

# Unit 20: Discrete Fourier Transform and Periodogram

Jeffrey Woo

Department of Statistics, University of Virginia

Spring 2020

## Readings for Unit 20

Textbook chapter 4.3.

## Last Unit

- 1 Autocovariance generating function for ARMA processes.
- 2 Rational spectrum representation for ARMA processes.

# This Unit

- 1 Discrete Fourier Transform
- 2 Periodogram

# Motivation

So far, we've looked at the spectral density, which is a population quantity. We'll next consider the sample version: the periodogram.

- 1 Discrete Fourier Transform
- 2 Periodogram
- 3 Sampling Distribution of Periodogram
- 4 Worked Example

# Discrete Fourier Transform

Given data  $x_1, x_2, \dots, x_n$ , we define the discrete Fourier transform (DFT) as

$$d(\omega_j) = n^{-1/2} \sum_{t=1}^n x_t e^{-2\pi i \omega_j t} \quad (1)$$

where  $\omega_j = j/n, j = 0, 1, \dots, n-1$ , and the frequencies  $\omega_j = j/n$  are called the Fourier or **fundamental frequencies**.

# Discrete Fourier Transform

The difference between the DFT (1) and the Fourier transform in Unit 18 is that the DFT does the computation for discrete frequency  $\omega_j$  while the Fourier transform does for all frequencies  $-1/2 \leq \omega_j \leq 1/2$ .



# Discrete Fourier Transform

Using Euler's formula  $e^{i\alpha} = \cos(\alpha) + i\sin(\alpha)$ , we could also write the DFT (1) as

$$d(\omega_j) = n^{-1/2} \sum_{t=1}^n x_t \cos(2\pi\omega_j t) - in^{-1/2} \sum_{t=1}^n x_t \sin(2\pi\omega_j t). \quad (2)$$

# Discrete Fourier Transform

The real part of (2) is identified as the cosine transform  $d_c(\omega_j)$  and the imaginary part as the sine transform  $d_s(\omega_j)$ . We could write the DFT (1) as

$$d(\omega_j) = d_c(\omega_j) - id_s(\omega_j).$$

- 1 Discrete Fourier Transform
- 2 Periodogram
- 3 Sampling Distribution of Periodogram
- 4 Worked Example

# Periodogram

The periodogram is defined to be

$$I(\omega_j) = |d(\omega_j)|^2. \quad (3)$$

i.e. the squared modulus of the DFT (1). Recall that  $|a + ib|^2 = a^2 + b^2$ . Therefore,

$$I(\omega_j) = (d_c(\omega_j))^2 + (-d_s(\omega_j))^2 = d_c^2(\omega_j) + d_s^2(\omega_j)$$

# A Few Notes about Complex Numbers

$$\overline{a + ib} = a - ib \quad (\text{complex conjugate, } \bar{a} = a \text{ for real number } a.)$$

$$|a + ib|^2 = a^2 + b^2 = (a + ib)\overline{(a + ib)}. \quad (|z| \text{ is the modulus of } z.)$$

$$\overline{z_1 z_2} = \bar{z}_1 \bar{z}_2, \quad \text{for two complex numbers } z_1, z_2.$$

# Periodogram as Sample Version of Spectral Density

# Inverse Fourier Transform

Another useful property is the inverse Fourier transform

$$\begin{aligned}x_t &= n^{-1/2} \sum_{j=0}^{n-1} d(\omega_j) e^{2\pi i \omega_j t} \\&= \\&= \\&= \\&= \end{aligned}$$

where  $a_0 = \bar{x} = (x_1 + \cdots + x_n)/n$ , and we have assumed for simplicity that  $n$  is odd and  $m = (n - 1)/2$ .

# Interpreting the Periodogram

We can think of the inverse Fourier transform as a regression of  $x_t$  on sines and cosines with the coefficients equal to  $2/\sqrt{n}$  times the sine part and the cosine part of the Fourier transforms respectively. Therefore,  $d_c(\omega_j)$  and  $d_s(\omega_j)$  measure the contribution the frequency  $\omega_j$  has in **explaining the variation** in the time series. The bigger  $d_c(\omega_j)$  and  $d_s(\omega_j)$ , the greater the contribution from the frequency  $\omega_j$ .



