

Nonstationarity Measure

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Outline

- Background and motivation
- Stationary vs nonstationary
- Nonstationarity measure
- Computation of Nonstationarity measure
- Conclusion

Complex system

- Complex system is “non”
 - nonlinear
 - nonstationary
 - nonequilibrium
 -

Stationary vs nonstationary

Stationary: statistical properties is invariant
strict stationary: all of the finite dimensional joint distributions are time-invariant
weak stationary: mean and covariance are time-invariant

Nonstationary:
trends (seasonal, scaling, cyclic, growth...)

Yes or No

Time and uncertainty are two most important factors when an economic agent makes a decision.

Economics and finance study the behaviors of economic agents in allocating their resources among different alternatives and over different time periods in an uncertain environment.

Stationary assumption

A and B are stationary process,
how can one say process A is
more stationary than process B?

- IID
- Zero serial dependence
- Short memory
- Long memory

A and B are nonstationary process,
how can one say process A is more
nonstationary than process B?

Challenge

In particular, for same time series or data stream, different detrend method reduce different trends, which one is more suitable?

Consider the **nonstationarity measure of residue series.**

General time series analysis

Given some observations on $\{X_t\}$, find a function $h(\cdot)$ such that $\{\varepsilon_t\}$ is an i.i.d. sequence, where

$$h(X_t, X_{t-1}, X_{t-2}, \dots) = \varepsilon_t.$$

The function $h(\cdot)$ captures the dynamic structure of $\{X_t\}$

Priestly, M.B.(1981), Time Series and Spectral Analysis.

Stable set---coarse grain

- Time series(data stream)

$$X = \{x_k, k = 1, 2, \dots\}$$

- Frequency sequence of A

$$f_n(A) = \frac{\#\{i : x_i \text{ lies in } A, i \leq n\}}{n}$$

- Stable set: Frequency sequence converges

$$f_n(A) \rightarrow f(A) \quad \text{as } n \rightarrow \infty.$$

Stable set admits reliable probability

- The phase space S that the data stream X lies is always stable.
- If the subset A is stable with respect to X , then the complementary set $S \setminus A$ is also stable with respect to X .
- Suppose two **disjoint** sets A and B are stable with respect to X , then the **union** of A and B is also stable with respect to X .

For given data stream X , we can assign the a **probability** on each **stable set** of it, and such a probability is **time-invariant**.

$$f_n(A) = \frac{\#\{i : x_i \text{ lies in } A, i \leq n\}}{n}$$

$$f_n(A) \rightarrow f(A) \quad \text{as } n \rightarrow \infty.$$

Information structure of data stream X

- **Partition** of the phase space S
- **Stable partition** of X is called a information structure of X
- Stationary process admits finest information structure----each subset of S is stable by **ergodic theorem** of stationary process
- Nonstationary process----some subsets of S may not stable

General ideas

*A data stream X is likely more **stationary** than Y if X admits **more** stable sets than Y .*

*Y is more **nonstationary** than X because Y admits **less** stable sets than X .*

More or less

mathematically?

Nonstationarity measure

- Information structure

$$IS = \begin{pmatrix} A_1 & A_2 & A_3 & \dots & A_k \\ p_1 & p_2 & p_3 & \dots & p_k \end{pmatrix}$$

- Shannon entropy

$$H(IS) = -\sum p_i \log p_i.$$

- Nonstationary measure $NS(X)$ of X is defined as the maximum Shannon entropy among all information structures of X .
- $NS(X) > NS(Y)$ implies X is more stationary than Y , or Y is more nonstationary than X .

Why and How

- Shannon entropy is widely used in many fields.....
- Shannon entropy can compare more or less in terms of finer or coarser of two partitions
- Shannon entropy captures probabilities in the information structure
-

Practical issues:

finite length of samples

- How to calculate the nonstationarity of X ?
Obtain the **finest** information structure (**stable** partition) of X
- Approximation
finest---- among a pre-assumed partition
stable---- criterion to ensure convergence

Procedures: approximate computation of nonstationarity measure

Given a data stream X in phase space S

1, Partition S into finite (small) pieces

Example: $S=[0, 1]$, partition S into 100 intervals with equal length

2, Select a criterion ----some kind of convergence rate. Example: A is stable if

$$|f_n(A) - f(A)| < C/n^r,$$

$f(A)$ is the average of last 20% of $f_n(A)$.

