

Spectral Analysis (Theory)

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Idea: Decompose a **stationary** time series $\{X_t\}$ into a combination of "sinusoids", with random (and uncorrelated) coefficients.

Example: A periodic time series

Consider

$$X_t = A\sin(2\pi wt) + B\cos(2\pi wt)$$

where A , B are uncorrelated random variables with mean zero and variance one.

Example: A periodic time series

- i) Find its mean.
- ii) Find its autocovariance function.
- iii) Find its autocorrelation function.

Solution i)

$$\begin{aligned} \text{i) } E(X_t) &= E[A\sin(2\pi wt) + B\cos(2\pi wt)] \\ &= \sin(2\pi wt)E(A) + \cos(2\pi wt)E(B) = 0 \end{aligned}$$

Solution ii)

Let us recall a result that will allow us to find the autocovariance function, more easily. Let X and Y be two independent random variables. Now we define $U_1 = aX + bY$ and $U_2 = cX + dY$, where a, b, c , and d are constants.

$$\begin{aligned} \text{cov}(U_1, U_2) &= \text{cov}[aX + bY, cX + dY] \\ &= ac \text{cov}(X, X) + ad \text{cov}(X, Y) + bc \text{cov}(Y, X) + bd \text{cov}(Y, Y) \\ &\text{(since } X \text{ and } Y \text{ are independent, } \text{cov}(X, Y) = 0) \\ \text{cov}(U_1, U_2) &= ac \text{cov}(X, X) + bd \text{cov}(Y, Y) = ac \text{Var}(X) + bd \text{Var}(Y). \end{aligned}$$

Solution ii)

Letting $X = A$, $Y = B$, $a = \cos(2\pi wt)$, $b = \sin(2\pi wt)$,
 $c = \cos[2\pi w(t - k)]$, $d = \sin[2\pi w(t - k)]$, we have that
 $\text{cov}(X_t, X_{t-k}) = \cos(2\pi wt)\cos[2\pi w(t - k)] + \sin(2\pi wt)\sin[2\pi w(t - k)]$
 $\text{cov}(X_t, X_{t-k}) = \cos\{2\pi[wt - (wt - wk)]\} = \cos[2\pi(wt - wt + wk)] =$
 $\cos[2\pi wk]$.

Thus,

$$\gamma(k) = \cos[2\pi wk].$$

Solution iii)

Since $\gamma(0) = \cos(0) = 1$, we have that

$$\rho(k) = \frac{\gamma(k)}{\gamma(0)} = \cos[2\pi wk].$$

Homework?

If $\{X_t\}$ and $\{Y_t\}$ are uncorrelated stationary sequences, i.e., if X_r and Y_s are uncorrelated for every r and s , show that $\{X_t + Y_t\}$ is stationary with autocovariance function equal to the sum of the autocovariance functions of $\{X_t\}$ and $\{Y_t\}$.

The autocovariance of the sum of two uncorrelated time series is the sum of their autocovariances. Thus, the autocovariance of a sum of random sinusoids is a sum of sinusoids with the corresponding frequencies:

$$X_t = \sum_{j=1}^m [A_j \sin(2\pi w_j t) + B_j \cos(2\pi w_j t)]$$

$$\gamma(h) = \sum_{j=1}^m \sigma_j^2 \cos(2\pi w_j h),$$

where A_j, B_j are uncorrelated, mean zero, and $\text{Var}(A_j) = \text{Var}(B_j) = \sigma_j^2$.

Spectrum of a stationary random process

Consider a stationary random sequence $\{Y_t\}$ with autocovariance function $\gamma(k) = \text{cov}(Y_t, Y_{t-k})$. The corresponding autocovariance generating function is the function

$$G(z) = \sum_{k=-\infty}^{\infty} \gamma(k)z^k,$$

whose argument, z , is a complex variable.

Spectrum of a stationary random process

If we choose $z = e^{-i(2\pi w)}$ where w is a real variable, we obtain the **spectrum** of $\{Y_t\}$ (a.k.a. **spectral density**)

$$f(w) = G(e^{-i(2\pi w)}) = \sum_{k=-\infty}^{\infty} \gamma(k) e^{-2\pi w; k}$$

Spectrum of a stationary random process

Because $\gamma(k) = \gamma(-k)$ and $e^{iw_*} + e^{-iw_*} = 2\cos(w_*)$ (we will justify this later) we can also write $f(w)$ as

$$f(w) = \gamma(0) + \sum_{k=1}^{\infty} \gamma(k)e^{-2\pi w_i k} + \sum_{k=1}^{\infty} \gamma(k)e^{2\pi w_i k}$$

or

$$f(w) = \gamma(0) + 2 \sum_{k=1}^{\infty} \gamma(k)\cos(2\pi kw)$$

Spectrum of a stationary random process

If σ^2 denotes the variance of Y_t we can similarly define a **normalized spectrum** (a.k.a. **normalized spectral density function**),

$$f^*(w) = \sum_{k=-\infty}^{\infty} \rho(k) e^{-2\pi w_i k}$$

or

$$f^*(w) = \frac{f(w)}{\sigma^2} = 1 + 2 \sum_{k=1}^{\infty} \rho(k) \cos(2\pi kw)$$

Example: White noise

Let Y_t be the process defined by $Y_t = W_t$ where W_t is white noise with mean 0 and variance σ_w^2 . Compute the spectral density and normalized spectral density of Y_t .