# Interpreting the Periodogram

Additionally, one can further show that

$$\sum_{t=1}^n (x_t - \bar{x})^2 = 2 \sum_{j=1}^m [d_c^2(\omega_j) + d_s^2(\omega_j)] = 2 \sum_{j=1}^m I(\omega_j).$$

# Interpreting the Periodogram

The sum of squares can be decomposed into 2 times the sum of the periodograms over frequencies  $\omega_j$ ,  $1 \leq j \leq m$ . In other words, the variation in the series  $x_t$  is distributed over frequencies  $\omega_j$ , where the amount of variation explained by frequency  $\omega_j$  is  $2I(\omega_j)$ .

# Interpreting the Periodogram

Thus, we can interpret the periodogram as the amount of variation at a certain frequency. This is how we also interpret the spectral density. The periodogram is the sample version of the spectral density, which is a population quantity.

- 1 Discrete Fourier Transform
- 2 Periodogram
- 3 Sampling Distribution of Periodogram
- 4 Worked Example

# Sampling Distribution of Periodogram

Let  $\omega_{j:n}$  denote a frequency of the form  $j_n/n$ , where  $\{j_n\}$  is a sequence of integers so that  $j_n/n \rightarrow \omega$  as  $n \rightarrow \infty$ . It turns out that

$$E[I(\omega_{j:n})] \rightarrow f(\omega) = \sum_{h=-\infty}^{\infty} \gamma(h) e^{-2\pi i \omega h}.$$

The spectral density is the **long run average** of the periodogram.

# Sampling Distribution of Periodogram

It turns out that if  $\{x_t\}$  is causal, then

$$\frac{2I(\omega_{j:n})}{f(\omega_j)} \xrightarrow{d} \text{i.i.d. } \chi^2_2. \quad (4)$$

provided  $f(\omega_j) > 0$  for  $j = 1, \dots, m$  for any collection of  $m$  distinct frequencies  $\omega_j$  with  $\omega_{j:n} \rightarrow \omega_j$ .

# Confidence Intervals

From (4), an approximate  $100(1 - \alpha)\%$  confidence interval for the spectral density takes the form

$$\frac{2I(\omega_{j:n})}{\chi_2^2(1 - \alpha/2)} \leq f(\omega) \leq \frac{2I(\omega_{j:n})}{\chi_2^2(\alpha/2)}. \quad (5)$$

# Comments

- $\chi^2_2$  distribution has mean 2, thus the expected value of  $I(\omega_j)$  is approximately  $f(\omega_j)$ , i.e. the periodogram is approximately **unbiased**.
- The variance of  $I(\omega_j)$  is approximately  $f^2(\omega_j)$ . For example, for Gaussian white noise, the variance of the periodogram is  $\sigma_w^4$  which **does not decrease with  $n$** .



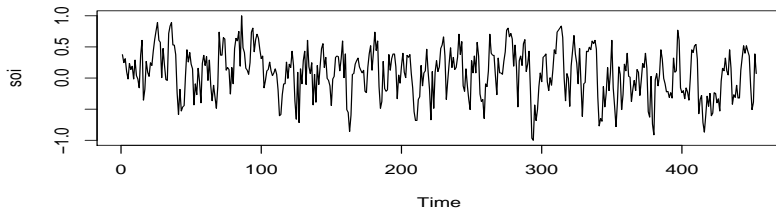
- 1 Discrete Fourier Transform
- 2 Periodogram
- 3 Sampling Distribution of Periodogram
- 4 Worked Example

## Worked Example

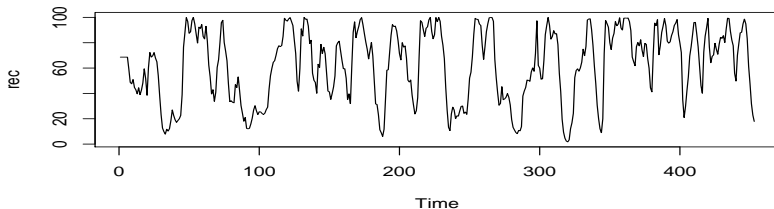
In this example, we will look at the Southern Oscillation Index and recruitment datasets, which contain monthly data on the changes in air pressure and estimated number of new fish in the central Pacific Ocean from 1950 to 1987. The central Pacific Ocean warms approximately every three to seven years due to El Niño.

