



# The power spectral density

The power spectral density (PSD) of a WSS stochastic process is defined as

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over the frequencies  $-\pi < \omega \le \pi.$  The inverse transform recovers  $r_y(k)$ 

$$r_y(k) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \phi_y(\omega) e^{i\omega k} d\omega$$

It is worth noting that

$$r_y(0) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \phi_y(\omega) d\omega$$

which, for a zero-mean process, measures the power of  $y_t$ , i.e.,

$$r_y(0) = E\{|y_t|^2\}$$

The PSD is  $\it real-valued$  and  $\it non-negative.$  For a real-valued process, the PSD is symmetric, whereas it is non-symmetric for a complex-valued process.

As a white noise is uncorrelated,  $r_x(k)=\sigma_x^2\delta_K(k)$ , where  $\delta_K(k)$  is the Kronecker delta. Thus, the PSD of a white noise is

$$\phi_x(\omega) = \sum_{-\infty}^{\infty} r_x(k) e^{-i\omega k} = \sigma_x^2$$



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Example: Consider a sinusoidal process

$$y_t = A \cos(\omega_0 t + \phi) + w_t$$

where  $\boldsymbol{w}_t$  is an AWGN. The ACF of  $\boldsymbol{y}_t$  is then

$$r_y(k) = \frac{A^2}{2} \cos(\omega_0 k) + \sigma_w^2 \delta_K(k)$$

This implies that the PSD of  $y_t$  is

$$\phi_y(k)=\frac{A^2}{4}\delta_D(\omega-\omega_0)+\frac{A^2}{4}\delta_D(\omega+\omega_0)+\sigma_w^2$$
 where  $\sigma_D(\omega)$  is the Dirac delta, satisfying

$$f(a) = \int_{-\infty}^{\infty} f(x)\delta_D(x - a)dx$$



# Estimating the power spectral density

Under the weak assumption that

$$\lim_{N \to \infty} \frac{1}{N} \sum_{k=-N}^{N} |k| |r_y(k)| = 0$$

the PSD can be expressed equivalently as

$$\phi_y(\omega) = \lim_{N \to \infty} E\left\{\frac{1}{N} \left| \sum_{t=1}^{N} y_t e^{-i\omega t} \right|^2\right\}$$



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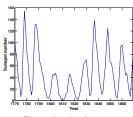
This suggests two natural estimators, namely as the periodogram

$$\hat{b}_{y}^{p}(\omega) = \frac{1}{N} \left| \sum_{t=1}^{N} y_{t}e^{-i\omega t} \right|^{2}$$

and the correlogram

$$\hat{\phi}_y^c(\omega) = \sum_{k=-N+1}^{N-1} \hat{r}_y(k)e^{-i\omega t}$$

Note that  $\hat{r}_y(k)$  should be the biased estimator



The annual number of sunspo



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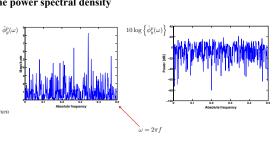
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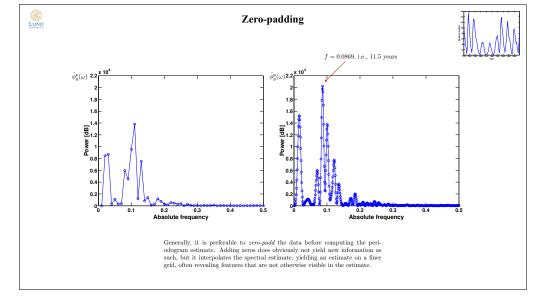
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These estimators are equivalent, and are asymptotically unbiased, but not consistent. Notably, the variance of the estimate is about the same as the true spectrum squared.







#### Windowing

It is often preferable to use some other time or lag window, for the periodogram

$$\hat{\phi}_y^p(\omega) = \frac{1}{N} \left| \sum_{t=1}^N y_t e^{-i\omega t} \right|^2 = \frac{1}{N} \left| \sum_{t=-\infty}^\infty v_t y_t e^{-i\omega t} \right|^2$$

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$$\hat{\phi}^c_y(\omega) = \sum_{k=-N+1}^{N-1} \hat{r}_y(k) e^{-i\omega k} = \sum_{k=-\infty}^{\infty} w_k \hat{r}_y(k) e^{-i\omega k}$$

where the time and lag windows,  $v_t$  and  $w_k$ , take the value 1 in the given range,

Such windows widens the main lobe, but lowers the sidelobes.



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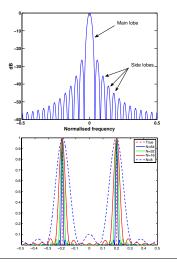
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Window Name	Defining Equation	Approx. Main Lobe Width (radians)	Sidelobe Level (dB)
Rectangular	w(k) = 1	$2\pi/M$	-13
Bartlett	$w(k) = \frac{M-k}{M}$	$4\pi/M$	-25
Hanning	$w(k) = 0.5 + 0.5\cos\left(\frac{\pi k}{M}\right)$	$4\pi/M$	-31
Hamming	$w(k) = 0.54 + 0.46\cos\left(\frac{\pi k}{M-1}\right)$	$4\pi/M$	-41
Blackman	$w(k) = 0.42 + 0.5 \cos \left(\frac{\pi k}{M-1}\right)$	$6\pi/M$	-57
	$+0.08 \cos \left(\frac{\pi k}{M-1}\right)$		



You can read more in the excellent textbook by Stoica and Moses (2005).