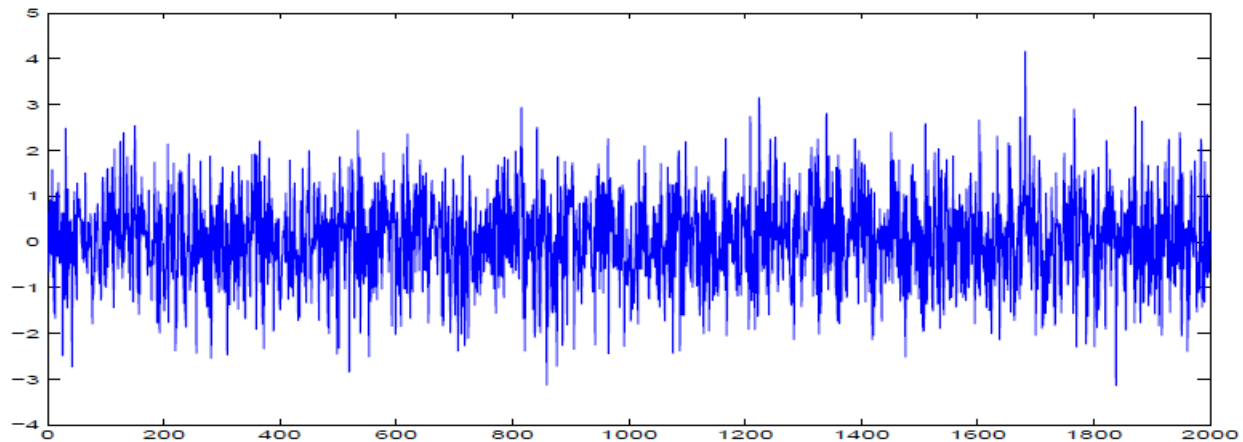
SP—finest stable partition among all partitions
coarser than the pre-assumed partition
 $H(SP)$ —good approximate of $NS(X)$

- 3, Use the criterion, check if the single piece in the pre-assumed partition is stable or not, take the stable one into the **stable partition SP**.
- 4, For the remain pieces (each one is not stable), check if the union of two pieces is stable or not, take those stable one into the **stable partition SP**.
- 5, Continue work on the remain piecese, check if the union of three pieces is stable or not.....
- 6, Stop when the weight of remain pieces is small.

