

Digital Communication Systems Engineering with Software-Defined Radio

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Lecture 05

Definition

- ▶ *Noise* is an unwanted signal that is superimposed on top of a useful information signal
- ▶ Introduction of noise to useful information signal usually occurs prior to interception and decoding by receiver
- ▶ Noise can be modeled as a *random variable*
 - ▶ Different noise-generating phenomena can be modeled by different random variables
 - ▶ Although noise signal not known precisely, its statistical characteristics can be used in the communication system design process

Noise Sources

- ▶ Noise originates from multiple sources within the communication system transmission environment
 - ▶ Natural sources
 - ▶ Solar flare activity and atmospheric conditions, e.g., ionosphere
 - ▶ Johnson-Nyquist noise, i.e., thermal noise in conductive materials
 - ▶ Lightning
 - ▶ Cosmic background noise
 - ▶ Human-built sources
 - ▶ Microwave ovens (2.4 GHz)
 - ▶ Power transmission lines (60 Hz)
 - ▶ Other wireless transmissions

Noise Blurs Intercepted Signals

- ▶ Randomness of noise makes detection and decoding of useful signals difficult

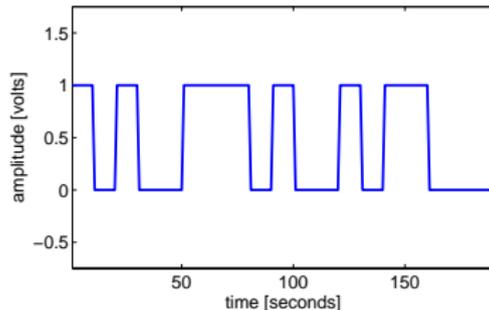


Figure : Signal Before Noise is Added. Notice Clear Pulse Shapes of Signal.

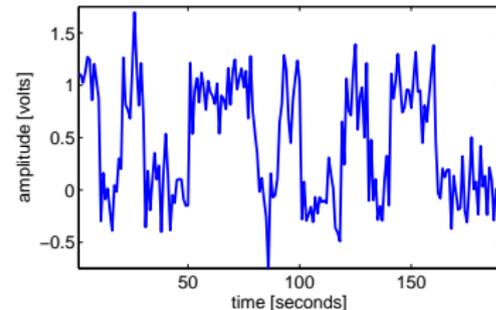


Figure : Signal After Noise is Added. Difficult for Receiver to Ascertain Pulse Shapes.

Gaussian Random Variable

- ▶ Often communication system engineers usually assume the noise to possess a *Gaussian* distribution
- ▶ Gaussian random variable $x \sim \mathcal{N}(\mu_x, \sigma_x^2)$ possesses a mean μ_x and variance σ_x^2
- ▶ Statistical behavior can be entirely characterize by the *probability density function* (PDF) of the random phenomenon
 - ▶ A Gaussian PDF is defined as:

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} e^{-\frac{(x-\mu_x)^2}{2\sigma_x^2}} \quad (1)$$

Computing Probabilities

- ▶ Probability density functions entirely capture the statistical behavior of a random variable
 - ▶ Although the exact random variable outcome is unknown, we can associate a probability to a specific outcome or range of outcomes occurring
- ▶ In general, to determine the probability that a random variable X will produce a value greater than or equal to a , solve:

$$P(X \geq a) = \int_a^{\infty} f_X(x) dx \quad (2)$$

- ▶ Since it is difficult to integrate a Gaussian PDF, we can alternatively use a *Q-function*, which is defined as:

$$Q(y) = \frac{1}{\sqrt{2\pi}} \int_y^{\infty} e^{-\frac{t^2}{2}} dt \quad (3)$$

- ▶ Use look-up tables to find actual numerical value

“White” = Uncorrelated

- ▶ *White noise* refers to a noise signal whose power spectral density (PSD) is flat across the entire frequency range
- ▶ The term “white” is used since white light possesses a PSD that spans the entire frequency range of visible light
- ▶ Another interpretation of white noise is that the outcome at every time sample is uncorrelated with every other time sample except itself
 - ▶ As we will see in Lecture 6, the PSD and autocorrelation function are related to each other
 - ▶ PSD $S(f) = \text{constant}$ is related via inverse Fourier transform to autocorrelation $R(\tau) = c\delta(\tau)$
- ▶ Only for Gaussian random variables does *uncorrelated* also mean *independent*

