

# Applied Time Series Analysis

## FS 2011 – Week 12

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# Applied Time Series Analysis

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### *Spectral Analysis*

**Idea:** Time series are interpreted as a combination of cyclic components, and thus, a linear combination of harmonic oscillations.

**Why:** As a descriptive means, showing the character and the dependency structure within the series.

**What:** It is in spirit, but also mathematically, closely related to the correlogram

**Where:**

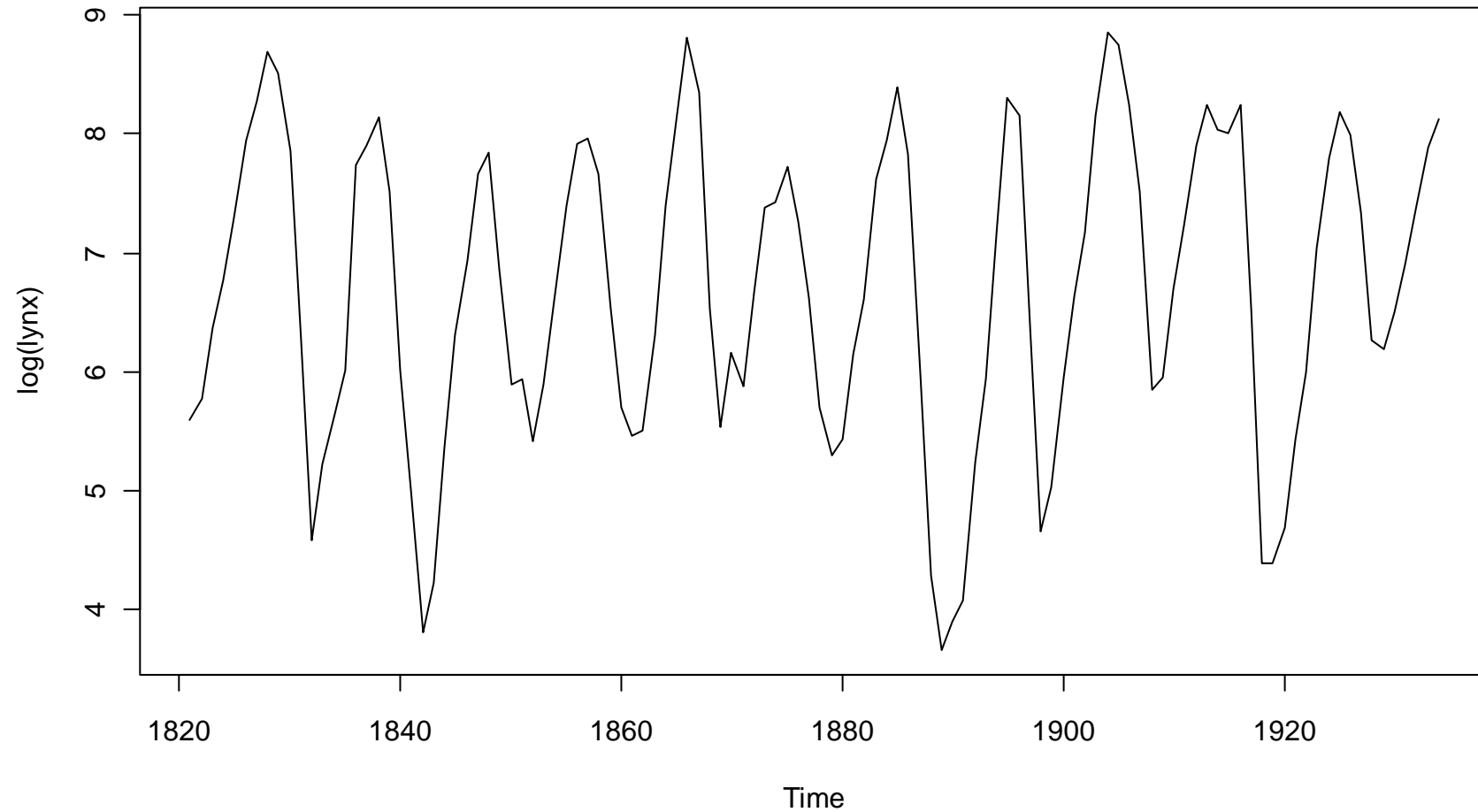
- engineering
- economics
- biology/medicine

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### *Lynx Data*

Log Lynx Data

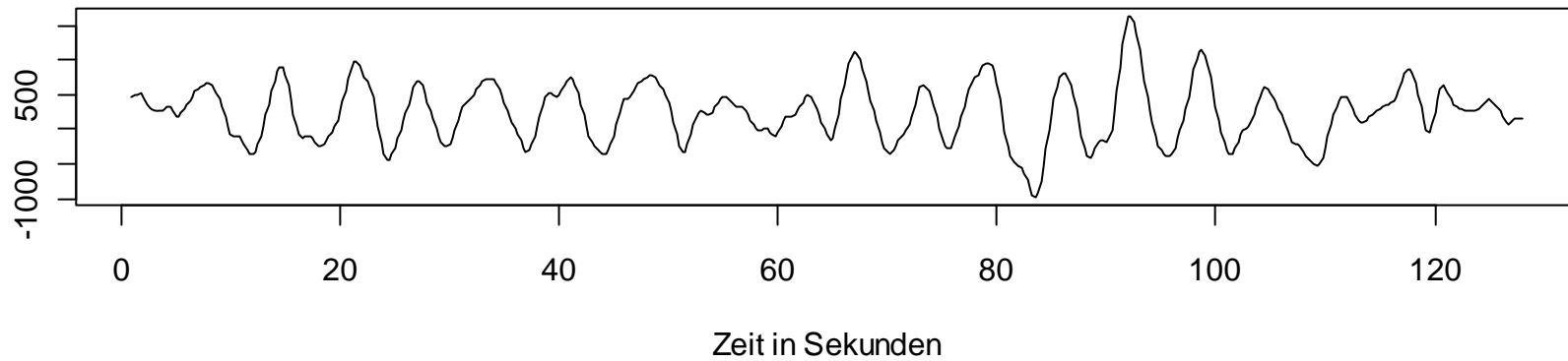


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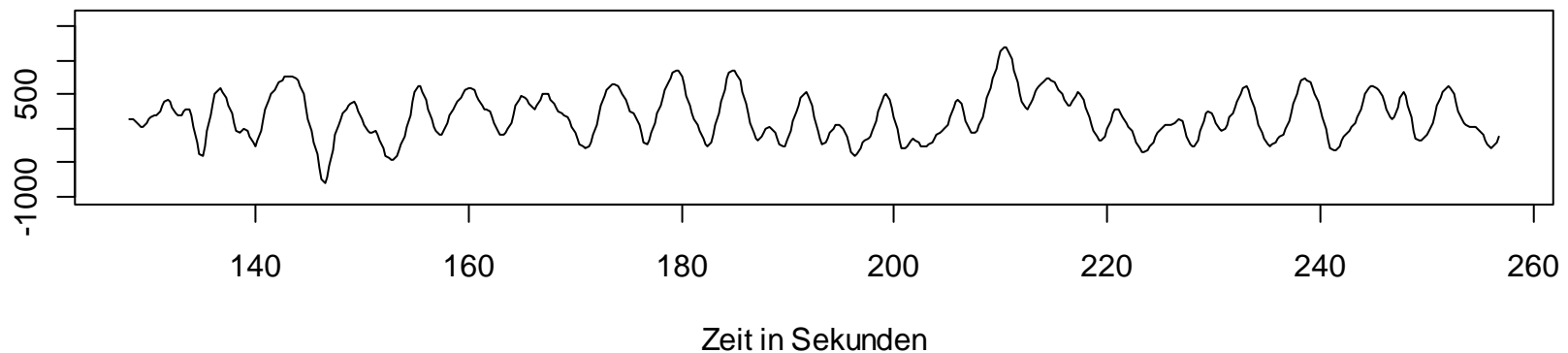
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### *Ocean Wave Data*

