Signal Theory
Semester 6

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Contents

Chapter 1

Definition of a Signal

A signal is a function that conveys information about a phenomenon. A signal can be a function of time, space, or any other variable. A signal can be continuous or discrete. Examples of signals include:

- Speech
- Images
- Audio
- Video
- Temperature
- Pressure
- Voltage
- Current
- etc.

1.1 Classification of Signals

Continuous-Time Signals A signal that is defined for all values of time.

Discrete-Time Signals A signal that is defined only at discrete values of time.

Analog Signals A signal that can take any value in a given range.

Digital Signals A signal that can take only a finite number of values.

Periodic Signals A signal that repeats itself after a certain period of time.

Energy Signals A signal that has finite energy.

$$E := \int_{-\infty}^{\infty} |x(t)|^2 \, dt \, .$$

Power Signals A signal that has finite power.

$$P_{(t_1, t_2)} := \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} |x(t)|^2 \, dt \, .$$

Even/Odd Signals A signal is even if $x(t) = x(-t)$ and odd if $x(t) = -x(-t)$. Any signal can be represented as the sum of an even and odd signal.

$$x(t) = \underbrace{\frac{x(t) + x(-t)}{2}}_{\text{Even}} + \underbrace{\frac{x(t) - x(-t)}{2}}_{\text{Odd}} \, .$$

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1.2 Operations on Signals

1.2.1 Unary Operations

Inverse The inverse of a signal $x(t)$ is $x(-t)$.

> **Note:-**
> The inverse of $x(t-1)$ is $x(-t-1)$.

Time Shift The time shift of a signal $x(t)$ by t_0 is $x(t+t_0)$. If t_0 is negative then the signal is shifted to the left.

> **Note:-**
> The time shift of $x(-t-1)$ by 1 is $x(-(t+1)-1)$.

Time Scaling The time scaling of a signal $x(t)$ by a is $x(at)$. If a is greater than 1 then the signal is compressed and if a is less than 1 then the signal is expanded.

> **Example 1.2.1**
>
> Given a signal $x(t)$, to get the signal $x(-2t+4)$ we have 2 methods:
>
> 1. $x(t) \xrightarrow{\text{Shift to the left}} x(t+4) \xrightarrow{\text{Inverse}} x(-t+4) \xrightarrow{\text{Time Scaling}} x(-2t+4)$
>
> 2. $x(t) \xrightarrow{\text{Time Scaling}} x(2t) \xrightarrow{\text{Inverse}} x(-2t) \xrightarrow{\text{Shift to the right}} x(-2(t-2)) = x(-2t+4)$

1.2.2 Binary Operations

Convolution
$$z(t) = x(t) * y(t) := \int_{-\infty}^{\infty} x(u)\, y(t-u)\, \mathrm{d}u \,.$$

The convolution operation is

1. Commutative

$$x(t) * y(t) = y(t) * x(t).$$

2. Associative

$$(x(t) * y(t)) * z(t) = x(t) * (y(t) * z(t)) = x(t) * y(t) * z(t).$$

3. Distributive over addition

$$x(t) * (y(t) + z(t)) = x(t) * y(t) + x(t) * z(t).$$

> **Note:-**
> Generally speaking,
> $$x(at) * y(at) \neq a(x(t) * y(t)) \quad \text{for } a \neq 1.$$

> **Note:-**
> When convolving two signals aim to have the signal $y(t-u)$ be the signal with the simplest t input.

Scalar Product

$$\langle x(t), y^*(t) \rangle := \int_{t_1}^{t_2} x(t)\, y^*(t)\, \mathrm{d}t\,.$$

If the scalar product is zero, then the two signals are orthogonal in the interval $[t_1, t_2]$. The scalar product is not commutative.

$$\langle x(t), y^*(t) \rangle \neq \langle y(t), x^*(t) \rangle = \int_{t_1}^{t_2} y(t)\, x^*(t)\, \mathrm{d}t = \left(\int_{t_1}^{t_2} x(t)\, y^*(t)\, \mathrm{d}t \right)^* = \langle x(t), y^*(t) \rangle^*\,.$$

Note:-

$$(A + B)^* = A^* + B^*$$
$$(AB)^* = A^* B^*$$

Correlation

$$\varphi_{xy}(\tau) := \int_{-\infty}^{\infty} x^*(u)\, y(u + \tau)\, \mathrm{d}u\,.$$

