



Cautionary remarks on the correlation analysis of non-Gaussian self-similar time series

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Questions

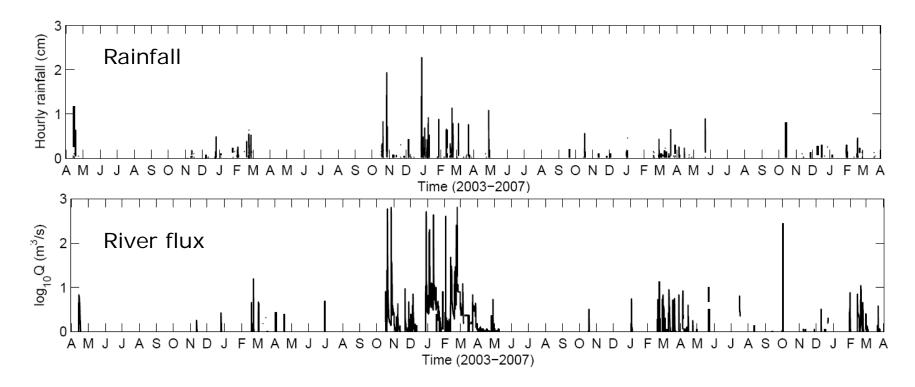
- When we describe the variability of the system or characterize the system, we may use decorrelation scales. Could we apply this approach to any types of data?
- Can we apply the correlation analysis to self-similar time/spatial series (or non-Gaussian variables)?
- Any other reasonable approaches to quantify the decorrelation scales?

Outline

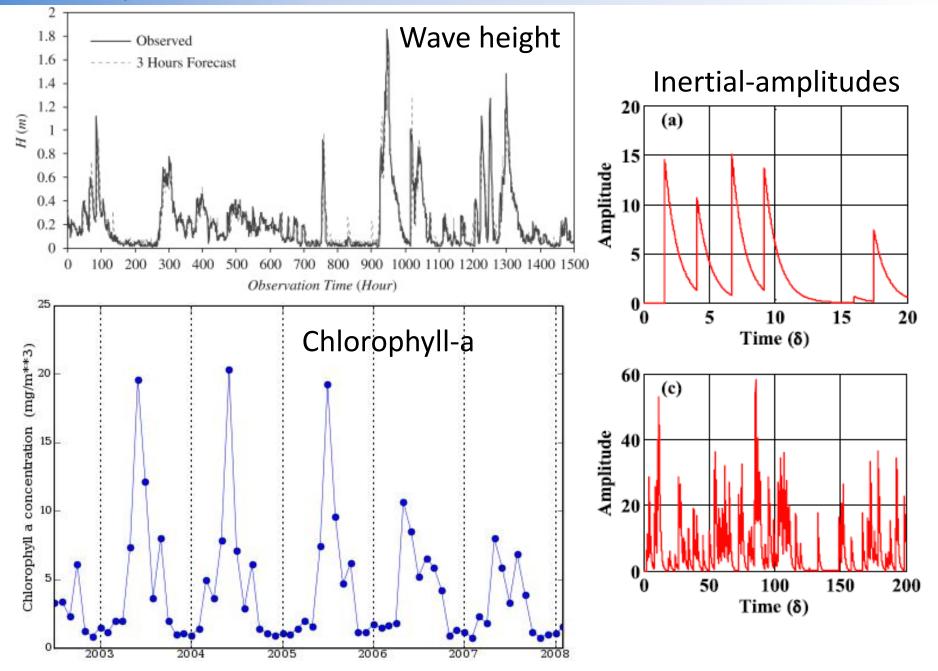
- Self-similar time series and correlation analysis
 - Examples of self-similar time series
 - Definition of correlation analysis
- Investigation with synthetic datasets
 - Generation of self-similar time series with exponential, Gaussian, and linearly decay patterns (single-sided and double-sided pulses)
 - Cross-correlation analysis to quantify the decay scales
 - Another approaches?
- Summary

Self-similar time series?