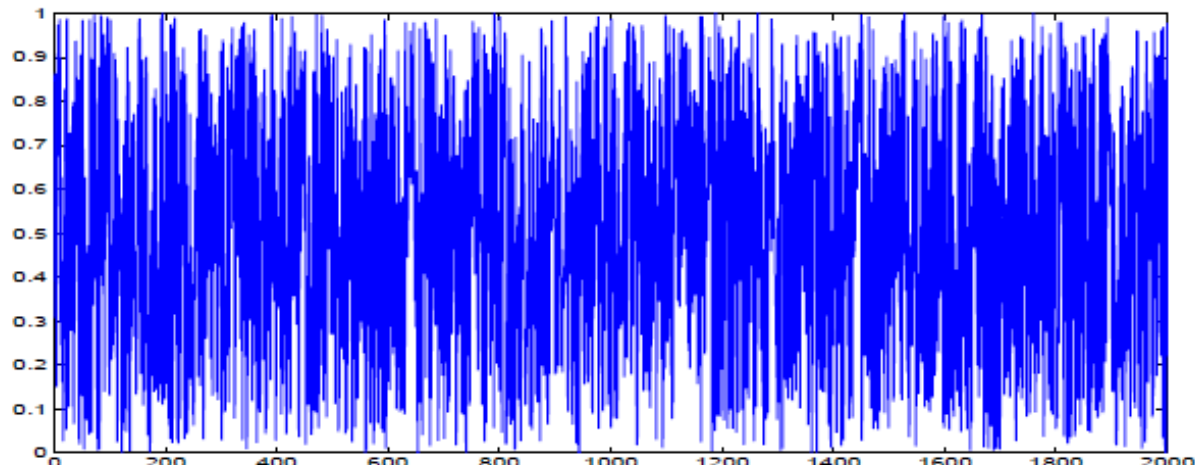
Data set

- IID $N(0, 1)$: length 2000
- IID $U(0, 1)$: length 2000
- Logistic map: length 2000 $x_t = 4x_{t-1}(1 - x_{t-1})$.
- FTSE 100: 2000.1.4----2007.12.31, length 2020
- NIKKEI 225: 2000.1.4----2007.12.28, length 1967
- S&P500: 2000.1.3----2007.12.31, length 2010
- SSE Composite index:2000.1.4----2007.12.31, length 2076.