The correlation operation is not commutative.

$$\varphi_{xy}(\tau) \neq \varphi_{yx}(\tau).$$

$$x(t) \neq y(t) \quad \text{intercorrelation}$$
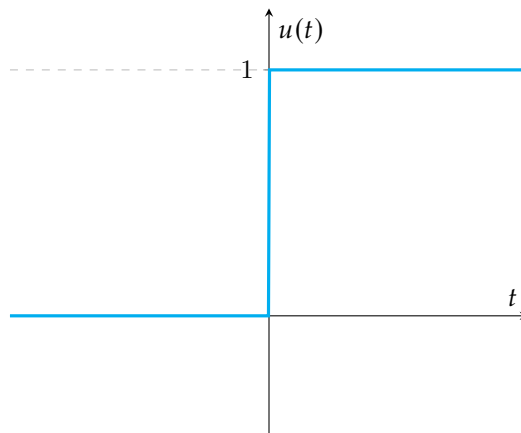$$x(t) = y(-t) \quad \text{autocorrelation}$$

Note:-

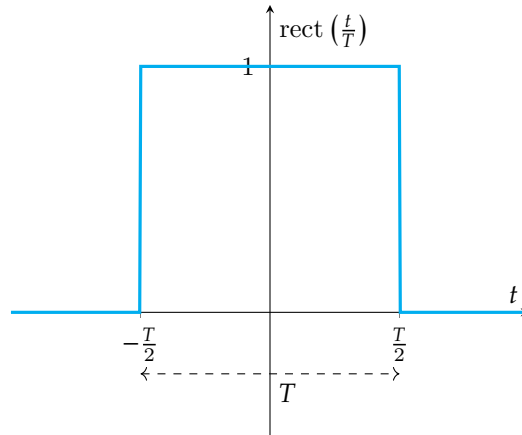$$\varphi_{xy}(\tau) = x^*(-\tau) * y(\tau).$$

1.3 Particular Signals

Unit Step Signal

$$u(t) := \begin{cases} 1 & t \geqslant 0 \\ 0 & t < 0 \end{cases}$$

Rectangular Signal

$$\text{rect}\left(\frac{t}{T}\right) := \begin{cases} 1 & -\frac{T}{2} \leqslant t \leqslant \frac{T}{2} \\ 0 & \text{otherwise} \end{cases}$$
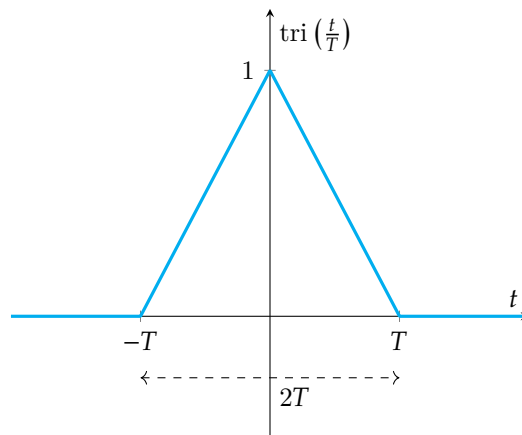
> **Note:-**
>
> The rect function is typically used to sample a signal over a period of time. The signal is sampled at the points where the rect function is equal to 1. For example
>
> $$x(t)\,\text{rect}\left(\frac{t-2}{4}\right).$$
>
> This signal is sampled in the interval $[2, 6]$.

