

## LECTURE FIFTEEN

### Spectrum Model

Periodogram:  $y_0, y_1, \dots, y_{n-1}$

DFT:  $b_0, b_1, \dots, b_{n-1}$

$$I\left(\frac{j}{n}\right) = \frac{|b_j|^2}{n}, \quad 0 < \frac{j}{n} < \frac{1}{2}$$

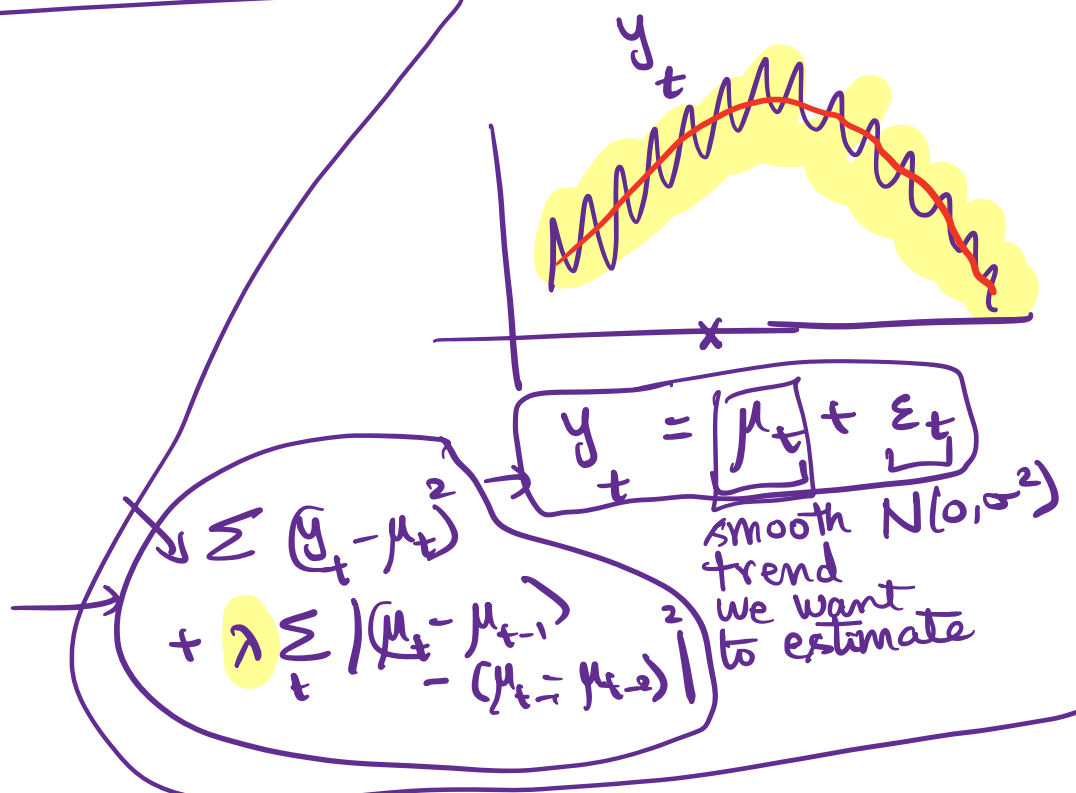
$n$ : odd,  $m = \frac{n-1}{2}$

$$\boxed{I\left(\frac{j}{n}\right), j=1, \dots, m}$$

: tells us sinusoids at which frequencies are contributing to the data

Most often in practice,  $I\left(\frac{j}{n}\right)$  will have many peaks, usually very noisy. especially looking at the log-scale.

Qn: How to SMOOTH PERIODOGRAM?



To smooth the periodogram, we shall write a model.

$$\rightarrow I\left(\frac{j}{n}\right) = \underbrace{f\left(\frac{j}{n}\right)} + \boxed{\varepsilon_j}, \quad \varepsilon_j \stackrel{\text{iid}}{\sim} N(0, \sigma^2)$$

WE WILL NOT  $\mathbb{E}(I(\frac{j}{n}))$ : mean of the periodogram  
USE THIS MODEL

BECAUSE ①  $I(\frac{j}{n})$  is always positive.

