

# Orthogonal Signal Transforms. Fourier Series summary

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- Vector calculus
- Orthogonal functions
- Trigonometric Fourier Series
- Exponential Fourier Series
- Triangular Fourier Series
- Discrete Orthogonal Signals
- Discrete Fourier Transform
- Discrete Cosine Transform

#### **Vector calculus**



A signal x(t), or x(n), n = 0, ..., N - 1 can be represented by vectors. This is very easy for discrete-time signals:  $\mathbf{x} = [x(0), ..., x(N - 1)]^T \in \mathbb{R}^N.$ 

There are major benefits for vectorial signal representations:

- Signal analysis is linked to geometry, e.g.,
  - signal similarity between two signals x, y can be inversely proportional to Euclidean vector distance ||x - y||.



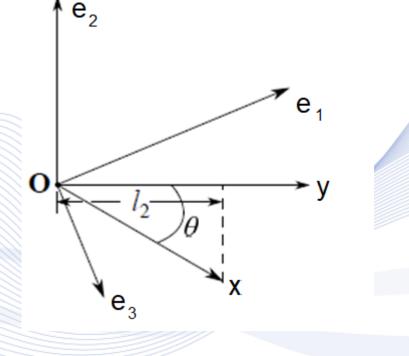
#### **Vector calculus**



**Inner vector product** or **dot product** in  $\mathbb{R}^N$ :

 $\mathbf{x}^T \mathbf{y} = \mathbf{x} \mathbf{y}^T = \sum_{n=1}^N x_n y_n = \|\mathbf{x}\| \cdot \|\mathbf{y}\| \cos \theta$ .

 $\theta$ : the angle formed by the two vectors.





#### **Vector calculus**



Properties of the inner vector product:

- It is a scalar value.
- Its value is equal to the product of the length of one vector  $||\mathbf{x}||$  and the length of the projection of the second vector  $l = ||\mathbf{y}|| \cos \theta$  on the first one.
- It is maximal for co-linear vectors of the same direction, i.e., for  $\theta = 0$ :

 $\mathbf{x}^T \mathbf{y} = \|\mathbf{x}\| \cdot \|\mathbf{y}\|.$ 





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#### **Orthogonal functions**



• Two vectors **x**, **y** are **orthogonal**, if their inner product equals 0:

$$\mathbf{x}^T \mathbf{y} = \mathbf{x} \mathbf{y}^T = \|\mathbf{x}\| \cdot \|\mathbf{y}\| \cos \theta = 0$$
,

or they are co-linear  $\theta = 0$ .

• The term orthogonality is used as a synonym for the term perpendicular.



#### **Orthogonal functions**



A vector **x** can be represented by its components  $x_n$ , n = 1, ..., N in the Euclidean space  $\mathbb{R}^N$ :

$$\mathbf{x} = \sum_{n=1}^{N} x_n \, \mathbf{e}_n$$

**Unit vectors**  $e_n$ , n = 1, ..., N are **perpendicular** to each other and orthonormal:

$$\mathbf{e}_n^T \mathbf{e}_m = \begin{cases} 1\\ 0 \end{cases}$$

n = m $n \neq m$ 

thus forming an **orthonormal coordinate system** in  $\mathbb{R}^N$ .

#### **Orthogonal functions**



Examples of orthogonal function systems:

 $\int_0^T \sqrt{\frac{2}{T}} \cos(n\Omega t) \sqrt{\frac{2}{T}} \cos(m\Omega t) dt = \begin{cases} 1, \\ 0, \end{cases}$ 

Trigonometric functions:

$$\int_{T}^{\frac{1}{T}} \sqrt{\frac{2}{T}} \cos(n\Omega t), \\ \sqrt{\frac{2}{T}} \sin(n\Omega t), \quad \Omega = \frac{2\pi}{T}, \quad n = 1, ...$$
  
Indeed, orthonormality holds:  
$$\int_{T}^{T} \sqrt{\frac{2}{T}} \int_{T}^{T} \int_{T$$

