Topic 3: Signal Representation in Communication Systems

Telecommunication Systems Fundamentals

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Concepts in this Chapter

- **Signal Classifications**
- **Band-Pass Signals and Systems**
- **Representations:**
  - Complex Envelope
  - Base-Band Equivalent
  - Discrete Base-Band Equivalent

Theory classes: 4 sessions (8 hours)
Problems resolution: 1 session (2 hours)
Bibliography


Sistemas de Comunicación. S. Haykin. Wiley

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Signal Classifications

- **Contiguous Time / Discrete Time**
  - Analogue signals / Digital Signals (amplitude)

- **Stochastic (random) / Deterministic signals**
  - For Deterministic: finite energy / power defined.
  - For Power Defined: periodic / non-Periodic

- **Time Limited / Time Unlimited**
  - For Time Unlimited: Periodic / Non-Periodic

- **Other Classifications:**
  - Real valued / Complex valued
  - Even / Odd
  - Hermitian / Anti-Hermitian
Continuous & Discrete-Time Signals

- **Continuous-Time Signals**
  - Most signals in the real world are continuous time, as the scale is infinitesimally fine.
  - E.g. voltage, velocity,
  - Denote by $x(t)$, where the time interval may be bounded (finite) or infinite

- **Discrete-Time Signals**
  - Some real world and many digital signals are discrete time, as they are sampled
  - E.g. pixels, daily stock price (anything that a digital computer processes)
  - Denote by $x[n]$, where $n$ is an integer value that varies discretely

- **Sampled continuous signal**
  \[ x[n] = x(nk) \quad k \text{ is sample time} \]
Signal Properties

- **Periodic signals:**
  - A signal is periodic if it repeats itself after a fixed period $T$, i.e. $x(t) = x(t+T)$ for all $t$. A $\sin(t)$ signal is periodic.

- **Even and odd signals:**
  - A signal is even if $x(-t) = x(t)$ (i.e. it can be reflected in the axis at zero). A signal is odd if $x(-t) = -x(t)$. Examples are $\cos(t)$ and $\sin(t)$ signals, respectively.

- **Exponential and sinusoidal signals:**
  - A signal is (real) exponential if it can be represented as $x(t) = Ce^{at}$. A signal is (complex) exponential if it can be represented in the same form but $C$ and $a$ are complex numbers.

- **Step and pulse signals:**
  - A pulse signal is one which is nearly completely zero, apart from a short spike, $\delta(t)$. A step signal is zero up to a certain time, and then a constant value after that time, $u(t)$.

- These properties define a large class of tractable, useful signals and will be further considered in the coming lectures.
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Band-Limited Signals

- We claim a signal is band-limited if its power/energy spectral density is non-zero on a limited frequency-range

\[ S_x(f) = 0, \quad \text{if } |f| \leq B \]
Band-Limited Signals (II)

- A band limited signal is considered band-pass if its power/energy spectral density is confined into a limited frequency range (bandwidth) around a given central frequency

$$S_x(f) = 0, \quad \text{if} \quad \begin{cases} |f| \leq f_c - B \\
|f| \leq f_c + B \end{cases}$$

- However, there are many other band limited signals which their non-zero spectrum is around DC – we call them low-pass signals (ex. human voice).
Bilateral and Unilateral Spectrum

- Real-valued signals are known to have hermitian spectrum, i.e. the positive values of the spectrum are equal to the complex conjugates of their negative counterparts – the spectrum is symmetric. So, if the signal is known to be real-valued, we can plot both positive and negative frequencies (bilateral spectrum) or we can only plot positive frequencies (unilateral spectrum)
Bilateral and Unilateral Spectrum

- Both bilateral and unilateral spectrum for real-value signals are equivalent, and provide exactly the same information.
- If we use unilateral spectrum and we want to maintain the same overall power of the signal, we have to multiply by two the amplitude of the unilateral spectrum respect to the bilateral one.
- Unilateral spectrum for complex-value signals do not make sense.
Transmission at Low Frequencies

Some of the problems encountered when transmitting at low frequencies are:

- Electromagnetic devices may have sizes comparable to the wavelength, so its size will increase for low frequencies.
- The noise and distortion for low frequencies usually is larger.
- It is not possible to simultaneously transmit several signals by differentiating their central frequency (FDMA).
- Many electromagnetic channels stop low frequencies.
Transmission at High Frequencies

On the other hand, transmission at high frequencies exhibit the following advantages:

- Smaller devices
- Better channels for transmission
- Lower level of noise and interferences
- Allowing simultaneous transmission of multiple signals in different frequencies (FDMA)
From Baseband to Passband

- Baseband signal = original signal without any processing.
  - Spectrum of the signal as “it is”. Bandwidth occupied depending on the nature of the signal

- How can we transform the Spectrum of the signal without changing the information contained in it
Example of Digital System working with Baseband signals
Example of Digital System working with Bandpass signals

Telecommunication Systems Fundamentals
Filtering

- Filtering concept
  - Every transmission system includes one or more “Filters” that contribute to reject all out-of-band undesired signals, clearing the desired signal
  - Filters are defined by their starting and ending passband frequencies, or equivalently their bandwidth and central frequency.

