

Introduction to Audio Content Analysis

module 3.1: input representation — signals

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 overview
 intro
 periodic signals
 random signals
 signal description
 summa

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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audio signals signal categories

- 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

audio signals signal categories



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)

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audio signals signal categories

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periodic signals: most prominent examples of deterministic signals

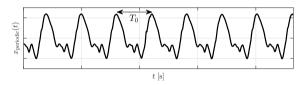
$$x(t) = x(t + T_0)$$

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

periodic signals: most prominent examples of deterministic signals

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

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

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

¹ Jean-Baptiste Joseph Fourier, 1768–1830

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audio signals periodic signals 2/5

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erview intro **periodic signals** random signals signal description summar

audio signals periodic signals 3/5



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



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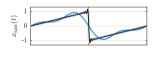
Fourier series

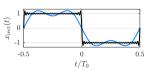
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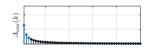


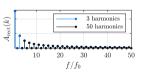
reconstruction of periodic signals with limited number of sinusoidals:

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

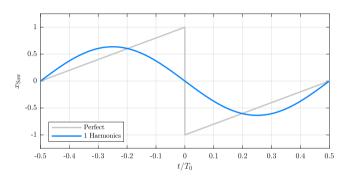




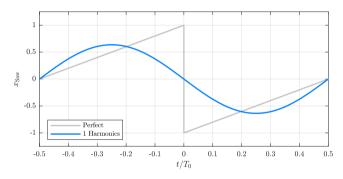




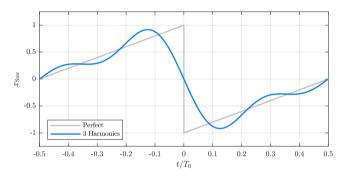




1 3 10 25 50 harmonics **◄)) ◄)) ◄)) ◄)) ◄))**

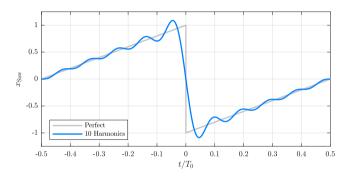




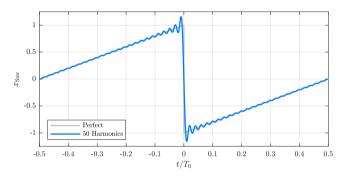


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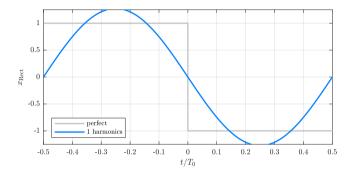




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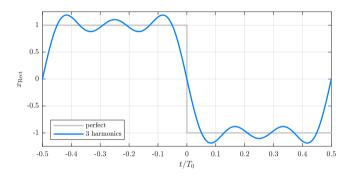




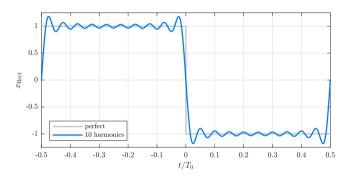


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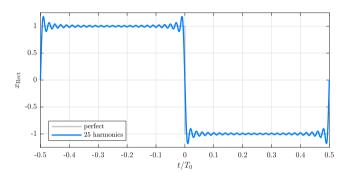




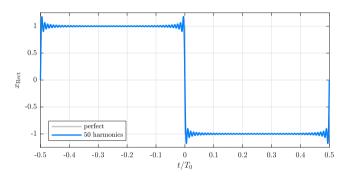


1 3 10 25 50 harmonics

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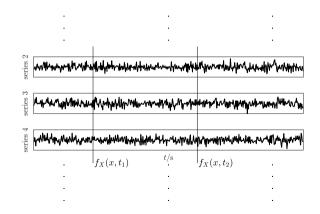




1 3 10 25 50 harmonics

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random process: ensemble of random series



audio signals random process 2/2



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



PDF $p_{x}(x)$

■ x-axis: possible (amplitude) values

y-axis: probability

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

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

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statistical signal description probability density function

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statistical signal description PDF examples

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What is the PDF of the following prototype signals:



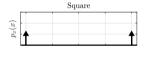
statistical signal description PDF examples

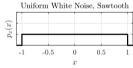


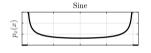
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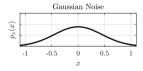
- square wave
- sawtooth wave
- sine wave
- white noise (uniform, gaussian)
- DC

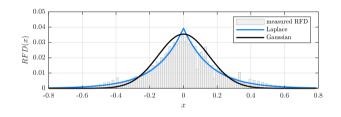
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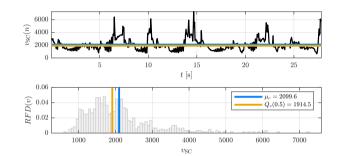
statistical signal description arithmetic mean

 \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

$$\mathrm{Mh}_{v} = \frac{\mathcal{N}}{\sum\limits_{0}^{\mathcal{N}-1} 1/v(n)}$$

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statistical signal description geometric & harmonic mean

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measure of spread of the signal around its mean

variance

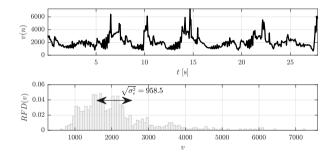
• from signal block:

$$\sigma_{x}^{2}(n) = \frac{1}{\mathcal{K}} \sum_{i=i_{s}(n)}^{i_{e}(n)} (x(i) - \mu_{x}(n))^{2}$$

from distribution:

$$\sigma_x^2(n) = \sum_{x=-\infty}^{\infty} (x - \mu_x)^2 \cdot \rho_x(x)$$

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



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measure of spread of the signal around its mean

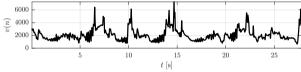
variance

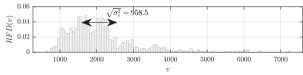
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from distribution:

$$\sigma_x^2(n) = \sum_{x=-\infty}^{\infty} (x - \mu_x)^2 \cdot p_x(x)$$





standard deviation

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

1000

2000

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measure of spread of the signal around its mean

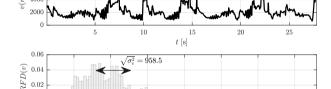
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4000

5000

6000

3000

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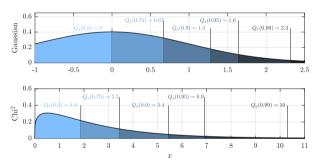
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statistical signal description quantiles & quantile ranges

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dividing the PDF into (equal sized) subsets

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



statistical signal description quantile examples

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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.

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

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summary lecture content



- 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

