Trigonometric functions in discrete time:



Clearly,

 $g(t) = \cos(\frac{2\pi}{2}t) = \cos(\pi t) = (-1)^t$ 

is the most rapid oscillation we can observe.

The frequency  $\pi$  (one half a cycle per sampling interval) is known as **Nyquist frequency**.

If one observation is sampled per unit of time, a periodic function with period *n* completes one full cycle in the observation period, a periodic function with period n/2 completes two full cycles in the observation period, etc. In general, a periodic function with period n/k completes *k* full cycles in the observation period. The frequency  $\omega_k = 2\pi k/n$  that is associated with the period n/k is called the k-th **Fourier frequency**.

We consider only those Fourier frequencies  $\omega_k$  that are less than or equal to the Nyquist frequency  $\pi$ , i.e.,  $k \le m = [n/2]$ .

If the improper Fourier frequency  $2\pi 0/n=0$  is also included, there are always *n* non-vanishing sines and cosines.

If *n* is even and m=n/2, we have 1,  $\cos(\frac{2\pi \cdot 1}{n}t)$ , ...,  $\cos(\frac{2\pi \cdot m}{n}t)$ ,  $\sin(\frac{2\pi \cdot 1}{n}t)$ , ...,  $\sin(\frac{2\pi \cdot (m-1)}{n}t)$ . If *n* is odd and m=(n-1)/2, we have 1,  $\cos(\frac{2\pi \cdot 1}{n}t)$ , ...,  $\cos(\frac{2\pi \cdot m}{n}t)$ ,  $\sin(\frac{2\pi \cdot 1}{n}t)$ , ...,  $\sin(\frac{2\pi \cdot m}{n}t)$ . Note that  $\cos(\frac{2\pi \cdot 0}{n}t)$  equals 1 and  $\sin(\frac{2\pi \cdot 0}{n}t)$  vanishes. If *n* is even, then  $\sin(\frac{2\pi \cdot m}{n}t)=\sin(\pi t)$  vanishes too.

Exericse: Show that 
$$\hat{b} = \sum_{t=1}^{n} x_t y_t / \sum_{t=1}^{n} x_t^2$$
 minimizes the sum of squared errors  $\sum_{t=1}^{n} (y_t - bx_t)^2$ . **CB**

Using only Fourier frequencies  $\omega_k = 2\pi k/n, k \in T \subseteq \{0,...,m\}$ , in the trigonometric regression model

$$y_t = \sum_{k \in T} (A_k \cos(\omega_k t) + B_k \sin(\omega_k t)) + u_t$$

has the advantage that the regressors are orthogonal<sup>1</sup>, i.e.,

$$C_k^{\mathrm{T}} S_j = 0, \text{ if } 0 \le j, k \le m,$$
  

$$S_k^{\mathrm{T}} S_j = C_k^{\mathrm{T}} C_j = 0, \text{ if } 0 \le j \ne k \le m,$$

where

$$C_{k} = \begin{pmatrix} \cos\left(\frac{2\pi k}{n} \cdot 1\right) \\ \vdots \\ \cos\left(\frac{2\pi k}{n} \cdot n\right) \end{pmatrix}, S_{j} = \begin{pmatrix} \sin\left(\frac{2\pi j}{n} \cdot 1\right) \\ \vdots \\ \sin\left(\frac{2\pi j}{n} \cdot n\right) \end{pmatrix}.$$

<sup>1</sup> See Appendix B.

In the case of orthogonal regressors, the LS estimates  

$$\begin{pmatrix} \hat{A}_{0}, \hat{A}_{1}, \hat{B}_{1}, ... \end{pmatrix}^{T} = \begin{pmatrix} (C_{0}, C_{1}, S_{1}, ...)^{T} (C_{0}, C_{1}, S_{1}, ...) \end{pmatrix}^{-1} (C_{0}, C_{1}, S_{1}, ...)^{T} y \\
= \begin{pmatrix} C_{0}^{T}C_{0} & C_{0}^{T}C_{1} & C_{0}^{T}S_{1} & \cdots \\ C_{1}^{T}C_{0} & C_{1}^{T}C_{1} & C_{1}^{T}S_{0} & \cdots \\ S_{1}^{T}C_{0} & S_{1}^{T}C_{1} & S_{1}^{T}S_{1} & \cdots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} \begin{pmatrix} C_{0}^{T}y \\ C_{1}^{T}y \\ S_{1}^{T}y \\ \vdots \end{pmatrix} \\
= diag(n, n/2, n/2, ...)$$

obtained from the full model are identical to the estimates

$$\hat{A}_{0} = \sum_{t=1}^{n} \cos(\omega_{0}t) y_{t} \Big/ \sum_{t=1}^{n} \cos^{2}(\omega_{0}t) = \frac{1}{n} \sum_{t=1}^{n} \cos(\omega_{0}t) y_{t} = \overline{y},$$
  

$$\hat{A}_{1} = \sum_{t=1}^{n} \cos(\omega_{1}t) y_{t} \Big/ \sum_{t=1}^{n} \cos^{2}(\omega_{1}t) = \frac{2}{n} \sum_{t=1}^{n} \cos(\omega_{1}t) y_{t},$$
  

$$\hat{B}_{1} = \sum_{t=1}^{n} \sin(\omega_{1}t) y_{t} \Big/ \sum_{t=1}^{n} \sin^{2}(\omega_{1}t) = \frac{2}{n} \sum_{t=1}^{n} \sin(\omega_{1}t) y_{t}, \dots$$
  
by botained from simple models with only one regressor. Classical statements of the simple models with only one regressor.