**Ocean Wave Height Data, Part 1**



**Ocean Wave Height Data, Part 2**

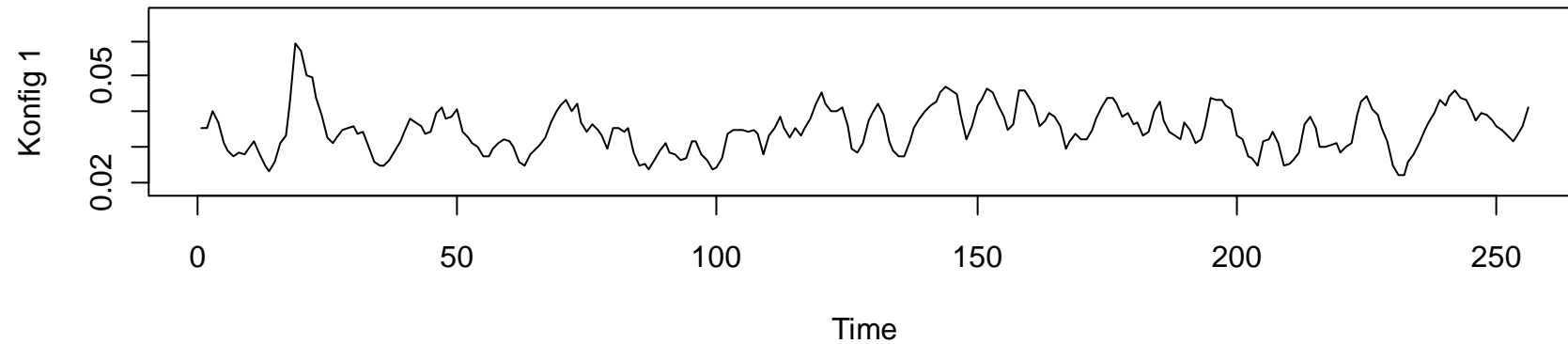


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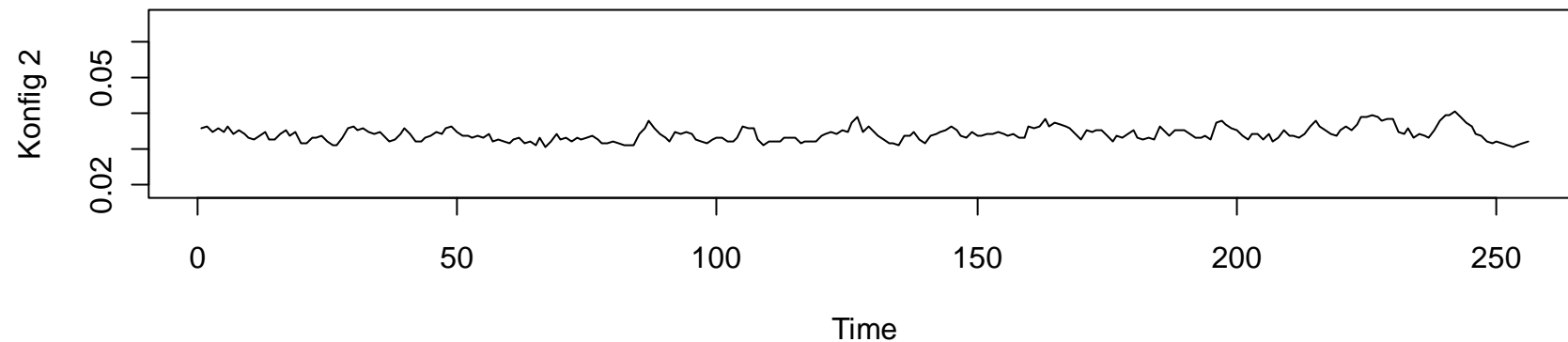
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### *2-Component-Mixture Data*

**2-Component-Mixture: Series 1**



**2-Component-Mixture: Series 2**



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### *Harmonic Oscillations*

The most simple periodic functions are sine and cosine, which we will use as the basis of our analysis.

A harmonic oscillation has the following form:

$$y(t) = \alpha \cos(2\pi\nu t) + \beta \sin(2\pi\nu t)$$

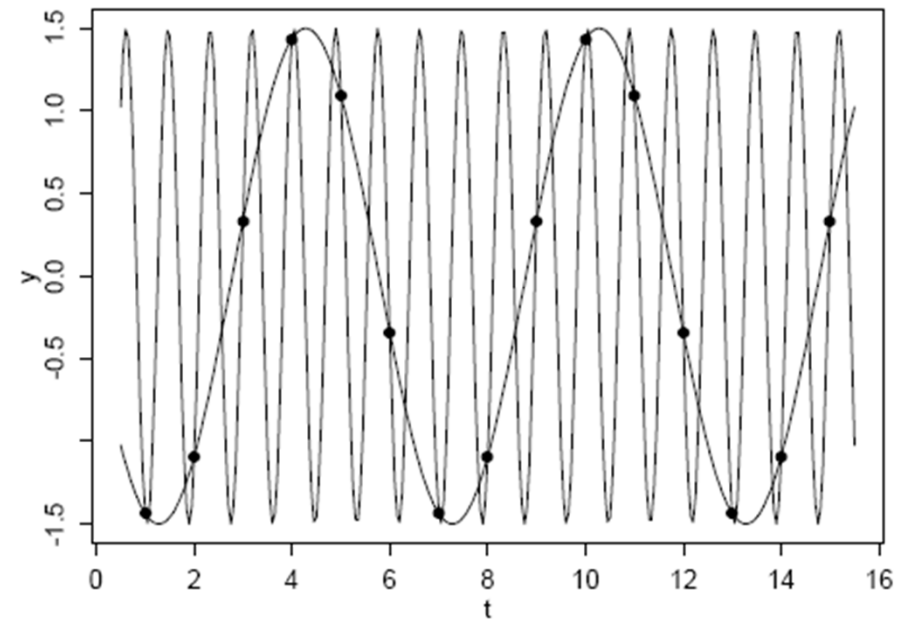
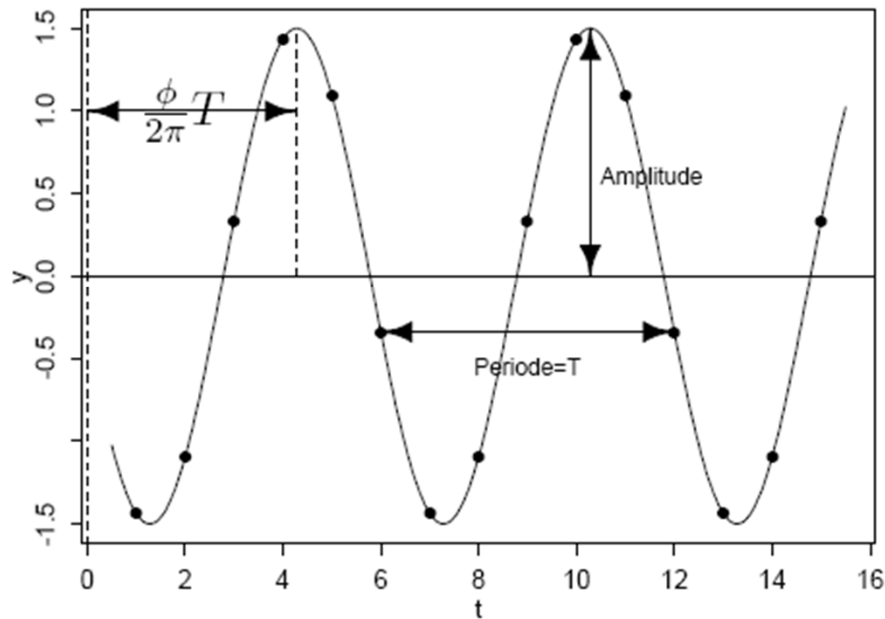
For the derivation, [see the blackboard...](#)