Gaussian Noise Example

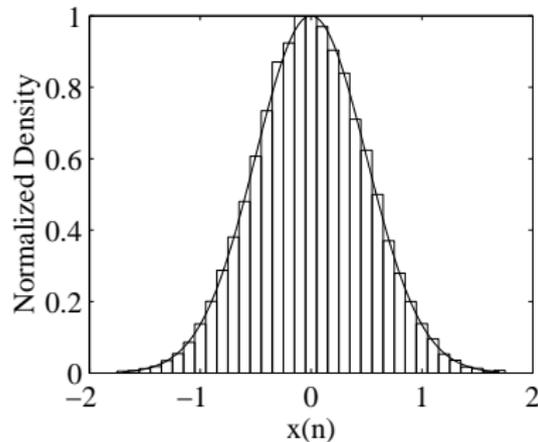


Figure : Example of a Gaussian Probability Density Function.

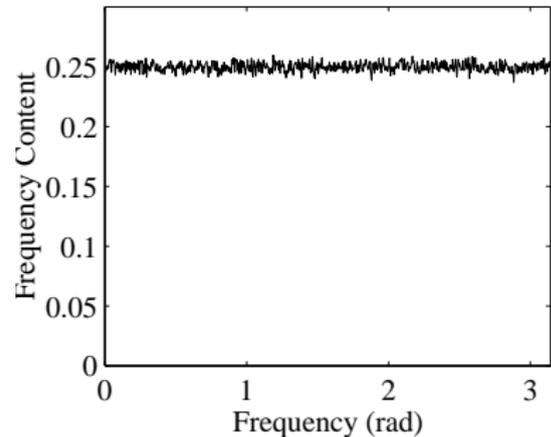


Figure : Example of a Gaussian Power Spectral Density.

Additive Noise Model

- ▶ Additive noise model is often used to describe how unwanted impairments are added to a useful information signal
- ▶ Usually $n(t)$ represents the culmination of all noise sources into a single variable
- ▶ Receiver only observes the corrupted version of $s(t)$ by $n(t)$, namely $r(t)$

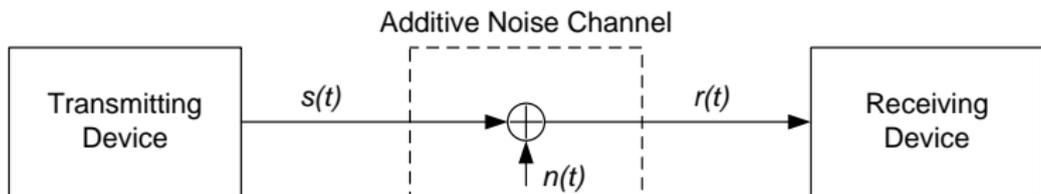


Figure : Simple Additive Noise Model.

Narrowband Noise Fundamentals

- ▶ Most transmissions are *bandlimited*
 - ▶ Constraint on amount of spectrum that can be used
 - ▶ Larger bandwidths yield more expensive wireless hardware
 - ▶ Narrowband filters at transmitter and receiver allow only modulated signal to pass
- ▶ A portion of the noise intercepted at the receiver also passes through the narrowband filter
 - ▶ If noise is white, then narrowband noise is shaped like a cosine-modulated bandpass filter response
 - ▶ How does this look like in the frequency domain? Time domain?