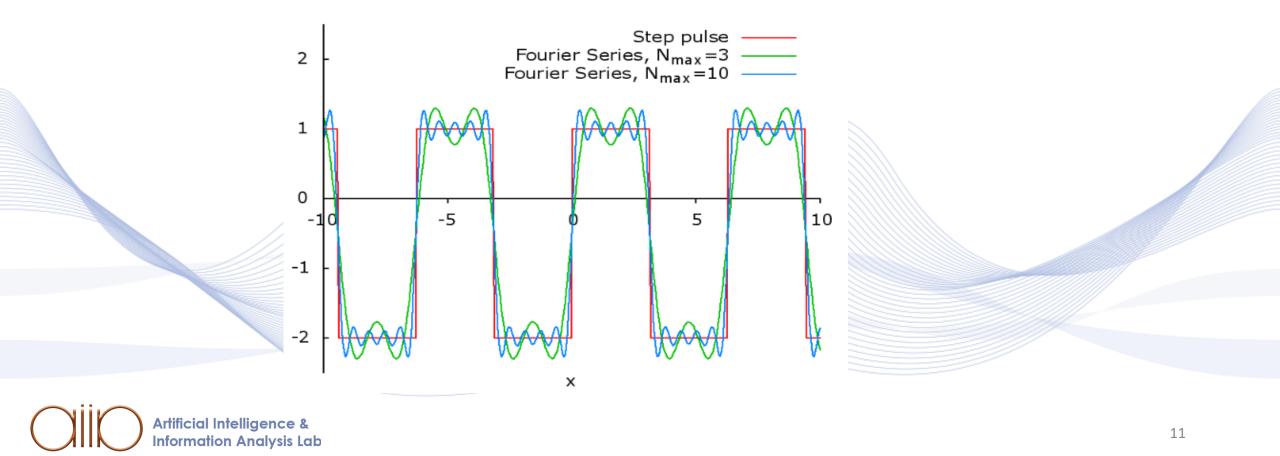
I. Pitas Digital Image Processing Fundamentals Digital Image Transform Algorithms  $n \neq m$ 



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Every periodic signal x(t), with period T can be represented by a series of sines and cosines which are harmonically related.



• Signal x(t) with period T can be represented as:

$$x(t) = \frac{a_0}{2} + \sum_{k=1}^{\infty} (a_k \cos k\Omega t + b_k \sin k\Omega t), \qquad \Omega = \frac{2\pi}{T}$$

- This forms the basis for *harmonic signal analysis*.
- Frequencies  $k\Omega = k \frac{2\pi}{T}$  are called *harmonics*.
- $a_k$  and  $b_k$  are the so-called **Fourier coefficients**.

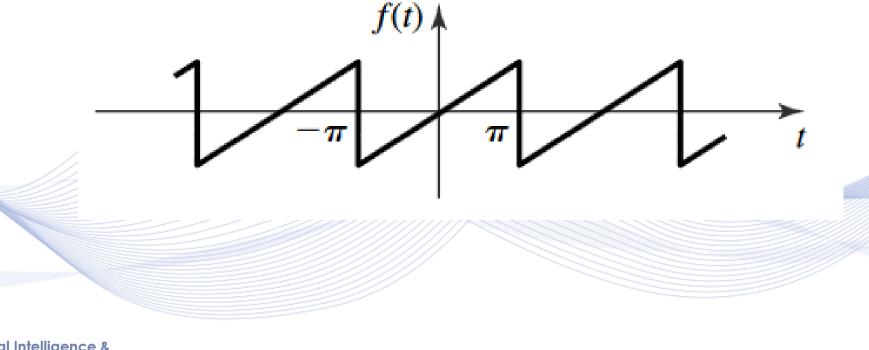
• We can calculate  $a_k$  and  $b_k$  as follows:

$$a_{0} = \frac{1}{T} \int_{T} x(t) dt, \qquad a_{k} = \frac{2}{T} \int_{T} x(t) \cos k\Omega t dt, \qquad k = 1, \dots$$

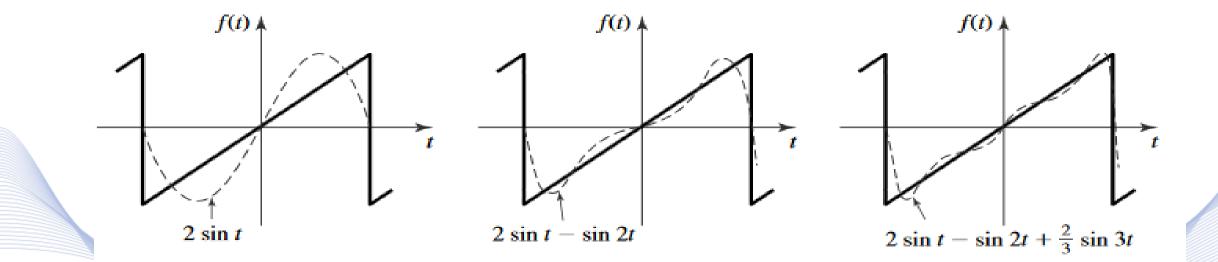
$$b_{k} = \frac{2}{T} \int_{T} x(t) \sin k\Omega t dt, \qquad k = 1, \dots$$
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**Example**: Find the Fourier Series representation of the periodic signal  $x(t) = t, -\pi < t < \pi$  in one period.