# Worked Example

**Time Series Plot of SOI**

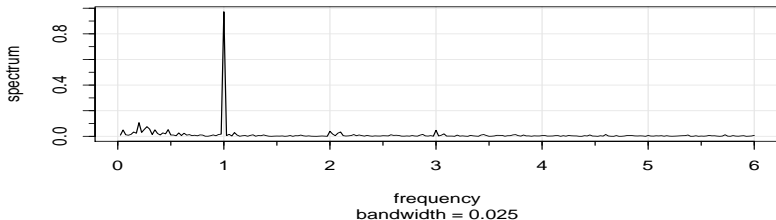


**Time Series Plot of Recruitment**

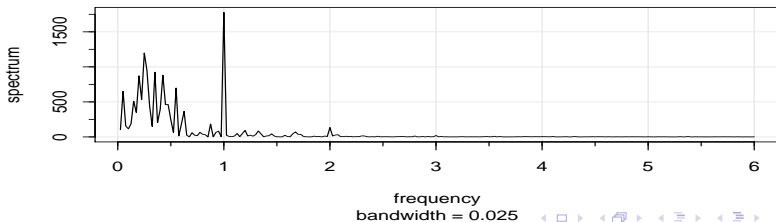


# Worked Example

**Series: soi**  
**Raw Periodogram**

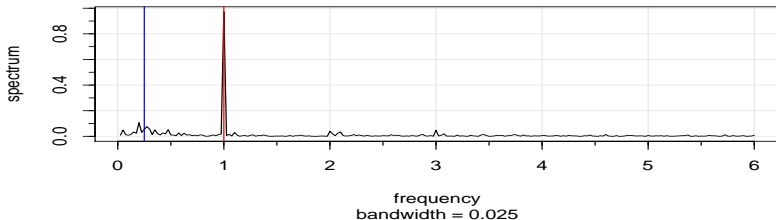


**Series: rec**  
**Raw Periodogram**

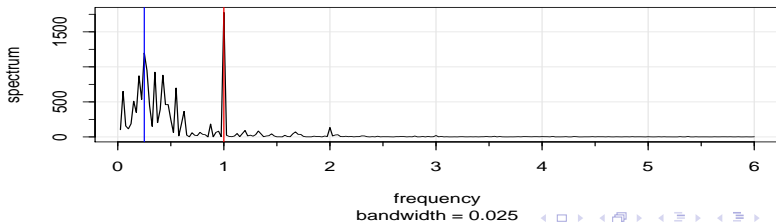


# Worked Example

**Series: soi**  
**Raw Periodogram**



**Series: rec**  
**Raw Periodogram**



## Worked Example

From the periodograms:

- obvious peak at  $\omega = 1/12$  for yearly cycle.
- some peaks at around  $\omega = 1/48$  for El Nino cycle. The wide band of activity suggests that this cycle is **not very regular**.

Note: the horizontal axis of the periodogram produced using the `mvspec()` function from the `astsa` package is in multiples of  $\frac{1}{12}$ . In Unit 6, the `spec.pgram()` function was used instead and the horizontal axis is the value of the frequency.

## Worked Example

From the SOI data, the value of the periodogram at  $\omega = \frac{1}{12}$  is  $I(\frac{1}{12}) = 0.9722$ . Since  $\chi_2^2(0.025) = 0.0506$  and  $\chi_2^2(0.975) = 7.3778$ , an approximate 95% confidence interval for the spectrum  $f(\frac{1}{12})$  is

$$\left( \frac{2(0.9722)}{7.3778}, \frac{2(0.9722)}{0.0506} \right) = (0.2636, 38.4011).$$

At  $\omega = \frac{1}{48}$ ,  $I(\frac{1}{48}) = 0.0537$ , therefore an approximate 95% confidence interval for the spectrum  $f(\frac{1}{48})$  is

$$\left( \frac{2(0.0537)}{7.3778}, \frac{2(0.0537)}{0.0506} \right) = (0.0146, 2.1222).$$

## Worked Example

**Question:** Use R to compute  $I(\frac{1}{12})$  and  $I(\frac{1}{48})$  for the recruit data, as well as derive their corresponding approximate 95% confidence intervals.