Triangular Signal

$$\text{tri}\left(\frac{t}{T}\right) := \begin{cases} \frac{t}{T} + 1 & -T \leqslant t \leqslant 0 \\ -\frac{t}{T} + 1 & 0 \leqslant t \leqslant T \\ 0 & \text{otherwise} \end{cases} = \begin{cases} 1 - \frac{|t|}{T} & -T \leqslant t \leqslant T \\ 0 & \text{otherwise} \end{cases}$$
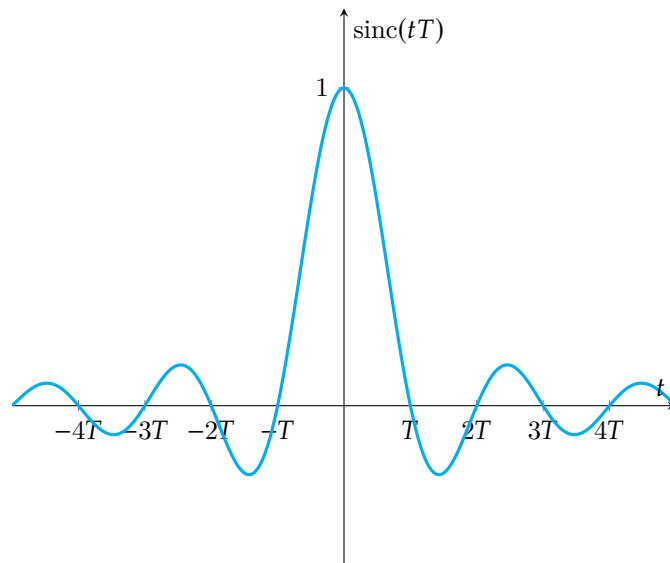

Sinc

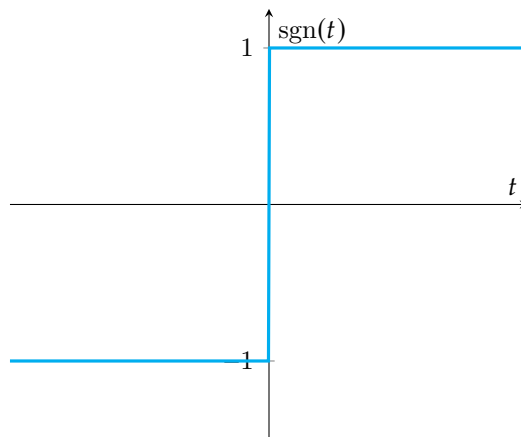$$\text{sinc}(t) := \begin{cases} \frac{\sin(\pi t)}{\pi t} & t \neq 0 \\ 1 & t = 0 \end{cases}$$

sinc(tT)

Sign Function

$$\text{sgn}(t) := \begin{cases} 1 & t > 0 \\ 0 & t = 0 \\ -1 & t < 0 \end{cases}.$$

sgn(t)

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1.4 Other Quantities

Mean

$$\bar{x} := \frac{1}{t_2 - t_1} \int_{t_1}^{t_2} x(t)\, dt \,.$$

Over the entire domain

$$\bar{x} := \lim_{T \to \infty} \int_{-T/2}^{T/2} x(t)\, dt \,.$$

Energy

$$E_{(t_1, t_2)} := \int_{t_1}^{t_2} |x(t)|^2\, dt \,.$$

Power

$$P_{(t_1, t_2)} := \frac{1}{t_2 - t_1} \int_{t_1}^{t_2} |x(t)|^2\, dt \,.$$

Example 1.4.1

1. For $u(t)$

$$\bar{u} = \lim_{T \to \infty} \frac{1}{T} \int_{-T/2}^{T/2} u(t)\, dt$$

$$= \lim_{T \to \infty} \frac{1}{T} \int_0^{T/2} 1\, dt$$

$$= \lim_{T \to \infty} \frac{1}{T} \frac{T}{2}$$

$$= \frac{1}{2}$$

$$\int_{-T/2}^{T/2} |u(t)|^2\, dt = \int_0^{T/2} 1\, dt = \frac{T}{2}$$

$$E_{(-\infty, \infty)} = \lim_{T \to \infty} \int_{-T/2}^{T/2} |u(t)|^2\, dt = \lim_{T \to \infty} \frac{T}{2} = \infty$$

$$P_{(-\infty, \infty)} = \lim_{T \to \infty} \frac{1}{T} \int_{-T/2}^{T/2} |u(t)|^2\, dt = \lim_{T \to \infty} \frac{1}{T} \frac{T}{2} = \frac{1}{2}$$

2. For $\mathrm{sgn}(t)$ Since the signal is odd, the mean is zero.

$$E_{(-\infty,\infty)} = \lim_{T\to\infty} \int_{-T/2}^{T/2} |\mathrm{sgn}(t)|^2 \, dt$$

$$= \lim_{T\to\infty} \int_{-T/2}^{T/2} 1 \, dt$$

$$= \lim_{T\to\infty} \frac{T}{2}$$

$$= \infty$$

$$P_{(-\infty,\infty)} = \lim_{T\to\infty} \frac{1}{T} \int_{-T/2}^{T/2} |\mathrm{sgn}(t)|^2 \, dt$$

$$= \lim_{T\to\infty} \frac{1}{T} \int_{-T/2}^{T/2} 1 \, dt$$

$$= \lim_{T\to\infty} \frac{1}{T} \frac{T}{2}$$

$$= \frac{1}{2}$$

Example 1.4.2 (Convolution)

