

Measurement uncertainties – GUM & Co.

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Right or wrong?

„The more often I repeat a
measurement, the more
accurate it gets.“

Right or wrong?

„The more often I repeat a
measurement, the more
accurate it gets.“

Wrong!

Some basic terms

Accuracy?
(Genauigkeit)

Systematic error?
(Systematische
Messabweichung)

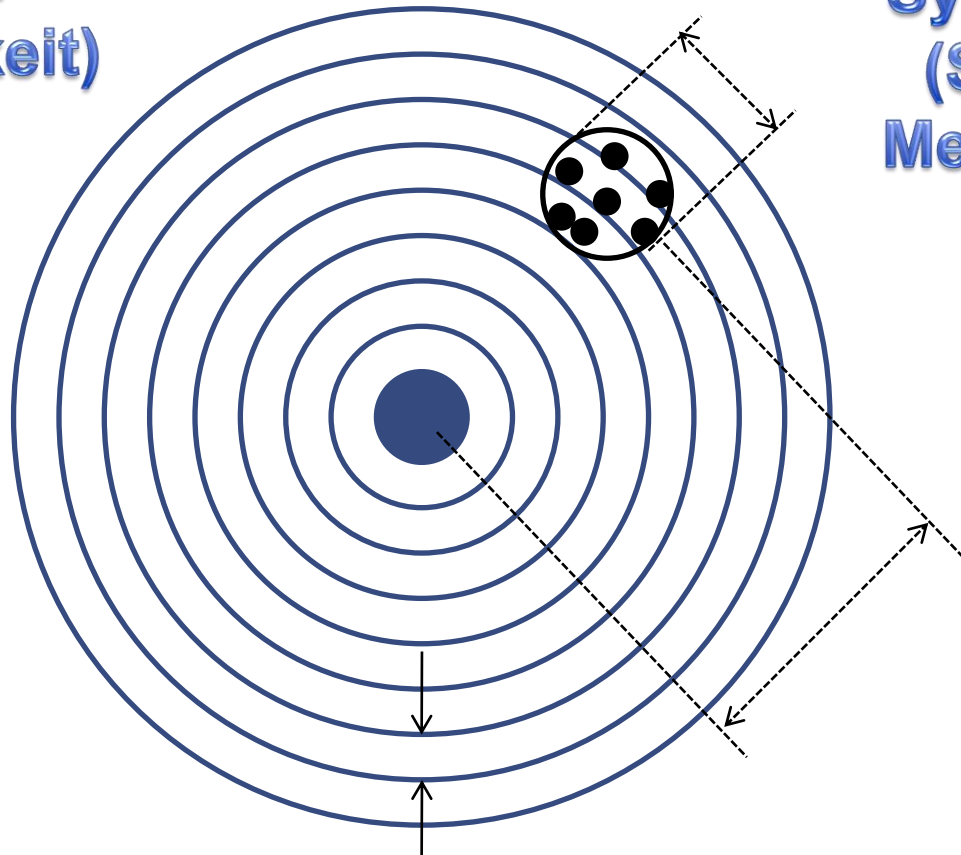
Precision?
(Präzision)

Random error?
(Zufällige Mess-
abweichung)

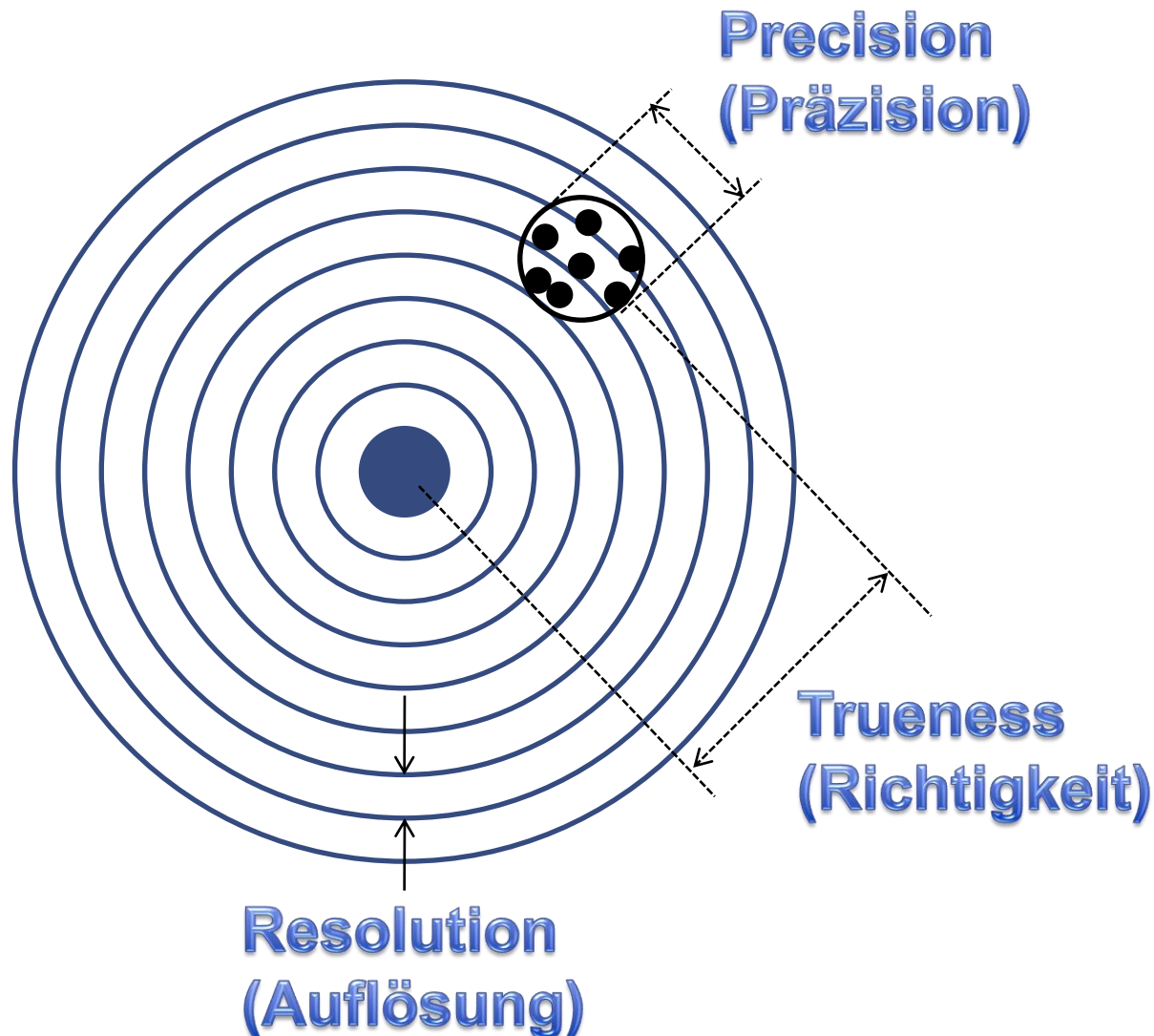
True value?
(Wahrer Wert)

