

Lecture on the MATLAB practical course

Experiment 3 Measurement data analysis

Felix Binder

10th June 2024 – H6









Overview of lecture content

I. Measurement technology

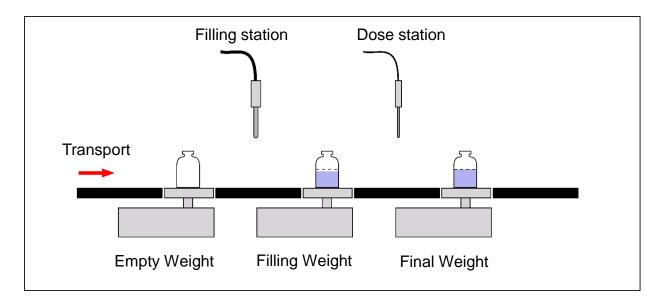
- a. Fourier analysis
- b. Evaluation of measurement results
- c. Determination of measurement uncertainty according to GUM

II. MATLAB

- a. Basics of programming with MATLAB
- b. Visualize data
- c. Symbolic Math Toolbox and useful functions

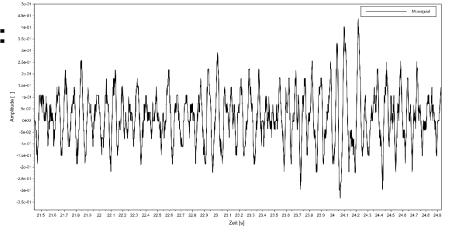


I.a: Example application - level measurement



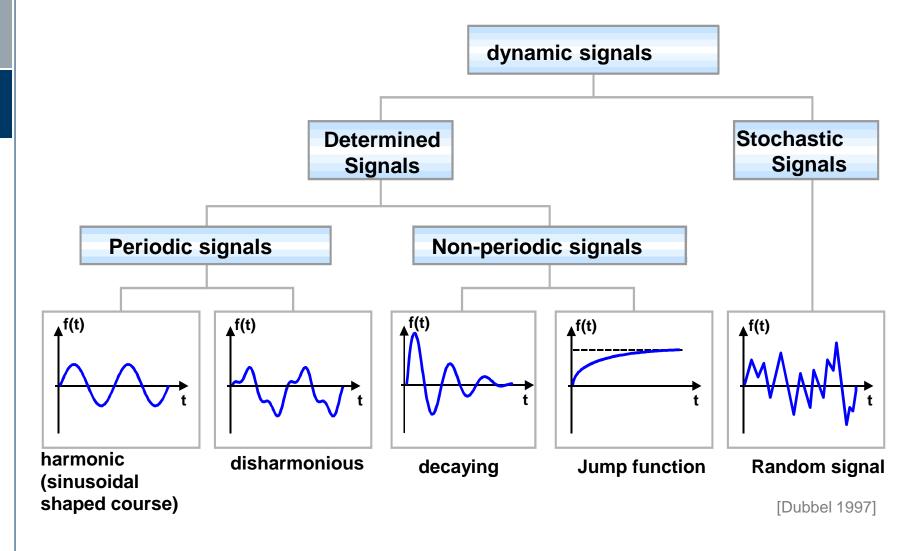
Possible main influencing factors:

- "laminar flow" ventilation
- Sorting pot
- Transported liquid





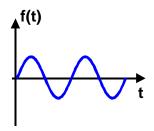
I.a.: Classification of dynamic signals





I.a: Signal description

Harmonic signals (periodic sinusoidal curve)



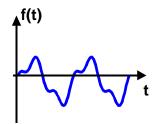
Parameters for the description:

- Offset
- Amplitude
- Frequency
- Phase

$$f(t) = A_0 + A\sin(\omega t + \varphi)$$

$$f(t) = A_0 + a\sin(\omega t) + b\cos(\omega t)$$

disharmonic signals (periodic non-sinusoidal course)

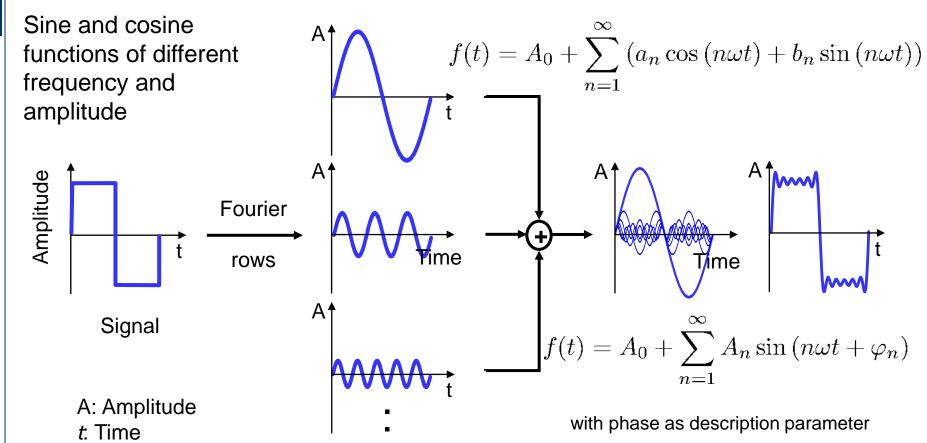


How to describe?



