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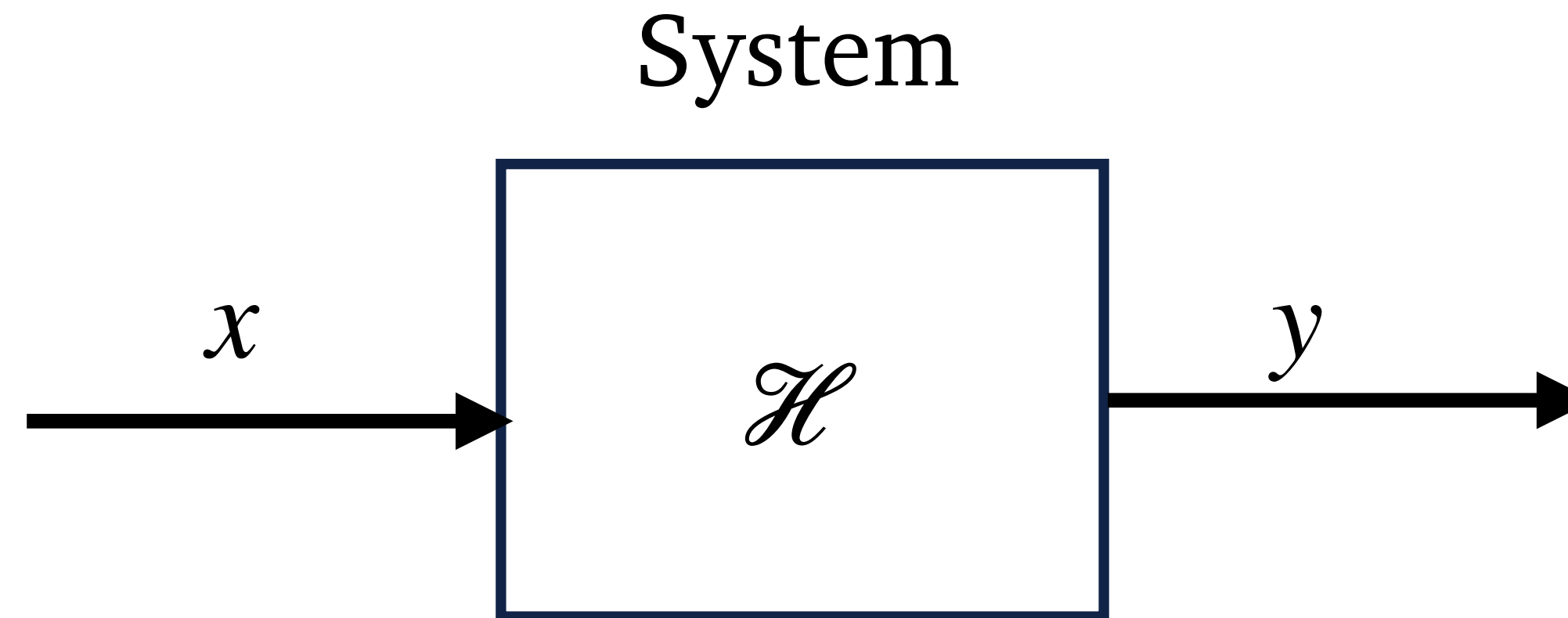
# Introduction to data processing and representation