Resolution?
(Auflösung)

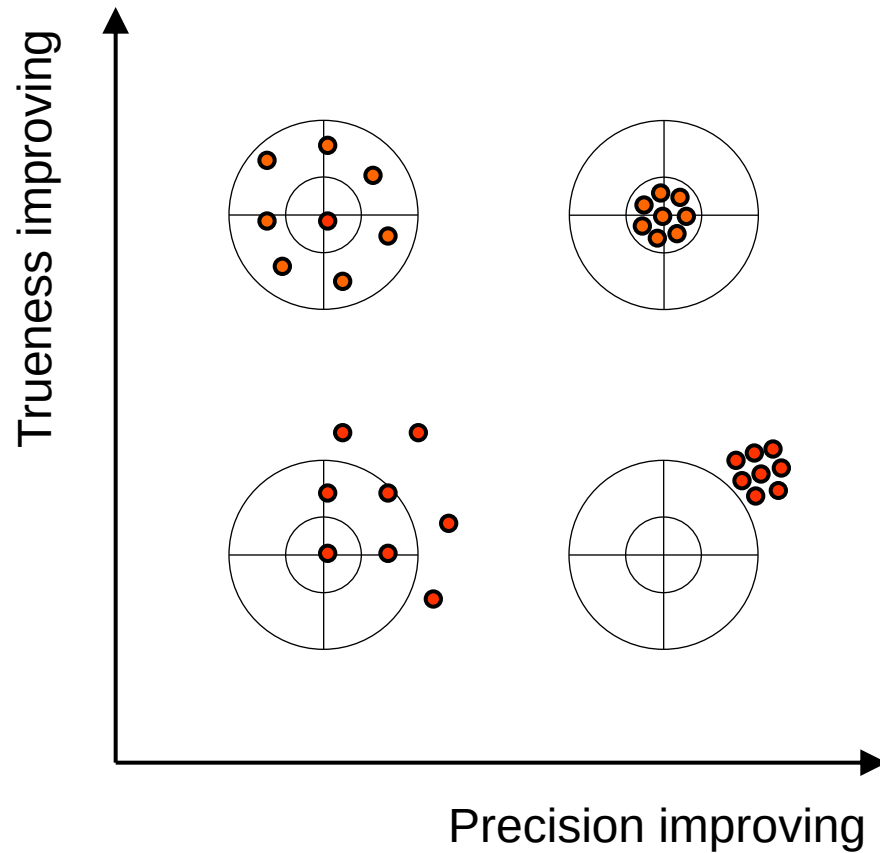
Trueness?
(Richtigkeit)



