

# Frequency Spectrum of a WSS RP

04/15  
(1 of 3)

\* Often useful to work w/ a RP in the frequency domain rather than the time domain

\* This differs from the characteristic & moment generating functions

Fourier Transform of the density function for a random variable

"freq. domain-esque" but does not relate to frequency in terms of how a signal varies over time

(Similar to using the frequency domain for deterministic signals)

\* Convolution may be avoided when finding  $R_Y(\tau) = (\tilde{h} * h * R_X)(\tau) \leftarrow \begin{matrix} \text{LTI} \\ \text{system} \end{matrix}$

\* Filter design for a deterministic signal is typically easiest when working with the frequency domain representation

(e.g., to filter noise, design a Low-Pass Filter & set cutoff frequency)

\* Noise in audio signals (e.g., white, pink, blue noise, etc.)

(defined in the freq. domain for audio signals)

How is the frequency spectrum defined for a Random Process?

⇒ Instead of finding the FT for each sample realization of the RP, use expectation to represent the frequency spectrum as a whole

(depends on the autocorrelation function)

\* Define the frequency spectrum in terms of the autocorrelation function of the process

# Power Spectral Density (PSD)

Def.: The PSD of a WSS RP  $X(t)$  is

$$S_x(\omega) = \int_{-\infty}^{\infty} R_x(\tau) e^{-i\omega\tau} d\tau \leftarrow (\text{FT of } R_x) \quad \omega \in \mathbb{R}$$

(frequency variable, not sample space)

\* Since  $R_x(-\tau) = R_x(\tau)$ ,  $S_x(\omega)$  is real-valued ("R<sub>x</sub> is an even function in  $\tau$ ")

\*  $R_x(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_x(\omega) e^{i\omega\tau} d\omega \leftarrow (\text{Inverse FT})$

\* Since  $R_x(\tau)$  is real,  $S_x(\omega)$  is an even function of  $\omega$ ,  $S_x(-\omega) = S_x(\omega)$ ,  $\forall \omega \in \mathbb{R}$

Def.: The Cross Power Spectral Density of jointly WSS RP  $X(t)$  &  $Y(t)$  is

$$S_{xy}(\omega) = \int_{-\infty}^{\infty} R_{xy}(\tau) e^{-i\omega\tau} d\tau$$

\* Recall that if  $X(t) \xrightarrow[\text{(WSS)}]{h(t)} Y(t) \xrightarrow[\text{(WSS)}]$  then  $R_y(\tau) = (\tilde{h} * h * R_x)(\tau)$

\* Let the frequency response be the FT of the impulse response:  $H(\omega) = \mathcal{F}\{h(t)\}$

$\Rightarrow \mathcal{F}\{\tilde{h}(t)\} = \mathcal{F}\{h(-t)\} = H^*(\omega) \leftarrow (\text{Complex conjugate})$

\* Even though  $S_x(\omega)$  &  $S_y(\omega)$  are real-valued,  $H(t)$  may not be "even" (where  $H(t) \neq H(-t)$ , thus  $H(\omega)$  would be complex)

$$\Rightarrow S_y(\omega) = \underbrace{H^*(\omega) \cdot H(\omega)}_{|H(\omega)|^2} \cdot S_x(\omega) \text{ or } |H(\omega)|^2 S_x(\omega), \quad \forall \omega \in \mathbb{R}$$

\* NOTE:  $R_{xy}(\tau) = (h * R_x)(\tau)$

Def.: A WSS RP is called White Noise if it is zero mean & the autocorrelation is a Dirac Delta function

$$\Rightarrow E[X(t)] = 0, \forall t \wedge R_x(\tau) = 0, \tau \neq 0$$

Thus, for white noise,  $E[X(t_1)X(t_2)] = 0, t_1 \neq t_2$

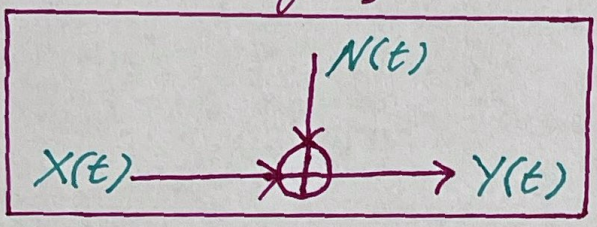
(for any 2 time instances that are non equal, the expectation of the product for these RVs is zero)

$\therefore X(t_1) \& X(t_2) \leftarrow$  UNCORRELATED (if  $t_1 \neq t_2$ )  
SINCE WHITE NOISE IS ZERO MEAN

\* Looking at 2 diff. time instances  $t_1$  &  $t_2$ , the RVs are uncorrelated

\* The Gaussian model & White Noise model are often combined to model noise

Example: WSS Signal,  $X(t)$

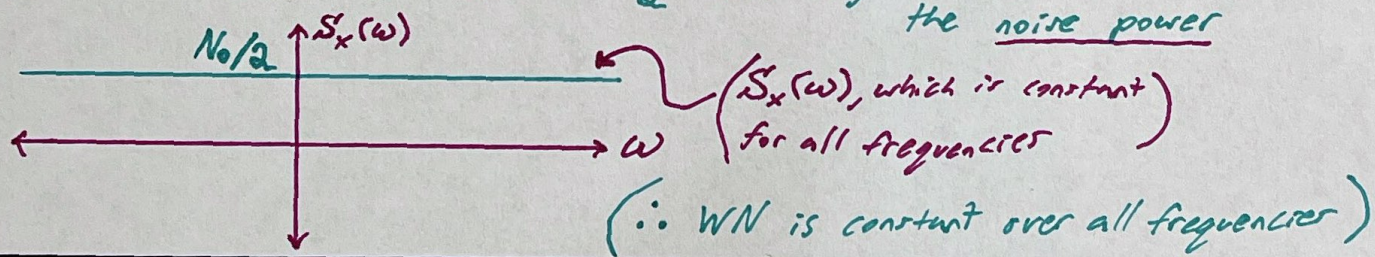


(Often assumed that  $N(t)$  is a Gaussian RP & is white noise)

$\Rightarrow$  Additive White Gaussian Noise model (AWGN), where white noise is an "idealization" - "true enough"

\* What happens when you filter white noise?

\* Let  $X(t)$  be WN w/  $R_x(\tau) = \frac{N_0}{2} \delta(\tau)$ , where  $N_0$  is constant & represents the noise power



\* Let  $h(t) = e^{-t} u(t) \leftarrow$  (decaying exponential)

\* What happens if WN is passed through filter  $h(t)$ ?

$$\mu_y = \mu_x \int_{-\infty}^{\infty} h(t) dt = 0 \quad \rightarrow \text{(Inverse FT)} \quad R_y(\tau) = \frac{N_0}{4} e^{-|\tau|}, \tau \in \mathbb{R}$$

$$H(\omega) = \frac{1}{1+j\omega} \Rightarrow S_y(\omega) = |H(\omega)|^2 S_x(\omega) = \frac{N_0/2}{1+\omega^2}$$

