# Energy and Noise

- Signal metrics: mean, power, energy
- Signal-to-Noise Ratio
- Random processes
- Probability Density Function, mean, variance

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Energy and Noise, Slide 1



 Assuming we get our demodulation frequency and phase just right, we can *perfectly* reconstruct the originally transmitted signal at the receiver

Is this true in the *real* world?

#### Lab 3 at a *Real* Lab Bench

Measured receive signal using monitor\_receive('rx\_a'):



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#### The Issue of Noise



- Noise is a non-predictable (i.e. random), corrupting signal that adds to the desired signal
  - For RF receiver, most of it comes from the analog circuits that amplify and demodulate the input signal
- An undesired signal is a *predictable*, corrupting signal which also adds to the desired signal
  - May be called noise if it is difficult to predict

#### **Energy Transfer in Wireless Communication**



- Receiver antenna is limited in its ability to capture transmitter energy according to its area and distance, r, from transmitter
  - Received signal energy is a function of these parameters
  - In free space, received energy is proportional to  $1/r^2$

#### Signal Versus Noise



- Moving the receiver closer to the transmitter increases desired signal energy
  - Noise from analog receiver circuits is unchanged

How is system performance impacted?

#### Development of *Metrics* for Analysis



- It is often useful to create a mapping between a signal waveform and a numerical value
  - Such a mapping is called a *metric*
  - Examples: energy, power, average, variance
- In this class, we prefer to do analysis on discrete-time signals
  - Our labs focus on Matlab sequences rather than analog signals in the underlying hardware
  - Key ideas transfer to analog signal analysis quite readily

#### Definition of Mean, Power, and Energy



• DC average or mean,  $\mu_x$ , is defined as

$$\mu_x = \frac{1}{N} \sum_{k=0}^{N-1} x[n]$$

• Power,  $P_x$ , and energy,  $E_x$ , are defined as

$$P_x = \frac{1}{N} \sum_{k=0}^{N-1} x[n]^2 \qquad \qquad E_x = \sum_{k=0}^{N-1} x[n]^2$$

- For communication systems, we often remove the mean since it is essentially irrelevant in terms of *information*:

$$\tilde{P}_x = \frac{1}{N} \sum_{k=0}^{N-1} (x[n] - \mu_x)^2 \qquad \tilde{E}_x = \sum_{k=0}^{N-1} (x[n] - \mu_x)^2$$

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#### Definition of Signal-to-Noise Ratio



 Signal-to-Noise ratio (SNR) indicates the relative impact of noise in system performance

$$SNR = \frac{\tilde{P}_{signal}}{\tilde{P}_{noise}}$$

• We often like to use units of dB to express SNR:

SNR (dB) = 
$$10 \log \left( \frac{\tilde{P}_{signal}}{\tilde{P}_{noise}} \right)$$

#### SNR Example



- Scaling the gain factor A leads to different SNR values
  - Lower A results in lower SNR
  - Signal quality steadily degrades with lower SNR



#### Analysis of Random Processes

- Random processes, such as noise, take on different sequences for different trials
  - Think of trials as different measurement intervals from the same experimental setup (as in Lab)
- For a given trial, we can apply our standard analysis tools and metrics
  - Fourier transform, mean and power calculations, etc...
- When trying to analyze the ensemble (i.e. all trials) of possible outcomes, we find ourselves in need of new tools and metrics



#### **Tools and Metrics for Random Processes**

- Assume that random processes we will deal with have the properties of being stationary and ergodic
  - True for noise in many practical communication systems
  - Greatly simplifies analysis 6.011 will provide details
- Examine in both time and frequency domains
  - Time domain
    - Introduce the concept of a *probability density function* (PDF) to characterize behavior of signals at a given sample time
    - Use PDF to calculate mean and variance
      - Similar to mean and power of non-random signals
  - Frequency domain
    - $\cdot$  We must wait for 6.011 to give you the proper framework
    - For now, we will simply use Fourier analysis on signals from individual trials as done in Labs
      - We will give some hints of *ensemble* behavior ...