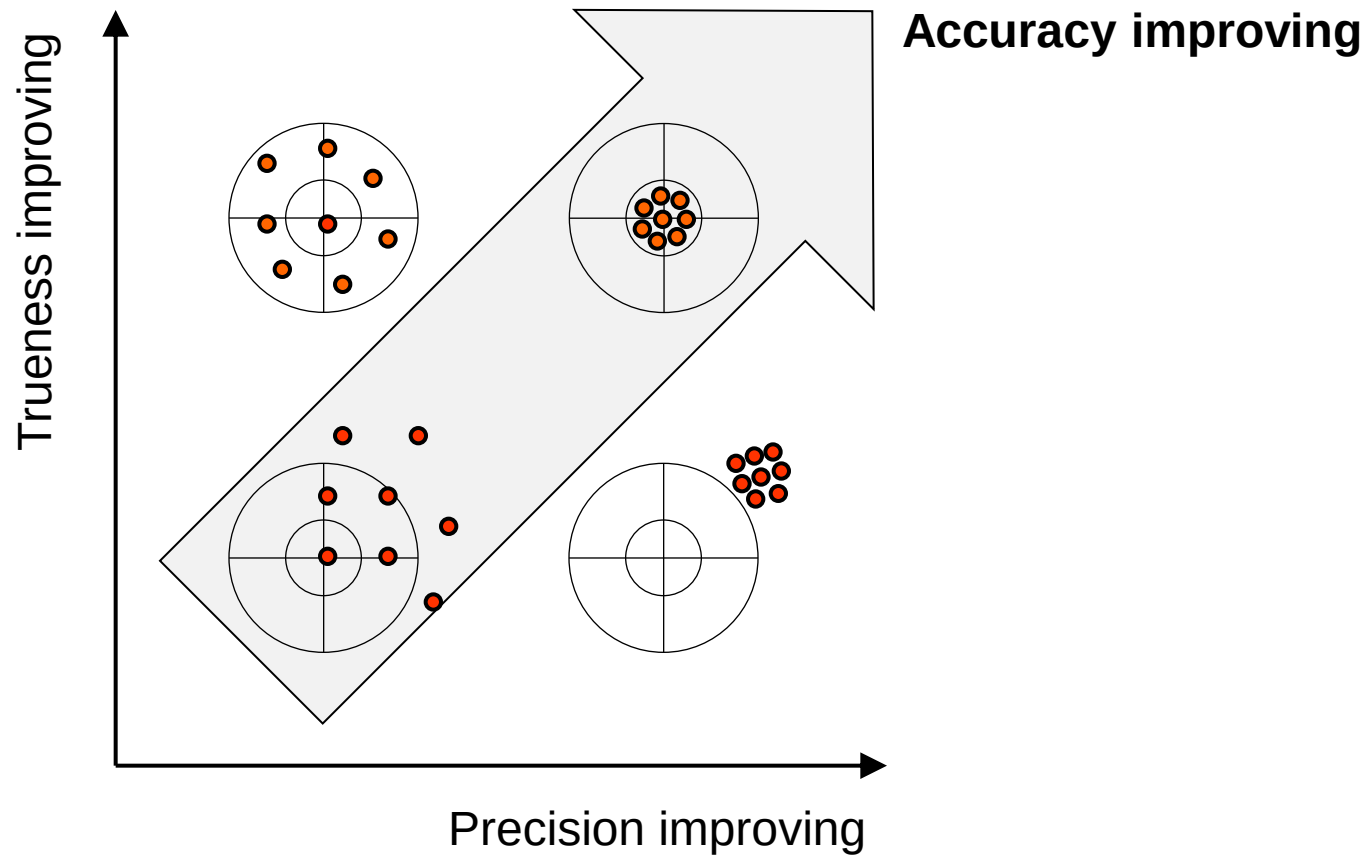
Some basic terms



And what about „accuracy“?



And what about „accuracy“?



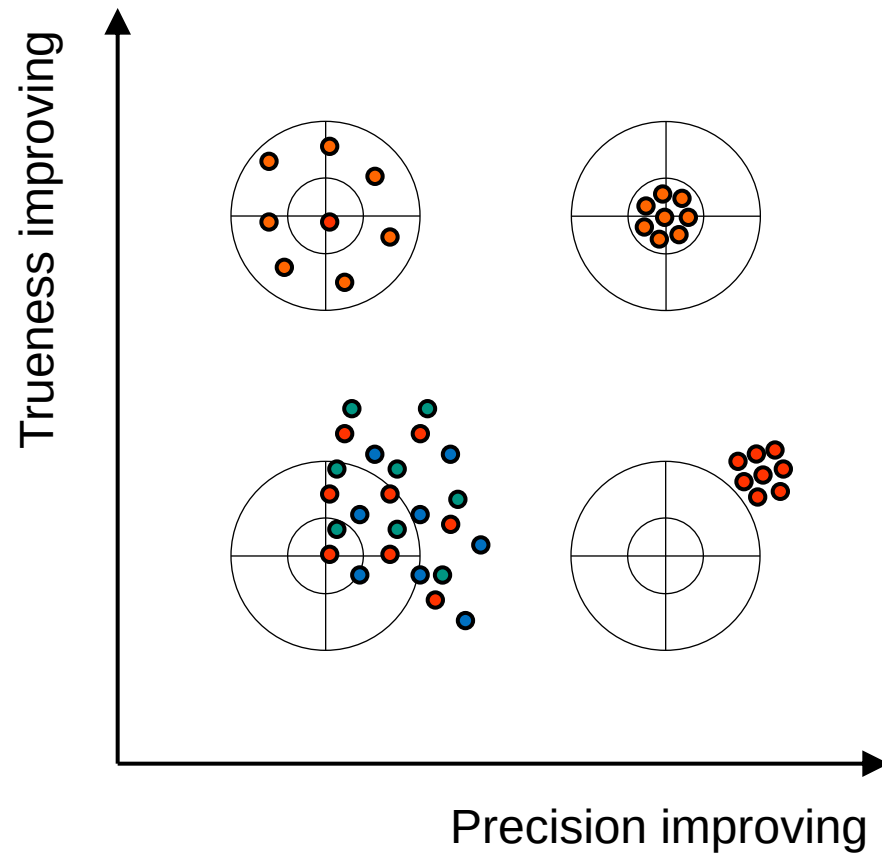
Right or wrong? Next try!

„The more often I repeat a
measurement, the more
precise it gets.“

Right or wrong? Next try!

„The more often I repeat a
measurement, the more
precise it gets.“

Wrong!



So what is it then?

„The more often I repeat a
measurement, ...“



GUM

Guide to the expression of Uncertainty in Measurement

→ „Guide to the Expression of Uncertainty in Measurement“

- Published 1993 by ISO (International Organization for Standardization), corrected versions 1995 and 2008
- Involved:
 - International Bureau of Weights and Measures (BIPM)
 - International Electrotechnical Commission (IEC)
 - International Federation of Clinical Chemistry (IFCC)
 - International Organization for Standardization (ISO)
 - International Union of Pure and Applied Chemistry (IUPAC)
 - International Union of Pure and Applied Physics (IUPAP)
 - International Organization of Legal Metrology (OIML)
- www.bipm.org/utis/common/documents/jcgm/JCGM_100_2008_E.pdf

Method for evaluating the measurement uncertainty according to GUM

- GUM distinguishes two cases in the evaluation of measurement uncertainties:
Type A and **Type B**
- **Type A**: Method of evaluation of uncertainty based on a series of observations (repeated measurements)
➡ **statistical analysis**
- **Type B**: Method of evaluation of uncertainty by means other than the statistical analysis of series of observations
➡ **if it's not type A, then it's type B**