- Continuous time series with similar shapes of disturbances, pluses, or amplitudes
- May be governed by non-Gaussian statistics
 - e.g., River flows, rain fall, wind speed, wave height, concentration of Chlorophyll, and inertial amplitudes



Examples of self-similar time series



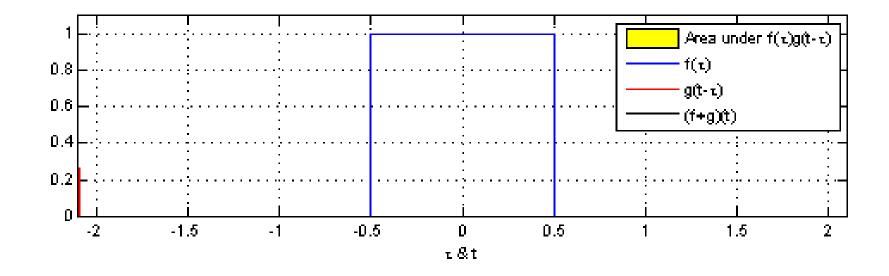
Decorrelation scales (or e-folding scales)

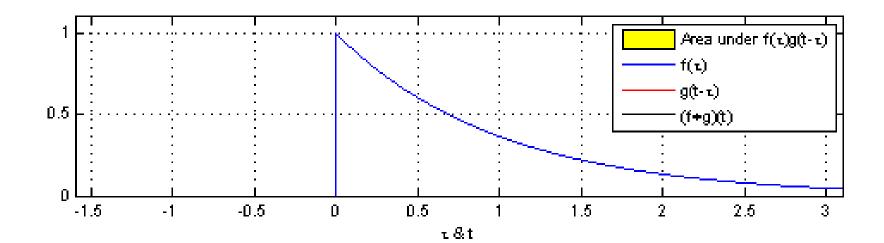
- Provide a convenient description on the structure of variables and a unique value for the system identification.
 - e.g.) decorrelation time scale is 3 days....
 E-folding length scales are 4 km and 10 km in the x- and y-directions.
- System design and analysis, sampling techniques and optimization
- How to quantify these scales?

Cross-correlations

• For the finite and evenly spaced data [d(t)]:

Cross-correlation (vs. convolution)

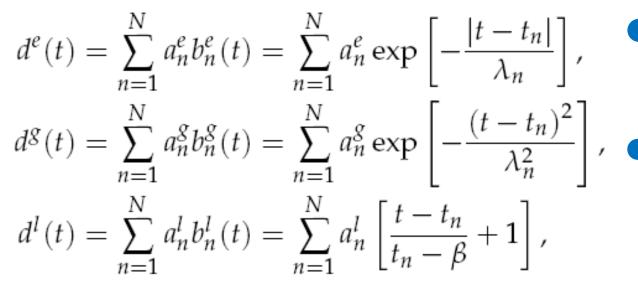




Issues of correlation analysis of self-similar time series

- However, the auto-correlation of self-similar times series produces spurious structure which is not relevant the raw time series.
- Let's take a look at what we examined with synthetic data sets:

Time series with multiple pulses

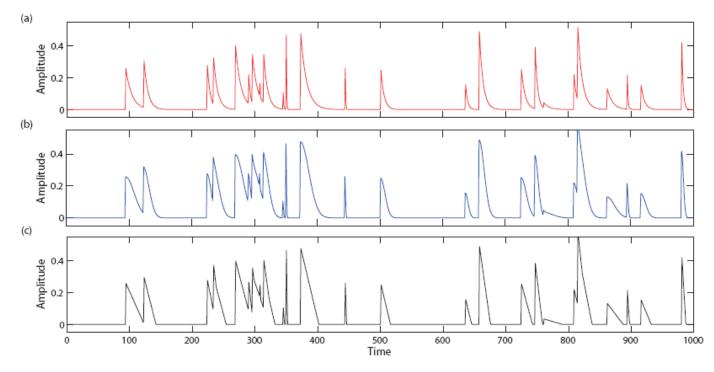


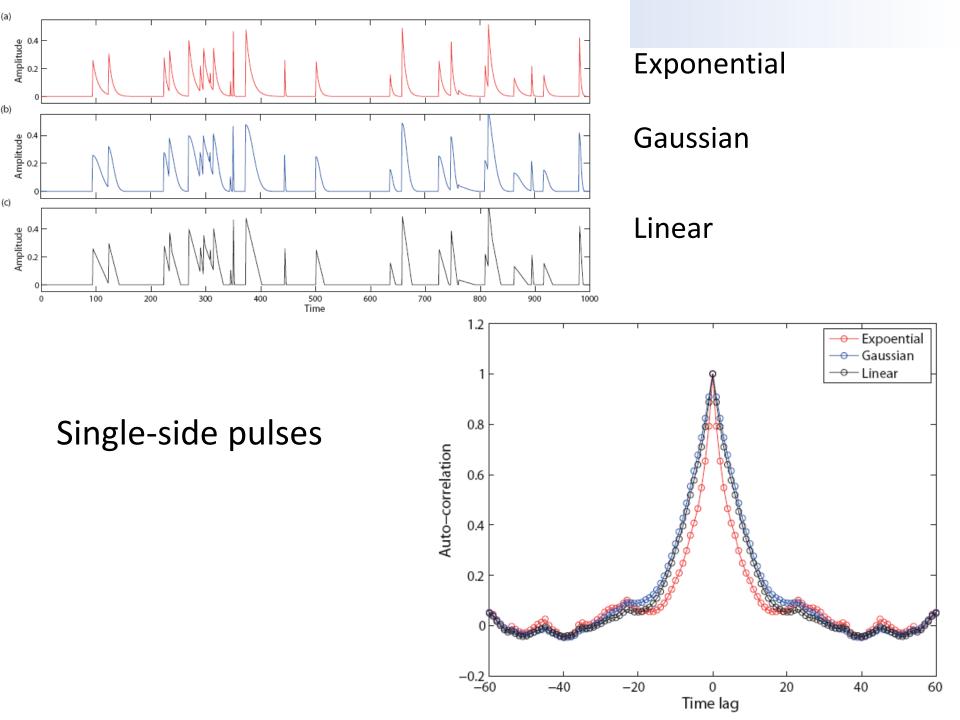
- **a**n: Amplitude of each pulse
- **b**n: Shape of decay of each pulse.

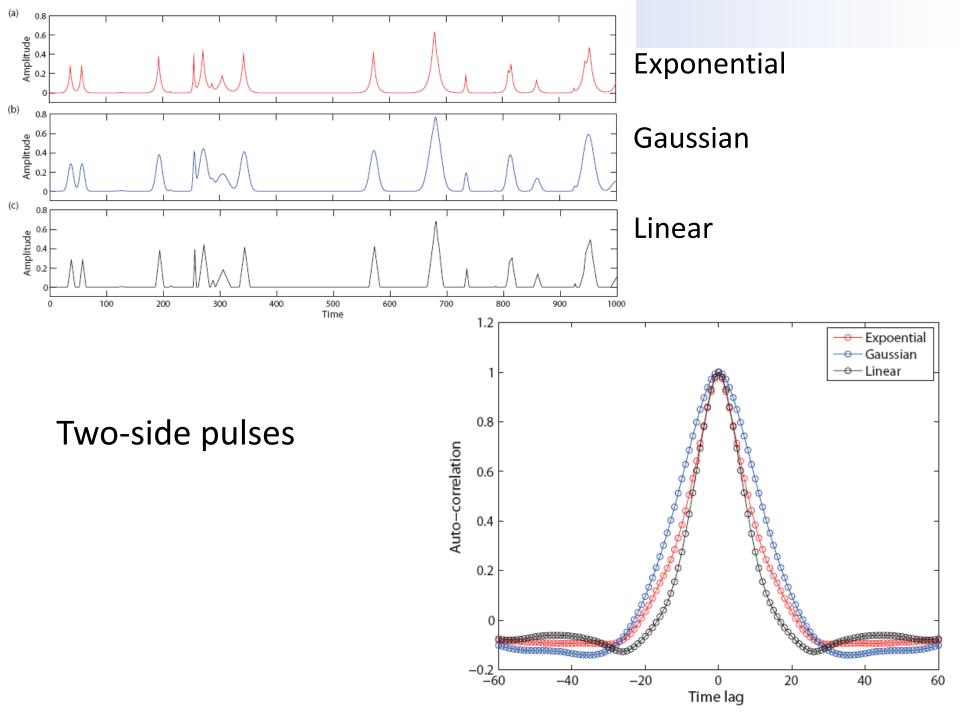




Linear







Another approach to estimate decorrelation scales?

- Direct correlation analysis can mislead the estimate of decorrelaiton scales.
- Decorrelation scales of Individual pulses can be averaged or their statistics can identify the system.
- Dynamical constraints can be used.
 - e.g., near-inertial amplitudes can be addressed with the response function estimated from observations of wind stress and currents instead of conducting the correlation analysis on the (self-similar) time series of amplitudes of near-inertial currents.

Summary

- Cross-correlation analysis of self-similar time series may generate the spurious results in the estimate of decorrelation scales because convolving two time series does not guarantee their shape based on the original datasets.
- Using cross-correlation analysis the time lag having a maximum correlation can be found.
- Dynamical data analysis or composite mean of individual pulses in the self-similar time series can be used to identify the system or quantify the decorrelation scales.