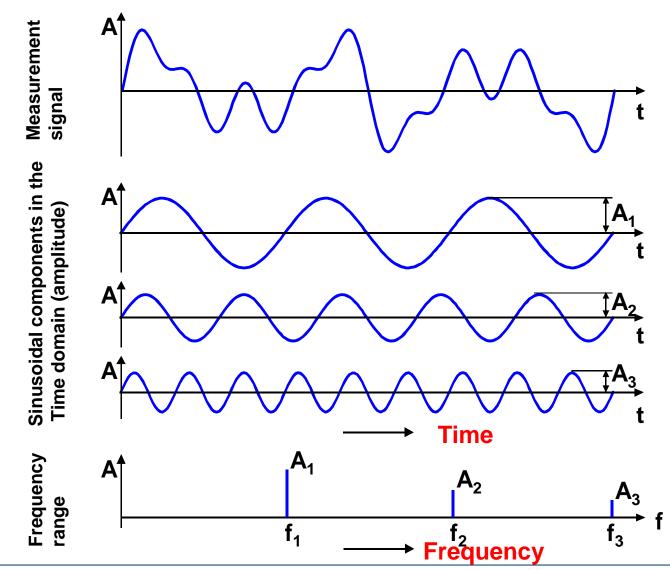
I.a: Signal description with Fourier series I

- Harmonic signals can be described very well
- Idea: describe a disharmonic signal as a composition of several harmonic signals





I.a: Signal description with Fourier series II



[Profos 1992]



I.a.: Fourier series and transformation I



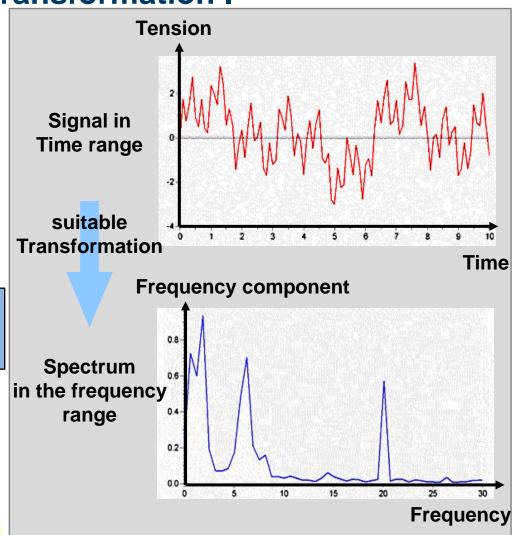
Jean Baptiste Fourier (1768 - 1830)

Fourier proved in 1822 that periodic functions can be split into a sum of angle functions

$$f(t) = A_0 + \sum_{n=1}^{\infty} A_n \sin(n\omega t + \varphi_n)$$

Possibility of breaking down a measured signal into individual frequency

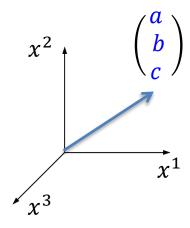
$$\mathfrak{F}{f(t)} = F(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-j\omega t} f(t) dt$$



Fourier transformation

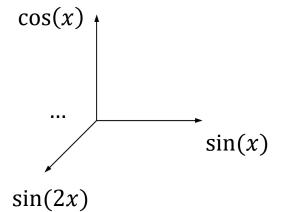


I.a.: Fourier series and transformation II



$$f(x) = ax^1 + bx^2 + cx^3$$

for 3-dimensions

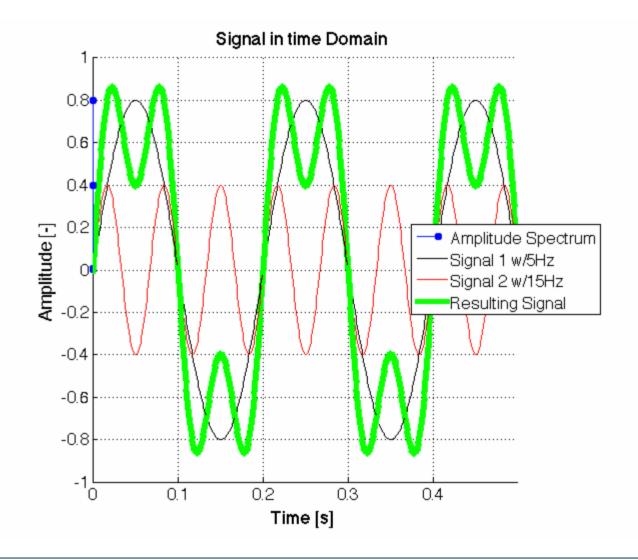


$$f(x) = a \sin(x) + b \cos(x) + c \sin(2x) + \dots$$

for N-dimensions



I.a.: Fourier series and transformation III





I.a: Determining the parameters of the Fourier series

Mathematical description of the signal curve

$$x(t) = a_0 + \sum_{k=1}^{\infty} a_k \cos(k\omega_0 t) + \sum_{k=1}^{\infty} b_k \sin(k\omega_0 t)$$

Circular frequency of the fundamental frequency:

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

Basic parameters:

 T_0 : Period duration of the fundamental oscillation

Fourier coefficients:

$$a_0 = \frac{1}{T_0} \int_{0}^{T_0} x(t) dt$$

$$a_k = \frac{2}{T_0} \int_0^{T_0} x(t) \cos(k\omega_0 t) dt$$

$$b_k = \frac{2}{T_0} \int_0^{T_0} x(t) \sin(k\omega_0 t) dt$$

[Lerchner 1990]



I.a.: Practical realization of the Fourier transform

- analytically solvable for given functions x(t)
- Mathematical description of the signal curve usually not available
 - → Signal is recorded discretely in terms of time and value (e.g. A/D converter)
- Calculation of Fourier coefficients on the basis of time-discretized values
 - → discrete Fourier transform (DFT) or fast Fourier transform (FFT)

Basic parameters:

 T_0 : Period duration of the fundamental oscillation

 f_{max} : maximum frequency

 k_{max} : maximum number of coefficients

T_A: Sampling time

M: Exponent of base 2

Practice-oriented equation set

$$T_A \le \frac{1}{5} \cdot \frac{1}{f_{\text{max}}}$$
 modified Nyquist of

$$N = \frac{T_0}{T_A}$$

$$N=2^M$$

$$k_{\text{max}} = f_{\text{max}} \cdot T_0$$

[Lerchner 1990]