Method A

Basic formulae

We have performed n independent repeat measurements of quantity x under the same conditions. Then:

- Best available estimate of the expected value μ
= arithmetic mean/average $\bar{X}$

$$\bar{X} = \frac{1}{n} \cdot \sum_{i=1}^n x_i$$

- Best estimate of the variance σ^2
= empirical variance $s^2(x)$

$$s^2(x) = \frac{1}{n-1} \cdot \sum_{i=1}^n (x_i - \bar{X})^2$$

- Best estimate for the variance of the mean
= empirical variance of the mean $s^2(\bar{X})$

$$s^2(\bar{X}) = \frac{s^2(x)}{n}$$

This is what we need!

normalizes variance to the sample size
→ comparability

„The more often I repeat a measurement, the smaller gets the empirical variance of the mean.“

Remarks (according to GUM)

- The **number of observations** n should be **large enough** to ensure that $\bar{x}$ provides a **reliable estimate** of the expectation μ of the random variable x , and $s^2(\bar{x})$ of the variance $\sigma^2(\bar{x}) = \sigma^2 / n$.

The **difference** between s^2 and σ^2 has to be considered when one constructs confidence intervals!

→ If the probability distribution of x is a **normal distribution**, the difference is taken into account through the **t -distribution**.

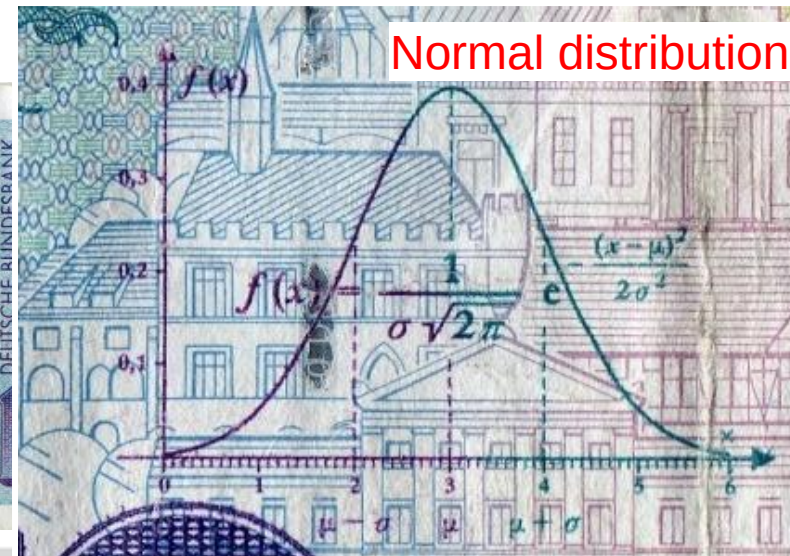
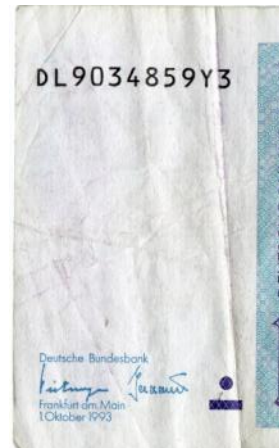
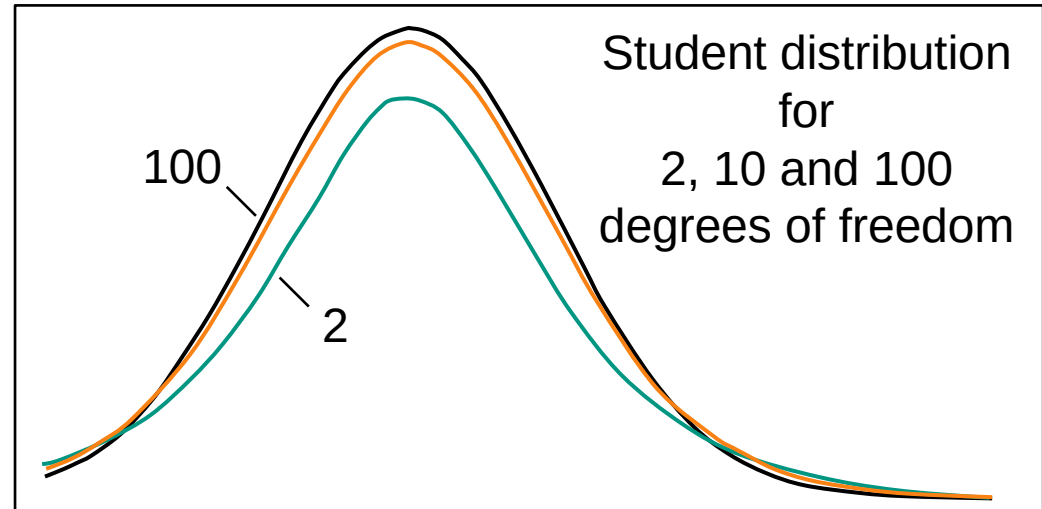
Student distribution / *t*-distribution

- *t*-distribution → normal distribution for ∞ degrees of freedom ν ($\nu = n-1$)

- **Normal distribution:**
68.3 % of measured values inside region $\pm \sigma$

***t*-distribution:**

Same probability for region $t_{68,3} \cdot s$ with t = student factor (*t*-factor)



Some values for t -factors (from GUM)

Table G.2 — Value of $t_p(v)$ from the t -distribution for degrees of freedom v that defines an interval $-t_p(v)$ to $+t_p(v)$ that encompasses the fraction p of the distribution