- In discrete time, we have aliasing, i.e. some frequencies cannot be distinguished ( $\rightarrow$  see next slide).
- The periodic analysis is limited to frequencies between 0 and 0.5, i.e. things we observe at least twice.

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### *Aliasing*



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### *Regression Model & Periodogram*

We try to write a time series with a regression equation containing sine and cosine terms at the fourier frequencies.

→ **see the blackboard**

The most important frequencies within the series, which when omitted, lead to pronounced increase in goodness-of-fit.

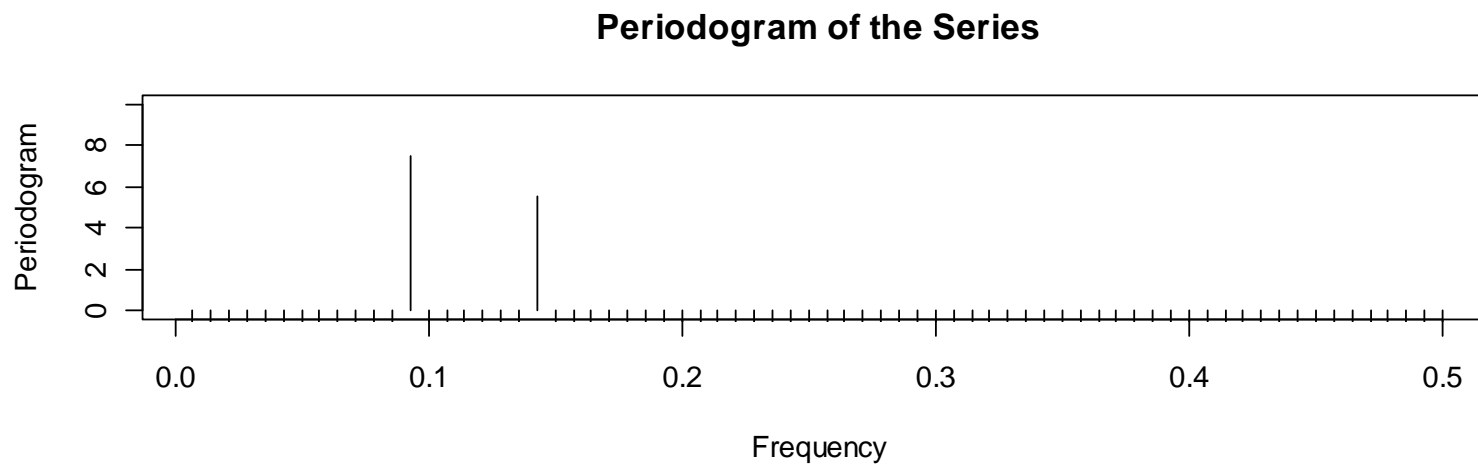
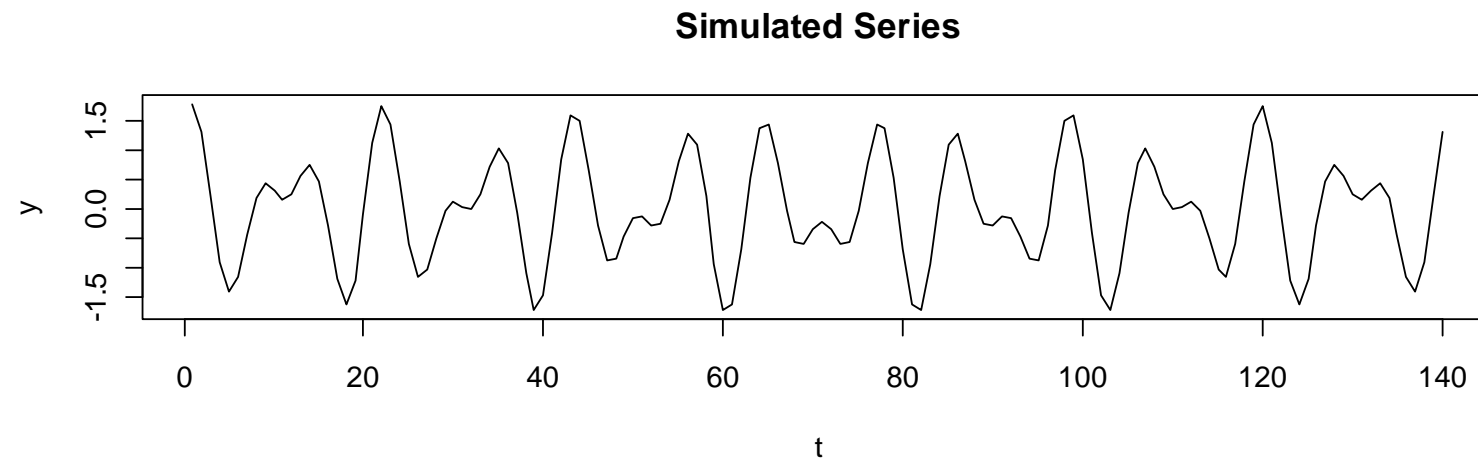
- This idea is used as a proxy for the periodogram,  
→ **see the blackboard...**
- However, if the „true“ frequency is not a fourier frequency, we have leakage (→ see next 2 slides).



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### *Periodogram of a Simulated Series*

