

CMPE323 Intro to Fourier Series

Chapter 7

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Lecture 1 1-1

Orthogonal Series



$$x(t) = \sum_{k=-\infty}^{\infty} c_k \phi_k(t)$$

• ...where $\left\{\phi_{_{\!k}}(t)\right\}$ are orthogonal...

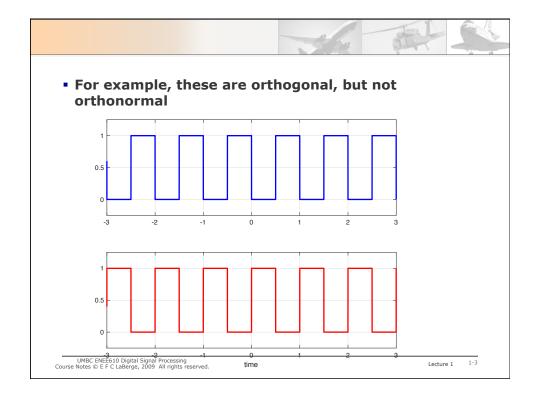
$$\int_{\alpha}^{\alpha+T} \phi_k(t) \phi_m^*(t) dt = \begin{cases} 0 & k \neq m \\ C & k = m \end{cases}$$

• ...and, ideally, orthonormal, where C=1 so that

$$\int_{\alpha}^{\alpha+T} \phi_k(t) \phi_k^*(t) dt = \int_{\alpha}^{\alpha+T} |\phi_k|^2 dt = 1$$

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Lecture 1



Complex exponentials as eigenfunctions

- We know the complex exponentials are eigenfunctions of LTI systems
- If we can decompose any generic input signal into the sum of complex exponentials...
- ...then we can write the output as the sum or superposition of scaled versions of the complex exponentials.
- This process of decomposition is called Fourier Analysis

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Lecture 1

LTI systems and eigenfunctions

The input
$$x(t) = Ae^{st}$$
 signal

The LTI
$$y(t) = H(s)Ae^{st}$$
 The output signal $H(s)$ = $H(s)e^{st}$

...and even more importantly, because the system is both linear and time invariant

$$x(t) = \underbrace{\sum \alpha_n e^{s_n(t-\tau_n)}}_{\text{system}} \underbrace{\begin{array}{c} \text{The LTI} \\ \text{system} \\ H(s) \end{array}}_{y(t) = \underbrace{\sum H(s_n)\alpha_n e^{s_n(t-\tau_n)}}_{}$$

$$y(t) = \sum_{n} H(s_n) \alpha_n e^{s_n(t-\tau_n)}$$

Now what happens if we have an arbitrary (but wellbehaved) input signal?

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Lecture 1 1-5

Fourier analysis of periodic signals

- Let be x(t) be "well-behaved" and periodic with period T
- We want to write $x(t) = \sum_{k=-\infty}^{\infty} c_k e^{j2\pi f_k t} = \sum_{k=-\infty}^{\infty} c_k e^{\frac{j2\pi kt}{T}}$
- Let's look at the following integral

$$I_{m} = \frac{1}{T} \int_{\alpha}^{\alpha+T} x(t) e^{-j\frac{2\pi mt}{T}} dt = \frac{1}{T} \int_{\alpha}^{\alpha+T} \left(\sum_{k=-\infty}^{\infty} c_{k} e^{\frac{j2\pi mt}{T}} \right) e^{-j\frac{2\pi kt}{T}} dt$$

$$= \frac{1}{T} \sum_{k=-\infty}^{\infty} c_{k} \left(\int_{\alpha}^{\alpha+T} e^{\frac{j2\pi mt}{T}} e^{-j\frac{2\pi kt}{T}} dt \right) = 0 \text{ if } m \neq k \text{ because the complex exponentials are orthogonal over one period!}$$

$$= \begin{cases} c_k & \text{if } m = k \\ 0 & \text{if } m \neq k \end{cases}$$