Degrees of freedom v	Fraction p in percent					
	68,27 1σ	90	95	95,45 2σ	99	99,73 3σ
1	1,84	6,31	12,71	13,97	63,66	235,80
2	1,32	2,92	4,30	4,53	9,92	19,21
3	1,20	2,35	3,18	3,31	5,84	9,22
4	1,14	2,13	2,78	2,87	4,60	6,62
5	1,11	2,02	2,57	2,65	4,03	5,51
6	1,09	1,94	2,45	2,52	3,71	4,90
7	1,08	1,89	2,36	2,43	3,50	4,53
8	1,07	1,86	2,31	2,37	3,36	4,28
9	1,06	1,83	2,26	2,32	3,25	4,09
10	1,05	1,81	2,23	2,28	3,17	3,96

Some values for t -factors (from GUM)

dof	1σ			2σ		3σ
20	1,03	1,72	2,09	2,13	2,85	3,42
25	1,02	1,71	2,06	2,11	2,79	3,33
30	1,02	1,70	2,04	2,09	2,75	3,27
35	1,01	1,70	2,03	2,07	2,72	3,23
40	1,01	1,68	2,02	2,06	2,70	3,20
45	1,01	1,68	2,01	2,06	2,69	3,18
50	1,01	1,68	2,01	2,05	2,68	3,16
100	1,005	1,660	1,984	2,025	2,626	3,077
∞	1,000	1,645	1,960	2,000	2,576	3,000

**t -factor important also for
more than 10 repeats!**

Example: Type A

- 5 measurements of a voltage

Table G.2 — Value of $t_p(v)$ from the t -distribution for an interval $-t_p(v)$ to $+t_p(v)$ that encompasses p

Degrees of freedom v	Fraction p		
	68,27 %	90	95
1	1,84	6,31	12,71
2	1,32	2,92	4,30
3	1,20	2,35	3,18
4	1,14	2,13	2,78

i	X_i in V	$(X_i - \mu)^2$ in V ²
1	6,7	0,04
2	6,4	0,01
3	6,3	0,04
4	6,6	0,01
5	6,5	0,00
Sum	32,5	0,10

Discussion:

What is the arithmetic mean, the empirical standard deviation, the uncertainty of the mean and the total uncertainty taking into account the student factor (68,27%)?

Example: Type A

- 5 measurements of a voltage

- Mean: $32,5 \text{ V} / 5 = 6,5 \text{ V}$

- Empirical varicance:
 $0,10 \text{ V}^2 / (5-1) = 0,025 \text{ V}^2$

- Empirical std. deviation:

$$s = \sqrt{s^2} = \sqrt{0,025 \text{ V}^2} = 0,156 \text{ V}$$

i	X_i in V	$(X_i - \bar{x})^2$ in V^2
1	6,7	0,04
2	6,4	0,01
3	6,3	0,04
4	6,6	0,01
5	6,5	0,00
Sum	32,5	0,10

- Uncertainty of the mean:

$$s(\bar{x}) = s / \sqrt{n} = 0,156 \text{ V} / \sqrt{5} = 0,070 \text{ V}$$

- Student factor for 68,3% probability: 1,14 ($\nu = 4$)

- Final result: $\mu \pm s(\bar{x}) \cdot t = 6,5 \text{ V} \pm 0,08 \text{ V}$

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Method B

Evaluation of the variance from other source („everything apart from statistical analysis“)

According to Hoffmann, Handbuch der Messtechnik

- Sometimes one can only perform one single measurement $\rightarrow$ variance?
 - Own experience (e.g. from former experiments)
 - Experience of others
 - literature values
- Often there are upper/lower bounds of the uncertainty given



What to do?

Evaluation of the variance if only upper/lower bounds of the uncertainty are known

■ First step:

Determination of the distribution function $f(x_i)$

■ Second step:

Calculation of the variance using these formulae:

Discrete distribution

$$u^2 = \sum_{i=1}^N (x_i - x_0)^2 f(x_i)$$

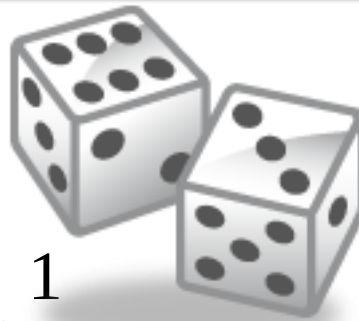
$$\text{with } x_0 = \sum_{i=1}^N x_i f(x_i)$$

Continuous distribution

$$u^2 = \int_{-\infty}^{\infty} (x - x_0)^2 f(x) dx$$

$$\text{with } x_0 = \int_{-\infty}^{\infty} x f(x) dx$$

Example: Dice



$$\begin{aligned}x_0 &= \sum_{i=1}^N x_i f(x_i) = \sum_{i=1}^6 x_i \cdot \frac{1}{6} \\&= (1 + 2 + 3 + 4 + 5 + 6) \cdot \frac{1}{6} \\&= 3,5\end{aligned}$$

$$u^2 = \sum_{i=1}^N (x_i - x_0)^2 f(x_i)$$

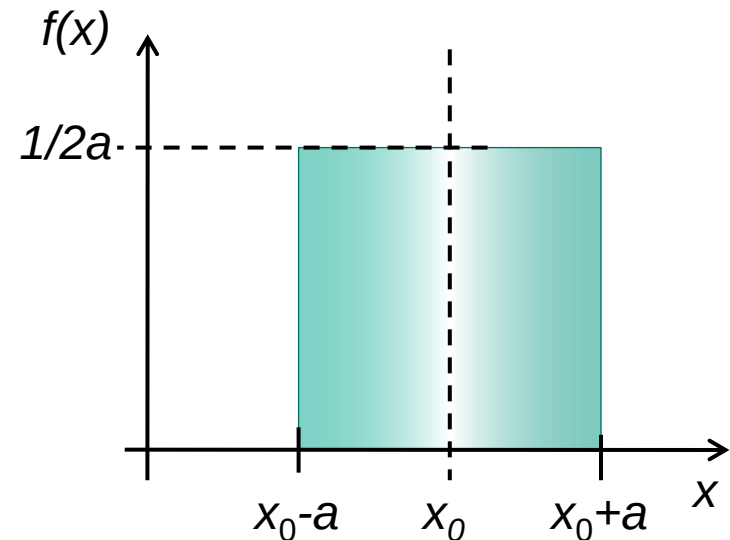
$$\text{with } x_0 = \sum_{i=1}^N x_i f(x_i)$$