I.a.: Practical realization of the Fourier transform

Fourier transform

$$\mathfrak{F}\{f(t)\} = F(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-j\omega t} f(t) dt$$



$$\mathfrak{F}\{f(t)\} = F(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-j\omega t} f(t) \, dt \qquad \bullet \bullet \bullet \qquad \mathfrak{F}^{-1}\{F(\omega)\} = f(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} F(\omega) e^{j\omega t} \, d\omega$$

Discrete Fourier transformation

$$X(n) = \frac{1}{N} \sum_{k=0}^{N-1} x(k) e^{-j2\pi \frac{nk}{N}}$$

Inverse discrete Fourier transform

$$x(k) = \sum_{n=0}^{N-1} X(n) e^{j2\pi \frac{kn}{N}}$$

- DFT = implementation of the Fourier transform for discrete signals
- Discrete-time equivalent of the Fourier series \sim
 - Line spectrum from 0 to $f_{\text{max}} = 0.5/T_A$ (from $0.5/T_A$ to $1/T_A$ mirrored, sampling theorem) with line spacing $\Delta f = 1/(NT)_A$
 - Spectrum of the DFT is periodic in contrast to the spectrum of the Fourier series
- FFT = numerically efficient implementation of the DFT for signals with length L=2 N



I.a: Prerequisites and calculation steps of the DFT

Prerequisites

- Signal must be sampled periodically
- Scanning row must be finite and form a window
- the cut-out time window continues periodically



Spectrum is completely described if the finite number of spectral lines of a period is known

DFT calculation steps

- Defining the parameters of the Fourier series
- Algorithmic calculation of the Fourier coefficients

```
public void dft(double[] in_real, double[] in_imag, double[] out_real, double[]
    out_imag){
    int N = in_real.length;

    for(int n = 0; n < N; n++){
        double sum_real = 0;
        double sum_imag = 0;

    for(int m = 0; m < N; m++){
            double angle = 2*Math.PI*n*m/N;

        sum_real += in_real[n]*Math.cos(angle)+in_imag[n]*Math.sin(angle);
        sum_imag += -in_real[n]*Math.sin(angle)+in_imag[n]*Math.cos(angle);
    }

    out_real[n] = sum_real;
    out_imag[n] = sum_imag;
}
</pre>
```

Implementierung der DFT als Java Funktion



I.a.: "Fast Fourier Transformation" I

FFT according to J. W. Cooley and J. W. Tukey ("An Algorithm for the Machine Calculation of Complex Fourier Series", 1964) reduces the complexity of to by utilizing the symmetry properties of the Fourier transform.

The data must therefore be reduced to or expanded with 0 ("zero padding"). The indices can be divided according to the following scheme.

$$X_k = \underbrace{\sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N/2} mk}}_{G_k: \ \mathrm{DFT} \ \mathrm{of} \ \mathrm{even} \ \mathrm{index}} + e^{-\frac{2\pi i}{N} k} \underbrace{\sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N/2} mk}}_{U_k: \ \mathrm{DFT} \ \mathrm{of} \ \mathrm{uneven} \ \mathrm{index}}$$

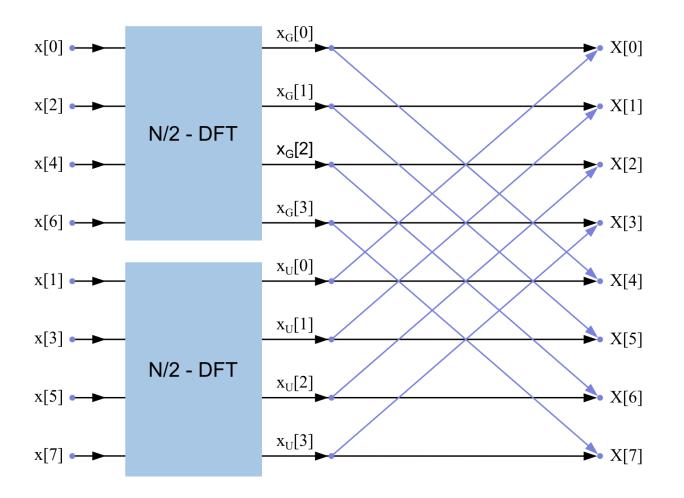
$$X_k = G_k + e^{-\frac{2\pi i}{N}k} \cdot U_k$$

$$X_{k+\frac{N}{2}} = G_k - e^{-\frac{2\pi i}{N}k} \cdot U_k$$



I.a: "Fast Fourier Transformation" II

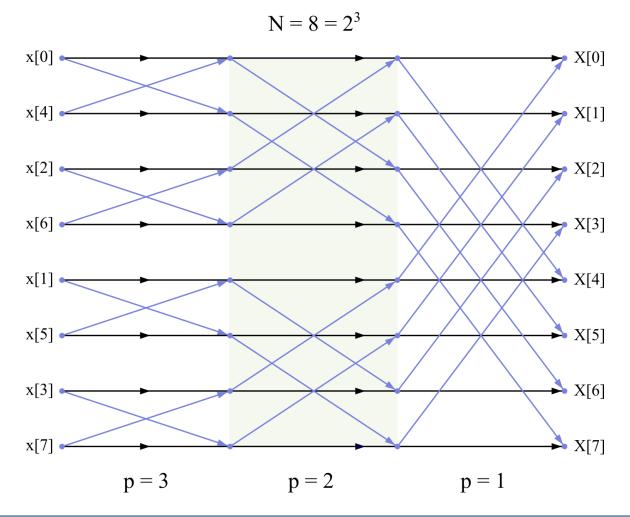
Signal flow diagram of the FFT for with halving of the DFT input data.