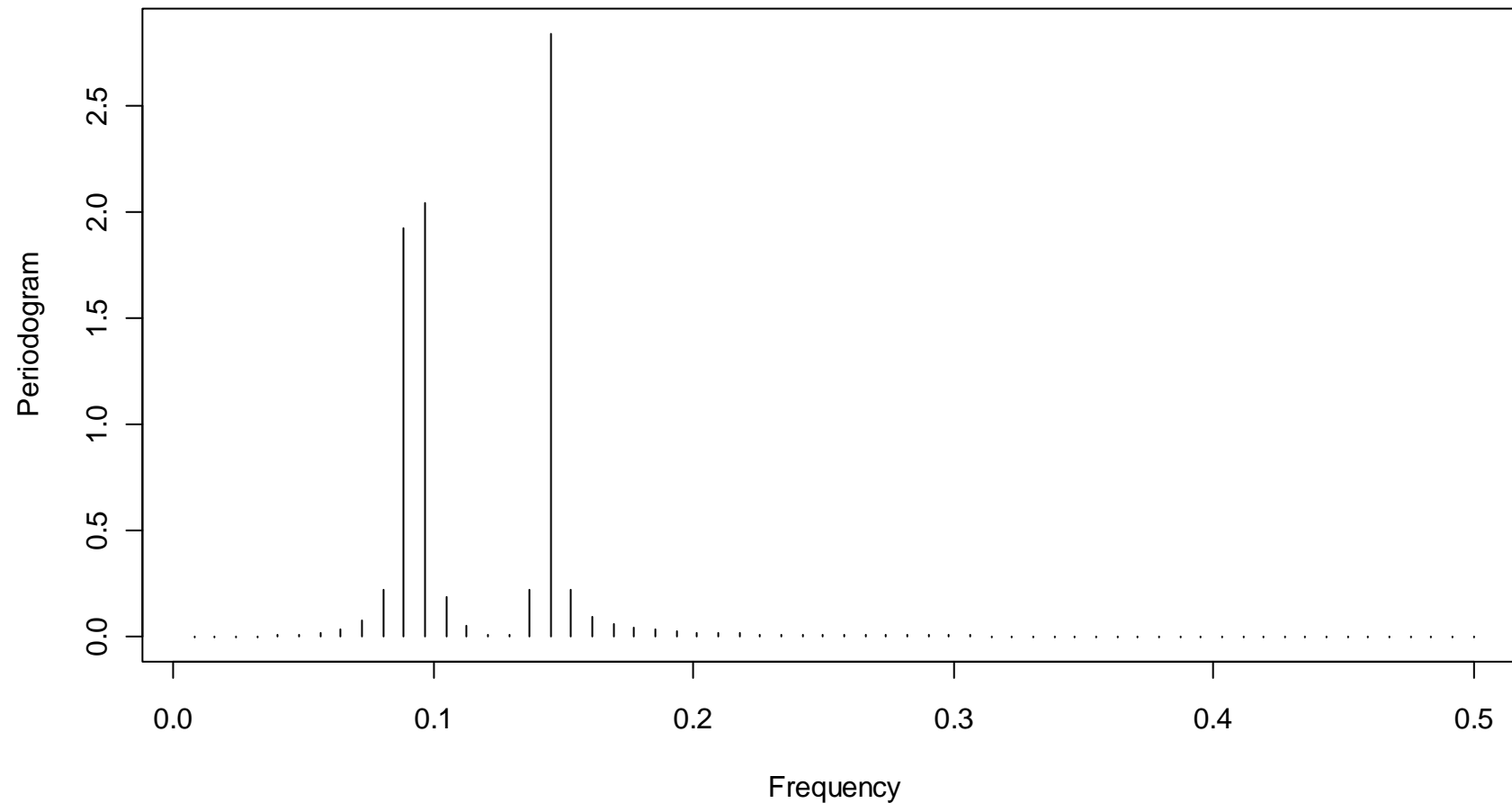


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### *Periodogram of the Shortened Series*

Periodogram of the Shortened Series



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### *Properties of the Periodogram*

Periodogram and correlogram are mathematically equivalent, the former is the **fourier transform** of the latter.

→ **see the blackboard** for the derivation

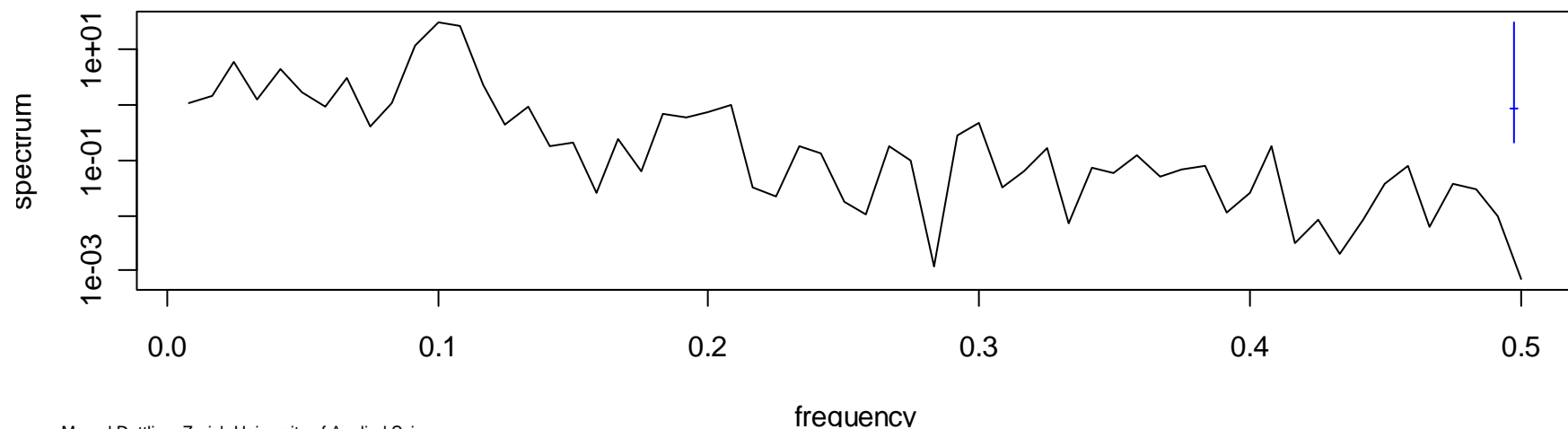
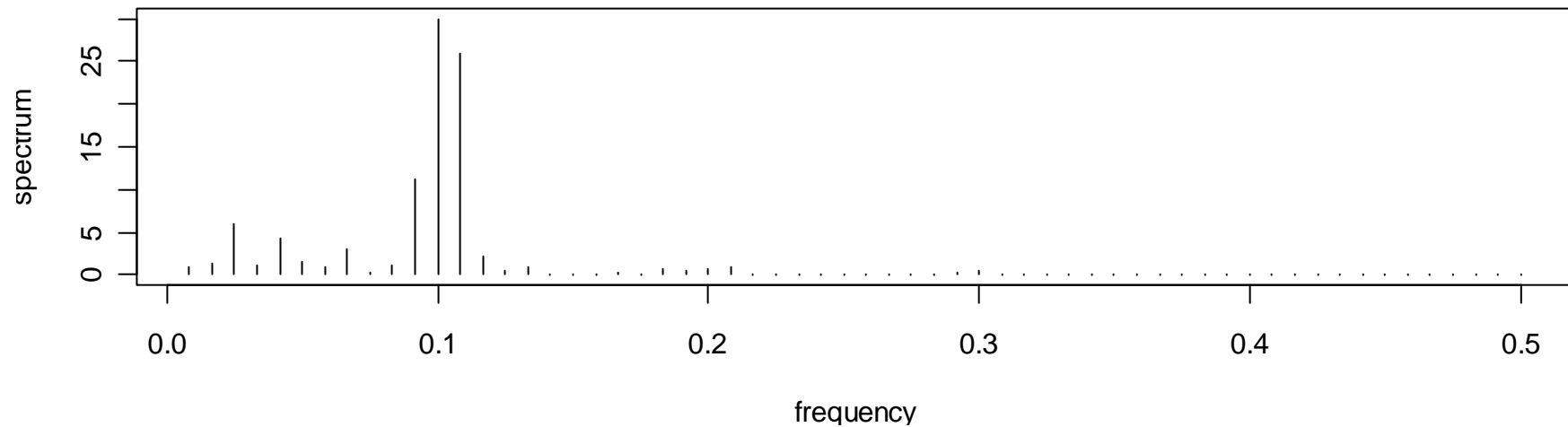
Note: this is a reason why we divided by  $1/n$  in the ACV.