$$u^2 = \sum_{i=1}^N (x_i - x_0)^2 f(x_i)$$

$$\begin{aligned}&= ((-2,5)^2 + (-1,5)^2 + (-0,5)^2 + 0,5^2 + 1,5^2 + 2,5^2) \frac{1}{6} \\&= 2,92\end{aligned}$$

Rectangular distribution

- Taken if it is only known that the uncertainty lies between certain bounds
- Example: next slide
- Variance:



$$u^2 = \int_{-\infty}^{\infty} (x - x_0)^2 f(x) dx$$

$$= \int_{x_0-a}^{x_0+a} \frac{1}{2a} (x - x_0)^2 dx = \frac{1}{2a} \frac{(x - x_0)^3}{3} \Big|_{x_0-a}^{x_0+a} = \frac{a^2}{3}$$

Example for rectangular distribution

- Given: Precision resistance of $100\ \Omega$ with a measurement uncertainty of $\pm 2\ \text{ppm}$ according to the calibration certificate
- Wanted: Variance and standard deviation

$$u^2 = \frac{1}{3} (2 \cdot 10^{-6})^2 = 1,33 \cdot 10^{-12} \quad (\text{relative})$$

$$\Rightarrow u = \sqrt{1,33 \cdot 10^{-12}} \cdot 100\Omega = 0,115\ \text{m}\Omega$$

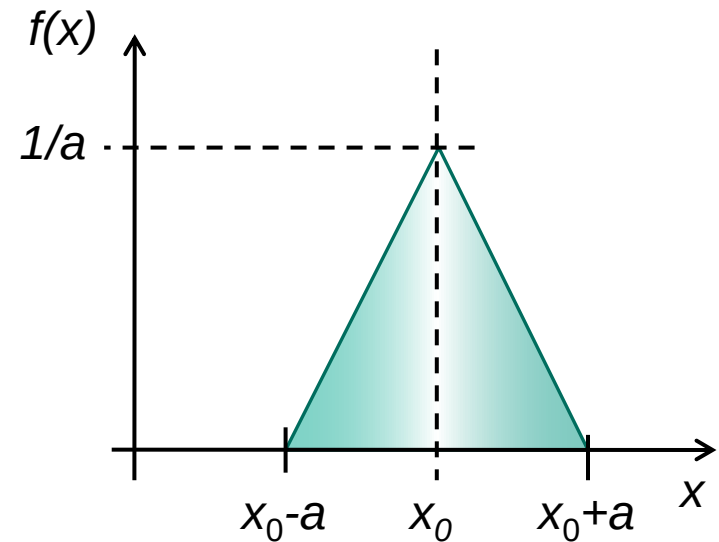
→ The resistance has a value of $(100,000000 \pm 0,000115)\ \Omega$.

Triangular distribution

F. Adunka, Messunsicherheiten

- Taken if it is known (e.g. from experience) that there is a preference for value x_0 between certain bounds
- Variance:

$$u^2 = \int_{-\infty}^{\infty} (x - x_0)^2 f(x) dx$$
$$= \frac{a^2}{6} \quad \text{with} \quad f(x) = \frac{1}{6} \left(1 - \frac{1}{a} |x - x_0|\right)$$



- Example: According to the manufacturer, the temperature coefficient can lie between -10 ppm and +10 ppm. A preference for small deviations is assumed from experience.

$$\rightarrow u^2 = \frac{1}{6} (10 \cdot 10^{-6})^2 = 1,67 \cdot 10^{-11} \quad (\text{relative})$$

Example for rectangular distribution

- Given: Precision resistance of $100\ \Omega$ with a measurement uncertainty of $\pm 2\ \text{ppm}$ according to the calibration certificate

and standard deviation

And why is this important?

$$u^2 = \frac{1}{3} (2 \cdot 10^{-8})^2 = 1,33 \cdot 10^{-12}$$

$$\Rightarrow u = \sqrt{1,33 \cdot 10^{-12}} \cdot 100\Omega = 0,115\ \text{m}\Omega$$

→ The resistance has a value of $(100,000000 \pm 0,000115)\ \Omega$.

Example for rectangular distribution

- Given: Precision resistance of $100\ \Omega$ with a measurement uncertainty of $\pm 2\ \text{ppm}$ according to the calibration certificate
- Wanted: Variance and standard deviation

$$u^2 = \frac{1}{3} (2 \cdot 10^{-6})^2 = 1,33 \cdot 10^{-12} \quad (\text{relative})$$

$$\Rightarrow u = \sqrt{1,33 \cdot 10^{-12}} \cdot 100\ \Omega = 0,115\ \text{m}\Omega$$

→ The resistance has a value of $(100,000000 \pm 0,000115)\ \Omega$.

better!

