

# The Wiener-Khinchin Theorem

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For any signal  $x(t)$ , let

$$x_T(t) = x(t) \operatorname{rect}\left(\frac{t}{T}\right)$$

denote a “time-windowed” projection of  $x(t)$  taking value zero outside of the interval  $[-T/2, T/2)$ , where  $T > 0$ . Assume, for each  $T$ , that the Fourier transform of  $x_T(t)$  exists, and is given by  $X_T(f)$ . To a power signal  $x$  we may associate the *power spectral density* given by

$$S_x(f) = \lim_{T \rightarrow \infty} \frac{1}{T} |X_T(f)|^2,$$

measured in units of W/Hz. Roughly speaking,  $S_x(f)$  measures the contribution to the power of  $x$  made by complex-exponential signal components at frequency  $f$ . The total power associated with  $x$  is then given by

$$P_x = \int_{-\infty}^{\infty} S_x(f) \, df.$$

If  $x$  is passed through an LTI system with frequency response  $H(f)$ , then the power spectral density of the output  $y$  is given by  $S_y(f) = S_x(f)|H(f)|^2$ , having total power

$$P_y = \int_{-\infty}^{\infty} S_x(f)|H(f)|^2 \, df.$$

In particular, note that if  $H(f)$  is a very narrow bandpass filter centered at some frequency  $f_0$ , then the power of the output is approximately proportional to  $S_x(f_0)$ ; thus, the function  $S_x(f)$  does indeed serve as a density function for power.

We would like to extend this notion of power spectral density to wide-sense stationary random processes. For any fixed power signal  $x$ , at any given frequency  $f$ , observe that  $S_x(f)$  is some

fixed non-negative real value that depends on  $x$ . For a wide-sense stationary random process  $X$  having power signals as sample functions, it makes sense to define the power spectral density via the expected value of that real value, i.e., to define

$$S_X(f) = \lim_{T \rightarrow \infty} \frac{1}{T} E[|X_T(f)|^2], \quad (1)$$

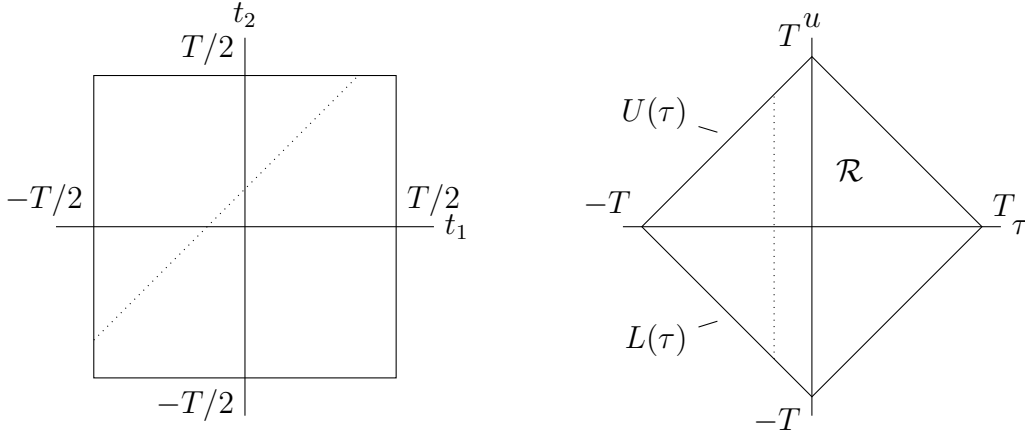
whenever the limit exists. We take (1) as the definition of the power spectral density.

Suppose now that the (complex-valued) random process has autocorrelation function  $R_X(\tau) = E[X(t)X^*(t - \tau)]$ , and that the Fourier transform of  $R_X(\tau)$  exists and is denoted  $\hat{R}_X(f)$ . The Wiener-Khinchin theorem states that, under mild conditions,  $S_X(f) = \hat{R}_X(f)$ , i.e., that the power spectral density associated with a wide-sense stationary random process is equal to the Fourier transform of the autocorrelation function associated with that process.

To see this, we follow [1, Sec. 11.2], and note that according to our definition (1), we have

$$\begin{aligned} S_X(f) &= \lim_{T \rightarrow \infty} \frac{1}{T} E[|X_T(f)|^2] \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} E \left[ \int_{-T/2}^{T/2} X(t_1) e^{-j2\pi f t_1} dt_1 \int_{-T/2}^{T/2} X^*(t_2) e^{j2\pi f t_2} dt_2 \right] \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \int_{-T/2}^{T/2} E[X(t_1)X^*(t_2)] e^{-j2\pi f(t_1-t_2)} dt_2 dt_1 \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \int_{-T/2}^{T/2} R_X(t_1 - t_2) e^{-j2\pi f(t_1-t_2)} dt_2 dt_1. \end{aligned}$$

The region of integration is the square region shown on the left (below).



