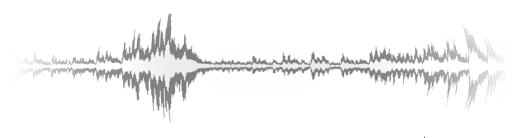
# Digital Signal Processing for Music Part 4: Signal Description

alexander lerch





introduction

- ergodic signals do not have a functional description
- other ways of describing these signals have to be found

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- we are looking for time-independent descriptions

## description of (random) signals

- ergodic signals do not have a functional description
- ⇒ other ways of describing these signals have to be found
- ergodic signal characteristics are not time variant
- ⇒ we are looking for time-independent descriptions
- these descriptions might also be convenient to use for some deterministic signals

probability and occurrence

N: number of overall observations  $N(x_i)$ : number of occurrences of symbol  $x_i$ 

relative number of occurrences:

$$\hat{p}_i = \frac{N(x_i)}{N}$$

probability:

$$p_i = \lim_{N \to \infty} \frac{N(x_i)}{N}$$

### properties

$$\sum_{i} p_{i} = 1$$

$$0 < p_{i} < 1$$

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## audio signal description probability and occurrence

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# audio signal description probability distribution example

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roll of a die

# audio signal description probability distribution example

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roll of a die

value 1 2 3 4 5 6 
$$p(x)$$
  $\frac{1}{6}$   $\frac{1}{6}$   $\frac{1}{6}$   $\frac{1}{6}$   $\frac{1}{6}$   $\frac{1}{6}$   $\frac{1}{6}$ 

probability distribution for the roll of two dice



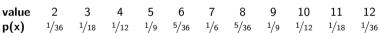
# audio signal description probability distribution example

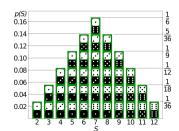


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### probability distribution for the roll of two dice







## audio signal description continuous probability density distribution

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 $i \rightarrow \text{continuous} \Rightarrow PDF$ 

$$\int_{-\infty}^{\infty} p_X(x) dx = 1$$

$$0 \le p_X(x)$$

•000000

$$\int_{0}^{x_{c}} p_{X}(x) dx$$

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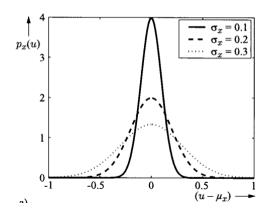
probability of x being a value smaller than or equal  $x_c$ 

$$\int_{-\infty}^{\infty} p_X(x) dx$$

## audio signal description

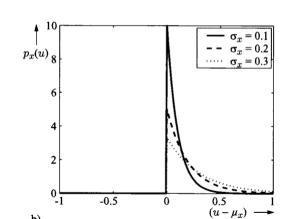
### example PDF: Gaussian

$$p_X(x) = \frac{1}{\sqrt{2\pi}\sigma_X} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}}$$



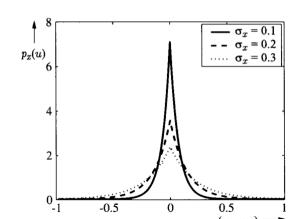
# audio signal description example PDF: Exponential

$$p_X(x) = \begin{cases} \frac{1}{\sigma_X} e^{-\frac{x}{\sigma_X}} & x > 0\\ 0 & \text{else} \end{cases}$$



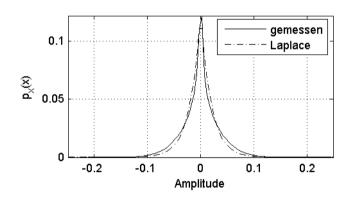
# audio signal description example PDF: Laplace (2-sided exp)

$$p_X(x) = \frac{1}{\sqrt{2}\sigma_X} e^{-\sqrt{2}\frac{|x-\mu_X|}{\sigma_X}}$$

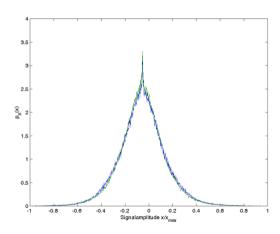


# audio signal description measured RDF





# audio signal description measured RDF



# audio signal description PDFs of generated signals 1/2

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describe the shape of the following PDFs

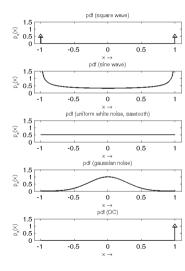


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## describe the shape of the following PDFs

- white noise (uniform)
- white noise (Gaussian)
- DC
- square
- sinusoidal
- sawtooth





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Example: average grade, five students, grades: 1, 2, 1, 3, 5

$$\hat{\mu}_X = \frac{1+2+1+3+5}{5} = 2.4$$

Grade	# occurrences	relative frequency
1	2	2/5
2	1	1/5
3	1	1/5
4		
5	1	1/5

# audio signal description expected value 1/3

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## audio signal description expected value 2/3

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$$\mu = \frac{2}{5} \cdot 1 + \frac{1}{5} \cdot 2 + \frac{1}{5} \cdot 3 + \frac{0}{5} \cdot 4 + \frac{1}{5} \cdot 5 = 2.4$$

$$\mu_X = \sum_{\forall x} p(x) \cdot x$$

$$\mu_X = \mathcal{E}\{X\} = \int x p_X(x) dx$$

## audio signal description expected value 2/3

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generalization:

$$\mathcal{E}\{f(X)\} = \sum_{i} f(x)p(x)$$

examples:

- mean: f(x) = x
  - quad. mean:  $f(x) = x^2$

### Georgia **Center for Music** Tech | Technology College of Design

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# audio signal description (central) moments 1/2

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kth moment

$$\mathcal{E}\{X^k\} = \int_{-\infty}^{+\infty} x^k p_X(x) dx$$

kth central momen

$$\mathcal{E}\{(X-\mu_X)^k\} = \int_{-\infty}^{+\infty} (x-\mu_X)^k p_X(x) dx$$

• example: 2nd order central moment: Variance

$$\sigma_X^2 = \mathcal{E}\{(X - \mu_X)^2\} = \int_0^{+\infty} (x - \mu_X)^2 p_X(x) dx$$

## audio signal description (central) moments 1/2

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### calculation of moments

(central) moments (mean, power, variance, etc.) can be computed from

- the signal
- the signal's PDF



## audio signal description central moments summary

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order	name	obs (cont)	pdf (cont)
1	$\mu_X$	$\frac{1}{T}\int\limits_{-T/2}^{T/2}x(t)dt$	$\int\limits_{-\infty}^{\infty} x p_X(x) dx$
2	$\sigma_X^2$	$\frac{1}{T}\int\limits_{-T/2}^{T/2}(x(t)-\mu_X)^2dt$	$\int_{-\infty}^{\infty} (x - \mu_X)^2 p_X(x) dx$

standard deviation  $\sigma_X = \sqrt{\sigma_X^2}$ 

- PDF can tell us many important details about a signal
- statistical measures can be used to describe signal properties
- statistical measures can be derived from both the time domain signal and its pdf
- often-used measures are:
  - mean and median
  - variance and standard deviation
  - higher order moments less frequently (skewness, kurtosis)
  - other pdf description possible (quartile-distances etc.)

### audio signal description summary



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## audio signal description



- Open PDF can tell us many important details about a signal
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