Mathematical models w/o covariance

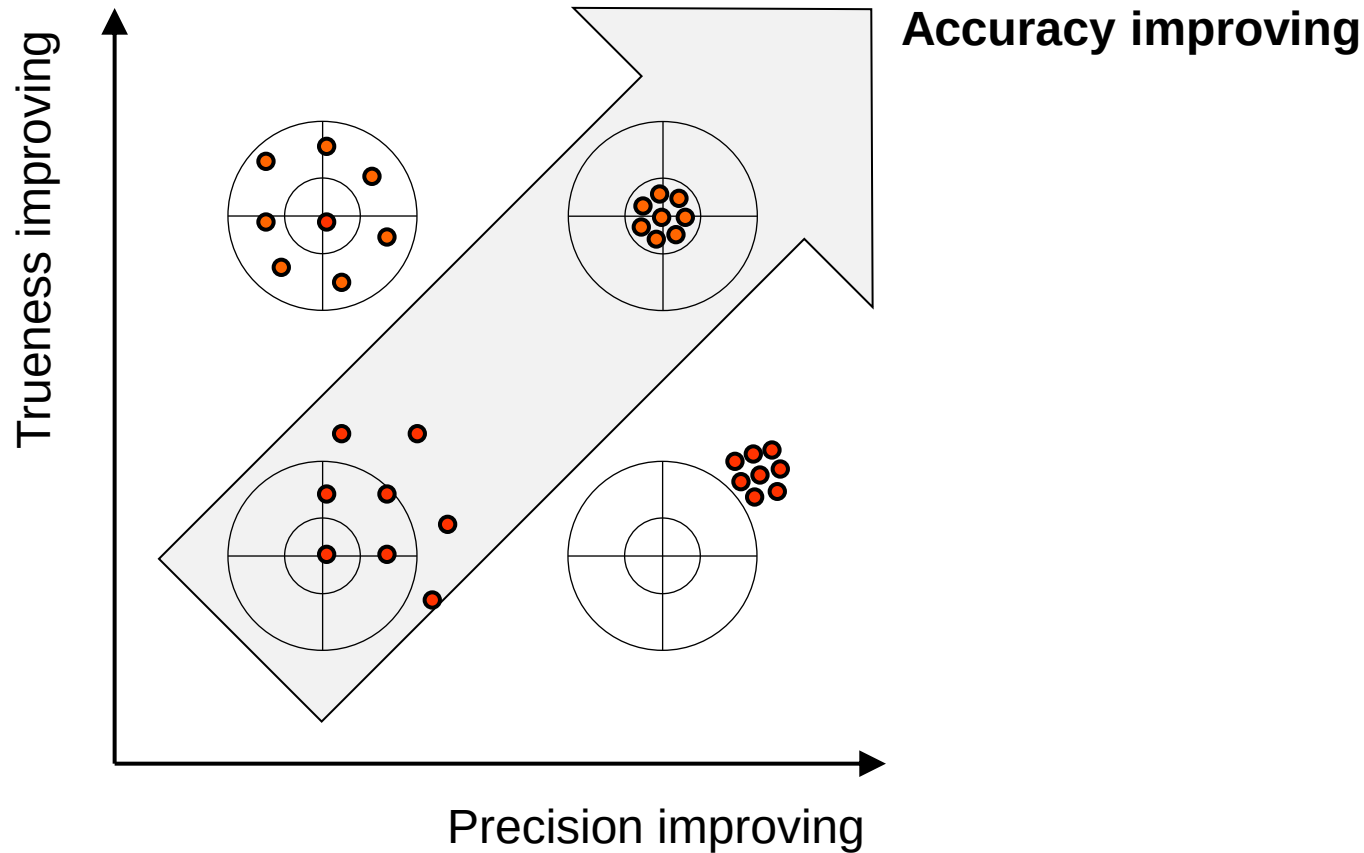
According to F. Adunka, Messunsicherheiten

- Measurement result constituted of several quantities
→ mathematical model $f(x_1, x_2, x_3, \dots, x_N)$ needed
- Example: $R=U/I$ (U, I = measured)
- The result of the measurement is then a function of the arithmetic means $x_{0,i}$: $y_0 = f(x_{0,1}, x_{0,2}, x_{0,3}, \dots, x_{0,N})$
- The (combined) variance is calculated via Gaussian propagation of uncertainties (here given for variances):

