

# Gaussian Random Processes

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# Gaussian Random Process

## Definition

A random process  $X(t)$  is Gaussian if its samples  $X(t_1), \dots, X(t_n)$  are jointly Gaussian for any  $n \in \mathbb{N}$  and distinct sample locations  $t_1, t_2, \dots, t_n$ .

Let  $\mathbf{X} = [X(t_1) \ \dots \ X(t_n)]^T$  be the vector of samples. The joint density is given by

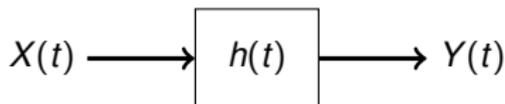
$$p(\mathbf{x}) = \frac{1}{\sqrt{(2\pi)^n \det(\mathbf{C})}} \exp\left(-\frac{1}{2}(\mathbf{x} - \mathbf{m})^T \mathbf{C}^{-1}(\mathbf{x} - \mathbf{m})\right)$$

where

$$\mathbf{m} = E[\mathbf{X}], \quad \mathbf{C} = E[(\mathbf{X} - \mathbf{m})(\mathbf{X} - \mathbf{m})^T]$$

# Properties of Gaussian Random Process

- The mean and autocorrelation functions completely characterize a Gaussian random process.
- Wide-sense stationary Gaussian processes are strictly stationary.
- If the input to a stable linear filter is a Gaussian random process, the output is also a Gaussian random process.



# White Gaussian Noise

## Definition

A zero mean WSS Gaussian random process with constant power spectral density

$$S_n(f) = \frac{N_0}{2}.$$

$\frac{N_0}{2}$  is termed the two-sided PSD and has units Watts per Hertz.

## Remarks

- Autocorrelation function  $R_n(\tau) = \frac{N_0}{2} \delta(\tau)$
- **Infinite Power!** Ideal model of Gaussian noise occupying more bandwidth than the signals of interest.

# White Gaussian Noise through Correlators

- Consider the output of a correlator with WGN input

$$Z = \int_{-\infty}^{\infty} n(t)u(t) dt = \langle n, u \rangle$$

where  $u(t)$  is a deterministic finite-energy signal

- $Z$  is a Gaussian random variable
- The mean of  $Z$  is

$$E[Z] = \int_{-\infty}^{\infty} E[n(t)] u(t) dt = 0$$

- The variance of  $Z$  is

$$\begin{aligned} \text{var}(Z) &= E[(\langle n, u \rangle)^2] = E\left[\int n(t)u(t) dt \int n(s)u(s) ds\right] \\ &= \int \int u(t)u(s)E[n(t)n(s)] dt ds \\ &= \int \int u(t)u(s)\frac{N_0}{2}\delta(t-s) dt ds \\ &= \frac{N_0}{2} \int u^2(t) dt = \frac{N_0}{2} \|u\|^2 \end{aligned}$$

# White Gaussian Noise through Correlators

## Proposition

Let  $u_1(t)$  and  $u_2(t)$  be linearly independent finite-energy signals and let  $n(t)$  be WGN with PSD  $S_n(f) = \frac{N_0}{2}$ . Then  $\langle n, u_1 \rangle$  and  $\langle n, u_2 \rangle$  are jointly Gaussian with covariance

$$\text{cov}(\langle n, u_1 \rangle, \langle n, u_2 \rangle) = \frac{N_0}{2} \langle u_1, u_2 \rangle.$$

## Proof

To prove that  $\langle n, u_1 \rangle$  and  $\langle n, u_2 \rangle$  are jointly Gaussian, consider a non-trivial linear combination  $a\langle n, u_1 \rangle + b\langle n, u_2 \rangle$

$$a\langle n, u_1 \rangle + b\langle n, u_2 \rangle = \int n(t) [au_1(t) + bu_2(t)] dt.$$

This is the result of passing  $n(t)$  through a correlator. So it is a Gaussian random variable.

# White Gaussian Noise through Correlators

## Proof (continued)

$$\begin{aligned}\text{cov}(\langle n, u_1 \rangle, \langle n, u_2 \rangle) &= E[\langle n, u_1 \rangle \langle n, u_2 \rangle] \\ &= E\left[\int n(t)u_1(t) dt \int n(s)u_2(s) ds\right] \\ &= \int \int u_1(t)u_2(s)E[n(t)n(s)] dt ds \\ &= \int \int u_1(t)u_2(s)\frac{N_0}{2}\delta(t-s) dt ds \\ &= \frac{N_0}{2} \int u_1(t)u_2(t) dt \\ &= \frac{N_0}{2} \langle u_1, u_2 \rangle\end{aligned}$$

If  $u_1(t)$  and  $u_2(t)$  are orthogonal,  $\langle n, u_1 \rangle$  and  $\langle n, u_2 \rangle$  are independent.

## Reference

- Chapter 3, *Fundamentals of Digital Communication*, Upamanyu Madhow, Cambridge University Press, 2008.

Questions?