I.a.: "Fast Fourier Transformation" III

Further reduction of the DFT input data for by swapping the indices in *p-steps*:

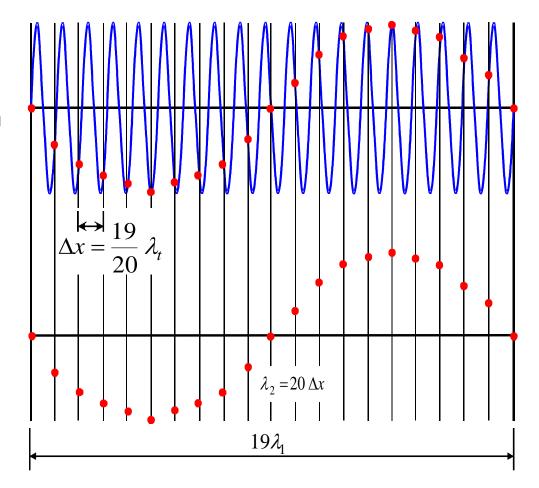




I.a: Sensing disorder I

 Error in signal acquisition due to non-compliance with the sampling theorem (sampling frequency too low)

 lead to digitized signal sequences that are not contained in the original signal



[Jähne 1991]



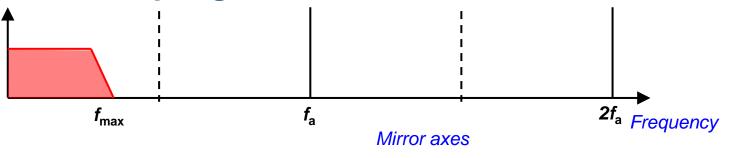
I.a: Sensing disorder II

Introduction of errors in the frequency spectrum of a sampled signal when components with frequencies too high to be analyzed at the sampling interval used contribute to the amplitudes of the components with lower frequencies [DIN IEC 60050-351:2013-07]

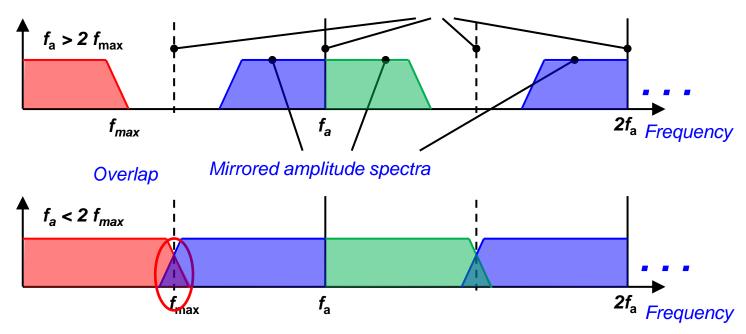


I.a.: Shannon's sampling theorem

Amplitude spectrum of an analog signal



Amplitude spectrum of a digital signal



 $f_a \ge 2 f_{\text{max}}$

 $f_a = 1/T_a$: Sampling frequency

f_{max}: maximum signal frequency



Overview of lecture content

I. Measurement technology

- a. Fourier analysis
- b. Evaluation of measurement results
- c. Determination of measurement uncertainty according to GUM

II. MATLAB

- a. Basics of programming with MATLAB
- b. Visualize data
- c. Symbolic Math Toolbox and useful functions



I.b.: Definition of measurement deviation according to VIM 2012

No measurement is completely controllable, nothing is completely known

Measurement deviation = measured value - reference value [VIM 2012 2.16]

Systematic measurement deviation:

"Component of the measurement error that remains constant or changes in a predictable manner with repeated measurements"

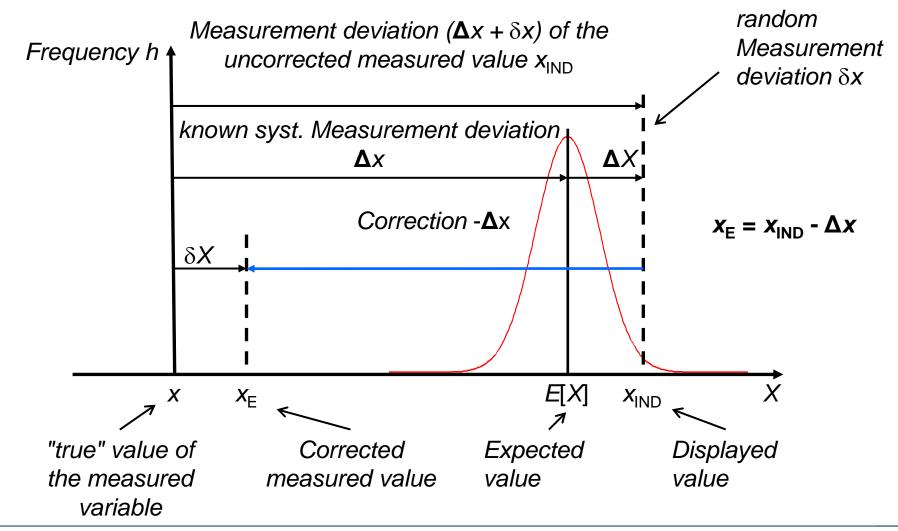
[VIM 2012, 2.17]

Random measurement deviation:

"Component of the measurement deviation that occurs with repeated measurements fluctuates in an unpredictable manner" [VIM 2012, 2.19]



I.b.: Dealing with measurement deviations





I.b.: How good is a measurement?

random and systematic deviations lead to uncertain measurement results



How good is the measurement?

How reliable are the measurement results?

How can the measurement result be quantitatively evaluated?