② Additive model is better suited for log-periodogram (not  $I(\frac{j}{n})$  directly)

Multiplicative Model:

$$I\left(\frac{j}{n}\right) = \underbrace{f\left(\frac{j}{n}\right)}_{\substack{\text{represents} \\ \text{smooth pattern} \\ \text{in the periodogram}}} \boxed{\eta_j} \rightarrow \text{multiplicative noise}$$

We want ①  $\mathbb{E} \eta_j = 1$  so that

②  $\eta_j$  to be positive valued

$$f\left(\frac{j}{n}\right) = \mathbb{E} I\left(\frac{j}{n}\right)$$

$$\eta_j \stackrel{\text{iid}}{\sim} \text{Exp}(1)$$

$$I\left(\frac{j}{n}\right) = \underbrace{f\left(\frac{j}{n}\right)}_{\text{smooth}} \eta_j \quad \eta_j \stackrel{\text{iid}}{\sim} \text{Exp}(1)$$

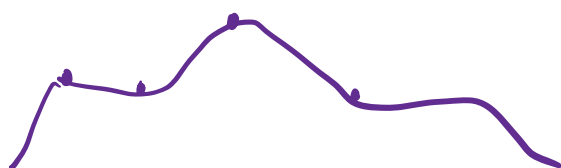
$$\text{Exp}(1) = \text{Gamma}(1, 1) = \frac{\chi^2_2}{2}$$

$$2 \eta_j \stackrel{iid}{\sim} \chi^2_2$$

$$f\left(\frac{j}{n}\right)$$

= mean of  $I\left(\frac{j}{n}\right) \rightarrow$

power of frequency  $\frac{j}{n}$



$\rightarrow$  spectral density ( $f$ )



$$I\left(\frac{j}{n}\right) = f\left(\frac{j}{n}\right) \eta_j, \quad \eta_j \stackrel{iid}{\sim} \text{Exp}(1)$$

$$\log I\left(\frac{j}{n}\right) = \underbrace{\log f\left(\frac{j}{n}\right)}_{\log f\left(\frac{j}{n}\right)} + \underbrace{\log \eta_j}_{\log \eta_j}$$

$$\mathbb{E}(\log \eta_j) \approx -0.529$$



$$y_t = \mu_t + \varepsilon_t, \quad \varepsilon_t \sim \text{N}(0, \sigma^2)$$

## Connection to DFT

$$y_0, \dots, y_{n-1}$$

$$b_0, \dots, b_{n-1}$$

$$\text{Re}(b_0), \text{Im}(b_j) \stackrel{iid}{\sim} N(0, \sigma_j^2)$$

$$\begin{aligned}
 I\left(\frac{j}{n}\right) &= \frac{|b_j|^2}{n} = \frac{\text{Re}(b_j)^2 + \text{Im}(b_j)^2}{n} \\
 &= \frac{\gamma_j^2 \underbrace{(N(0,1))^2}_{\text{Independent}} + \gamma_j^2 \underbrace{(N(0,1))^2}_{\text{Independent}}}{n} \\
 &= \frac{\gamma_j^2}{n} \underbrace{(N(0,1)^2 + N(0,1)^2)}_{\text{independent}} \\
 &= \frac{\gamma_j^2}{n} \chi_2^2 \\
 &= \left( \frac{2\gamma_j^2}{n} \right) \left( \frac{\chi_2^2}{2} \right) \text{Exp}(1)
 \end{aligned}$$

$$f\left(\frac{j}{n}\right) = \frac{2\gamma_j^2}{n}$$

### Spectrum Model

- ①  $I\left(\frac{j}{n}\right) = f\left(\frac{j}{n}\right) \eta_j$ ,  $\eta_j \stackrel{\text{iid}}{\sim} \text{Exp}(1)$
- ②  $\text{Re}(b_j), \text{Im}(b_j) \stackrel{\text{iid}}{\sim} N(0, \gamma_j^2)$

$$\frac{2\gamma_j^2}{n} = f\left(\frac{j}{n}\right)$$

Data:  $y_t$

Representation of the Model in terms of the data  $\{y_t\}$

Inverse DFT formula:

$$y_t = \frac{1}{n} \sum_{j=0}^{n-1} b_j \exp\left(\frac{2\pi i j t}{n}\right) \rightarrow \text{IDFT}$$

$$b_j = \sum_{t=0}^{n-1} y_t \exp\left(-\frac{2\pi i j t}{n}\right) \rightarrow \text{DFT}$$

$$\begin{aligned} y_t &= \frac{1}{n} \sum_{j=0}^{n-1} b_j \exp\left(\frac{2\pi i j t}{n}\right) \quad \underbrace{\begin{matrix} \text{Re}(b_j) \\ \text{Im}(b_j) \end{matrix}}_{\sim N(0, \sigma_j^2)} \\ &= \frac{1}{n} \sum_{j=0}^{n-1} \left( \text{Re}(b_j) + i \text{Im}(b_j) \right) \left( \cos \frac{2\pi j t}{n} + i \sin \frac{2\pi j t}{n} \right) \\ &= \frac{1}{n} \sum_{j=0}^{n-1} \left( \text{Re}(b_j) \cos \frac{2\pi j t}{n} - \text{Im}(b_j) \sin \frac{2\pi j t}{n} \right) \\ &\quad + \frac{1}{n} \sum_{j=0}^{n-1} \left( \text{Re}(b_j) \sin \frac{2\pi j t}{n} + \text{Im}(b_j) \cos \frac{2\pi j t}{n} \right) \\ &\quad \parallel \quad b_{n-j} = \overline{b_j} \\ &\quad \text{because data is real.} \end{aligned}$$