Mathematical Framework

- ▶ Need a convenient mathematical framework to represent narrowband noise
 - ▶ In-phase/Quadrature Representation
 - ▶ Envelope/Phase Representation
- ▶ Given $x = Ae^{j\phi} = a + jb$, where $x \in \mathbb{C}$ and $A, \phi, a, b \in \mathbb{R}$, we have the following relationships:
 - ▶ $A = \sqrt{a^2 + b^2}$ and $\phi = \tan^{-1}(b/a)$
 - ▶ $a = A \cos(\phi)$ and $b = A \sin(\phi)$

In-phase/Quadrature Representation

- ▶ Suppose we define the complex envelope (complex baseband) of a narrowband noise signal as:

$$\tilde{n}(t) = n_I(t) + jn_Q(t) \quad (4)$$

- ▶ By definition, the bandpass version of the narrowband noise signal is:

$$n(t) = \text{Real} \left\{ \tilde{n}(t) e^{j2\pi f_c t} \right\} \quad (5)$$

- ▶ Using Euler's Relationship, namely $e^{j\omega} = \cos(\omega) + j \sin(\omega)$, we get the following expression:

$$n(t) = n_I(t) \cos(2\pi f_c t) - n_Q(t) \sin(2\pi f_c t) \quad (6)$$

- ▶ How do you generate/extract $n_I(t)$ and $n_Q(t)$?

Several Important Properties of I/Q Representation

- ▶ Both $n_I(t)$ and $n_Q(t)$ have zero mean
- ▶ If $n(t)$ is Gaussian, so are $n_I(t)$ and $n_Q(t)$
- ▶ If $n(t)$ is stationary, $n_I(t)$ and $n_Q(t)$ are jointly stationary
- ▶ The PSD of $n_I(t)$ and $n_Q(t)$ are equal to:

$$S_{N_I}(f) = S_{N_Q}(f) = \begin{cases} S_N(f - f_c) + S_N(f + f_c), & -B \leq f \leq B \\ 0, & \text{otherwise} \end{cases} \quad (7)$$

- ▶ Both $n_I(t)$ and $n_Q(t)$ have the same variance as $n(t)$
- ▶ If $n(t)$ is Gaussian and its PSD symmetric, then $n_I(t)$ and $n_Q(t)$ are statistically independent
- ▶ The cross spectral density between $n_I(t)$ and $n_Q(t)$ is purely imaginary, and for $-B \leq f \leq B$ it is equal to (zero otherwise):

$$S_{N_I N_Q}(f) = -S_{N_Q N_I}(f) = j(S_N(f + f_c) - S_N(f - f_c)) \quad (8)$$

Envelope/Phase Representation

- ▶ We could also express $n(t)$ in terms of envelope and phase:

$$n(t) = r(t) \cos(2\pi f_c t + \phi(t)) \quad (9)$$

where $r(t) = \sqrt{n_I(t)^2 + n_Q(t)^2}$ is the envelope and $\phi(t) = \tan^{-1}(n_Q(t)/n_I(t))$ is the phase

- ▶ Given a joint PDF for $n_I(t)$ and $n_Q(t)$ equal to:

$$f_{N_I N_Q}(n_I, n_Q) = \frac{1}{2\pi\sigma^2} e^{-\frac{n_I^2 + n_Q^2}{2\sigma^2}}, \quad (10)$$

and using the relationships $n_I = r \cos(\phi)$ and $n_Q = r \sin(\phi)$ as well as a *Jacobian* in order to get:

$$f_{R\Phi}(r, \phi) = \begin{cases} \frac{r}{2\pi\sigma^2} e^{-\frac{r^2}{2\sigma^2}}, & r \geq 0 \text{ and } 0 \leq \phi \leq 2\pi \\ 0, & \text{otherwise} \end{cases} \quad (11)$$

- ▶ Notice the combination of *Rayleigh* and *Uniform* PDFs