Consider the 2 signals

$$x(t) = \mathrm{rect}\,(t) \quad \text{and} \quad y(t) = \mathrm{rect}\,(t).$$

$$z(t) = x(t) * y(t) = \int_{-\infty}^{\infty} \mathrm{rect}\,(u)\, \mathrm{rect}\,(t-u)\, du.$$

Since the rect function is even

$$\int_{-\infty}^{\infty} \mathrm{rect}\,(u)\, \mathrm{rect}\,(u-t)\, du.$$
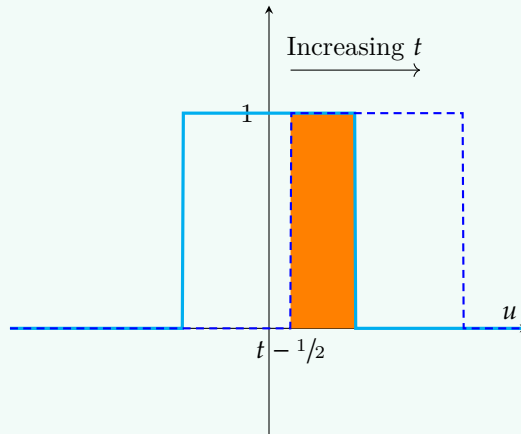
The integral only really takes a value when the two rect functions overlap.

The boundaries of the integral are $t - 1/2$ and $1/2$ when $0 \leqslant t \leqslant 1$.

$$z(t) = \int_{t-1/2}^{1/2} 1 \, du = 1 - t.$$

Decreasing t

$t + {}^{1}\!/_{2}$

u
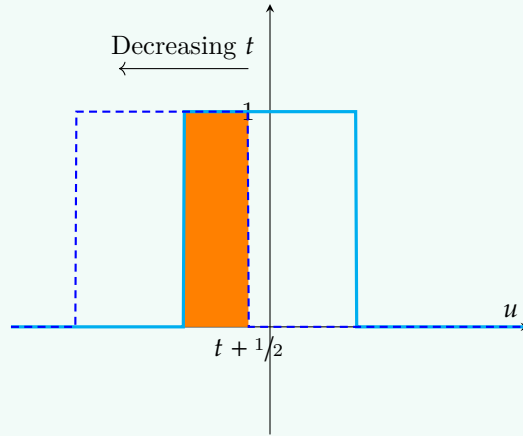
Similarly, the boundaries of the integral are $t + {}^{1}/_{2}$ and $-{}^{1}/_{2}$ when $-1 \leqslant t \leqslant 0$.

$$z(t) = \int_{-1/2}^{t+1/2} 1 \, du = 1 + t.$$

Thus

$$z(t) = \begin{cases} 1 - t & 0 \leqslant t \leqslant 1 \\ 1 + t & -1 \leqslant t \leqslant 0 = \text{tri}(t). \\ 0 & \text{otherwise} \end{cases}$$

Chapter 2

Deterministic Signals

2.1 Fourier Transform

Recall that a Fourier series is a representation of a periodic signal as a sum of sines and cosines.

$$x(t) = \frac{a_0}{2} + \sum_{n=1}^{\infty} \underbrace{\left(a_n \cos\left(\frac{2\pi nt}{T} \right) + b_n \sin\left(\frac{2\pi nt}{T} \right) \right)}_{A_n \cos(n\omega_0 t + \phi_n)}.$$

$$a_n = \frac{2}{T} \int_T x(t) \cos\left(\frac{2\pi nt}{T} \right) dt$$

$$b_n = \frac{2}{T} \int_T x(t) \sin\left(\frac{2\pi nt}{T} \right) dt$$

where $\omega_0 = \frac{2\pi}{T}$.