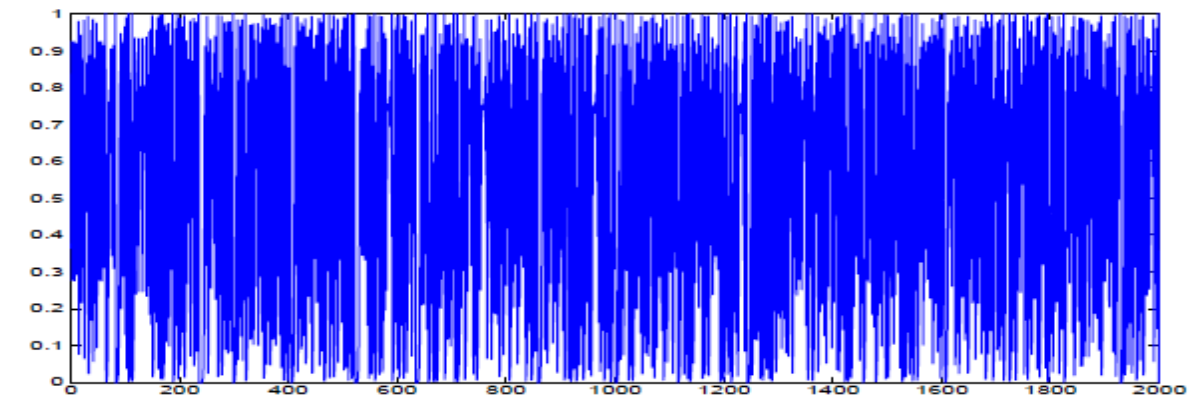
IID $N(0,1)$

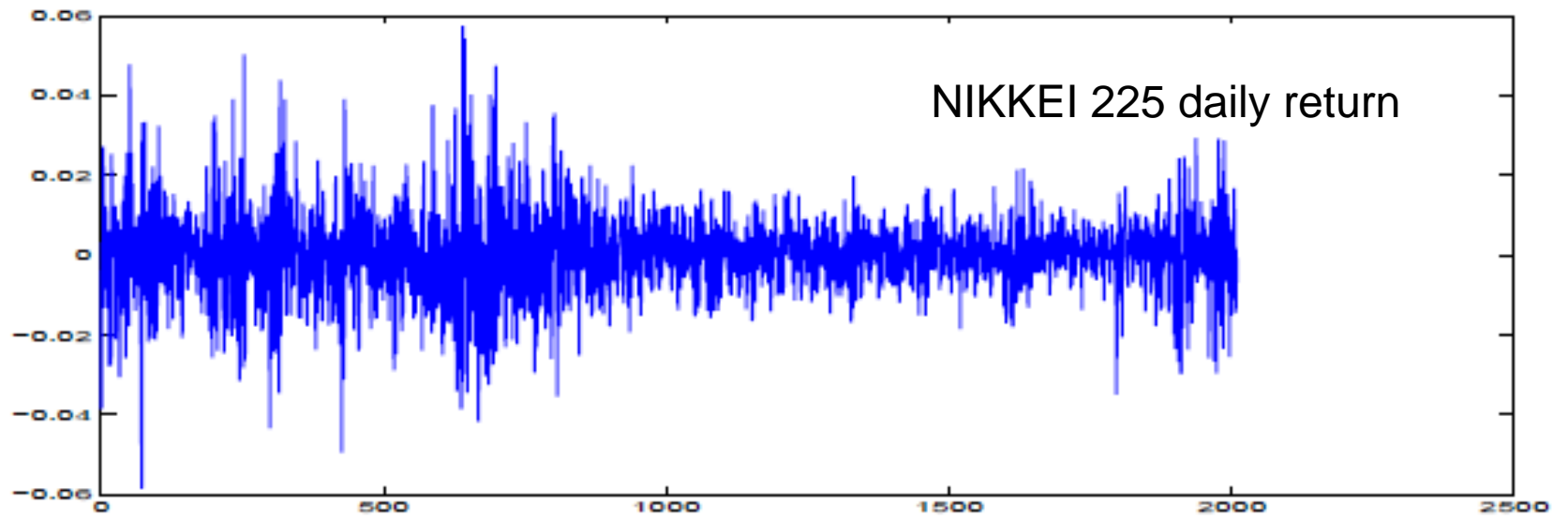
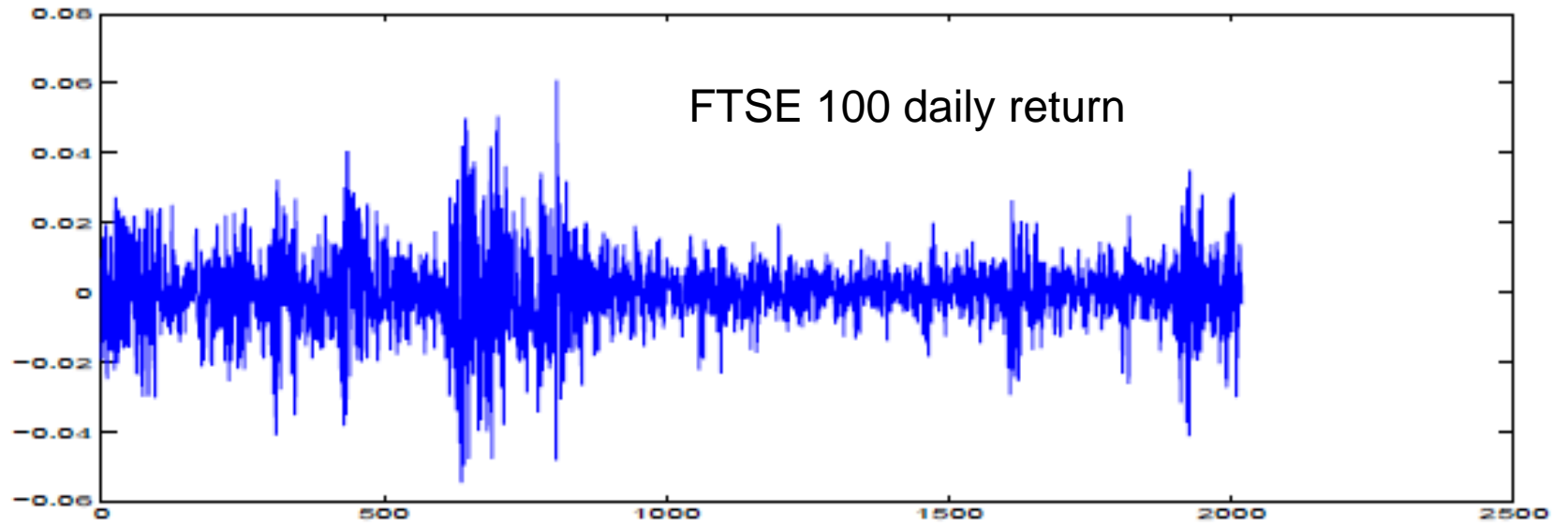


