

Advanced Signals and Systems

Part 1: Introduction

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Digital Signal Processing and System Theory



Contents of the Lecture

Today:

- ❑ Boundary conditions of the lecture
 - ❑ Contents
 - ❑ Literature hints
 - ❑ Exams
- ❑ Start with first part (discrete signals and processes) of the lecture

Contents of the Lecture

Entire Semester:

- ❑ Introduction
- ❑ Discrete signals and random processes
- ❑ Spectra
- ❑ Discrete systems
- ❑ Idealized linear, shift-invariant systems
- ❑ Hilbert transform
- ❑ State-space description and system realizations
- ❑ Generalizations for signals, systems, and spectra

English (and German) Books:

Statistical signal theory:

- ❑ A. Papoulis: ***Probability, Random Variables, and Stochastic Processes***, McGraw Hill, 1965
- ❑ E. Hänsler: ***Statistische Signale: Grundlagen und Anwendungen***, Springer, 2001
(in German)

Discrete signal processing:

- ❑ A. V. Oppenheim, R. W. Schaffer: ***Discrete-Time Signal Processing***, 2nd edition, Prentice Hall, 1999
- ❑ J. G. Proakis, D. K. Manolakis: ***Digital Signal Processing***, 4th edition, Prentice Hall, 2006
- ❑ K. D. Kammeyer, K. Kroschel: ***Digitale Signalverarbeitung - Filterung und Spektralanalyse mit MATLAB-Übungen***, Teubner, 2002 (in German)

Signal processing:

- ❑ A. V. Oppenheim, A. S. Willsky, S. Hamid: ***Signals and Systems***, 2nd edition, Prentice Hall, 1996
- ❑ S. Haykin, B. Van Veen: ***Signals and Systems***, 2nd edition, Wiley, 2002

Boundary Condition of the Lecture

Credit Points, Exams, Exercises, and Lecture Notes

Credit points:

- ❑ 7 ECTS points

Written exam:

- ❑ 90 minutes test
- ❑ In the exams period

Exercises:

- ❑ Every week two hours (45 min) during the semester

Lecture notes:

- ❑ Printed versions will be spread at the beginning of each lecture
- ❑ In the internet as pdf files via www.dss.tf.uni-kiel.de

Introduction

Origin of this Lecture

Thanks to ...

*... Prof. Dr.-Ing. Ulrich Heute
(slides are based on his script)*



Prof. Heute has given this lecture until 2010.

The DSS Team

People that you can ask for help

- ❑ The DSS team has no “question hours” ... just come over, someone will have time for you.
- ❑ Details (where we are located, etc.) can be found via www.dss.tf.uni-kiel.de.

Exercises

- ❑ M.Sc. Anne Theiß



Some General Words about Signals and Systems

“Advanced” Signals and Systems:

We will treat the basic theory of **continuous, deterministic, one-dimensional signals**. We assume that **basic knowledge about continuous, stochastic, one-dimensional signals** as well as theory of **continuous, one-dimensional systems with one input and one output are known**. What we will treat here are:

- ❑ **Signals** (discrete signals, sequences) representing (in an abstract manner) any entity depending on any independent variable (e.g. pixel brightness on a 2-D grid). Please note that we call signals to be **digital**, if a discrete signal has also a discrete (quantized) amplitude.
- ❑ **Systems** are **operators** that are excited by input sequences, creating internal (state) sequences and output sequences. **Digital systems** are operators that are working on **digital inputs** with **digital parameters** (e.g. quantized coefficients) and quantized states yielding **digital outputs**.

In the following ...

... we will restrict ourselves **to discrete signals and systems** with a **short extension** towards **digital signals** and some references to **continuous (analog) signals and systems**.