And in complex form

$$x(t) = \sum_{n=-\infty}^{\infty} \lambda_n e^{jn\omega_0 t}.$$

Where

$$\lambda_n = \frac{1}{T} \int_T x(t) e^{-jn\omega_0 t} \, dt \,.$$

Consider an interval $[-T, T]$ for the signal $x(t)$. We then assume that the function is periodic with period $2T$ over the rest of the domain (the section $[-T, T]$ will repeat infinitely). Since the signal is periodic we can represent it using a Fourier series $x_{\text{rep}}(t, T)$. As T approaches infinity then we obtain the original signal periodic over it's entire domain

$$\lim_{T \to \infty} x_{\text{rep}}(t, T) = x(t).$$

$$x_{\text{rep}}(t, T) = \sum_{n=-\infty}^{\infty} \frac{1}{2T} \int_{-T}^{T} x_{\text{rep}}(t, T) e^{-j\frac{n\pi t}{T}} \, dt \, e^{j\frac{n\pi t}{T}}.$$

Since we have the amplitude if the signal at each frequency, we can represent the signal in the frequency domain (i.e. the sum of frequencies that make up the signal). This is the Fourier transform.

> **Definition 2.1.1: Fourier Transform**
>
> The Fourier transform of a signal $x(t)$ is defined as
>
> $$\mathcal{F}_t\{x(t)\} = \hat{x}(f) = X(f) := \int_{-\infty}^{\infty} x(t)e^{-j2\pi ft}\,\mathrm{d}t\,.$$
>
> The inverse Fourier transform is
>
> $$\mathcal{F}_t^{-1}\{X(f)\} = x(t) := \int_{-\infty}^{\infty} X(f)e^{j2\pi ft}\,\mathrm{d}f\,.$$

A signal has a Fourier transform if the signal belongs to $\mathbb{L}^2$ (i.e. the signal has finite energy).

$$x(t) \in \mathbb{L}^2 \quad \text{if} \quad \begin{cases} E < \infty \\ E = \infty \text{ and } P < \infty \end{cases}.$$

> **Example 2.1.1**
>
> Consider the signal $x(t) = \mathrm{rect}\,(t)$.
>
> $$\begin{aligned} \mathcal{F}_t\{x(t)\} = \mathrm{Rect}\,(f) &= \int_{-\infty}^{\infty} \mathrm{rect}\,(t)\,e^{-j2\pi ft}\,\mathrm{d}t \\ &= \int_{-1/2}^{1/2} e^{-j2\pi ft}\,\mathrm{d}t \\ &= \frac{e^{-j\pi f} - e^{j\pi f}}{j2\pi f} \\ &= \mathrm{sinc}(f) \end{aligned}$$
>
> Since we have a division by f, we have to take a special case for $f = 0$.
>
> $$\begin{aligned} \mathcal{F}_t\{x(t)\} &= \int_{-\infty}^{\infty} \mathrm{d}t \\ &= 1 \end{aligned}$$
>
> The signal is mostly constant so the Fourier transform is highest closest to zero and the farther away from zero the lower the amplitude of the associated frequency.

2.1.1 Properties of the Fourier Transform

Here we assume that the signals $x(t), y(t), \ldots \in \mathbb{L}^2$

Linearity
$$\mathcal{F}_t\{ax(t) + by(t)\} = a\hat{x}(f) + b\hat{y}(f) \quad \forall a, b \in \mathbb{R}.$$

Inverse
$$\mathcal{F}_t\{x(-t)\} = \hat{x}(-f).$$

Time Shift
$$\mathcal{F}_t\{x(t - t_0)\} = e^{-j2\pi ft_0}\hat{x}(f).$$

Modulation
$$\mathcal{F}_t\{e^{j2\pi f_0 t}x(t)\} = \hat{x}(f - f_0).$$

Differentiation

$$\mathcal{F}_t\left\{\frac{\mathrm{d}x(t)}{\mathrm{d}t}\right\} = j2\pi f\hat{x}(f)$$

$$\mathcal{F}_t\left\{\frac{\mathrm{d}^n x(t)}{\mathrm{d}t^n}\right\} = (j2\pi f)^n \hat{x}(f)$$

