

fourier

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1 Introduction

1.1 Signals and Systems in Continuous Time

The input signal $x(t)$ is defined for $t \in (-\infty, \infty)$ and is a periodic extension of a signal defined over a limited interval with period T . The output signal $y(t) = \mathcal{H} \circ x(t)$ is also defined for $t \in (-\infty, \infty)$.

1.2 Properties of Systems

The system \mathcal{H} is said to be *linear* if :

$$\mathcal{H}(k_1x_1(t) + k_2x_2(t)) = k_1\mathcal{H}(x_1(t)) + k_2\mathcal{H}(x_2(t)) = k_1y_1(t) + k_2y_2(t) \quad (1)$$

The system \mathcal{H} is said to be *shift-invariant* if :

$$\mathcal{H}(\mathcal{T}_{t_0}(x(t))) = \mathcal{T}_{t_0}(y(t)) \quad \text{where} \quad \mathcal{T}_{t_0}(x(t)) = x(t - t_0) \quad (2)$$

1.3 Examples

System	Linear	Shift-Invariant
$y(t) = x^2(t)$		
$y(t) = x(t) + a$		
$y(t) = ax(t)$		
Optimal sampling		

2 Linear Shift-Invariant Systems

2.1 Input Signal and System

The input signal $x(t)$ is defined for $t \in (-\infty, \infty)$ and is a periodic extension of a signal defined over a limited interval with period 1.

The system \mathcal{H} has an impulse response $\tilde{h}(t)$ with a basic period such that $\tilde{h} : [0, 1) \rightarrow \mathbb{R}$. This impulse response is periodically extended into $h(t)$.

3 Convolution Systems

Consider an input signal $x(t)$ defined for $t \in [0, 1]$. The convolution system is given by :

$$y(t) = \int_0^1 x(\tau)h(t - \tau)d\tau \quad (3)$$

We use here the orthonormal family of Fourier functions, defined over $[0, 1)$ for $k \in \mathbb{Z}$:

$$\beta_k^F(t) = \exp(i2\pi kt) \quad (4)$$

4 Fourier-based Analysis of Systems

Using the Fourier family, the representations are :

$$x(t) = \sum_{k=-\infty}^{\infty} a_k \beta_k^F(t) \quad (5)$$

$$a_k = \langle \beta_k^F, x \rangle = \int_0^1 x(t) \exp(-i2\pi kt) dt \quad (6)$$

$$h(t) = \sum_{k=-\infty}^{\infty} b_k \beta_k^F(t) \quad (7)$$

$$b_k = \langle \beta_k^F, h \rangle = \int_0^1 h(t) \exp(-i2\pi kt) dt \quad (8)$$

4.1 Convolution Systems

The convolution system is expressed as :

$$y(t) = \int_0^1 x(\tau) h(t - \tau) d\tau \quad (9)$$

$$= \sum_{k=-\infty}^{\infty} a_k b_k \beta_k^F(t) \quad (10)$$

$$c_k = a_k b_k \quad (11)$$

5 Derivation System

Consider an input signal $x(t)$ defined for $t \in [0, 1]$. The derivation system is :

$$y(t) = \frac{d}{dt} x(t) \quad (12)$$

with boundary values obeying the cyclic continuity : $x(0) = x(1)$.

$$c_k = i2\pi k a_k \quad (13)$$

6 DFT

The Discrete Fourier Transform (DFT) matrix is given by :

$$[\text{DFT}] = \frac{1}{\sqrt{N}} \begin{bmatrix} W^{*0 \cdot 0} & W^{*1 \cdot 0} & \dots & W^{*(N-1) \cdot 0} \\ W^{*0 \cdot 1} & W^{*1 \cdot 1} & \dots & W^{*(N-1) \cdot 1} \\ \vdots & \vdots & \ddots & \vdots \\ W^{*0 \cdot (N-1)} & W^{*1 \cdot (N-1)} & \dots & W^{*(N-1) \cdot (N-1)} \end{bmatrix} \quad (14)$$

where $W = \exp\left(\frac{j2\pi}{N}\right)$. Note that the DFT matrix is symmetric and unitary.

7 Representation of a Discrete Signal in the DFT Domain

The representation of the discrete signal $\phi \in R^N$ is :

$$\phi^F = [\text{DFT}]\phi \quad (15)$$

Observe that :

$$[\text{DFT}]^* \phi^F = [\text{DFT}]^* [\text{DFT}]\phi \quad (16)$$

$$[\text{DFT}]^* \phi^F = \phi \quad (17)$$

Above is the inverse DFT procedure.

8 Example

Consider the following discrete signal of N samples :

For $n = 0, 1, 2, \dots, (N-1)$, $\phi(n) = \cos\left(\frac{2\pi k_0}{N}n\right)$ where $k_0 \in \{0, 1, 2, \dots, (N-1)\}$.

The k^{th} component of the DFT-domain representation of the above signal is :

$$\phi^F(k) = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} W^{*k \cdot n} \phi(n) \quad (18)$$

$$= \frac{1}{\sqrt{N}} \left(\frac{1}{2} \sum_{n=0}^{N-1} W^{*k \cdot n} (W^{k_0 \cdot n} + W^{-k_0 \cdot n}) \right) \quad (19)$$

When $k = k_0$:

$$\sum_{n=0}^{N-1} W^{*k \cdot n} (W^{k_0 \cdot n}) = N \quad (20)$$

When $k \neq k_0$:

$$\sum_{n=0}^{N-1} W^{*k \cdot n} (W^{k_0 \cdot n}) = \frac{(W^{-(k-k_0)})^N - 1}{W^{-(k-k_0)} - 1} = 0 \quad (21)$$

Similarly :

$$\phi^F(k) = \frac{1}{\sqrt{N}} \left(\frac{\sqrt{N}}{2} (\delta_{k,k_0} + \delta_{k,-k_0}) \right) \quad (22)$$

For example, for $N = 9$ and $k_0 = 3$, the vector of DFT coefficients is :

$$\left[0, 0, 0, \frac{3}{2}, 0, 0, \frac{3}{2}, 0, 0 \right]^\top \quad (23)$$