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Complex exponentials are an orthonormal set

$$\left|\phi_{n}\right|^{2} = \frac{1}{T_{0}} \int_{0}^{0+T_{0}} \underbrace{e^{jn\omega_{0}t}}_{\phi_{n}(t)} \underbrace{e^{-jn\omega_{0}t}}_{\phi_{n}^{*}(t)} dt = \frac{1}{T_{0}} \int_{0}^{0+T_{0}} 1 dt = 1, \ \omega_{0} \triangleq \frac{2\pi}{T_{0}}$$

And similarly for the normal equation $\frac{1}{T_0} \int_{\alpha}^{\alpha + T_0} \phi_n(t) \phi_m^*(t) = 0, \ n \neq m$

$$\phi_n \cdot \phi_k = \frac{1}{T_0} \int_0^{0+T_0} e^{jn\omega_0 t} e^{-jk\omega_0 t} dt = \frac{1}{T_0} \int_0^{0+T_0} e^{j(n-k)\omega_0 t} dt$$

$$=\frac{1}{T_0}\int_0^{0+T_0} \cos(\underbrace{(n-k)\omega_0}_{\substack{n\neq k,\ n-k\in\mathbb{Z},\\\text{so this is an}\\\text{integer number}}} t) + j\sin(\underbrace{(n-k)\omega_0}_{\substack{n\neq k,\ n-k\in\mathbb{Z},\\\text{so this is an}\\\text{integer number}}} t) dt = 0$$

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The Fourier series

• So, we can compute this interval for each k, and write

$$c_k = x(t) \cdot \phi_n(t) = \frac{1}{T} \int_{-\alpha}^{\alpha + T} x(t) e^{-j\omega_0 kt} dt, \quad \omega_0 = \frac{2\pi}{T} \quad \text{Coefficients}$$

$$x(t) = \sum_{k=-\infty}^{\infty} c_k e^{j\omega_0 kt}$$
 Fourier Series or

$$x(t) = \frac{a_0}{2} + \sum_{n=1}^{\infty} \left(a_n \cos(n\omega_0 t) + b_n \sin(n\omega_0 t) \right)$$

$$a_n = \frac{2}{T} \int_{\alpha}^{\alpha + T} x(t) \cos(n\omega_0 t) dt, \ b_n = \frac{2}{T} \int_{\alpha}^{\alpha + T} x(t) \sin(n\omega_0 t) dt$$

 ...and, under reasonable conditions, the infinite sum is not an approximation, but an equality

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Let's try it

 This particular waveform is both real and even

real:
$$Re[x(t)] = x(t)$$

even:
$$x(t) = x(-t)$$

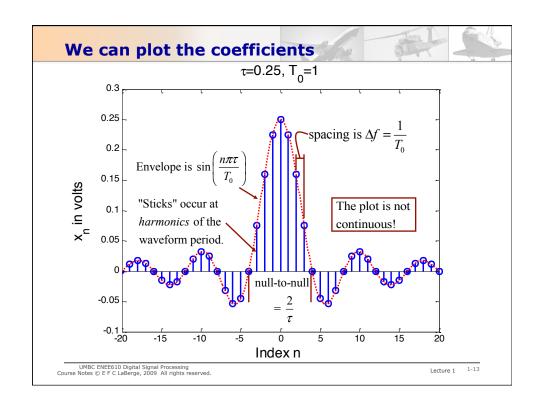
$$c_k = \frac{1}{T_0} \int_{-T_0/2}^{T_0/2} x(t) e^{-j\frac{2\pi kt}{T}} dt = \frac{A}{T_0} \int_{-\tau/2}^{\tau/2} e^{-j\frac{2\pi kt}{T}} dt$$

$$c_{k} = \frac{A}{T_{0}} \left(\frac{1}{-j\omega_{0}k} e^{-j\omega_{0}kt} \right)_{-\tau/2}^{\tau/2} = \frac{A}{T_{0}} \left(\frac{e^{j\omega_{0}k\tau/2} - e^{-j\omega_{0}k\tau/2}}{j\omega_{0}k} \right)$$

$$= \frac{A}{T_0} \left(\frac{2j\sin(\omega_0 k\tau/2)}{j\omega_0 k} \right) \left(\frac{\tau/2}{\tau/2} \right)$$

$$= \frac{A\tau}{T_0} \frac{\sin(n\pi\tau / T_0)}{(n\pi\tau / T_0)} @ \frac{A\tau}{T_0} \operatorname{sinc} \left(\frac{n\tau}{T_0}\right) \longrightarrow \frac{\text{notice the } \pi \text{ in the definition}}{\text{the definition}}$$