First, we check that Y_t is stationary.

$$E(Y_t) = 0$$

Let $t \neq s$

$$\text{cov}(Y_t, Y_s) = E[Y_t Y_s] = E[W_t W_s] \text{ (since } W_s \text{ and } W_t \text{ are independent)}$$

$$\text{cov}(Y_t, Y_s) = (0)(0) = 0.$$

If $t = s$

$$\text{cov}(Y_t, Y_t) = E[Y_t^2] = E[W_t^2] = \sigma_w^2$$

$$\gamma(k) = \begin{cases} \sigma_w^2, & \text{if } k = 0, \\ 0, & \text{otherwise.} \end{cases}$$

We know that the spectral density of Y_t is given by

$$f(w) = \sum_{k=-\infty}^{\infty} \gamma(k) e^{-2\pi w_j k}.$$

In this case,

$$f(w) = \gamma(0) e^{-0} + 0 = \gamma(0) = \sigma_w^2.$$

Then, we have that the normalized spectral density of Y_t is

$$f^*(w) = \sum_{k=-\infty}^{\infty} \rho(k) e^{-2\pi w_i k} = \rho(0) e^{-0} + 0 = \rho(0) = 1.$$

This is the origin of the name white noise: it is like white light, which is a uniform mixture of all frequencies in the visible spectrum.

Example. AR(1)

Let Y_t be the process defined by $Y_t = \phi Y_{t-1} + W_t$ (where W_t $|\phi| < 1$ is white noise with mean 0 and variance σ_W^2). Compute the spectral density and normalized spectral density of Y_t .

Since we have an AR(1) process with $|\phi| < 1$, we know it is stationary. It can be shown that

$$\gamma(0) = \frac{\sigma_w^2}{1-\phi^2}$$

$$\gamma(1) = \frac{\phi\sigma_w^2}{1-\phi^2}$$

$$\dots$$
$$\gamma(k) = \frac{\phi^k\sigma_w^2}{1-\phi^2}$$

(actually, we have done it a couple of times together, in class)

Thus

$$\begin{aligned}f(w) &= \sum_{k=-\infty}^{\infty} \gamma(k) e^{-2\pi w_i k} \\&= \frac{\sigma_w^2}{1-\phi^2} \left[1 + \sum_{k=1}^{\infty} \phi^k e^{-2\pi i w k} + \sum_{k=1}^{\infty} \phi^k e^{2\pi i w k} \right] \\&= \frac{\sigma_w^2}{1-\phi^2} \left[1 + \sum_{k=1}^{\infty} (\phi e^{-2\pi i w})^k + \sum_{k=1}^{\infty} (\phi e^{2\pi i w})^k \right] \\&= \frac{\sigma_w^2}{1-\phi^2} \left[1 + \frac{\phi e^{-2\pi i w}}{1-\phi e^{-2\pi i w}} + \frac{\phi e^{2\pi i w}}{1-\phi e^{2\pi i w}} \right]\end{aligned}$$

(After a little bit of algebra... Remember? We did it together)

$$\begin{aligned} &= \frac{\sigma_w^2}{1-\phi^2} \left[\frac{1-\phi^2}{(1-\phi e^{-2\pi i w})(1-\phi e^{2\pi i w})} \right] \\ &= \frac{\sigma_w^2}{1-\phi[e^{2\pi i w} + e^{-2\pi i w}] + \phi^2} \end{aligned}$$

(next, I am going to remind you something that you know from Calc, that will allow us to simplify the last expression)

Recalling that

$$\sin(x) = \sum_{k=1}^{\infty} \frac{(-1)^{k+1} x^{2k-1}}{(2k-1)!}$$

$$\cos(x) = \sum_{k=0}^{\infty} \frac{(-1)^k x^{2k}}{(2k)!} \text{ and}$$

$$\exp(x) = \sum_{k=0}^{\infty} \frac{x^k}{k!}$$

$$\begin{aligned} \exp(ix) &= \sum_{k=0}^{\infty} \frac{(ix)^k}{k!} \\ &= \left[1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \frac{x^6}{6!} + \dots \right] \\ &\quad + i \left[x - \frac{x^3}{3!} + \frac{x^5}{5!} - \frac{x^7}{7!} + \dots \right] \end{aligned}$$

Thus,

$$\exp(ix) = \cos(x) + i \sin(x)$$

Euler's Formula.

Solution(spectral density)

Using Euler's Formula

$$\exp(2\pi iw) = \cos(2\pi w) + i \sin(2\pi w)$$

$$\exp(-2\pi iw) = \cos(2\pi w) - i \sin(2\pi w).$$

$$\text{So, } \exp(2\pi iw) + \exp(-2\pi iw) = 2\cos(2\pi w)$$

Finally!!,

$$f(w) = \frac{\sigma_w^2}{1 - 2\phi\cos(2\pi w) + \phi^2}$$

Solution(normalized spectral density)

We know that

$$f^*(w) = \sum_{k=-\infty}^{\infty} \rho(k) e^{-2\pi w_i k}.$$

So ... Homework? (It should be easy, Right? Just use $\rho(k)$ instead of $\gamma(k)$).

Homework Problem

Let Y_t be the process defined by

$$Y_t = W_t + \beta W_{t-1}$$

(where $W_t \sim WN(0, \sigma_W^2)$). Compute the spectral density and normalized spectral density of Y_t .