 - Classically, mean, variance (skew, kurtosis, etc.)
 - As the metric changes, the change has occurred
- Can combine with linear modeling
 - Segmented regression
- Changepoints in R

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Singular Spectrum Transformation

- For each data point
 - Left (previous) and right (future) windows
 - SSA on previous and future, separately
 - Get eigenvectors from k largest eigenvalues on left; these define a (hyper)plane
 - Project the largest eigenvector on the right onto the hyperplane, and find the distance from hyperplane to eigenvector
 - Large distance indicates a break in dynamics
- Ugh
- SST in R

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Multivariable Connections (Linear)

- Correlations
 - Pick your favorite flavor: Pearson, Spearman, Kendall, etc.
- Correlation is not causality
 - Tyler Vigen's *Spurious Correlations*
- But (lagged) correlation is a necessary condition for causality
 - Cause must occur before effect
- Partial correlations
- Cross-lagged panel models
 - Must be stationary

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Correlations (Nonlinear Type 1)

- Causation without (linear) correlation
 - E.g., Pumping a swing; $\text{cor}(\text{pumping force, swing position}) = 0$
- Distance correlation
 - `energy::dcor()`
- Partial distance correlation
 - Take away the linear first
 - `ndstools::pdcor_ci()`

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Mutual Information (Nonlinear Type 2)

- What is correlation?
 - Answer: How much information about variable y is contained in variable x .
- In nonlinear systems, *mutual information* does this
 - Recall our constant delay embedding procedure; AMI and delay

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Model Selection (Linear)

- More variables (almost) always make for better R^2
 - Never, ever use R^2 to choose which model is “better”
- Balance the desire for higher R^2 against the danger of overfitting
 - Information Criteria
- Many flavors
 - Akaike (AIC), AIC with small sample correction (AICc), Bayesian IC (BIC), Deviance IC, Focused IC, Hannan-Quinn IC, and many, many others
- IC in R

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Linear Causality

- Significant correlation: Necessary
 - Not sufficient
- Granger causality
 - Well, really *Granger precedence* or *Granger predictive causality*
 - Requires separability (i.e., linearity)
 - Two conditions
 - Cause occurs before the effect
 - Cause has unique information about effect

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Nonlinear Granger

- Circular, not linear
- Vector AutoRegression Neural Network (VARNN)
 - Details: Hmamouche (2020)
 - Uses machine learning to see if there is precedence
- Transfer entropy
 - Entropy, but with delays
 - Do earlier data contain information about later data?
- Granger in R

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Nonlinear Intertwining

- Nonlinear systems lose separability
- Intertwined Metrics
 - Correlations
 - Causality
 - Model Selection
- Convergent cross mapping

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Reconstructed State Space

- Assume there is a state space, attractor, and trajectory, A
- Reconstruct delay state space, attractor, and trajectory, B , from variable v
- Reconstruct delay state space, attractor, and trajectory, C , from variable w
- v is “causal” on w iff
 - One-to-one mapping of the points of trajectory A onto B onto C (which implies vice versa)
 - Delay of C is larger than the delay of B (i.e., v changes first)

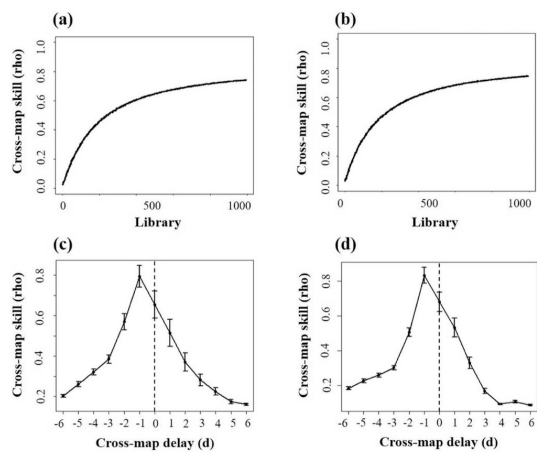
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Convergent Cross Mapping (CCM)

- CCM
 1. Constructs shadow state spaces
 2. Checks for the one-to-one mapping
 - Both directions!
 3. Checks for a delay (cause precedes effect)
- If the data fulfill those conditions, then the two variables are “causally” linked
 - Causality and correlation and model selection intertwined
- Note: correlation should improve as more points are used
 - Hence, do with varying number of data points

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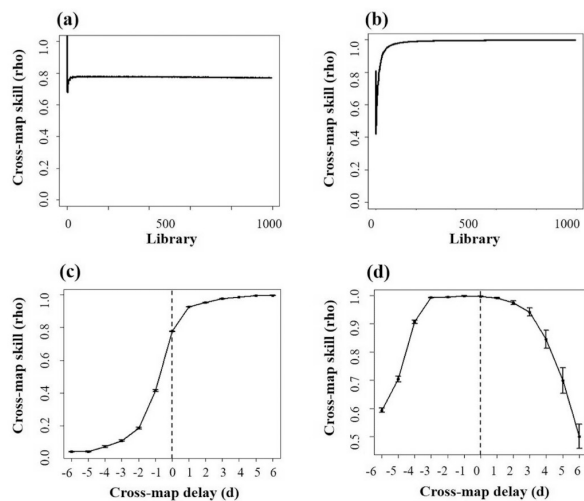
True Positive



HBR Figure 8.2

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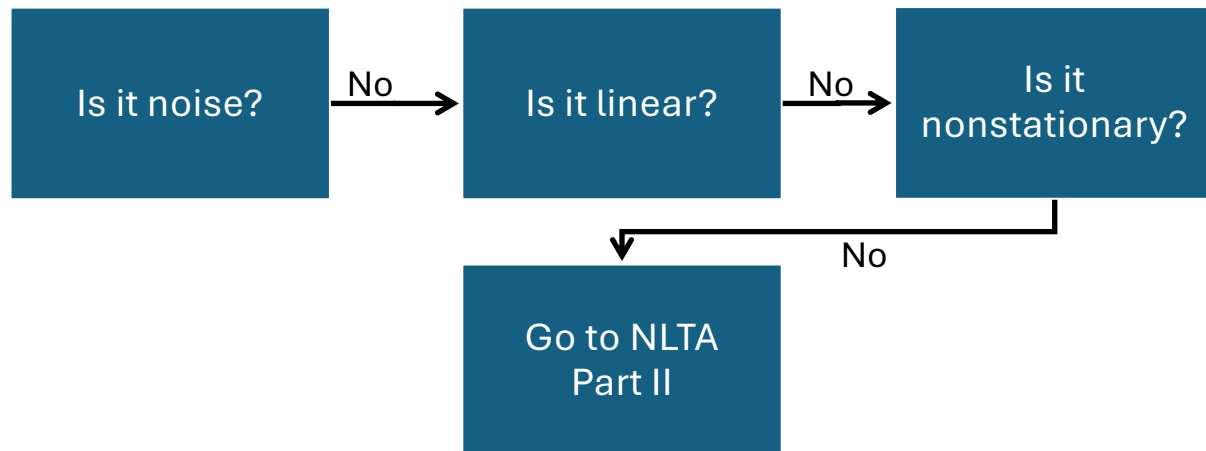
False Positive



HBR Figure 8.3

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NLTA Part 2: Phenomenological Modeling



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Forum

- Remember, the online Forum at [Instats.org](https://instats.org) is available to you for the next 30 days
- I'll be checking it
 - I may be a little slow 5 April-12 April

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Questions

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