#### Stationary and Ergodic Random Processes

#### Stationary

- Statistical behavior is independent of *shifts* in *time* in a given trial:
  - Implies noise[k] is statistically indistinguishable from noise[k+N]
- · Ergodic
  - Statistical sampling can be performed at one sample time (i.e., n=k) across different trials, or across different sample times of the same trial with no change in the measured result



## Examples

Non-Stationary

Stationary, but Non-Ergodic





#### Experiment to see Statistical Distribution



- Create histograms of sample values from trials of increasing lengths
- Assumption of stationarity implies histogram should converge to a shape known as a probability density function (PDF)



#### Formalizing the PDF Concept

- Define x as a random variable whose PDF has the same shape as the histogram we just obtained
- Denote PDF of x as  $f_{x}(x)$ 
  - Scale  $f_{\chi}(x)$  such that its overall area is 1

$$\Rightarrow \int_{-\infty}^{\infty} f_X(x) = 1$$



#### Formalizing Probability

• The probability that random variable x takes on a value in the range of  $x_1$  to  $x_2$  is calculated from the PDF of x as:



- Note that probability values are always in the range of 0 to 1
  - Higher probability values imply greater likelihood that the event will occur

#### Example Probability Calculation



• Verify that overall area is 1:

$$\int_{-\infty}^{\infty} f_X(x) dx = \int_0^2 0.5 \, dx = 1$$

 Probability that x takes on a value between 0.5 and 1.0:

$$Prob(0.5 \le x \le 1.0) = \int_{0.5}^{1.0} 0.5 \, dx = 0.25$$

#### Examination of Sample Value Distribution



 Assumption of ergodicity implies the value occurring at a given time sample, noise[k], across many different trials has the same PDF as estimated in our previous experiment of many time samples and one trial

• We can model *noise[k]* as the random variable X M.H. Perrott © 2007 Energy and Noise, Slide 19

#### **Probability Calculation**



• In a given trial, the *probability* that *noise[k]* takes on a value in the range of  $x_1$  to  $x_2$  is computed as

$$Prob(x_1 \le x \le x_2) = \int_{x_1}^{x_2} f_X(x) dx$$

We will return to this when we analyze performance of digital modulation systems

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• The mean of random variable x,  $\mu_x$ , corresponds to its average value

- Computed as 
$$\mu_x = \int_{-\infty}^{\infty} x f_X(x) dx$$

- The variance of random variable x,  $\sigma^2_x$ , gives an indication of its variability
  - Computed as

$$\sigma_x^2 = \int_{-\infty}^{\infty} (x - \mu_x)^2 f_X(x) dx$$

- Similar to power of a signal

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### Visualizing Mean and Variance from PDF



- Changes in mean shift the *center of mass* of PDF
- Changes in variance narrow or broaden the PDF
  - Note that area of PDF must always remain equal to one

#### **Example Mean and Variance Calculation**



• Mean:

$$\mu_x = \int_{-\infty}^{\infty} x f_X(x) dx = \int_0^2 x \frac{1}{2} dx = \left. \frac{1}{4} x^2 \right|_0^2 = \boxed{1}$$

• Variance:

$$\sigma_x^2 = \int_{-\infty}^{\infty} (x - \mu_x)^2 f_X(x) dx = \int_0^2 (x - 1)^2 \frac{1}{2} dx$$
$$= \frac{1}{6} (x - 1)^3 \Big|_0^2 = \frac{1}{6} + \frac{1}{6} = \frac{1}{3}$$

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## Frequency Domain View of Random Process

- It is valid to take *fft* of sequence from a given trial
  - We did this in lab
- Notice that
  the *fft* result
  changes for
  different trials
  - We saw this in the *spectrum* plots of lab 3
  - Fourier Transform is undefined !





#### White Noise

#### White Noise



- When the *fft* result (i.e. *spectrum*) looks relatively flat, we refer to the random process as being white
  - Note: this type of noise source is often used for calibration of advanced stereo systems

#### Shaped Noise



- Shaped noise occurs when white noise is sent into a filter
  - *fft* of shaped noise will have frequency content according to the type of filter
    - Example: highpass filter yields shaped noise with only high frequency content

#### Summary

- A useful metric characterizing the performance of communication systems is Signal-to-Noise Ratio
  - High SNR values are desirable
  - SNR often varies in a wireless system according to the distance between transmitter and receiver
- Analysis of random processes (such as noise) requires additional tools
  - Concepts of stationarity and ergodicity
  - Random variables and their associated PDF functions
  - Metrics such as mean and variance
- Take 6.011 to learn more about random processes
  - Proper framework for frequency domain analysis
  - Advanced topics such as estimation