Symmetry

$$\hat{x}(f) = |\hat{x}(f)|e^{j\varphi(f)}.$$

$$\left.\begin{array}{ll} \text{even:} & |\hat{x}(f)| = |\hat{x}(-f)| \\ \text{odd:} & \varphi(f) = -\varphi(-f) \end{array}\right\} \implies \underset{\text{Hermitian Symmetry}}{\hat{x}(f) = \hat{x}^*(-f)}$$

or

$$\begin{array}{l} \text{even:}\quad \mathrm{Re}(f) = \mathrm{Re}(-f) \\ \text{odd:}\quad \mathrm{Im}(f) = -\mathrm{Im}(-f) \end{array}$$

Convolution

$$\mathcal{F}_t\{x(t) * y(t)\} = \hat{x}(f) \cdot \hat{y}(f).$$

Product

$$\mathcal{F}_t\{x(t)y(t)\} = \hat{x}(f) * \hat{y}(f).$$

Scaling

$$\mathcal{F}_t\{x(at)\} = \frac{1}{|a|}\hat{x}\left(\frac{f}{a}\right).$$

Theorem 2.1.1 Parseval's Theorem

Given a signal $x(t)$ with Fourier transform $\hat{x}(f)$. The energy of the signal in the time domain is equal to the energy of the signal in the frequency domain.

$$E = \int_{-\infty}^{\infty} |x(t)|^2 \,\mathrm{d}t = \int_{-\infty}^{\infty} |\hat{x}(f)|^2 \,\mathrm{d}f.$$

$$\underset{\text{Energy Density}}{\frac{\mathrm{d}E}{\mathrm{d}f} = |\hat{x}(f)|^2.}$$

2.2 Dirac Delta Function

Definition 2.2.1: Dirac Delta Function

The Dirac delta function is a function that is zero everywhere except at $t = 0$. There are two ways to define the Dirac delta function:

1. A method useful for computing integrals

$$\delta(t) = \lim_{T\to 0}\frac{1}{T}\,\mathrm{rect}\left(\frac{t}{T}\right) = \begin{cases} \infty & t = 0 \\ 0 & t \neq 0 \end{cases}.$$

2. A method useful for general use

$$\delta(t) = \lim_{T\to 0}\,\mathrm{tri}\left(\frac{t}{T}\right) = \begin{cases} 1 & t = 0 \\ 0 & t \neq 0 \end{cases}.$$

The Dirac delta function has the following properties:

Signal	Fourier Transform
$u(t)$	$\frac{1}{2}\delta(f) + \frac{1}{j\omega}$
$\delta(t)$	1
$\text{rect}\left(\frac{t}{T}\right)$	$T\,\text{sinc}\,(fT)$
$\text{tri}\left(\frac{t}{T}\right)$	$T\,\text{sinc}^2\,(fT)$
$\text{sinc}(t)$	$\text{rect}\,(f)$
$\text{sgn}(t)$	$\frac{1}{j\pi f}$
$e^{-j2\pi f_0 t}$	$\delta(f - f_0)$
$e^{j2\pi f_0 t}$	$\delta(f + f_0)$

1.

$$\int_{-\infty}^{\infty} \delta(t)x(t)\,\mathrm{d}t = x(0).$$

$$\int_{a}^{b} \delta(t - t_0)x(t)\,\mathrm{d}t = \begin{cases} x(t_0) & t_0 \in [a,b] \\ 0 & \text{otherwise} \end{cases}.$$

$$\int_{-\infty}^{\infty} x(t - a)\delta(t - b)\,\mathrm{d}t = x(b - a).$$

2.

$$x(t)\delta(t - t_0) = x(t_0)\delta(t - t_0).$$

3.

$$x(t) * \delta(t) = x(t).$$

$$x(t - t_0) * \delta(t - t_1) = x(t - t_0 - t_1).$$

4.

$$\delta(at) = \frac{1}{|a|}\delta(t).$$

5.

$$\mathcal{F}_t\{\delta(t)\} = 1.$$

6.

$$\mathcal{F}_t\{e^{j2\pi f_0 t}\} = \delta(f - f_0).$$

Note:-

$$\mathcal{F}_t\{x(t)\} = \frac{\mathcal{F}_t\{x'(t)\}}{j\omega} + a\delta(f).$$

where

$$a = \lim_{T \to \infty} \frac{1}{T} \int_{-T/2}^{T/2} x(t)\,\mathrm{d}t\,.$$

2.3   Fourier Transform of Some Signals

2.4   Fourier Transform of Periodic Signals

Consider a signal $x(t)$ that is periodic with period T.