**Tutorial 6:**

**Fourier Analysis of Linear Shift-Invariant Systems & DFT**

Winter 2024-2025

# Signals and Systems in Continuous Time



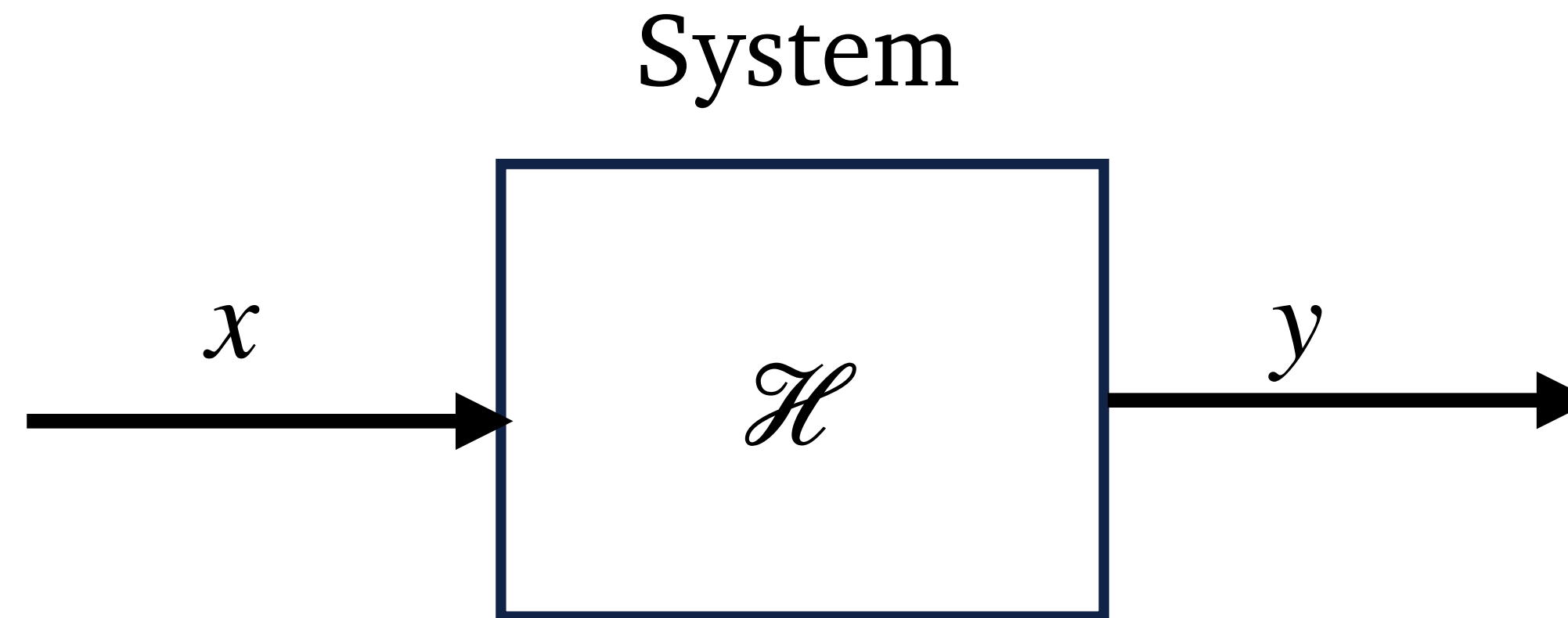
Input signal:

- $x(t)$  is defined for  $t \in (-\infty, \infty)$
- $x(t)$  is periodic extension of a signal defined over a limited interval with period  $T$

Output signal:

- $y(t) = \mathcal{H} \circ x(t)$  is defined for  $t \in (-\infty, \infty)$

# Signals and Systems in Continuous Time



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

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

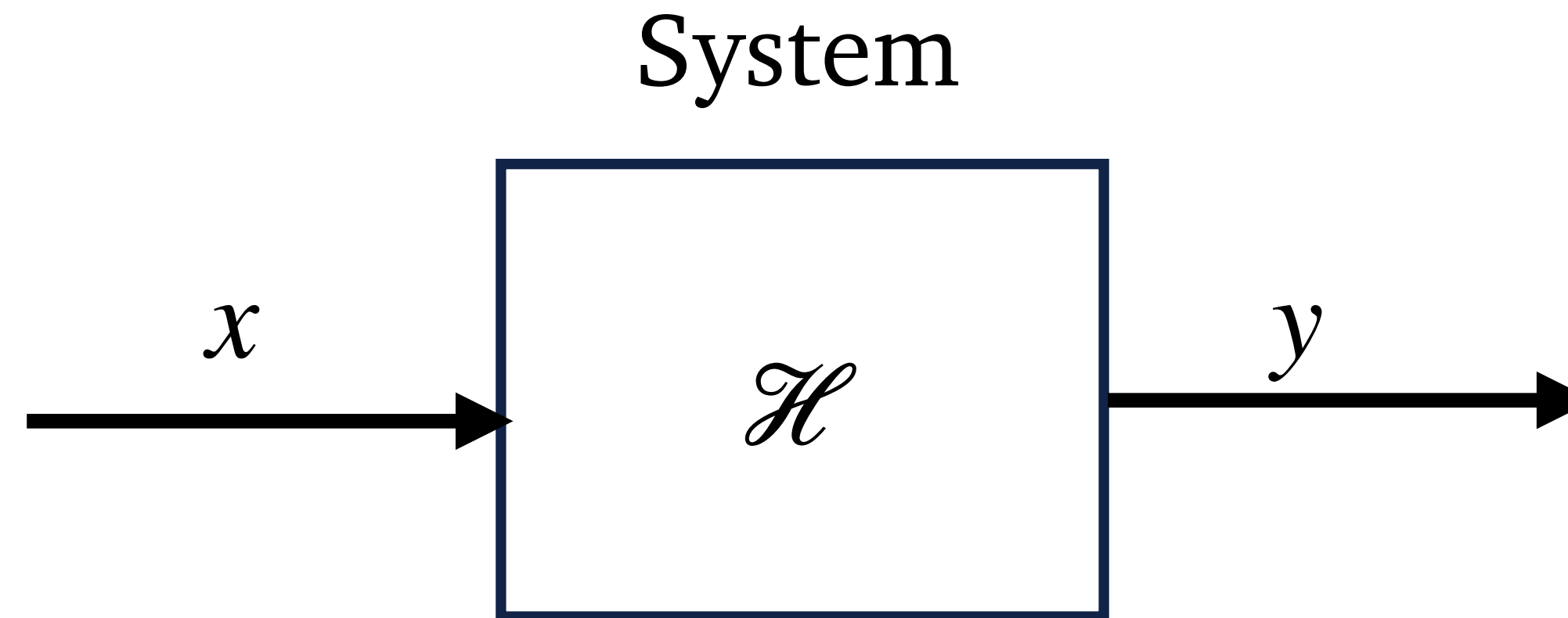
The system  $\mathcal{H}$  is said to be **shift-invariant** if

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

# Examples

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

# Linear Shift-Invariant Systems



Input signal:

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

System  $\mathcal{H}$

- Let  $\tilde{h}(t)$  be the basic period of its **impulse response** such that  $\tilde{h} : [0,1) \rightarrow \mathbb{R}$
- Periodically extend  $\tilde{h}(t)$  into  $h(t)$

# Convolution Systems

Consider an input signal  $x(t)$  defined for  $t \in [0,1)$

Consider the convolution system:

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

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

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

# Fourier-based Analysis of Systems

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

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

We use this Fourier family for the following representations

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

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

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

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

# Convolution Systems

The convolution system:

$$\begin{aligned} y(t) &= \int_0^1 x(\tau)h(t - \tau)d\tau \\ &= \int_0^1 \left( \sum_{k=-\infty}^{\infty} a_k \beta_k^F(\tau) \right) \left( \sum_{l=-\infty}^{\infty} b_l \beta_l^F(t - \tau) \right) d\tau = \int_0^1 \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} a_k b_l \beta_k^F(\tau) \beta_l^F(t - \tau) d\tau \\ &= \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} a_k b_l \int_0^1 \beta_k^F(\tau) \beta_l^F(t - \tau) d\tau = \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} a_k b_l \beta_l^F(t) \int_0^1 \exp(i2\pi(k - l)\tau) d\tau \\ &= \sum_{k=-\infty}^{\infty} a_k b_k \beta_k^F(t) \end{aligned}$$



# Convolution Systems

The convolution system:

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

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

$$c_k = a_k b_k$$

# Derivation System

Consider an input signal  $x(t)$  defined for  $t \in [0,1)$

Consider the derivation system:

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

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

Is it linear? Is it shift invariant?