Note however that the integrand is constant along contours where  $t_1 - t_2$  is a constant (e.g., along the dotted line shown). This motivates us to apply a change of variables, defining  $\tau = t_1 - t_2$  and  $u = t_1 + t_2$ . Under this change of variables, after noting that

$$\begin{bmatrix} \tau \\ u \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \end{bmatrix},$$

so that we have a Jacobian matrix of determinant 2, we get

$$S_X(f) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{\mathcal{R}} R_X(\tau) e^{-j2\pi f\tau} \frac{du d\tau}{2},$$

where  $\mathcal{R}$  is the rotated-and-scaled region shown on the right (in the figure above). Note that other changes of variables are possible; for example one might take  $\tau = t_1 - t_2$ ,  $u = t_1$ . The region of integration will be different, but the final answer will be the same.

For each value of  $\tau$ , let us denote the value of  $u$  along lower boundary of the region  $\mathcal{R}$  as  $L(\tau)$ , and let us denote the value of  $u$  along the upper boundary as  $U(\tau)$ , as indicated in the figure. This gives us

$$\begin{aligned} S_X(f) &= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{L(\tau)}^{U(\tau)} R_X(\tau) e^{-j2\pi f\tau} du d\tau \\ &= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_X(\tau) e^{-j2\pi f\tau} \left( \int_{L(\tau)}^{U(\tau)} du \right) d\tau \\ &\stackrel{(a)}{=} \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_X(\tau) e^{-j2\pi f\tau} (2T - 2|\tau|) d\tau \\ &= \lim_{T \rightarrow \infty} \int_{-T}^T R_X(\tau) \left( 1 - \frac{|\tau|}{T} \right) e^{-j2\pi f\tau} d\tau, \end{aligned} \tag{2}$$

where (a) follows from the fact that, when  $|\tau| \leq T$ , we have  $U(\tau) - L(\tau) = 2T - 2|\tau|$ .

We recognize that the integral in (2) is simply taking the Fourier transform of the product of  $R_X(\tau)$  and the triangle function  $\frac{1}{T} \text{rect}(\tau/T) \star \text{rect}(\tau/T)$ . Applying the modulation theorem (i.e., the Fourier transform of a product is the convolution of the Fourier transforms), we get

$$\begin{aligned} S_X(f) &= \lim_{T \rightarrow \infty} \hat{R}_X(f) \star T \text{sinc}^2(fT) \\ &= \lim_{T \rightarrow \infty} \int_{-\infty}^{\infty} \hat{R}_X(f - x) T \text{sinc}^2(xT) dx. \end{aligned}$$

For any fixed value of  $f$ , observe that the convolution is carrying out a “weighted average” of the values of  $\hat{R}_X$ , with values near  $f$  (corresponding to small  $|x|$ ) given larger weight and with values farther from  $f$  (corresponding to large  $|x|$ ) given smaller weight. Indeed, the function  $T \text{sinc}^2(fT)$  has unit area, is everywhere non-negative, and, for  $f \neq 0$ , converges to zero as  $T \rightarrow \infty$ . In other words,  $T \text{sinc}^2(fT)$  seems to have, when  $T$  is large, the properties of a Dirac delta. Thus, we are tempted to write

$$S_X(f) = \lim_{T \rightarrow \infty} \hat{R}_X(f) \star T \text{sinc}^2(fT) = \hat{R}_X(f) \star \delta(f) = \hat{R}_X(f).$$

To justify this, let us abstract the situation. Suppose that we have a family of real-valued functions  $s_T(x)$ , parameterized by a positive real number  $T$ , with the properties that

1.  $s_T(x) \geq 0$  for all  $x$  and all  $T > 0$ ,
2.  $\int_{-\infty}^{\infty} s_T(x) dx = 1$  for all  $T > 0$ , and
3. for every  $\epsilon > 0$  and for every  $\delta > 0$ , we can find a  $T_0 > 0$  such that  $\int_{-\delta}^{\delta} s_T(x) dx \geq 1 - \epsilon$  for every  $T \geq T_0$ .

Such a family is called an “approximate identity under convolution.” It is not hard to show that  $\{T \operatorname{sinc}^2(xT) : T > 0\}$  is indeed such a family.

**Lemma 1.** *Let  $g(x)$  be a given real-valued function that is continuous at  $x = 0$ , and is bounded by some  $B \geq 1$ , i.e., for all  $x$ ,  $|g(x)| \leq B$ . Let  $\{s_T(x) : T > 0\}$  be an approximate identity under convolution, and furthermore suppose that*

$$\int_{-\infty}^{\infty} g(x)s_T(x) dx$$

*exists for all  $T > 0$ . We then have that*

$$\lim_{T \rightarrow \infty} \int_{-\infty}^{\infty} g(x)s_T(x) dx = g(0). \quad (3)$$

