Introduction to Audio Content Analysis

Module 3.1: Input Representation — Signals

alexander lerch



introduction overview

corresponding textbook section

Section 3.1

lecture content

- deterministic & periodic signals
- Fourier Series
- random signals
- statistical signal description
- digital signals

■ learning objectives

- name basic signal categories
- discuss the nature of periodic signals with respect to harmonics
- give a short description of meaning and use of the Fourier Series
- list common descriptors for properties of a random signal



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deterministic signals:

predictable: future shape of the signal can be known (example: sinusoidal)

random signals: unpredictable: no knowledge can help to predict what is coming next (example white noise)

"real-world" audio signals can be modeled as time-variant combination of

- (quasi-)periodic parts
- (quasi-)random parts

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audio signals periodic signals 1/5

periodic signals: most prominent examples of deterministic signals

$$x(t) = x(t + T_0)$$

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

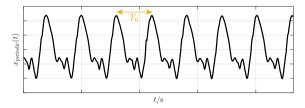
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periodic signals: most prominent examples of deterministic signals

$$x(t) = x(t + T_0)$$

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$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{j\omega_0 kt}$$

- $\omega_0 = 2\pi \cdot f_0$
- $\mathbf{k}\omega_0$: integer multiples of the lowest frequency
- $= e^{j\omega_0kt} = \cos(\omega_0kt) + i\sin(\omega_0kt)$
- a_k: Fourier coefficients amplitude of each component

$$a_k = rac{1}{T_0} \int\limits_{-T_0/2}^{T_0/2} x(t) \mathrm{e}^{-\mathrm{j}\omega_0 kt} \, dt$$

¹ Jean-Baptiste Joseph Fourier, 1768–1830

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Fourier series

- **every** periodic signal can be represented in a Fourier series
- \blacksquare a periodic signal **contains only** frequencies at integer multiples of the fundamental frequency f_0
- Fourier series can only be applied to periodic signals
- Fourier series is analytically elegant but only of limited practical use as the fundamental period has to be known



Fourier series

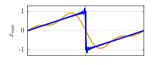
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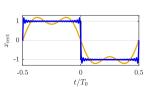


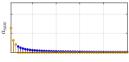
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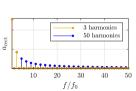
reconstruction of periodic signals with limited number of sinusoidals:

$$\hat{x}(t) = \sum_{k=-\mathcal{K}}^{\mathcal{K}} a_k e^{\mathrm{j}\omega_0 kt}$$



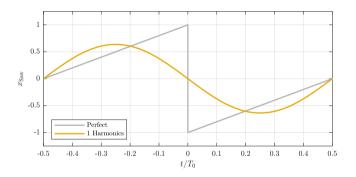






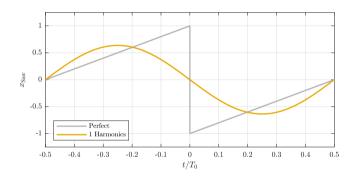
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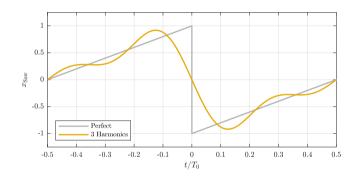














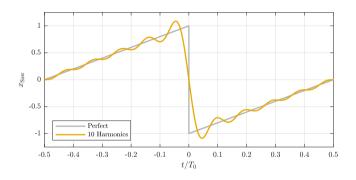
















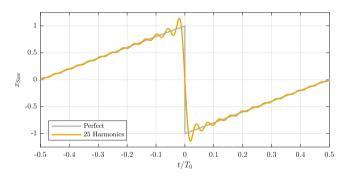








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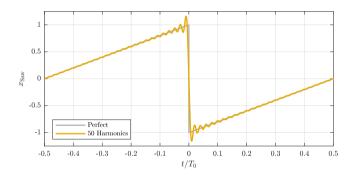












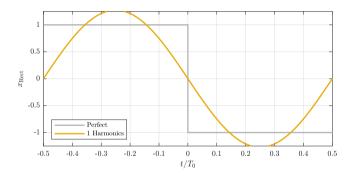








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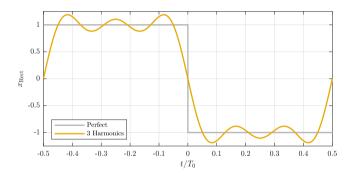












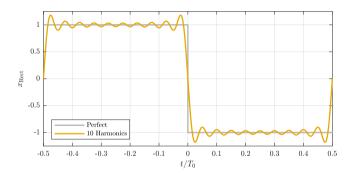














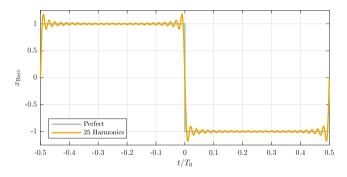






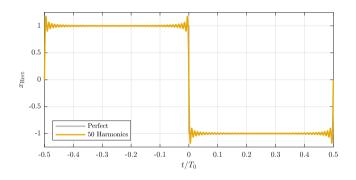








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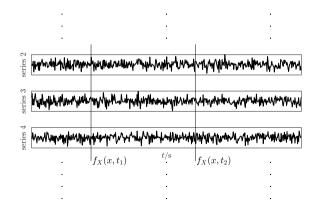


