

Time Series Wrap-up

04/26/2022

- Is there variability in our data over some timescale?
 - if not, then nothing interesting is going on.
 - if so, then we need ways to search for and quantify that variability.

- What model do we use for the data?

→ TEMPLATE: if we know the physics of the signal, then we fit the parameters as usual.

→ FOURIER ANALYSIS: still a model, but more agnostic to the physics.

MODEL $y_i(t_i) = Y_0 + \sum_{m=1}^M \beta_m \sin(M\omega t_i + \phi_m) + \varepsilon_i$

constant reconstruct my periodic function. measurement noise

→ β_m values may be linked to physics or just agnostically describe variability.

... great, but how do we implement a Fourier analysis?

(i) Fourier transform \rightarrow easy for evenly sampled data.

(ii) Lomb-Scargle Periodogram

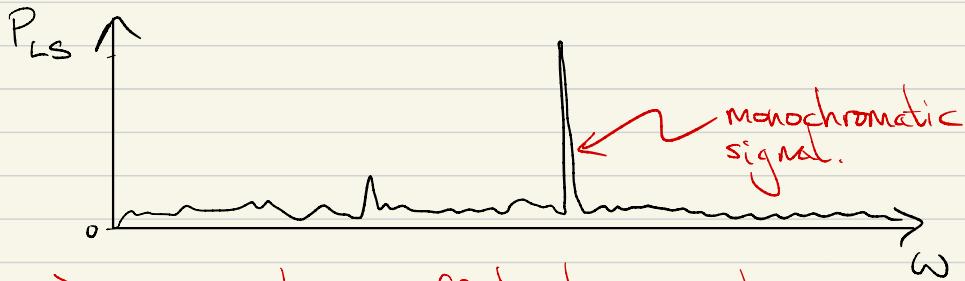
power at different frequencies

unevenly sampled data

heteroscedastic uncertainties.

$$P_{LS}(\omega) = \frac{1}{2} \left[\frac{\left(\sum_{j=1}^N y_j \sin[\omega(t_j - \bar{t})] \right)^2}{\sum_{j=1}^N \sin^2[\omega t_j]} + \frac{\left(\sum_{j=1}^N y_j \cos[\omega(t_j - \bar{t})] \right)^2}{\sum_{j=1}^N \cos^2[\omega t_j]} \right]$$

where $\bar{t} = \frac{1}{2\omega} \tan^{-1} \left(\frac{\sum \sin[2\omega t_j]}{\sum \cos[2\omega t_j]} \right)$ ensures time-shift invariance



\rightsquigarrow LS periodogram effectively computes a weighted inner product of data against sines and cosines ... no need to be regular.

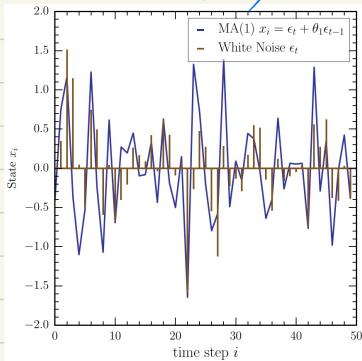
better for very
poorly sampled data

random process

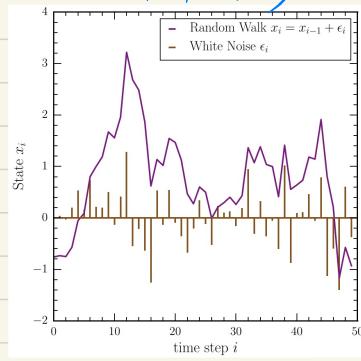
varies over multiple
timescales!

(iii) (c) ARMA Models

MA(1)



AR(1)



short timescale structure + long timescale structure

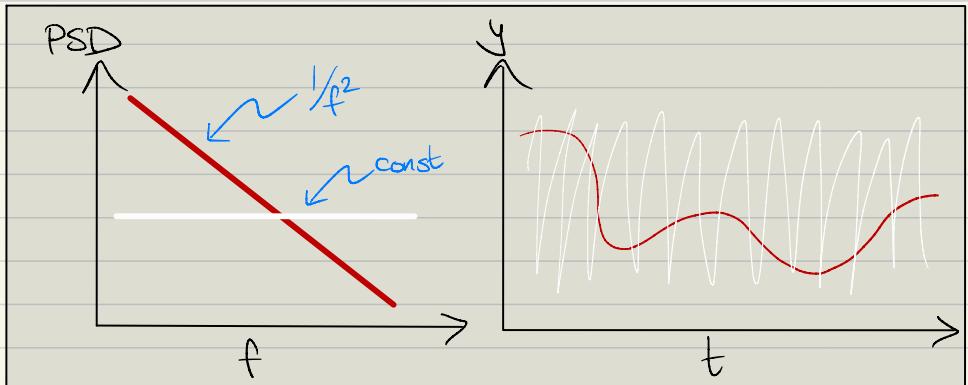
$$y_i = \varepsilon_i + \sum_{j=1}^p a_j y_{i-j} + \sum_{j=1}^q b_j \varepsilon_{i-j}$$

ARMA(p, q) Process

ARMA assumes evenly sampled data.

Continuous ARMA (CARMA) extends this to unevenly sampled data.

⇒ defined by a stochastic differential equation



AUTOCORRELATION
FUNCTION
→ (TIME)

$$ACF(\Delta t) = \frac{\lim_{T \rightarrow \infty} \int_T g(t)g(t+\Delta t) dt}{\sigma_g^2}$$

AUTOCOVARIANCE
FUNCTION
→ (TIME)

$$ACVF(\Delta t) = \sigma_g^2 \times ACF(\Delta t)$$

POWER SPECTRAL
DENSITY
⇒ (FREQUENCY)

$$PSD(f) = \int_{-\infty}^{\infty} ACVF(\Delta t) \times e^{-2\pi i f \Delta t} d(\Delta t)$$

Wiener - Khinchin Theorem