- $I(\nu_k)$  or  $\log(I(\nu_k))$  are plotted against  $k/n$
- Estimates seem rather instable and noisy
- On the log-scale, most frequencies are present
- It seems as if smoothing is required for interpretation.

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### *Periodogram of the Log Lynx Data*

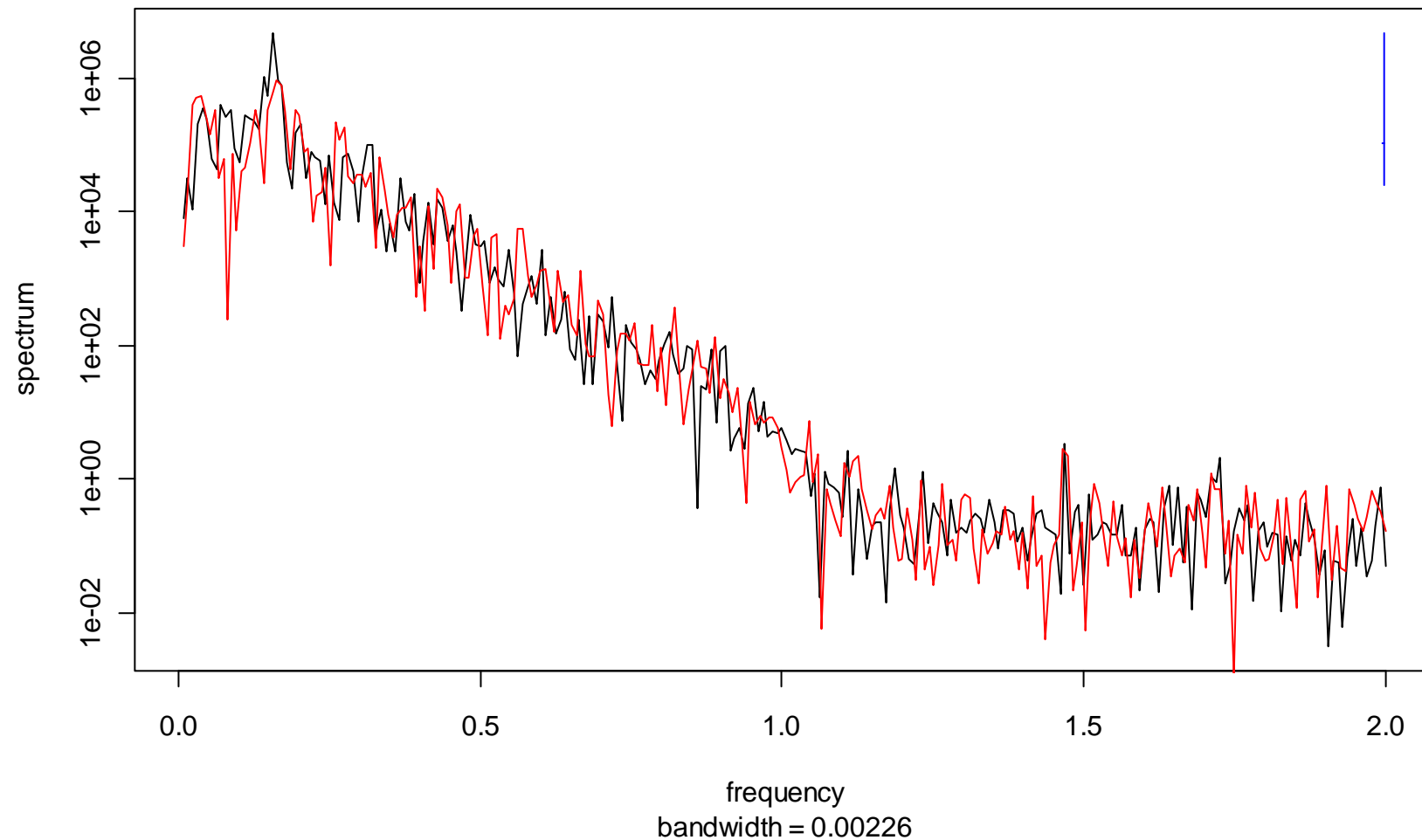


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# *Periodogram of the Ocean Wave Data*

Periodogram of the Ocean Wave Data

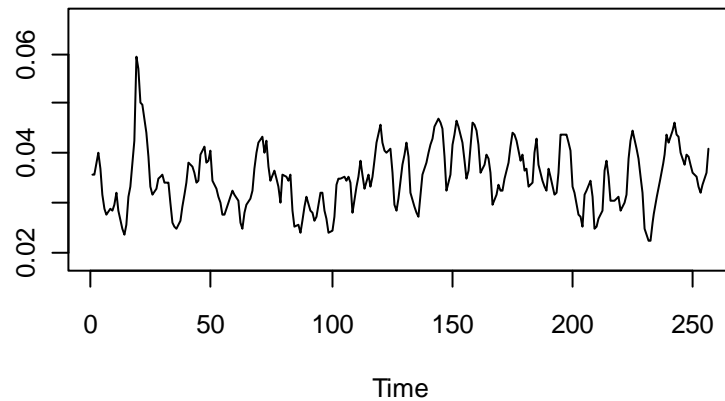


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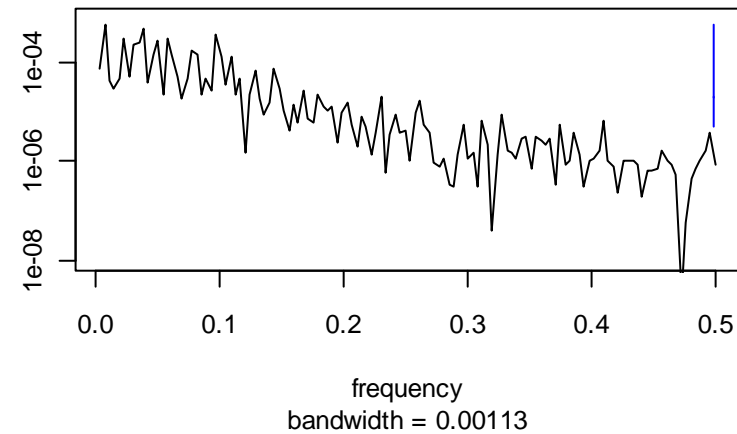
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### *Periodogram of the 2-Component-Mixture*