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audio signals random process 1/2

random process: ensemble of random series



random process

- ensemble of random series
- each series represents a sample of the process
- the following value is *indetermined*, regardless of any amount of knowledge
- special case: stationarity statistical properties such as the mean are time invariant
- example: white noise



statistical signal description probability density function

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PDF $p_{x}(x)$

- x-axis: possible (amplitude) values
- y-axis: probability

$$p_X(x) \geq 0$$
, and
$$\int_{-\infty}^{\infty} p_X(x) dx = 1$$

RFD—Relative Frequency Distribution (sample of PDF) histogram of (amplitude) values

statistical signal description probability density function

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statistical signal description

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statistical signal description PDF examples

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What is the PDF of the following prototype signals:



statistical signal description PDF examples

0

What is the PDF of the following prototype signals:

- square wave
- sawtooth wave
- sine wave
- white noise (uniform, gaussian)
- DC

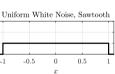
statistical signal description PDF examples

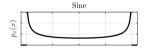
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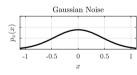
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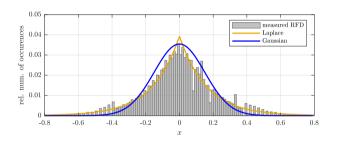








statistical signal description RFD: real world signals



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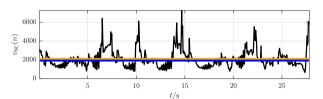
statistical signal description arithmetic mean

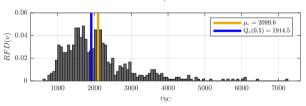
 \blacksquare from time series x:

$$\mu_{\mathsf{x}}(n) = \frac{1}{\mathcal{K}} \sum_{i=i_{\mathsf{s}}(n)}^{i_{\mathsf{e}}(n)} \mathsf{x}(i)$$

• from distribution p_x :

$$\mu_{x}(n) = \sum_{x=-\infty}^{\infty} x \cdot p_{x}(x)$$





statistical signal description geometric & harmonic mean

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■ geometric mean

$$\begin{aligned} \mathrm{Mg}_{v} &= & \sqrt[N]{\prod_{0}^{\mathcal{N}-1} v(n)} \\ &= & \exp\left(\frac{1}{\mathcal{N}} \sum_{0}^{\mathcal{N}-1} \log\left(v(n)\right)\right). \end{aligned}$$

harmonic mean

$$\mathrm{Mh}_{v} = \frac{\mathcal{N}}{\sum\limits_{n=1}^{N-1} 1/v(n)}$$

signal description 00000000

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statistical signal description geometric & harmonic mean

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statistical signal description variance & standard deviation

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measure of spread of the signal around its mean

variance

• from signal block:

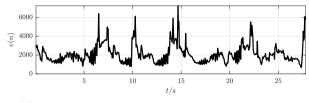
$$\sigma_x^2(n) = \frac{1}{\mathcal{K}} \sum_{i=i_s(n)}^{i_e(n)} (x(i) - \mu_x(n))^2$$

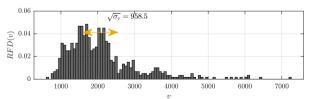
• from distribution:

$$\sigma_{x}^{2}(n) = \sum_{x=-\infty}^{\infty} (x - \mu_{x})^{2} \cdot \rho_{x}(x)$$

standard deviation

$$\sigma_{\mathsf{x}}(n) = \sqrt{\sigma_{\mathsf{x}}^2(n)}$$





statistical signal description variance & standard deviation

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measure of *spread* of the signal around its mean

variance

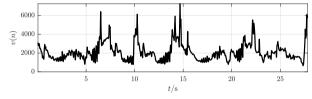
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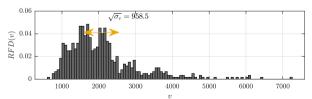
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from distribution:

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$$\sigma_{\mathsf{x}}(n) = \sqrt{\sigma_{\mathsf{x}}^2(n)}$$





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statistical signal description variance & standard deviation

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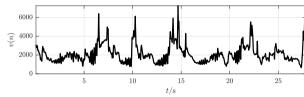
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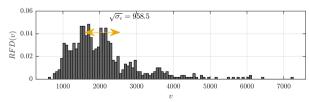
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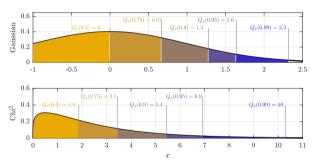


statistical signal description quantiles & quantile ranges

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dividing the PDF into (equal sized) subsets

$$Q_{\mathrm{X}}(c_{p}) = \operatorname{argmin} \left(F_{\mathrm{X}}(x) \leq c_{p}
ight)$$
 with $F_{\mathrm{X}}(x) = \int\limits_{-\infty}^{x} p_{\mathrm{x}}(y) \, dy$



■ median

$$Q_{\rm X}(0.5) = \operatorname{argmin} \left(F_{\rm X}(x) \le 0.5 \right)$$

- **quartiles**: $Q_X(0.25)$, $Q_X(0.5)$, and $Q_X x(0.75)$
- **quantile range**, e.g.

$$\Delta Q_{\rm X}(0.9) = Q_{\rm X}(0.95) - Q_{\rm X}(0.05)$$

- signals can be categorized into **deterministic and random signals**
 - deterministic signal can be described in a mathematical function
 - random processes can only be described by their general properties

periodic signals

- periodic signals are probably the most music-related deterministic signal
- any periodic (pitched) signal is a sum of weighted sinusoidals
- frequencies only at the fundamental frequency and integer multiples
- random signals
 - noise, unpredictable
- real-world signals
 - can be seen as a time-varying mixture of these two signal categories

statistical features

- summarize technical signal characteristics in few numerical values
- may be used on a time domain, frequency domain, or feature domain signal