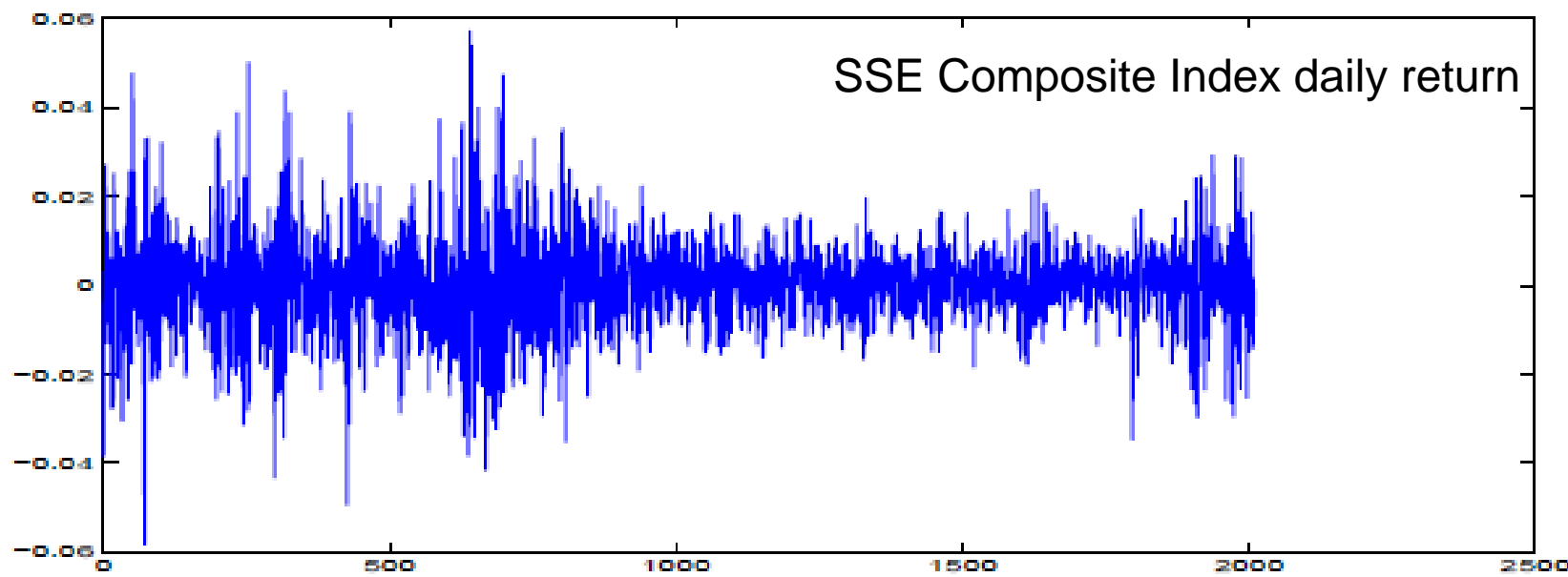
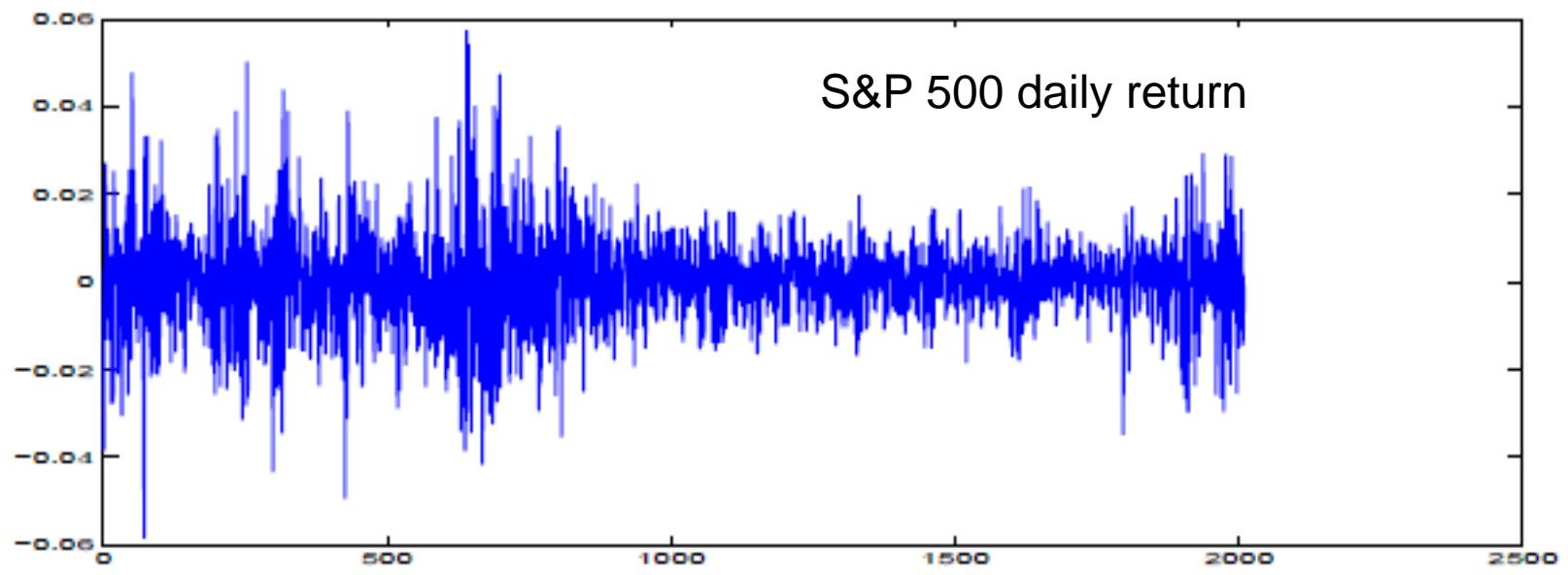
IID $U(0, 1)$