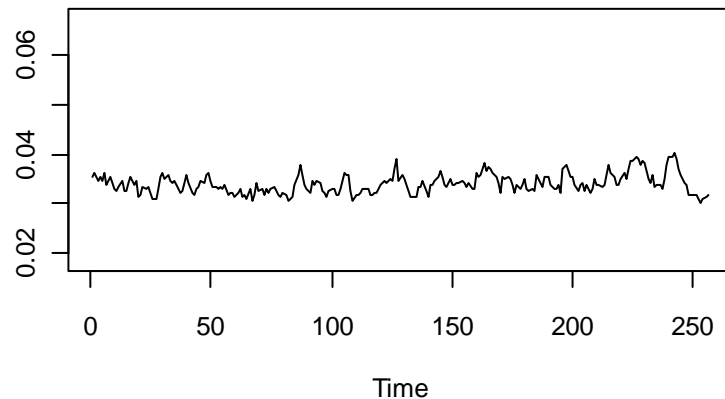
**2-Component-Mixture: Config 1**



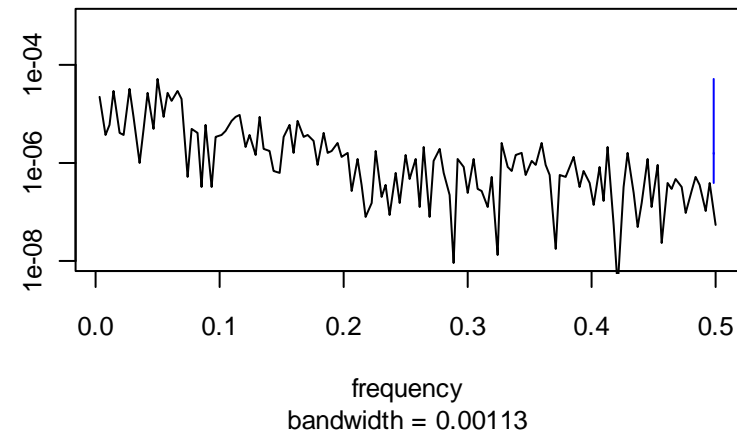
**Periodogram of Config 1**



**2-Component-Mixture: Config 2**



**Periodogram of Config 2**



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### *The Spectrum*

Observed time series	↔	Stochastic process
Empirical ACF	↔	Theoretical ACF
Periodogram	↔	Spectrum

There is a link between ACF and periodogram/spectrum

$$f(\nu) = \sum_{k=-\infty}^{+\infty} \gamma(k) \cos(2\pi\nu k)$$

and

$$\gamma(k) = \int_{-0.5}^{+0.5} f(\nu) \cos(2\pi\nu k) d\nu$$

respectively. The spectrum is thus the Fourier transformation of the ACF.

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### *What's the Spectrum Good For?*

#### Theorem: **Cramer Representation**

Every stationary process can be written as the limit of a linear combination consisting of harmonic oscillations with random, uncorrelated amplitudes.

- The spectrum characterizes the variance of all these random amplitudes.
- Or vice versa:  $\int_{\nu_1}^{\nu_2} f(\nu) d\nu$  is the variance between the frequencies that make the integration limits.
- The spectrum takes only positive values. Thus, not every ACF sequence defines a stationary series.



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### *A Few Particular Spectra*

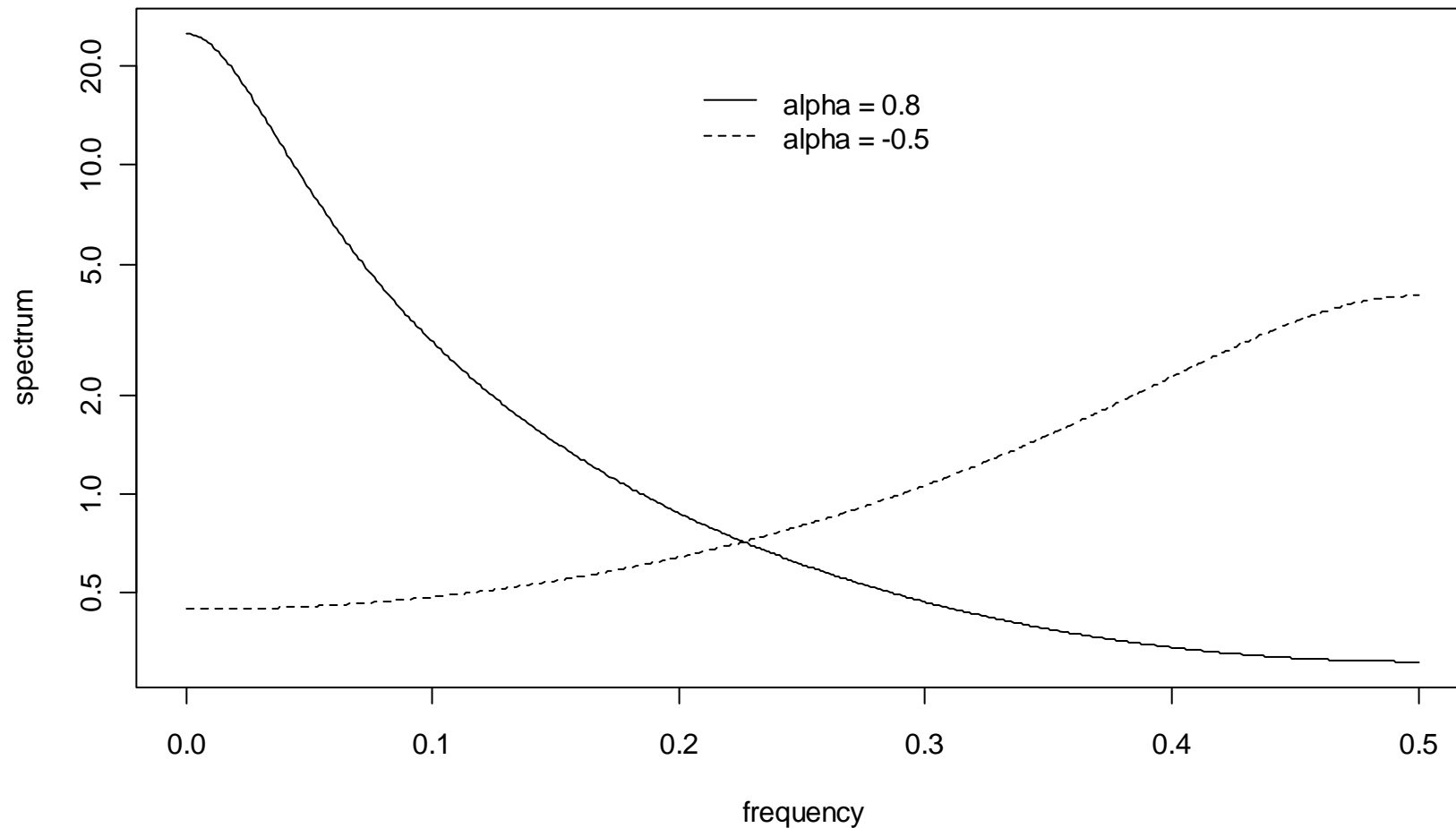
- **White noise**  
→ the spectrum is constant over all frequencies.
- **AR(1)**, [see next slide](#)  
→ already quite a complicated function  $\alpha_1$
- **ARMA (p,q)**  
→ the characteristic polynoms determine the spectrum
$$f(\nu) = \sigma_E^2 \frac{|\Theta(\exp(-i2\pi\nu))|}{|\Phi(\exp(-i2\pi\nu))|}$$
- Note: to generate  $m$  maxima in the spectrum, we require an AR-model, where the order is at least  $2m$ .