$$x(t) = \sum_{n=-\infty}^{\infty} \lambda_n e^{jn\omega_T t} \quad \omega_T = \frac{2\pi}{T}.$$

The Fourier transform of this signal is

$$\mathcal{F}_t\{x(t)\} = \sum_{n=-\infty}^{\infty} \lambda_n \delta(f - n f_T).$$

where

$$\lambda_n = \frac{1}{T} \int_T x(t) e^{-jn\omega_T t} \, \mathrm{d}t \,.$$

2.5   Temporary and Frequential Duration

The mean and standard deviation of a signal in the time domain are given by

$$t_0 = \frac{\int_{-\infty}^{\infty} t |x(t)|^2 \, \mathrm{d}t}{\int_{-\infty}^{\infty} |x(t)|^2 \, \mathrm{d}t} = \frac{\int_{-\infty}^{\infty} t |x(t)|^2 \, \mathrm{d}t}{E}$$

$$\sigma_0^2 = \frac{\int_{-\infty}^{\infty} (t - t_0)^2 |x(t)|^2 \, \mathrm{d}t}{E}$$

The mean and standard deviation of a signal in the frequency domain are given by

$$f_0 = \frac{\int_{-\infty}^{\infty} f |\hat{x}(f)|^2 \, \mathrm{d}f}{\int_{-\infty}^{\infty} |\hat{x}(f)|^2 \, \mathrm{d}f} = \frac{\int_{-\infty}^{\infty} f |\hat{x}(f)|^2 \, \mathrm{d}f}{E}$$

$$\sigma_f^2 = \frac{\int_{-\infty}^{\infty} (f - f_0)^2 |\hat{x}(f)|^2 \, \mathrm{d}f}{E}$$

Theorem 2.5.1 Heisenberg Uncertainty Principle

$$\sigma_0 \sigma_f \geqslant \frac{1}{4\pi}.$$

This means that a compressed signal in the time domain will have a wider frequency domain and vice versa.

2.6   Bounded Signals

A signal is bounded if

$$|x(t)| \neq 0 \quad \forall t \in [a, b].$$

Theorem 2.6.1 Riemann-Lebesgue Lemma

2.7 Energy Spectral Density

Note:-

$$\mathcal{F}_t\{\varphi_{xy}\} = \mathcal{F}_t\{x^*(-\tau) * y(\tau)\} = \hat{x}(f)\hat{y}(f) \implies \varphi_x = |\hat{x}(f)|^2 = \phi_x.$$

The Spectral Energy at finite energy is given by

$$\varphi_x = |\hat{x}(f)|^2.$$

The Spectral Power at finite power is given by

$$\varphi_{xy} = \int_{-\infty}^{\infty} x^*(\tau)y(\tau + u)\, du\,.$$

$$\frac{dP}{df} = \phi_x(f).$$

2.8 Hilbert Transform

Definition 2.8.1: Hilbert Transform

The Hilbert transform of a signal $x(t)$ is defined as

$$\mathcal{H}\{x(t)\} = \frac{1}{\pi}\text{P.V.}\int_{-\infty}^{\infty} \frac{x(\tau)}{t - \tau}\, d\tau.$$

The Hilbert transform is a result of the convolution of the signal with the function $\frac{1}{\pi t}$.

$$\mathcal{H}\{x(t)\} = x(t) * \frac{1}{\pi t}.$$

Theorem 2.8.1 Relationship with the Fourier transform

$$\mathcal{F}_t\{\mathcal{H}\{x(t)\}\} = -j\,\text{sgn}(f)\hat{x}(f).$$

The Hilbert transform has the following properties:

Linearity
$$\mathcal{H}\{ax(t) + by(t)\} = a\mathcal{H}\{x(t)\} + b\mathcal{H}\{y(t)\}\,.$$

Orthogonality $x(t)$ and $\mathcal{H}\{x(t)\}$ are orthogonal signals.

Analytic Signal
$$z(t) = r(t)e^{j\theta(t)}.$$

where $r(t)$ is the envelope of the signal and $\theta(t)$ is the phase of the signal. The instantaneous frequency of the signal is given by

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$$\omega(t) = \frac{\mathrm{d}\theta(t)}{\mathrm{d}t}.$$