- Ideal filters:

\[
H(f) = \begin{cases} 
  Ke^{-j\omega t} & f_L \leq |f| \leq f_U \\
  0 & \text{otherwise}
\end{cases}
\]

\[
B = f_u - f_L
\]
Filtering (II)

- Actual filters
  - Ideal filters cannot be implemented within finite impulse response. Actual approximations are used to approximate ideal response.

- Filter types:
  - Low-Pass
  - High-Pass
  - Band-Pass
  - Band-Stop
Equivalent Bandwidth for Actual Systems

- Example: \( x(t) = \Pi(t/T) \Rightarrow X(f) = T \cdot \text{sinc}(fT) \)
  \[ T = 10^{-4} \text{ seg} \]
  Is BandWidth (BW) = \( \infty \)?

- BW to first null:
  - Ej.: BW = 1/T = 10KHz

- Noise Equivalent BW:
  - Same power than ideal filter of BW
  - Ex.: BW = 1/(2T) = 5KHz

- 3dB BW:
  - \( G(f_{3dB}) = G(0) / 2 \)
  - Ex.: BW = 4.5KHz
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Signal Representation

- We look for simple models to describe signals
- Requirements for the models:
  - It must include all the valuable information of the signal – no information loss
  - It should be as simplest as possible
- Example: to represent a straight line -> duple(point, slope)
Phasors

- Assuming a Sin waveform:
  - We only need to know: amplitude, frequency and phase

\[ v(t) = A \cos (\omega_0 t + \phi) \]

- Recall the Euler’s theorem:

\[ e^{\pm j\theta} = \cos \theta \pm j \sin \theta \]
Phasors (II)

- Phasor’s description of a tone:

\[ A \cos (\omega_0 t + \phi) = A \Re\left[ e^{j(\omega_0 t + \phi)} \right] = \Re\left[ Ae^{j\phi} e^{j\omega_0 t} \right] \]

- The term between brackets can be understood as a vector spinning around.

- The projection of the phasor over the real axis is the Sine wave.

- Conventions for Phasors:
  - The angle/phase is measured respect the cosine.
  - The amplitude is always considered positive.

\[ \sin \omega t = \cos (\omega t - 90^\circ) \]

\[ -A \cos \omega t = A \cos (\omega t \pm 180^\circ) \]
Low Pass Equivalent: Objective

- Low Pass Equivalent intends to be a representation of an actual signal that:
  - Be equivalent -> contains the same information
  - Low Pass -> it Power/Energy Spectral Density is low pass, i.e. centered around \( f=0 \)

- Graphically
  - We start with the original signal
  - Is there redundant info – Analytical Signal
  - Now, we shift the spectrum towards \( f=0 \)
Any band-pass signal centered around $f_0$ can be re-written as:

$$x(t) = A_c(t) \cos(2\pi f_0 t + \phi(t)) =$$

$$= A_c(t) \cos(\phi(t)) \cos(2\pi f_0 t) - A_c(t) \sin(\phi(t)) \sin(2\pi f_0 t) =$$

$$= x_i(t) \cos(2\pi f_0 t) - x_q(t) \sin(2\pi f_0 t)$$

Where we get two components that are orthogonal to each other.

So, we define two components:

- **In-Phase** $x_i(t)$: which is the result of projecting $x(t)$ over cosine carrier.
- **Quadrature** $x_q(t)$: projection of $x(t)$ over sine carrier.

Note that for mathematical neatness quadrature component appears with a (-) minus sign.
Starting Point: Band-Pass Signals (II)

\[ x(t) = x_i(t) \cos(2\pi f_0 t) - x_q(t) \sin(2\pi f_0 t) \]
In-Phase and Quadrature Components

- It can be defined the complex valued signal $z(t)$ as $z(t) = x_i(t) + j x_q(t)$.

- Graphically we can think of this signals as two components in a complex plane.