Notation – Part 1

Scalars and Vectors

Scalars:

□ Signals: $v(n)$ *Discrete time index*

□ Impulse responses (time-variant): $h_i(n)$ *Coefficient index*

□ Example for a (real) convolution:
$$y(n) = \sum_{i=0}^{N-1} v(n-i) h_i(n)$$

Vectors:

□ Signal vectors: $\mathbf{v}(n) = [v(n), v(n-1), \dots, v(n-N+1)]^T$ *Boldface and lowercase*

□ Impulse response vectors (time-variant): $\mathbf{h}(n) = [h_0(n), h_1(n), \dots, h_{N-1}(n)]^T$

□ Example for a real convolution: $y(n) = \mathbf{v}^T(n) \mathbf{h}(n) = \mathbf{h}^T(n) \mathbf{v}(n)$

Matrices:

$\mathbf{A}(n) = \begin{bmatrix} a_{00}(n) & a_{01}(n) & \dots & a_{0N}(n) \\ a_{10}(n) & a_{11}(n) & \dots & a_{1N}(n) \\ \vdots & \vdots & & \vdots \\ a_{M0}(n) & a_{M1}(n) & \dots & a_{MN}(n) \end{bmatrix}$ *Boldface and uppercase*

Notation – Part 2

Signals – Part 2

Signals (Details):

□ Notation:

$$\{v(n), n \in \mathbb{Z}\} \xrightarrow{\text{sloppy}} v(n)$$

with

$$v(n) \in \mathbb{C}.$$

Non-quantized, complex quantity!

□ Signal vector versus M –dimensional signals:

$$\mathbf{v}(n) \quad \text{E.g. speed of an object (x, y, and z-direction)}$$

$$v(\mathbf{n}) = v(n_0, n_1, \dots, n_{M-1})$$

E.g. brightness of a picture (M=2)

During the lecture we will focus on one-dimensional scalar signals!

Signals and Random Processes

Random variables and processes:

□ Notation: $x(n), x_1(n), x_2(n)$

↙ **No differences between deterministic signals and random processes – different writing styles: $x(\eta, n), x(\omega, n), X(n)$**

□ Probability density function: $f_x(x, n), f_{x_1 x_2}(x_1, x_2, n_1, n_2)$

□ Stationary random processes: $f_x(x, n) = f_x(x, n + n_0) = f_x(x)$
 $f_{x_1 x_2}(x_1, x_2, n_1, n_2) = f_{x_1 x_2}(x_1, x_2, n_1 + n_0, n_2 + n_0)$
 $= f_{x_1 x_2}(x_1, x_2, n_2 - n_1)$

□ Expected values of stationary random processes:

$$\mathbb{E}\{x(n)\} = \int_{x=-\infty}^{\infty} x f_x(x) dx = \mu_x^{(1)} = \mu_x$$

$$\mathbb{E}\{x^2(n)\} = \int_{x=-\infty}^{\infty} x^2 f_x(x) dx = \mu_x^{(2)}, \quad \mathbb{E}\{g(x(n))\} = \int_{x=-\infty}^{\infty} g(x) f_x(x) dx$$

Auto and cross correlation for real, stationary random processes:

- Auto-correlation function:

$$\mathbb{E}\{v(n)v(n+l)\} = s_{vv}(l)$$

- Cross-correlation function:

$$\mathbb{E}\{v(n)y(n+l)\} = s_{vy}(l)$$

- (Auto) power spectral density:

$$S_{vv}(\Omega) = \sum_{l=-\infty}^{\infty} \mathbb{E}\{v(n)v(n+l)\} e^{-j\Omega l} = \sum_{l=-\infty}^{\infty} s_{vv}(l) e^{-j\Omega l}$$

- (Cross) power spectral density:

$$S_{vy}(\Omega) = \sum_{l=-\infty}^{\infty} \mathbb{E}\{v(n)y(n+l)\} e^{-j\Omega l} = \sum_{l=-\infty}^{\infty} s_{vy}(l) e^{-j\Omega l}$$

White Noise

Stationary white noise:

- Auto-correlation function:

$$s_{vv}(l) \Big|_{\text{white noise}} = \begin{cases} \sigma_v^2, & \text{if } l = 0, \\ 0, & \text{else.} \end{cases}$$

- Auto power spectral density:

$$S_{vv}(\Omega) \Big|_{\text{white noise}} = \sigma_v^2$$

A First Exercise

Please try on your own:

- A linear, causal, and (time-) shift-invariant system – specified by its impulse response h_i – is excited with zero-mean white noise $v(n)$ with variance σ_v^2 . What is the output power of the system?

Summary and Outlook

This part:

- Boundary conditions of the lecture
 - Contents
 - Literature hints
 - Exams, credit points, etc.
- Notation

Next part:

- Discrete signals and random processes