Fourier Series representation using increasing number of Fourier coefficients.





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#### **Exponential Fourier Series**



Every periodic signal x(t), with period T, can be represented as a sum of complex exponentials:

$$x(t) = \sum_{-\infty}^{\infty} X_k e^{ik\Omega t}$$
,  $\Omega = \frac{2\pi}{T}$ 

•  $X_k$  are the so-called **Fourier coefficients.** 

Relation between trigonometric and exponential Fourier Series:  $x(t) = \frac{a_0}{2} + \sum_{k=1}^{\infty} (a_k \cos k\Omega t + b_k \sin k\Omega t) = \sum_{-\infty}^{\infty} c_k e^{ik\Omega t}.$ 



#### **Exponential Fourier Series**



2

-1

-1

**Example**: Find the exponential series for the following rectangular wave with period T = 2:

Solution:

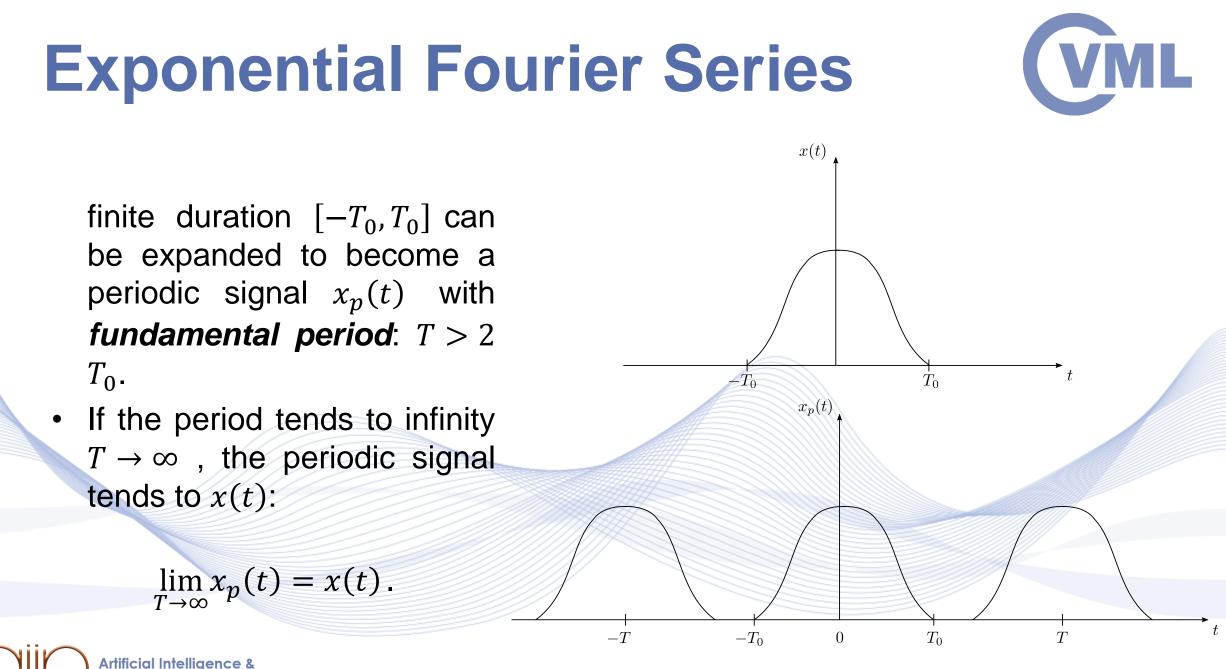
$$\Omega = \frac{2\pi}{T} = \pi.$$

$$X_0 = \frac{1}{2} \int_{-1}^{1} x(t) dt = \frac{1}{2} \int_{-1}^{0} dt - \frac{1}{2} \int_{0}^{1} dt = 0.$$