$$\begin{aligned} y_t &= \frac{1}{n} \sum_{j=0}^{n-1} \left( \text{Re}(b_j) \cos \frac{2\pi j t}{n} - \text{Im}(b_j) \sin \frac{2\pi j t}{n} \right) \\ &\quad \text{Re}(b_j), \text{Im}(b_j) \stackrel{\text{i.i.d.}}{\sim} N(0, \sigma_j^2), \\ &\quad j=1, \dots, m \\ &= \frac{b_0}{n} + \frac{1}{n} \sum_{j=1}^m \left[ \text{Re}(b_j) \cos \frac{2\pi j t}{n} - \text{Im}(b_j) \sin \frac{2\pi j t}{n} \right] \end{aligned}$$

$$+ \frac{1}{n} \sum_{j=m+1}^{n-1} (\underbrace{\operatorname{Re}(b_j)}_{b_{n-j}}) \cos \frac{2\pi j t}{n} - \underbrace{\operatorname{Im}(b_j)}_{b_j} \sin \frac{2\pi j t}{n}$$

Check: last two terms coincide

$$b_{n-j} = \overline{b_j}$$

$$y_t = \frac{b_0}{n} + \sum_{j=1}^m \left[ \frac{2 \operatorname{Re}(b_j)}{n} \cos \frac{2\pi j t}{n} + \frac{-2 \operatorname{Im}(b_j)}{n} \sin \frac{2\pi j t}{n} \right]$$

→ Real IBFT formula

$$y_t = \beta_0 + \sum_{j=1}^m \left( \beta_{1j} \cos \frac{2\pi j t}{n} + \beta_{2j} \sin \frac{2\pi j t}{n} \right)$$

where  $\beta_0$  is a constant

$\beta_{1j}, \beta_{2j}$  independent across  $j$

$\beta_{1j}, \beta_{2j} \stackrel{iid}{\sim} N(0, \tau_j^2)$

$$y_t = \beta_0 + \beta_1 \cos 2\pi f t + \beta_2 \sin 2\pi f t + \epsilon_t$$

$\epsilon_t$  has no additive error term.

### Spectrum Model

$y_0, \dots, y_{n-1}$  : Data

$b_0, \dots, b_{n-1}$  : DFT

$I(\frac{j}{n}), j=1, \dots, m$  : Periodogram

Model can be described in terms of either of these

① Periodogram:  $I(\frac{j}{n}) = \underbrace{f(\frac{j}{n})}_{\text{parameters}} \eta_j$   
 $\eta_j \stackrel{iid}{\sim} \text{Exp}(1)$

② DFT:  $b_1, \dots, b_m$  independent  
 $\text{Re}(b_j), \text{Im}(b_j) \stackrel{iid}{\sim} N(0, \sigma_j^2)$

$$f(\frac{j}{n}) = \frac{2\sigma_j^2}{n}$$

③  $y_t = \beta_0 + \sum_{j=1}^m \left( \beta_{1j} \cos \frac{2\pi j t}{n} + \beta_{2j} \sin \frac{2\pi j t}{n} \right)$   
 $\beta_{1j}, \beta_{2j} \stackrel{iid}{\sim} N(0, \tau_j^2) \rightarrow \text{all Fourier frequency sinusoids}$

Parameters:  $\tau_1^2, \dots, \tau_m^2$

$$\beta_{1j} = \frac{2 \text{Re}(b_j)}{n} \sim N(0, \frac{4\sigma_j^2}{n^2})$$

$$\tau_j^2 = \frac{2\sigma_j^2}{n}$$

Estimation of parameters

$$\prod_{j=1}^m \frac{f(I(j/n))}{I(j/n)} = \prod_{j=1}^m \frac{f(I(j/n))}{f(\frac{j}{n}) \eta_j}$$

$$I(\frac{j}{n}) = f(\frac{j}{n}) \eta_j$$

$$= \prod_{j=1}^m \frac{1}{f(j/n)} \exp\left(-\frac{I(j/n)}{f(j/n)}\right)$$

WHITTLE  
LIKELIHOOD

$$\sum_{j=1}^m -\frac{I(j/n)}{f(j/n)} - \log f(j/n) \quad \nearrow$$

neg log:

$$\sum_{j=1}^m \left[ \underbrace{\log f(j/n)} + \frac{I(j/n)}{f(j/n)} \right]$$

$$\log f(j/n) = g_j$$

$$\sum_j \left[ g_j + e^{-g_j} I(j/n) \right]$$

$$+ \lambda \sum_{j=3}^m \left( (g_j - g_{j-1}) - (g_{j-1} - g_{j-2}) \right)^2$$