I.b.: Measuring precision

Measurement precision:

"Extent of agreement of **indications** or **measured values** obtained by **repeated measurements** on the same or similar objects under specified conditions"

[VIM 2012, 2.15]

Parameters for measurement precision:

Standard deviation s, variance s² or coefficient of variation c under specified measurement conditions

$$c = \frac{s}{\bar{x}}$$



I.b.: Measuring accuracy

Measuring accuracy:

"Extent of approximation of a measured value to a true value of a measurand" [VIM 2012, 2.13]

- is not a variable and is not expressed quantitatively
- It is said that a measurement is more accurate if it has a smaller measurement deviation



I.b.: Measurement accuracy

Measurement accuracy:

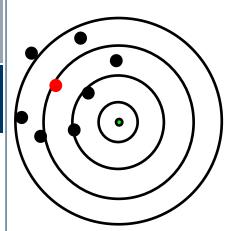
"Extent of approximation of the mean value of a infinite number of repeated measured values a reference value"

[VIM 2012, 2.14]

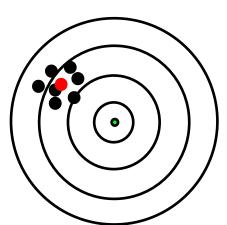
- is not a variable and is not expressed quantitatively
- is inversely related to the systematic error of measurement



I.b.: Example: Shooting at a target

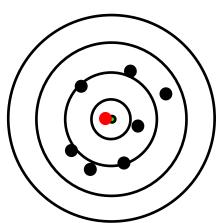


- Precision
- Measurement accuracy
- Accuracies

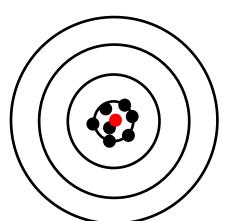


- Precision
- Measurement accuracy
- Accuracies

• single value mean value true walue



- Precision
- Measurement accuracy
- Accuracies



- Precision
- Measurement accuracy
- Accuracies



Overview of lecture content

I. Measurement technology

- a. Fourier analysis
- Evaluation of measurement results



c. Determination of measurement uncertainty according to GUM

II. MATLAB

- a. Basics of programming with MATLAB
- b. Visualize data
- c. Symbolic Math Toolbox and useful functions



I.c.: Error propagation law (old concept)

The Gaussian error propagation law is used to determine how large the expected error of a measurement result calculated from deviating measured values is.

according to [Hoffmann 2000]

$$y = f(x_1, \dots, x_N)$$

 $s_{
m y}$ Standard deviation of the overall result

for random deviations.

 x_j Individual sizes

 s_j Standard deviation of the individual variables

$$s_{y} = \sqrt{\sum_{j=1}^{N} \left(\frac{\partial f}{\partial x_{j}} s_{j}\right)^{2}}$$

 $\frac{\partial f}{\partial x_i}$ partial derivatives

N Number of input variables



I.c.: Error propagation law (old concept)

for systematic deviations:

$$\Delta y = \sum_{j=1}^{N} \left(\frac{\partial f}{\partial x_j} \Delta x_j \right)$$

 Δy

Systematic deviation of the overall result

Individual sizes

 x_j

Average deviation of the individual variables

 Δx_j

partial derivatives

 ∂x_j

Number of input variables

N



I.c.: Disadvantages of classic error analysis

Disadvantages:

- Systematic deviations are usually not taken into account
- Correlations not taken into account

Other reasons:

- Quantification of measurement deviations ultimately requires knowledge of a "(single) true value"
- 2. Overemphasis on statistically determined deviations
- 3. Separate determination of measurement result and uncertainty
- 4. Measurement model ("measurement equation") plays practically no role
- 5. There are neither completely known systematic measurement deviations nor completely unknown
- 6. No generally accepted definition of uncertainty

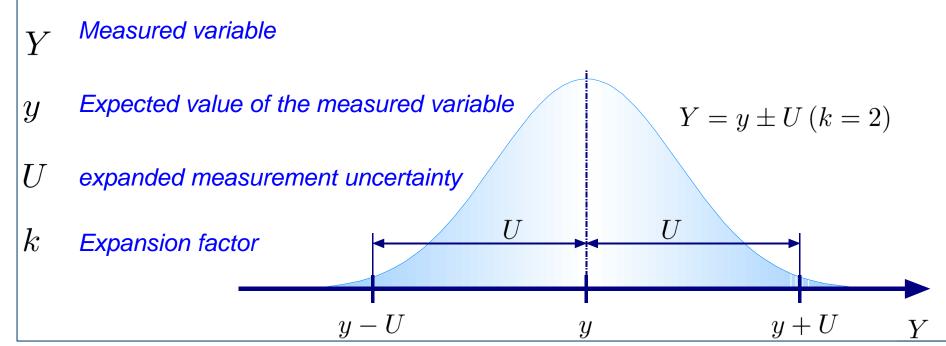


I.c.: Measurement uncertainty

Measurement uncertainty:

"non-negative parameter **characterizing** the dispersion of **values** associated with **the measurand on the** basis of the information used"

[VIM 2012, 2.26]





I.c.: Structure of the standard procedure of the GUM

- Collecting knowledge about the measuring process and the involved input quantities / Gather knowledge about the measurement process and the input variables involved
- Create a model for evaluation of the measurement / X $X_{\rm IND}$ $X_{\rm IND}$
- Evaluating the input quantities / Estimating the input variables Methods: Type A and B / Methods: Type A and B
- Calculation of the measured value and combining the uncertainties / Calculating measured values and combining uncertainties $\Rightarrow y, u_{c}$
- Calculation of the expanded uncertainty Calculating the expanded measurement uncertainty $\Rightarrow U, k \Rightarrow U_{\rm p}, k_{\rm p}, \nu_{\rm eff}$
- Indication and evaluation of the measurement result $\Rightarrow y \pm U \ (k = \ldots)$



I.c.: Step 1 of the standard GUM procedure

1

Gather knowledge about the measurement process and the input variables involved

Collecting information about the measurement process:

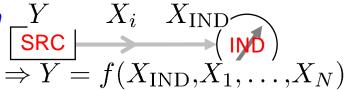
- Description of the measurement task
- Measuring principle used
- Measurement methods used
- Supplementary descriptions of the measurement procedure



I.c.: Step 2 of the standard GUM procedure

Setting up the model for the measurement evaluation

$$\Rightarrow X_{\text{IND}} = g(Y, X_1, \dots, X_N)$$



possibly:

Setting up the mathematical

Model of the measurement and subsequent inversion

or the same

Setting up the mathematical Model for measurement evaluation

$$X_{\mathrm{IND}}$$

$$Y = f(X_{\mathrm{IND}}, X_1, \dots, X_N)$$

$$X_i$$

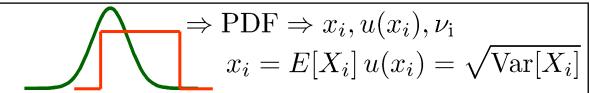
Measurement

 $X_{\text{IND}} = g(Y, X_1, \dots, X_N)$



I.c.: Step 3 of the standard GUM procedure

3 Estimating the input variables Methods: Type A and B



Sources for the quantitative evaluation of input variables:

- Results of direct measurements, comparative measurements
- Empirical values, results of previous evaluations
- Values from calibration certificates and other certificates
- Manufacturer's data, literature values



I.c.: Step 3 of the standard GUM procedure

Method type A:

Statistical analysis (series of measurements)

$$\bar{q} = \frac{1}{n} \sum_{j=1}^{n} q_j$$

$$s(q_j) = \sqrt{\frac{1}{n-1} \sum_{j=1}^{n} (q_j - \bar{q})^2}$$

$$x_i = \bar{q}$$

$$u(x_i) = s(q_j) / \sqrt{n}$$

$$\nu_i = n-1$$

approximation for large n) for statistical analyses: n ≥ 4

Method type B: other information

$$\nu_i = \infty$$

- a single measured value $x_i, u(x_i)$ or $x_i, u(x_i), \nu_i$
- Specification of a lower and upper limit (e.g. temperature range)

$$x_i = (a_+ + a_-)/2$$

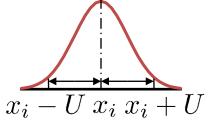
$$u(x_i) = \Delta a/\sqrt{3}$$

$$2 \cdot \Delta a$$

$$a_- x_i \quad a_+$$

Specification in a calibration certificate

$$\begin{aligned}
 x_i \\
 u(x_i) &= U/k
 \end{aligned}$$



Source: http://www.ptb.de/cms/fileadmin/internet/publikationen/kessel.pdf (as at: 11.04.2013)



I.c.: Step 4 of the standard GUM procedure

Calculation of the measured value and combining the uncertainties y, u_c

A) Measured value calculation:

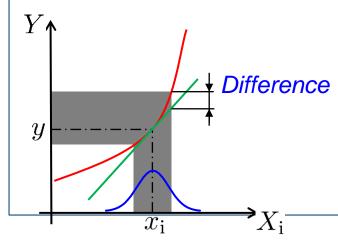
$$y = f(x_{\text{IND}}, x_1, \dots, x_N)$$

B) Combining the uncertainties:

Basic idea of the error propagation law.

Taylor series development — Linearization for operating

Taylor series development \rightarrow Linearization for operating point x_i



$$c_i = \left. \frac{\partial f}{\partial X_i} \right|_{x_i} = \left. \frac{\partial f}{\partial x_i} \right|_{x_i}$$

only possible for sufficiently linear models

Sensitivity coefficient



I.c.: Step 4 of the standard GUM procedure

for uncorrelated input variables:

$$u_{\mathrm{c}}^2(y) = \sum_{i=1}^N \left(\frac{\partial f}{\partial x_i}\right)^2 u^2(x_i) = \sum_{i=1}^N c_i^2 u^2(x_i)$$
 $u(x_i)$ Standard uncertainty

for correlated input variables:

$$u_{\mathrm{c}}^{2}(y) = \sum_{i=1}^{N} \sum_{j=1}^{N} \frac{\partial f}{\partial x_{i}} \frac{\partial f}{\partial x_{j}} u(x_{i}, x_{j})$$

$$u(x_{i}, x_{j}) \implies \text{Covariance}$$

$$u(x_{i}, x_{j}) \implies \text{Covariance}$$

$$u_{c}^{2}(y) = \sum_{i=1}^{N} \left(\frac{\partial f}{\partial x_{i}}\right)^{2} u^{2}(x_{i}) + 2 \sum_{i=1}^{N-1} \sum_{j=i+1}^{N} \frac{\partial f}{\partial x_{i}} \frac{\partial f}{\partial x_{j}} u(x_{i}, x_{j})$$

Correlation coefficient
$$r(x_i,x_j) = \frac{u(x_i,x_j)}{u(x_i)u(x_j)}$$

$$u_{c}^{2}(y) = \sum_{i=1}^{N} c_{i}^{2} u^{2}(x_{i}) + 2 \sum_{i=1}^{N-1} \sum_{j=i+1}^{N} c_{i} c_{j} r(x_{i}, x_{j}) u(x_{i}) u(x_{j})$$



I.c.: Step 5 of the standard GUM procedure



Calculating the expanded measurement uncertainty $\Rightarrow U, k$

$$\Rightarrow U, k$$

$$\Rightarrow U_{\rm p}, k_{\rm p}, \nu_{\rm eff}$$

for determination method type A:

Degree of freedom v_i

- = number of observations number of parameters to be determined only for a small number of observations
- → Consideration of the reliability of the standard measurement uncertainty

effective degree of freedom:

$$u_{ ext{eff}} = rac{u(y)^4}{\sum_{i=1}^N rac{u(x_i)^4}{
u(x_i)^4}} \qquad ext{Welch - Satterthwaite}$$