$$x_t = 4x_{t-1}(1 - x_{t-1})$$







- Partition each phase space into 50 equal pieces
- Convergence criterion:

$Y = \{y_1, y_2, \dots, y_T\}$ converges to y_T if

$$\#\{y_i \in Y : \frac{|y_i - y_T|}{T - i} \frac{T}{y_T} \leq \gamma, i = 1, 2, \dots, T\} \geq \alpha T.$$

Parameters: $\alpha = 0.9, \gamma = 0.1, 0.2, \dots, 1.0.$

90% of points in a frequency sequence lie in a linear strip controlled by gamma around the limit.

Data	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
Gaussian ^a	2.3488	3.5650	3.7768	3.8860	4.1495	4.2553	4.2737	4.4416	4.5606
Uniform ^a	0.8211	4.1421	4.5596	4.6819	4.8420	4.9251	5.0171	4.9802	4.9231
Logistic ^a	1.8721	4.0068	4.3632	4.5966	4.7423	4.7713	4.9012	4.9314	4.9911

Data	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
FTSE 100 ^a	2.1066	2.8647	2.8317	3.2178	3.4306	3.5420	3.6301	3.8069	3.7698
S&P 500 Index ^a	1.8299	2.6397	2.8180	2.9966	3.4156	3.5371	3.7178	3.8777	3.9269
NIKKEI 225 ^a	2.1662	2.8458	3.2706	3.6163	3.7211	3.8862	3.9573	3.9471	3.9452
E Composite Index ^a	1.0399	1.9708	2.6294	2.7553	2.6989	2.6630	2.8766	2.8736	2.9607

Conclusions

- It is necessary and possible to compare the (non)stationary levels of different data streams
- Use coarse grain and stability of frequency sequence to obtain stable partition (information structure)
- Use Shannon entropy of the information structure to compare the fitness of different information structures
- Effective approximate methods

Thank you

for your attention!

Comments or debates?

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