- The complex signal is regarded as “Low Pass Equivalent” or “Complex Envelope”.
In-Phase and Quadrature Components (II)

The “Low-Pass-Equivalent” or “Complex Envelope” approach allows to easily find the carrier instantaneous envelope/amplitude and carrier instantaneous phase:

\[
\phi(t) = \arctan\left(\frac{x_q(t)}{x_i(t)}\right)
\]

\[
A(t) = \sqrt{x_q^2(t) + x_i^2(t)}
\]

\[
x_i(t) = A(t)\cos(\phi(t))
\]

\[
x_q(t) = A(t)\sin(\phi(t))
\]
The analytical Signal associated to a real-valued signal, \( x(t) \), is to satisfy the following properties:

- Complex valued (\( \rightarrow \) two components: real and imaginary axis).
- Real axis should coincide with real-valued signal, \( x(t) \).
- Its spectrum is null for negative values of frequencies.

So, the quadrature component has to be computed, such that:

\[
\hat{z}(t) = x(t) + j\hat{x}(t) = A_z(t)e^{j\varphi_z(t)}
\]

- The quadrature component is to be the Hilbert Transform of the real-valued signal.
Hilbert Transform

- It can be proven that by defining the quadrature component as:
  \[ \hat{x}(t) = \frac{1}{\pi t} * x(t) \]
  all the restrictions on the analytical signal are satisfied.

- This signal is obtained as the Hilbert Transform of \( x(t) \):

  \[ z(t) = x(t) + j\hat{x}(t) = x(t) + j \frac{1}{\pi t} * x(t) = x(t) * \left( \delta(t) + j \frac{1}{\pi t} \right) \]

- In other words, from a real valued signal, a different real valued signal can be obtained using the Hilbert Transform and combination of both signals gives the analytical signal.
Hilbert Transform (II)

Consequently, Hilbert Transform is defined as:

\[ \hat{x}(t) = \frac{1}{\pi t} \ast x(t) = \frac{1}{\pi} \int_{-\infty}^{\infty} x(\tau) \frac{1}{t-\tau} d\tau \]

and therefore, in frequency domain, it is defined through the following frequency response:

\[ TF \left\{ \frac{1}{\pi t} \right\} = \begin{cases} -j, & f > 0 \\ 0, & f = 0 \\ j, & f < 0 \end{cases} = \begin{cases} e^{-j\pi/2}, & f > 0 \\ 0, & f = 0 \\ e^{j\pi/2}, & f < 0 \end{cases} = -j \cdot \text{sign}(f) \]

\[ \hat{X}(f) = [-j \cdot \text{sign}(f)]X(f) \]
Hilbert Transform (III)

- Properties of the HT:
  - It changes phase of signal, but it does not alter its module
    \[- j \cdot \text{sign}(f) = 1\]
  - Any signal and its HT are orthogonal
    \[x(t) \perp \hat{x}(t)\]
  - \(\hat{x}(t) = -x(t)\)
  - If \(x(t)\) is a band-pass signal and \(m(t)\) a low-pass one, and their spectrums do not overlap
    \[c(t) = m(t)x(t) \Rightarrow \hat{c}(t) = m(t)\hat{x}(t)\]

- Well known HT pairs
  \[x(t) = \cos(2\pi f_0 t) \Rightarrow \hat{x}(t) = \sin(2\pi f_0 t)\]
  \[x(t) = \sin(2\pi f_0 t) \Rightarrow \hat{x}(t) = -\cos(2\pi f_0 t)\]
Analytical Signal (II)

- Given a real valued signal, \( x(t) \), which Power/Energy Spectral Density is defined by \( S_x(f) \), the corresponding Analytical Signal has an spectrum defined by:

\[
S_z(f) = 2S_x(f)U(f)
\]

where \( U(f) \) is the Fourier Transform of:

\[
U(f) = \mathcal{F}\left\{\delta(t) + j \frac{1}{\pi t}\right\} = \begin{cases} 1 & f > 0 \\ 0, & \text{elsewhere} \end{cases}
\]

satisfying:

\[
U(f) = |U(f)|^2
\]
Analytical Signal (III)

- Intuitively, the plot of the Power Spectral Density are:

\[ x(t) \]

\[ z(t) = x(t) + j\hat{x}(t) \]

- Analytical signal and HT have been proposed on previous slides for real valued signal. However they can be generalized for complex signals.

  - Generally speaking, HT provides a 90° phase shifted version of the original signal
Low-Pass Equivalent Signal

From the Analytical Signal, what is missing to get the Low-Pass Equivalent Signal?

- Frequency shift by $f_0$ to the left
- In frequency domain that is to convolve the spectrum with $\delta(f - f_0)$

or, equivalently, to multiply by $e^{j2\pi f_0 t}$ in the time domain
Recall of Complex Exponential Signal

- Euler’s rule:
  \[ e^{j(\omega t + \phi)} = \cos(\omega t + \phi) + j \sin(\omega t + \phi) \]

  - So, cosine and sine can be seen as the in-phase and quadrature components of the complex exponential.