→ Assessment of the reliability of the combined standard uncertainty of measurement



I.c.: Step 5 of the standard GUM procedure

Selection of the expansion factor k or k_p :

$$U_{\rm p}(y) = k_{\rm p} \cdot u_{\rm c}(y)$$
 for $k_{\rm p} = t_{\rm p}(\nu_{\rm eff})$ for p = 68.3 %; p = 95.5 %; p = 99.73 %

- depending on the effective degree of freedom
- from the t-distribution

$ u_{eff}$	3	4	5	10	20	50	∞
$k_{\rm p} \ (p = 95 \ \%)$	3,18	2,78	2,57	2,23	2,09	2,01	1,96

for method type B and a large number of observations

typical values k = 2 for p = 95% and k = 3 for p = 99%

$$U(y) = k \cdot u_{\rm c}(y)$$
 for $k = t_{\rm p}(\infty)$



1.c.: Step 6 of the standard GUM procedure

Specifying and evaluating the measurement result

$$\Rightarrow y \pm U (k=2)$$

general form:

$$y \pm U (k=2)$$

 $L = (22.17 \pm 0.11) \text{ mm } (k = 2)$

or

 $L = 22.17 \text{ mm } \pm 0.11 \text{ mm } (k = 2)$

"The standard measurement uncertainty has been determined in accordance with EA-4/02 or DKD-3 (DIN V ENV 13005)."

- Specify measurement uncertainty with a maximum of 2 significant digits
- Round mathematically! (Rounding only by a maximum of 5 %)
- Specify the value of the result according to the significant digits of the measurement uncertainty (DIN 1333)



I.c.: Step 4 of the standard GUM procedure

Measurement uncertainty balance:

"Specification of a **measurement uncertainty**, the components this measurement uncertainty and its calculation and combination" [VIM 2012, 2.33]

$$u_i(y) = c_i \cdot u(x_i)$$

Size	Value	Uncertai nty	Distrib ution	Degree of	Sensitivity coefficient	Uncertainty contribution
X _i	X _i	$u(x)_i$		freedom v _i	\boldsymbol{c}_i	u _i (y)
<i>X</i> ₁	<i>x</i> ₁ =	u(x) ₁	Form 1	ν_1	$c_1 = \partial f/\partial x_1$	u ₁ (y)
X_2	$x_2 =$	$u(x)_2$	Form 2	ν_2	$c_2 = \partial f/\partial x_2$	u ₂ (y)
		•	•			
X_N	$X_N = \dots$	$u(x)_N$	Form N	ν_N	$c_N = \partial f/\partial x_N$	и _N (у)
V	<i>y</i> =					g Metro Gay 12024-06-10 44/



I.c.: Determination of measurement uncertainty by computer simulation

Determination of the measurement uncertainty using the Monte Carlo method

(DIN EN 13005 Supplement 1:2010)

- Application for.
 - non-linear model of the measurement
 - Deviation of the probability density function from a Gaussian distribution
- always returns a result for the probability density function
- Random experiments carried out very frequently (law of large numbers)



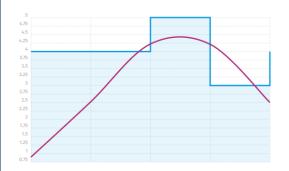
I.c.: Central limit theorem

"The mean values of the samples of a population are normally distributed."

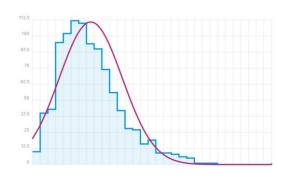
Example: Sampling distributions of the exponential distribution

Number of samples: *n*

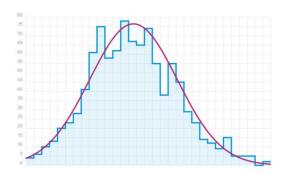
Sample size: *m*



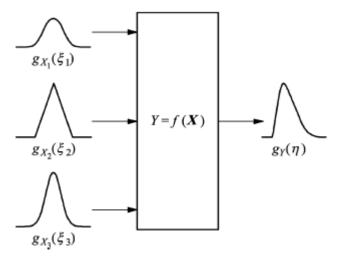
$$n = 20, m = 5$$



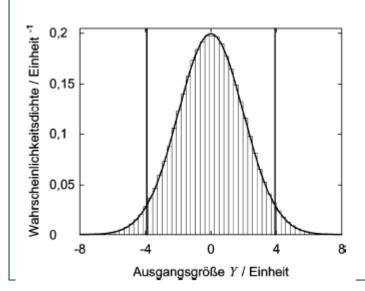
$$n = 1000, m = 5$$

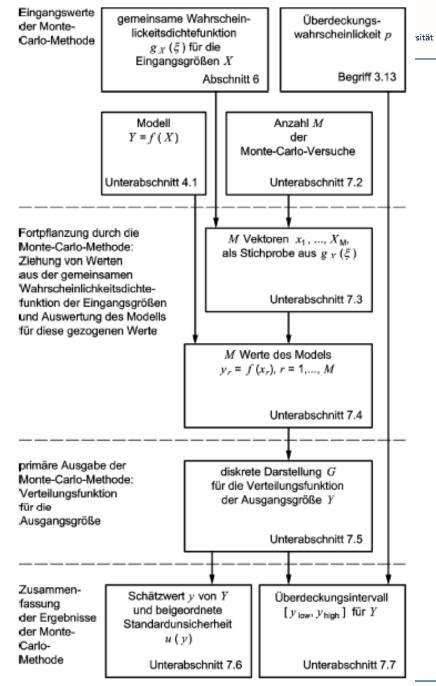


$$n = 1000, m = 30$$



[DIN V ENV 13005 Supplement 1:20]







Overview of lecture content

Measurement technology

- a. Fourier analysis
- b. Evaluation of measurement results
- Determination of measurement uncertainty according to GUM