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### *Spectrum of AR(1)-Processes*

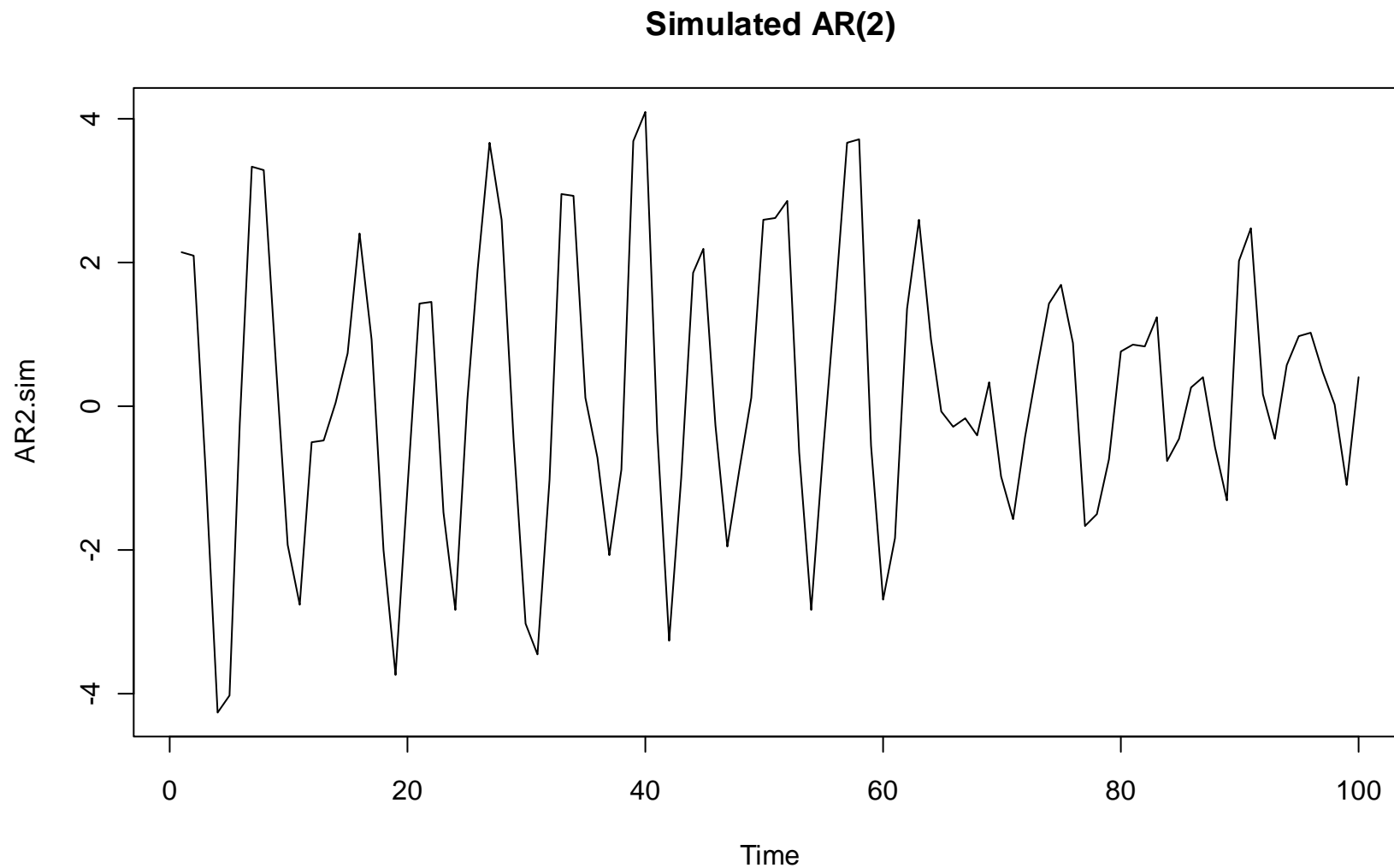
Spectrum of Simulated AR(1)-Processes



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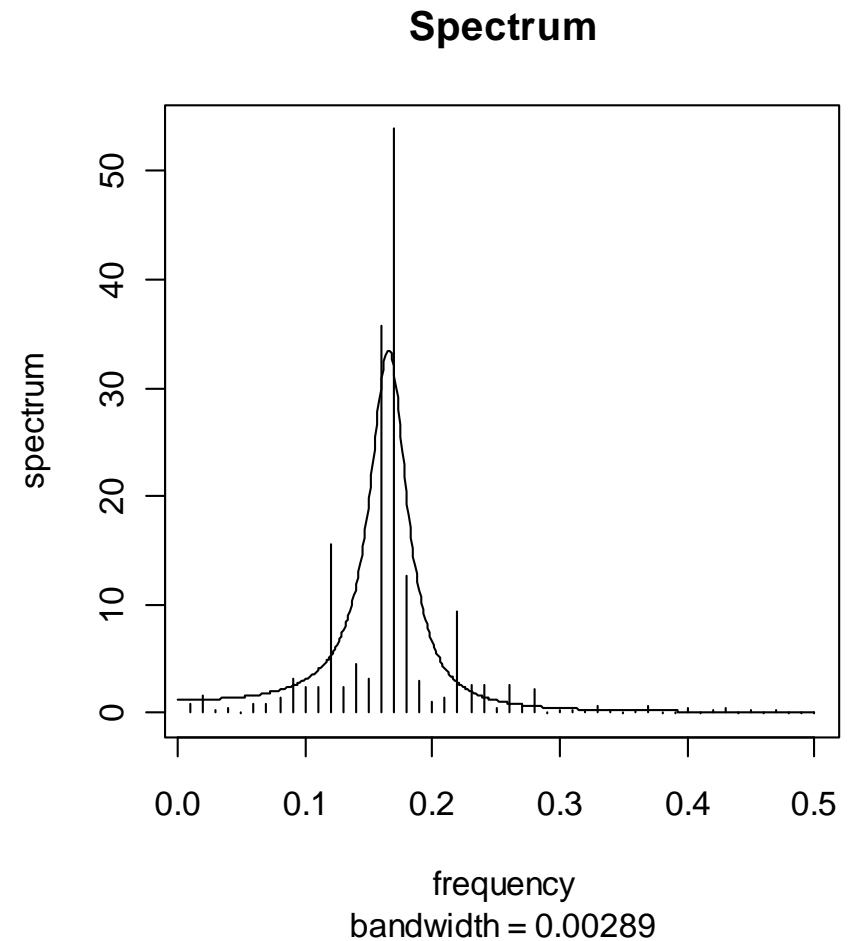
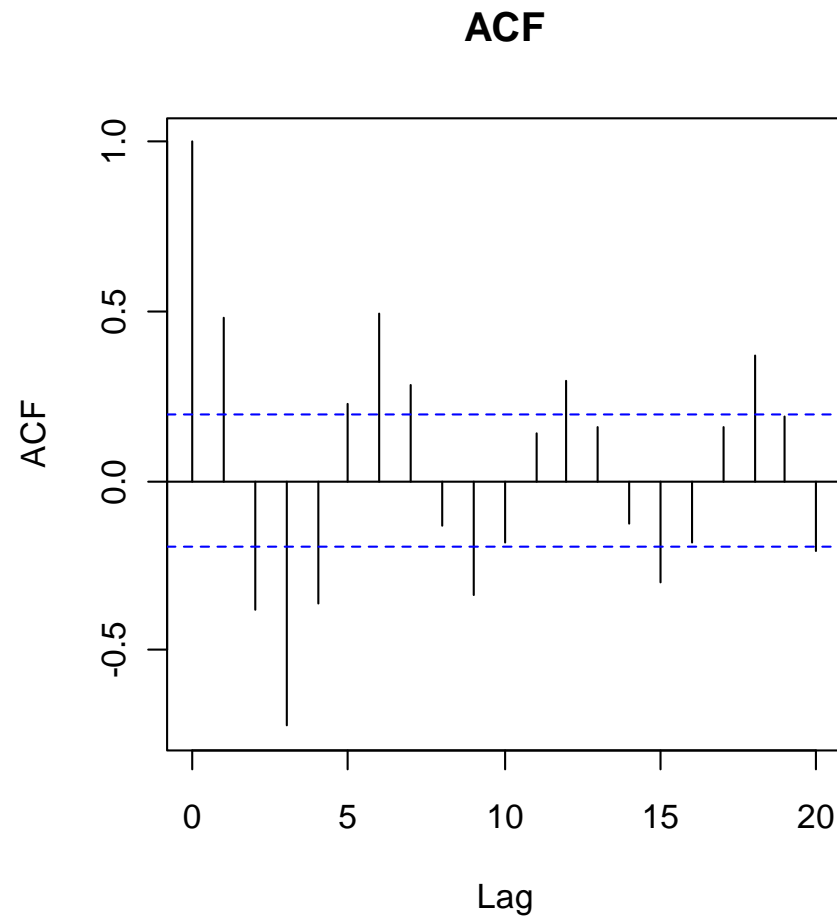
### *Simulated AR(2)-Process*