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So this means



$$x(t) = \sum_{k = -\infty}^{\infty} c_k e^{j\frac{2\pi kt}{T_0}} = \frac{A\tau}{T_0} \sum_{k = -\infty}^{\infty} \operatorname{sinc}\left(\frac{n\tau}{T_0}\right) e^{j\frac{2\pi kt}{T_0}}$$

...which is defined only at integer multiples of

$$f_0 = \frac{\omega_0}{2\pi} = \frac{1}{T_0}$$

 This expression is known as the Fourier Series, and it relates the time domain (x(t)) to the frequency domain

 $\{c_k\}$ or $X(f) = \sum_{k=-\infty}^{\infty} c_k \delta \left(f - \frac{k}{T} \right)$

And the inverse relationship hold as well

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- But what if the signal is not periodic?

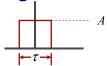
 There is an equivalent result for non-periodic signals
- A non-periodic signal can be viewed as the limit of a periodic signal as $T_0 \rightarrow \infty$

$$c_{k} = \frac{1}{T_{0}} \int_{-T_{0}/2}^{T_{0}/2} x(t) e^{-j\frac{2\pi kt}{T_{0}}} dt \quad \frac{k}{T_{0}} \to f \quad \frac{T_{0}}{2} \to \infty \quad \frac{1}{T_{0}} \to df$$

$$X(f) \bigotimes_{-\infty}^{\infty} x(t)e^{-j2\pi ft}dt$$
 contribution between f and $f + df$

- X(f) is called the Fourier Transform, or the voltage spectrum of x(t),
- ...and we have $x(t) = \int_{-\infty}^{\infty} X(f)e^{j2\pi ft}df = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega)e^{j\omega t}d\omega$
- Notice $c_n = \frac{1}{T_0} X(f)_{f = \frac{n}{T_0}} = \frac{1}{T_0} X(\frac{n}{T_0})$

But what if the signal is not periodic?



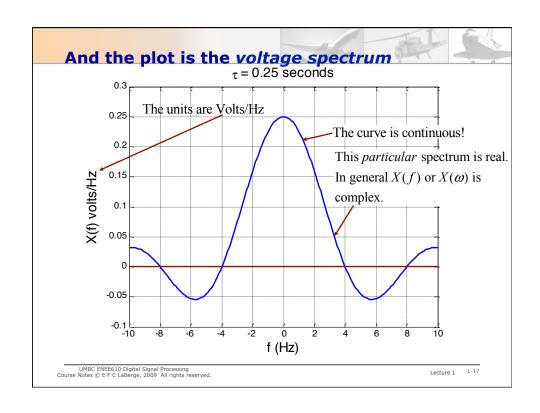
We define the Fourier Transform of x(t)

$$X(f) = \int_{-\infty}^{\infty} x(t)e^{-j2\pi ft}dt \text{ or } \int_{-\infty}^{\infty} x(t)e^{-j\omega t}dt = X(\omega)$$

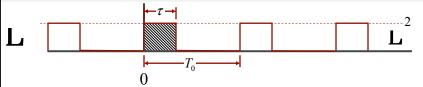
• In this case
$$X(f) = \int_{-\infty}^{\infty} x(t)e^{-j2\pi ft}dt = \int_{-\tau/2}^{\tau/2} Ae^{-j2\pi ft}dt = \frac{A}{-j2\pi f}e^{-j2\pi ft}\Big|_{-\tau/2}^{\tau/2}$$

$$= \frac{A}{j2\pi f} \Big(e^{j2\pi f\pi^2} - e^{-j2\pi f\pi^2}\Big) = \frac{A}{\pi f} \frac{\Big(e^{j\pi f\tau} - e^{-j\pi f\tau}\Big)}{j2} = A\tau \frac{\sin(\pi f\tau)}{\pi f\tau} = A\tau \text{sinc}(f\tau)$$

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Now you try (in class example) (10 minutes)



Group 1A: Compute a_n

Group 1B: Compute b_{r}

Group 2: Compute x_n (Hint factor out $e^{-j\frac{\pi n}{T_0}}$ and simplify)

Group 3: Compute the Fourier Transform, X(f) of the cross-hatched waveform

Detailed solutions will be posted on Blackboard tonight!

