



# Select number of harmonics

Let  $y_t$  denote time series data with multiple frequencies. For example, if  $y_t$  represents hourly data, then it will have two frequencies:  $m_1 = 24$  (daily) and  $m_2 = 24 * 7$  (weekly). Let  $i = 1, 2$ .

**Step 1:** Decompose the series for each frequency

Apply a  $2 \times m_i$  moving average to estimate the trend  $f_{i,t}$  and then compute the detrended series  $z_{i,t} = y_t - f_{i,t}$ .

**Step 2:** Model each season component

For each series, we'll model the corresponding detrended series using a Fourier series. For  $i = 1, 2$ ,

$$z_{i,t} \approx \sum_{j=1}^{k_i} a_{i,j} \cos\left(\frac{2\pi jt}{m_i}\right) + b_{i,j} \sin\left(\frac{2\pi jt}{m_i}\right) \quad (1)$$

Here  $k_i$  denotes the number of terms in the Fourier series, also known as harmonics.

**Step 3:** Estimate coefficients of Fourier series

For each frequency and a given harmonic  $k_i$ , perform multiple linear regression to estimate the coefficients. The regression model for each frequency  $i = 1, 2$  is given in equation (2)

$$z_{i,t} = \beta_0 + \sum_{j=1}^{k_i} \left( \beta_{1,j} \cos\left(\frac{2\pi jt}{m_i}\right) + \beta_{2,j} \sin\left(\frac{2\pi jt}{m_i}\right) \right) + \epsilon_{i,t} \quad (2)$$

Here

- $\beta_0$  is the intercept.
- $\beta_{1,j}$  and  $\beta_{2,j}$  are the coefficients for the cosine and the sine terms of the  $j$ th harmonic.
- $\epsilon_{i,t}$  is the error term.

Performing separate regressions for each frequency is a common approach when dealing with multiple frequencies since this allows us to better isolate the effects of each frequency. It can also simplify the interpretation of results.

#### Step 4: Combine the models

Combine the models for each frequency to obtain a model for the seasonal component  $z_t$ .

$$z_t = \sum_{i=1}^2 \sum_{j=1}^{k_i} a_{i,j} \cos\left(\frac{2\pi jt}{m_i}\right) + b_{i,j} \sin\left(\frac{2\pi jt}{m_i}\right) \quad (3)$$

#### How to select the number of harmonics $k_i$

To select the number of harmonics, we can use an information criteria like the AIC. We can test different values for  $k_i$ , starting with 1, and then select the model with the lowest AIC.

When dealing with periodic functions with period  $M$ , the exact representation is achieved when

$$k = \begin{cases} \frac{M}{2} & \text{if } M \text{ is even} \\ \frac{M-1}{2} & \text{if } M \text{ is odd} \end{cases} \quad (4)$$

This is due to the Nyquist-Shannon sampling theorem.

In practice, however, a smaller number of harmonics is enough to model the seasonality.

## References

chatGPT: <https://chat.openai.com/share/1c20b6ce-0bde-4c6a-b9a6-b993342bb51c>

- Fourier Analysis of Time Series: An Introduction, by Peter Bloomfield.

## Search words

- Fourier decomposition of a time series.