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

# Derivation System

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Consider the derivation system:

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

$$\begin{aligned} x(t) &= \sum_{k=-\infty}^{\infty} a_k \beta_k^F(t) & a_k &= \int_0^1 x(t) \exp(-i2\pi kt) dt \\ y(t) &= \sum_{k=-\infty}^{\infty} c_k \beta_k^F(t) & c_k &= \int_0^1 y(t) \exp(-i2\pi kt) dt \end{aligned}$$

# Derivation System

$$\begin{aligned}c_k &= \int_0^1 y(t) \exp(-i2\pi kt) dt = \int_0^1 \frac{dx}{dt}(t) \exp(-i2\pi kt) dt \\&= \left[ x(t) \exp(-i2\pi kt) \right]_0^1 - \int_0^1 x(t) \left( \frac{d}{dt} \exp(-i2\pi kt) \right) dt \\&= - \int_0^1 x(t) (-i2\pi k) \exp(-i2\pi kt) dt = i2\pi k \int_0^1 x(t) \exp(-i2\pi kt) dt\end{aligned}$$

$$c_k = i2\pi k a_k$$

# DFT

$$[\text{DFT}] = \frac{1}{\sqrt{N}} \begin{bmatrix} W^{*0 \cdot 0}, & W^{*1 \cdot 0}, & \dots, & W^{*(N-1) \cdot 0} \\ W^{*0 \cdot 1}, & W^{*1 \cdot 1}, & \dots, & W^{*(N-1) \cdot 1} \\ \vdots, & \vdots, & \dots, & \vdots \\ W^{*0 \cdot (N-1)}, & W^{*1 \cdot (N-1)}, & \dots, & W^{*(N-1) \cdot (N-1)} \end{bmatrix} \in \mathbb{C}^{N \times N}$$

where

$$W = \exp \left( \frac{i2\pi}{N} \right)$$

Note that the DFT matrix is symmetric and unitary.

# Representation of a Discrete Signal in the DFT Domain

The representation of the discrete signal  $\phi \in \mathbb{R}^N$  is

$$\phi^F = [\text{DFT}] \phi$$

Observe that

$$[\text{DFT}]^* \phi^F = [\text{DFT}]^* [\text{DFT}] \phi$$

$$[\text{DFT}]^* \phi^F = \phi$$

Above it's the inverse DFT procedure.

# 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  $k^{\text{th}}$  component of the DFT-domain representation of the above signal is

$$\begin{aligned}\phi^F(k) &= \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} W^{*k \cdot n} \phi(n) = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} W^{*k \cdot n} \frac{1}{2} (W^{k_0 \cdot n} + W^{-k_0 \cdot n}) \\ &= \frac{1}{\sqrt{N}} \left( \frac{1}{2} \sum_{n=0}^{N-1} W^{*k \cdot n} (W^{k_0 \cdot n}) + \frac{1}{2} \sum_{n=0}^{N-1} W^{*k \cdot n} (W^{-k_0 \cdot n}) \right)\end{aligned}$$



# Example

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

When  $k = k_0$

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

When  $k \neq k_0$

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

# Example

$$\sum_{n=0}^{N-1} W^{*k \cdot n} (W^{k_0 \cdot n}) = N\delta_{k,k_0} = \begin{cases} 1, & k = k_0 \\ 0, & \text{otherwise} \end{cases}$$

Similarly,

$$\sum_{n=0}^{N-1} W^{*k \cdot n} (W^{-k_0 \cdot n}) = N\delta_{k,-k_0} = \begin{cases} 1, & k = -k_0 \\ 0, & \text{otherwise} \end{cases}$$

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

# Example

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

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

$$[0, 0, 0, \frac{3}{2}, 0, 0, \frac{3}{2}, 0, 0]^\top$$