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Properties of Fourier Transform

- **Duality:** $X(f) = \mathbf{F}(x(t)) \Rightarrow x(f) = \mathbf{F}(X(-t))$ and $x(-f) = \mathbf{F}(X(t))$
- Linearity: $z(t) = ax(t) + by(t) \Rightarrow Y(f) = aX(f) + bY(f)$
- Time Shift: $\mathbf{F}\left(x(t-t_0)\right) = e^{-j2\pi f t_0} \mathbf{F}\left(x(t)\right)$
- Scaling: For $a \neq 0 \in {}^{\circ}$, $\mathbf{F}(x(at)) = \frac{1}{|a|} X\left(\frac{f}{a}\right)$
- Modulation: $\mathbf{F}\left(x(t)e^{j2\pi f_0t}\right) = X(f-f_0)$
- Conjugation: $\mathbf{F}\left(x^*(t)\right) = X^*(-f)$
- Parseval: $\int_{-\infty}^{\infty} x(t)y^*(t)dt = \int_{-\infty}^{\infty} X(f)Y^*(f)df$
- Rayleigh $\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} |X(f)|^2 df$

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Advanced Properties of Fourier Transform

- Integration: $\mathbf{F}\left(\int_{-\infty}^{t} x(t)dt\right) = \frac{X(f)}{j2\pi f} + \frac{1}{2}X(0)\delta(f)$
- Differentiation: $\mathbf{F}\left(\frac{d}{dt}x(t)\right) = j2\pi fX(f)$
- Moments: $\int_{-\infty}^{\infty} t^n x(t) dt = \left(\frac{j}{2\pi}\right)^n \frac{d^n}{df^n} X(f)$

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Lecture 1 1-20

The convolution theorem (very important!)

• The output of a LTI system with transfer function H(f)

$$Y(f) = X(f)H(f) \qquad y(t) = \int_{-\infty}^{\infty} Y(f)e^{j2\pi ft}df = \int_{-\infty}^{\infty} X(f)H(f)e^{j2\pi ft}df$$

$$\text{Write } H(f) = \int_{-\infty}^{\infty} h(\tau)e^{-j2\pi f\tau}d\tau$$

$$y(t) = \int_{-\infty}^{\infty} X(f) \left(\int_{-\infty}^{\infty} h(\tau)e^{-j2\pi f\tau}d\tau\right)e^{j2\pi ft}df$$

$$= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} X(f)h(\tau)e^{j2\pi f(t-\tau)}df\right)d\tau$$

$$= \int_{-\infty}^{\infty} h(\tau) \left(\int_{-\infty}^{\infty} X(f)e^{j2\pi f(t-\tau)}df\right)d\tau$$

$$= \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau = \int_{-\infty}^{\infty} h(t-\tau)x(\tau)d\tau$$

$$\frac{\text{UMBC ENEE 61 0 Digital Signal Processing}}{\text{Course Notes (0 E F C LaBerue, 2009 All rights reserved,}} @x(t)*h(t) \text{ the CONVOLUTION} \frac{1}{\text{scture 1}} = \frac{1-21}{1-21}$$

Summary

- We can decompose a periodic signal into the weighted sum of complex exponentials,...
- ...or, equivalently, to the weighted sum of sines and cosines.
- We write the weighted sum as a Fourier Series
- The frequency-domain representation consists of a series of harmonically-related terms, with separation equal to the period of the signal,
- The coefficients have units of volts (or amps)
- We can decompose a non-periodic signal into the weighted integral of complex exponentials,...
- ...or, equivalently, to the weighted integral of sines and cosines
- The frequency-domain representation has a continuous spectrum with units of volts/Hz (or amps/Hz).

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