 $x(t) = \begin{cases} 1, & 0 < t < 1 \\ -1, & 1 < t < 2 \end{cases}$ 



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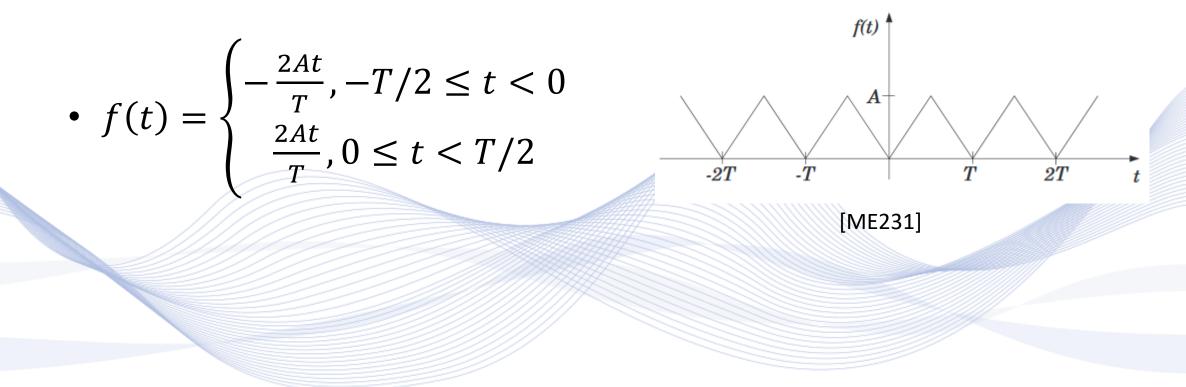


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#### **Triangular Fourier Series**

• We have the triangular wave function:







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#### **Orthogonal signals**



Two discrete-time signals x(n) and x(n) are **orthogonal** if:

$$\sum_{m=-\infty}^{\infty} x(n)y(n) = 0.$$
Discrete-time functions  $\varphi_k(n)$ ,  $k, n = 0, ..., N-1$ , are orthonormal if:  

$$\sum_{n=0}^{N-1} \varphi_k(n)\varphi_l(n) = \begin{cases} 1, & k = l \\ 0, & k \neq l \end{cases}$$



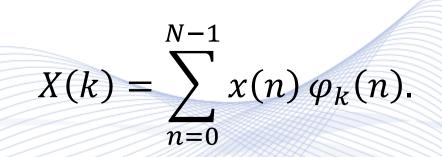
#### **Discrete Orthogonal Signals**



They form an orthonormal basis function system, to be used in function x(n) decomposition:

$$x(n) = \sum_{k=0}^{N-1} X(k) \varphi_n(k),$$

where:



#### Together they form an orthogonal signal transform pair.



#### **Discrete Orthogonal Signals**



Examples of orthogonal function systems:

• Discrete-time complex exponential signals:

$$\varphi_k(n) = e^{-i\frac{2\pi nk}{N}}, \qquad k, n = 0, ..., N-1$$

#### **Discrete-time cosine signals**:

$$\varphi_k(n) = \sqrt{\frac{2}{N} \cos \frac{(2n+1)k\pi}{2N}},$$

$$k,n=0,\ldots,N-1.$$



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Forward DFT:

$$X(k) = \sum_{n=0}^{N-1} x(n) W_N^{nk}$$

 $2\pi$ 

•  $W_N$  are the N – th complex roots of unity:

$$W_N = e^{-i\frac{2\pi}{N}},$$

Inverse DFT (IDFT):

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-nk}.$$

 $W_{N}^{N} = 1.$ 

• It is extensively used in digital signal processing/analysis.



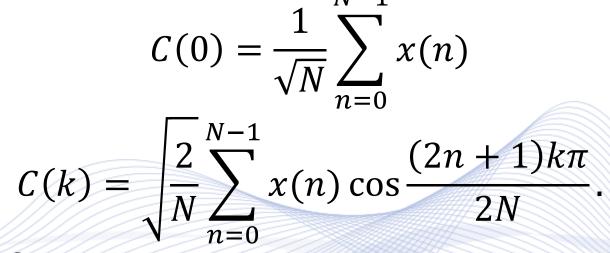


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#### **Discrete Cosine Transform**



- •DCT is used in the JPEG and MPEG standards.
- Forward DCT transform:



• Inverse DCT:

$$x(n) = \frac{1}{\sqrt{N}}C(0) + \sqrt{\frac{2}{N}}\sum_{k=1}^{N-1}C(k)\cos\frac{(2n+1)k\pi}{2N}.$$

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#### Thank you very much for your attention!

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