*Proof.* We need to show that for every  $\epsilon > 0$  there exists  $T_0 > 0$  such that  $T \geq T_0$  implies

$$\left| \int_{-\infty}^{\infty} g(x)s_T(x) dx - g(0) \right| \leq \epsilon.$$

Thus, let us fix  $\epsilon > 0$ . As a first step, let us choose a value  $\delta > 0$  so that whenever  $|x| \leq \delta$ , we have that  $|g(x) - g(0)| \leq \epsilon/3$ ; since  $g(x)$  is continuous at  $x = 0$ , such a  $\delta$  certainly exists. Next, let us choose  $T_0$  such that for all  $T \geq T_0$  we have  $\int_{-\delta}^{\delta} s_T(x) dx \geq 1 - \epsilon/(3B)$ ; by the third property of approximate identities, such a  $T_0$  certainly exists. Now, let us write

$$\int_{-\infty}^{\infty} g(x)s_T(x) dx = \int_{-\delta}^{\delta} g(x)s_T(x) dx + \int_{-\infty}^{-\delta} g(x)s_T(x) dx + \int_{\delta}^{\infty} g(x)s_T(x) dx$$

and bound the contributions made to the left-hand integral by the terms on the right-hand side. With the choices for  $\delta$  and  $T_0$  just made, we observe that

$$\int_{-\delta}^{\delta} g(x)s_T(x) dx \leq \int_{-\delta}^{\delta} (g(0) + \epsilon/3)s_T(x) dx \leq \int_{-\infty}^{\infty} (g(0) + \epsilon/3)s_T(x) dx = g(0) + \epsilon/3.$$

Furthermore

$$\begin{aligned} \int_{-\delta}^{\delta} g(x)s_T(x) dx &\geq \int_{-\delta}^{\delta} (g(0) - \epsilon/3)s_T(x) dx \geq (g(0) - \epsilon/3)(1 - \epsilon/(3B)) \\ &\stackrel{(a)}{\geq} (g(0) - \epsilon/3)(1 - \epsilon/3) \geq g(0) - 2\epsilon/3, \end{aligned}$$

where inequality (a) follows from the fact that  $B \geq 1$ . Next let

$$I_2 = \int_{-\infty}^{-\delta} g(x)s_T(x) dx + \int_{\delta}^{\infty} g(x)s_T(x) dx,$$

and observe that

$$|I_2| \leq \int_{-\infty}^{-\delta} B s_T(x) dx + \int_{\delta}^{\infty} B s_T(x) dx \leq B\epsilon/(3B) = \epsilon/3.$$

In sum, we see that for all  $T \geq T_0$  we have

$$g(0) - \epsilon \leq \int_{-\infty}^{\infty} g(x)s_T(x) dx \leq g(0) + 2\epsilon/3 \leq g(0) + \epsilon,$$

which is what we set out to show. □

Observe that Lemma 1 extends to complex-valued functions  $g(x)$  that are bounded in magnitude and continuous at  $x = 0$ , as the real and imaginary parts of the integrand can be treated separately.

Note that, even though  $s_T(t)$  does not approach a well-defined *function* as  $T \rightarrow \infty$ , the limit in (3) nevertheless converges to a well-defined linear *functional*: namely, the one that maps  $g(x)$  to its value  $g(0)$  at  $x = 0$ . This is the essence of the definition of the Dirac delta as a distribution; see, e.g., [2].

Now suppose that  $R_X(\tau)$  is absolutely integrable, i.e., that  $B = \int_{-\infty}^{\infty} |R_X(\tau)| d\tau$  exists. We then have, for any  $f$ , that

$$|\hat{R}_X(f)| = \left| \int_{-\infty}^{\infty} R_X(\tau)e^{-j2\pi f\tau} d\tau \right| \leq \int_{-\infty}^{\infty} |R_X(\tau)e^{-j2\pi f\tau}| d\tau = \int_{-\infty}^{\infty} |R_X(\tau)| d\tau = B.$$

Thus, if  $R_X(\tau)$  is absolutely integrable, its Fourier transform  $\hat{R}_X(f)$  (if it exists) is bounded.

Now let us put the pieces together. First, we already have that

$$S_X(f) = \lim_{T \rightarrow \infty} \int_{-\infty}^{\infty} \hat{R}_X(f - x)T \operatorname{sinc}^2(xT) dx$$

and we know that  $\{T \operatorname{sinc}^2 xT) : T > 0\}$  is an approximate identity for convolution. Furthermore, assuming that  $R_X(\tau)$  is absolutely integrable implies that  $\hat{R}_X(f)$  is bounded. Thus, Lemma 1 applies, and we get that

$$S_X(f) = \hat{R}_X(f)$$

at all points  $f$  where  $\hat{R}_X(f)$  is continuous. This is the Wiener-Khinchin theorem.

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## References

- [1] B. P. Lathi, *Modern Digital and Analog Communication Systems*, 3rd Edition. Oxford University Press, 1998.
- [2] R. S. Strichartz, *A Guide to Distribution Theory and Fourier Transforms*. CRC Press, 1994.