Exercise: Show that 
$$\sum_{t=1}^{n} \cos^2(\frac{2\pi k}{n}t) = n$$
, if  $k = \frac{n}{2}$ . CN

<u>Exercise</u>: If *n* is even, then m=n/2 and  $\omega_m=\pi$ . But since  $\sin(\pi t)=0$ , only  $\cos(\pi t)$  can be used as a regressor. Show that the LS estimate of the parameter  $A_m$  is given by

$$\hat{A}_m = \frac{1}{n} \sum_{t=1}^n y_t (-1)^t$$
. CM

The **periodogram**  $I(\omega)$  of a time series  $y_1,...,y_n$  is for a Fourier frequency  $\omega_k$  with  $k \in \{1,2,...,\lfloor \frac{n-1}{2} \rfloor\}$  defined by

$$I(\omega_{k}) = \frac{n}{8\pi} (\underbrace{\hat{A}_{k}^{2} + \hat{B}_{k}^{2}}_{\hat{R}_{k}^{2}})$$
  
=  $\frac{n}{8\pi} ((\frac{2}{n} \sum_{t=1}^{n} y_{t} \cos(\omega_{k}t))^{2} + (\frac{2}{n} \sum_{t=1}^{n} y_{t} \sin(\omega_{k}t))^{2})$   
=  $\frac{1}{2\pi n} ((\sum_{t=1}^{n} y_{t} \cos(\omega_{k}t))^{2} + (\sum_{t=1}^{n} y_{t} \sin(\omega_{k}t))^{2}).$ 

Apart from an unimportant scaling factor,  $I(\omega_k)$  is just the squared estimate of the amplitude of a sinusoid with frequency  $\omega_k$ , hence its size indicates how important that particular frequency is.

Defining the periodogram for an arbitrary frequency  $\omega$  by

$$I(\omega) = \frac{1}{2\pi n} \left( \left( \sum_{t=1}^{n} y_t \cos(\omega t) \right)^2 + \left( \sum_{t=1}^{n} y_t \sin(\omega t) \right)^2 \right),$$

it can be written in complex form as

$$I(\omega) = \frac{1}{2\pi n} \left| \sum_{t=1}^{n} y_t \cos(\omega t) + i \sum_{t=1}^{n} y_t \sin(\omega t) \right|^2$$
$$= \frac{1}{2\pi n} \left| \sum_{t=1}^{n} y_t (\cos(\omega t) + i \sin(\omega t)) \right|^2$$
$$= \frac{1}{2\pi n} \left| \sum_{t=1}^{n} y_t e^{i\omega t} \right|^2.$$
CF

<u>Exercise</u>: Show that it is sufficient to consider  $I(\omega)$  on the interval  $0 \le \omega \le \pi$  by proving that  $I(\omega)$  is periodic with period  $2\pi$  and symmetric about the *y*-axis. CS

**Exercise:** Revisit the not seasonally adjusted **Retail and Food Services Sales**. Plot the differenced log series.

d <- y-lag(y,k=-1); d <- na.omit(d,method="r") par(mar=c(2,2,1,1)); plot(d,type="l")



The sequence

$$\sum_{t=1}^{n} y_t \ e^{-i\omega_k t}, k=0,...,n-1,$$

is called the **discrete Fourier transform (DFT)** of the sequence  $y_1, \ldots, y_n$ .

The R function **fft** uses a fast algorithm (the **fast Fourier transform**) to calculate the slightly different version

$$\sum_{t=1}^{n} y_t \ e^{-i\omega_k(t-1)}, \ k=0,...,n-1.$$

However, we have

$$\sum_{i=1}^{n} y_t e^{-i\omega_k (t-1)} \Big|^2 = \left| e^{i\omega_k} \sum_{t=1}^{n} y_t e^{-i\omega_k t} \right|^2$$
$$= \left| \frac{e^{i\omega_k}}{e^{-i\omega_k}} \right|^2 \left| \sum_{t=1}^{n} y_t e^{-i\omega_k t} \right|^2.$$

Exercise: Calculate the periodogram of **d** at the Fourier frequencies  $\omega_k$ , k=1,...,[n/2] and plot it.

n <- length(d); m <- floor(n/2) # number of frequencies f <- (2\*pi/n)\*(1:m) # vector of frequencies ft <- fft(d)[2:(m+1)] # excl. k=0,m+1,m+2,...,n-1 pg <- (1/(2\*pi\*n))\*(Mod(ft))^2 # Mod = modulus plot(f,pg,type="l")



<u>Exercise</u>: Redo the last exercise, but this time omit some observations to guarantee that the seasonal frequencies  $2\pi j/12, j=1,...,6$ , are Fourier frequencies.

r12 <- n%%12 # n modulo 12

 $\#\,$  the remainder we get when we divide n by 12 n12 <- n - r12

d12 <- d[(r12+1):n] # omit the first r12 observations

# length of new series d12 is divisible by 12 m12 <- floor(n12/2); f12 <- (2\*pi/n12)\*(1:m12) ft12 <- fft(d12); ft12 <- ft12[2:(m12+1)] pg12 <- (1/(2\*pi\*n12))\*(Mod(ft12))^2 plot(f12,pg12,type="l")



The peaks at the seasonal frequencies are more pronounced than the slightly displaced peaks in the last exercise.