II. MATLAB



- a. Basics of programming with MATLAB
- b. Visualize data
- c. Symbolic Math Toolbox and useful functions



II.a.: Styleguide I

```
n = 10; l = 3;
A = (n / 4) * 1^2 * cot(pi / (2*n));

numberEdges = 10; edgeLength = 3;
areaPolygon = (numberEdges/4)*edgeLength^2*cot(pi/(2*numberEdges));
```

Suggestions for the realization of meaningful naming:

- Assign meaningful names, paying attention to the scope!
- Write variables in lower case, classes and functions (scripts) in upper case
- Use compound words with alternating capitalization
- Good names increase the understanding of code enormously
- Good names are part of a good documentation of the program
- Only use English-language names

Robert C. Martin: "Clean Code: A Handbook of Agile Software Craftsmanship" MATLAB Programming Fundamentals:

https://www.mathworks.com/help/pdf doc/matlab/matlab prog.pdf



II.a.: Styleguide II

```
vectorLength = sqrt(x^2 + y^2); % formula documented at ...
```

Use the comment function to provide added value for comprehensibility, e.g. on derivations for formulas or other contexts used.

Do not document any superfluous information!

Write comments in English!

```
qualityParameterZaxis = offsetParameterYaxis + weightingZaxis +
...
```

baseQualityParamterZaxis;

Long lines can be wrapped by '...' to structure code clearly. Spaces can be placed anywhere in the code between individual elements.

Style guides are always optional, but it is strongly recommended to adopt a suitable style and not to change it for a project.



II.b.: figure class

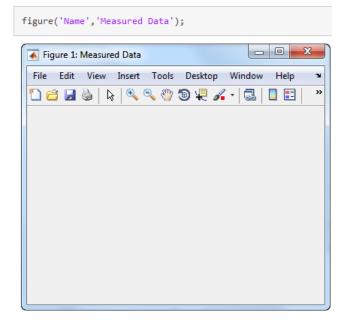
MATLAB figure class https://de.mathworks.com/help/matlab/ref/figure.html

figure Creates new figure with default settings
figure(Name, Value) Modifies the settings of the figure
f = figure(___) Returns the figure object
figure(f) Sets figure f as current figure

Current figure:

Target for graphic instructions such as axes

axes are automatically created in the *current* figure when a data visualization function is executed



https://de.mathworks.com/help/matlab/ref/figure.html#bvjs6cb-5



II.b.: plot() function

Overview of functions provided for visualizing data:

https://de.mathworks.com/help/matlab/creating_plots/types-of-matlab-plots.html

```
plot(X,Y) Creates 2-D line plot with x and y values (vectors)
plot(X,Y,LineSpec) Defines the line representation
plot(X1,Y1,...,Xn,Yn) Represents several x/y pairings in a plot
https://de.mathworks.com/help/matlab/ref/plot.html
```

```
hold on prevents existing axes from being deleted when a new one is created title Defines a title for the selected axes object xlabel | ylabel | zlabel Adds axis labeling xlim | ylim | zlim Selects display range for axes manually legend Defines legend entries for axes objects cla deletes any existing data in the selected axes object
```



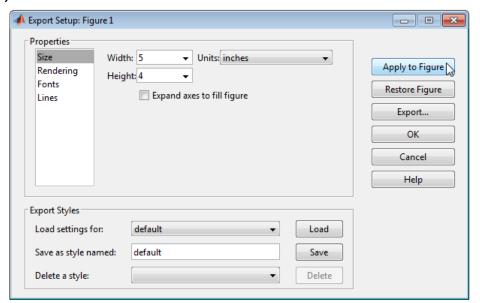
I.b.: Export of plots

After plots have been created and neatly formatted, the results must be suitably formatted and exported, e.g. for final theses

Settings for the size of the plot (e.g. adjustment to the page width of the format template of a Word document possible)

Resolution of the resulting image is adjustable to create high quality graphics

Recommendation: Save the figure as a .fig file (MATLAB format) and (uncompressed) .tiff for the respective application



https://de.mathworks.com/help/matlab/creating_plots/customize-figure-before-saving.html



II.c.: Introduction to the Symbolic Math Toolbox

Symbolic computing enables the direct implementation and processing of mathematical relationships, but generally at the expense of processing speed.

```
syms x y f Creates symbolic variables or functions f = x^2 + y^3; Defines symbolic function
```

```
g = int(f,x); Integrates function f according to variable x symbolically, (function diff() for differentiating)
```

solution = solve(g == 0,x); Finds the zeros of g according to variable x

f = subs(f, [x y], [1 4]); Replaces the variables x and y in the function f with the numerical values 1 and 4 and stores the result in f again.

result = **double(f)**; Converts the symbolic result of $f(1^2 + 4^3)$ into a floating point number



II.c.: Useful functions for the practical part

```
transformedSignal = fft(signalData); Fourier Transformation
transformedSignal = ifft(signalData); inv. Fourier Transformation
dx = gradient(x); numerical gradient of the vector x
y = round(x,n); Rounds x to the next integer value or decimal place, n>0
rounds place to the right of the decimal point
c = unique(a); Unique values in the vector/array
y = sqrt(x); root of x
y = randn(sz1,...,szn); Matrix of normally distributed random numbers
sz1 x ... x szn
y = std(x); standard deviation of x
y = range(x); Range of x
y = struct(); create new struct
y.a = 10; adds a new field to y with the name a and value 10
y.('a') = 10; adds a new field with the name a and value 10 to y
y. (a) = 10; adds a new field to y with the value of a and the value 10
```



Thank you for your attention!

17.06.2024 Group A, 12:00 AM - 6:00 PM, CIP-Pool-MB

24.06.2024 Group B, 12:00 AM - 6:00 PM, CIP-Pool-MB