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### *ACF/Spectrum of Simulated AR(2)-Process*



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### *Spectral Analysis*

- **Spectral analysis** is a descriptive technique, where the time series is interpreted as a linear combination of harmonic oscillations.
- The **periodogram** shows empirically, which frequencies are „important“, i.e. lead to a substantial increase in RSS when omitted from the linear combination.
- The **spectrum** is the theoretical counterpart to the periodogram. It can also be seen as the Fourier transformation of the theoretical autocovariances.
- The periodogram is a poor **estimator** for the spectrum: it's not smooth and inconsistent.

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## *Improving the Raw Periodogram*

### 1) Smoothing with a running mean

- + simple approach
- choice of the bandwidth

### 2) Smoothing with a weighted running mean

- + choice of the bandwidth is less critical
- difficulties shift to the choice of weights

### 3) Weighted plug-in estimation

- + weighted Fourier trsf. of estimated autocovariances
- choice of weights

### 4) Piecewise periodogram estimation with averaging

- + can serve as a check for stationarity, too

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### *Improving the Raw Periodogram*

#### 5) **Spectrum of an estimated model**

- + fundamentally different from 1)-4)
- only works for „small“ orders  $p$

#### 6) **Tapering**

- + further modification of periodogram estimation
- + reduces the bias in the periodogram
- + should always be applied

#### 7) **Prewhitening and Rescoloring**

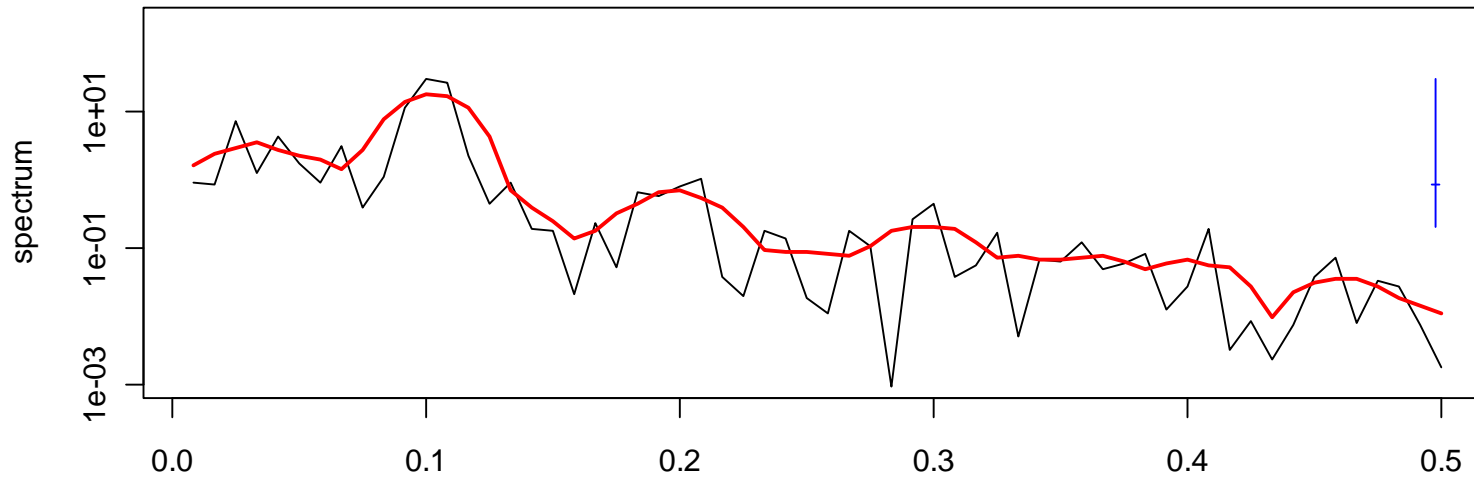
- + model fit and periodogram estimation on residuals
- + the effect of the model will be added again

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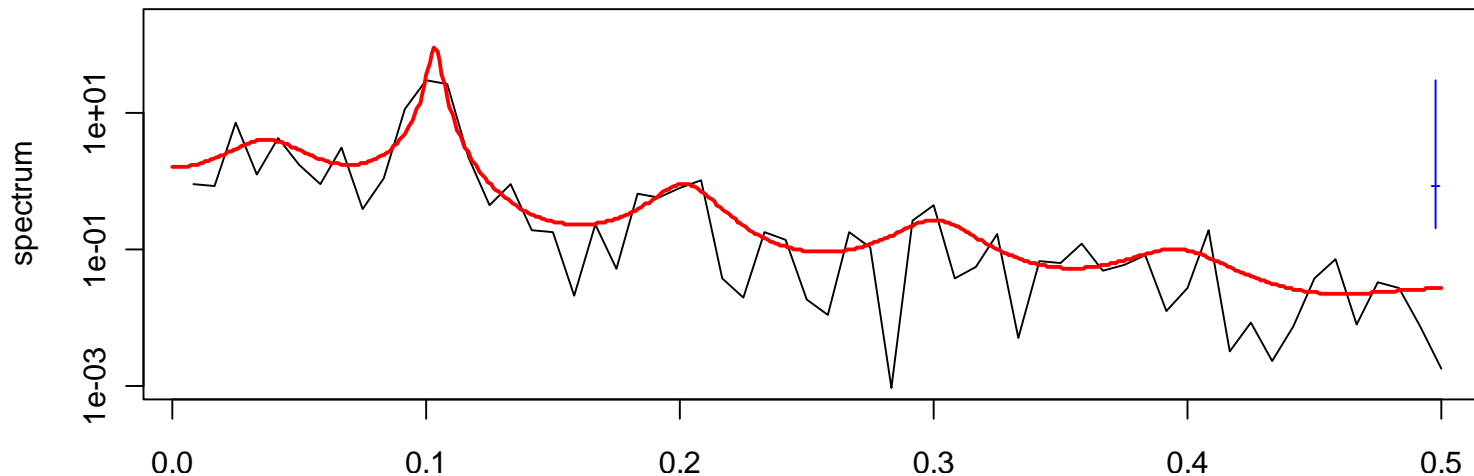
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### *Modified Periodogram of log(Lynx) Data*

Raw and Smoothed Periodogram



Raw and Model Based Periodogram





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### *Modified Periodogram of log(Lynx) Data*

Piecewise periodogram of ocean wave data