  - Neat separation between amplitude and phase.
The Power Spectral Density of the Complex Exponential signal has only a positive frequency contribution.
Recall of Complex Exponential Signal (III)

- Summarizing:
  - A real valued sine signal can be seen as one of the components of a complex exponential signal.
  - A complex exponential is the natural extension of sine and cosine waves into the complex plane.
  - Complex exponentials allows us to easily separate phase and amplitude in communication signals.
  - Positive Spectrum of a sine (pass-band) signal and complex exponential (analytical) match each other.
  - Complex exponential (analytical) has not power at negative frequencies, while sine signals (real valued) do.
Low-Pass Equivalent

• Band-Pass (Real Valued)
  \[ x(t) = x_i(t) \cos(2\pi f_0 t) - x_q(t) \sin(2\pi f_0 t) \]

• Analytical Signal
  \[ z(t) = x(t) + j\hat{x}(t) = x(t) \ast \left( \delta(t) + \frac{j}{\pi t} \right) \]

• Low-Pass Equivalent
  \[ x_l(t) = x_i(t) + jx_q(t) = z(t)e^{-j2\pi f_0 t} \]
In-Phase and Quadrature Modulation

- In a Band-Pass modulation, both components (In-Phase and Quadrature) can be modulated with independent information.

\[ x_i(t) = m_1(t) \]

\[ x_q(t) = m_2(t) \]
In-Phase and Quadrature Modulation (II)

The transmitted signal, $x(t)$
- Is band-pass
- Is real-valued
- Transmits double information while it occupies the same bandwidth

$$x(t) = m_1(t) \cos(2\pi f_0 t) - m_2(t) \sin(2\pi f_0 t)$$

$$\hat{x}(t) = m_2(t) \cos(2\pi f_0 t) + m_1(t) \sin(2\pi f_0 t)$$
Channel’s Model

- We know that transmitted signals arrive to the receiver suffering some kinds of degradation
  - Attenuation
    \[ y(t) = \frac{x(t)}{L} \]
  - Delay
    \[ y(t) = x(t - \tau) \]
  - Noise
    - From different sources
    - Sometimes it can be modeled as Gaussian
      \[ y(t) = x(t) + n(t) \]
  - Time dispersion
    - General concept of linear convolution
      \[ y(t) = x(t) \ast h(t) \]
Channel’s Model

- We have a general model for the channel:

\[ y(t) = x(t) \ast h(t) + n(t) \]

- We can further simplify it by:
  - Working with Base Band Equivalent
  - Working on Discrete Time Equivalent by sampling signals and channel response, \( T_s \).
  → We get the Low-Pass Discrete Equivalent model
Low-Pass Discrete Equivalent Channel

- Assuming the signal and the channel response are band-pass
  - Signals can be expressed by their Low-Pass Equivalent Channel
Low-Pass Discrete Equivalent Channel

- A Low-Pass Equivalent model for the overall transmitter-channel-receiver can be defined as in previous slide

\[ x_b(t) \rightarrow h_b(t) \rightarrow y_b(t) \]

- Where \( h_b(t) \) is the low-pass equivalent of \( h(t) \)
  - All these signals are band-limited, therefore they can be sampled following the Nyquist rule and there is not loss of information

\[ x_b(t) = \sum_n x_b[n] \text{sinc}(W_s t - n) \]
Low-Pass Discrete Equivalent Channel

Physical Channel

Low-Pass Equivalent Channel (complex value)

Low-Pass Discrete Equivalent Channel (complex value)
Low-Pass Discrete Equivalent Channel

- A model for the noise has to be added too
  - The signals are complex and discrete ➔ noise should be complex and discrete, with the same characteristics than a AWGN sampled process
    \[ w[n] \rightarrow G(0, N_0 / 2) + j \cdot G(0, N_0 / 2) \rightarrow G_{\text{Comp}}(0, N_0) \]
- So, the Low-Pass Discrete Equivalent signal at the receiver is
  \[ y_b[n] = h_b \ast x_b[n] + w[n] \]
Any Band-Pass signal can be decomposed into its In-Phase and Quadrature components
  – So, any Band-Pass signal can be decomposed into two orthogonal signals
  – That can be easily separated at the receiver
  – And they can be modulated independently (different info)

The Low-Pass Equivalent of a real value signal is a complex signal:
  – Low-Pass equivalent does not provide information about the carrier frequency. It is supposed to be known

All the above signal representation are of great value when analyzing communication signals and systems.

A whole communications link can be analyzed using their Low-Pass Discrete Equivalent models for the transmitted signal, the channel, noise and for receiver.