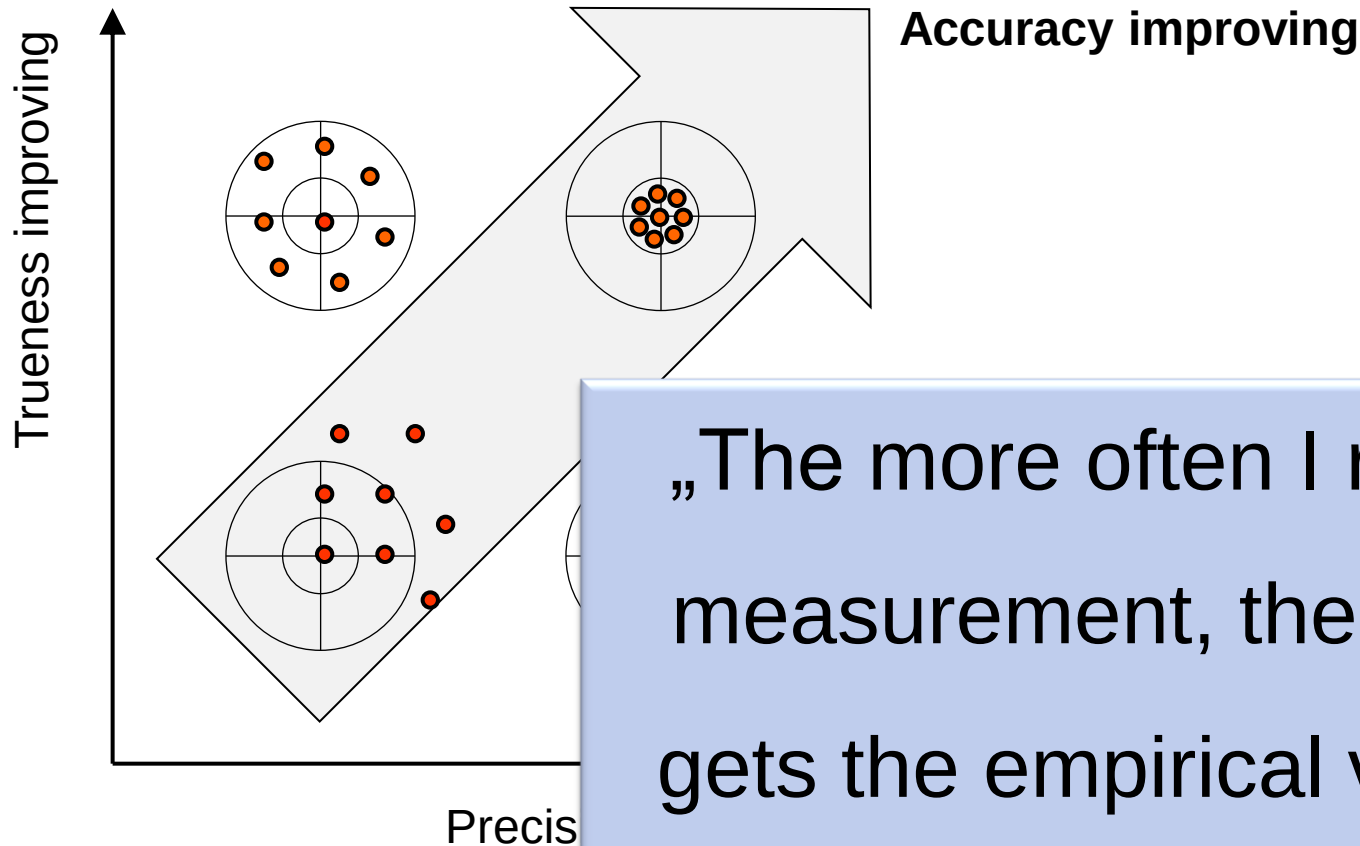
$$u^2(y) = \sum_{i=1}^N \left(\frac{\partial f}{\partial x_i} \right)^2 u^2(x_i)$$

Note: Formula derived via Taylor series (to $n = 1$); just valid for small u_i

Take-home-messages



Take-home-messages



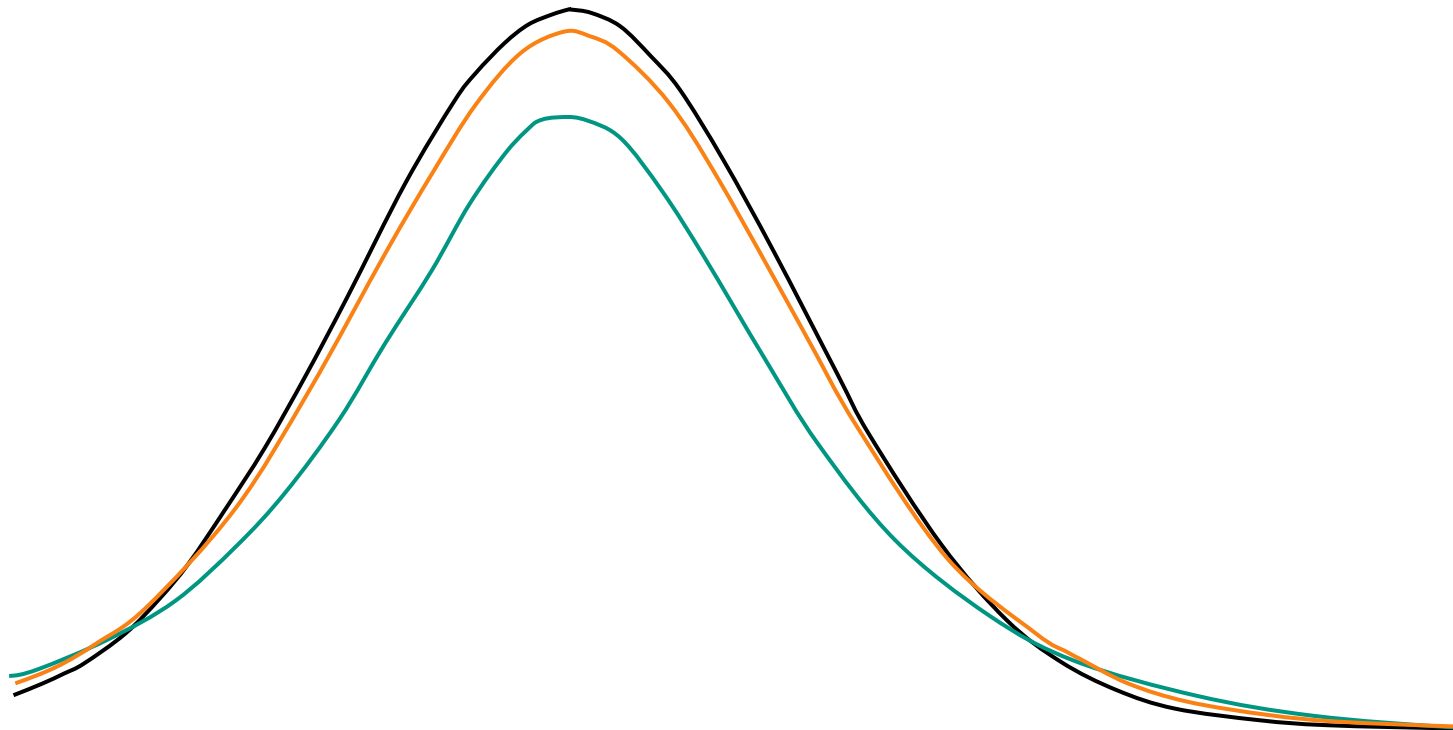
„The more often I repeat a measurement, the smaller gets the empirical variance of the mean.“

GUM

Guide to the expression of Uncertainty in Measurement

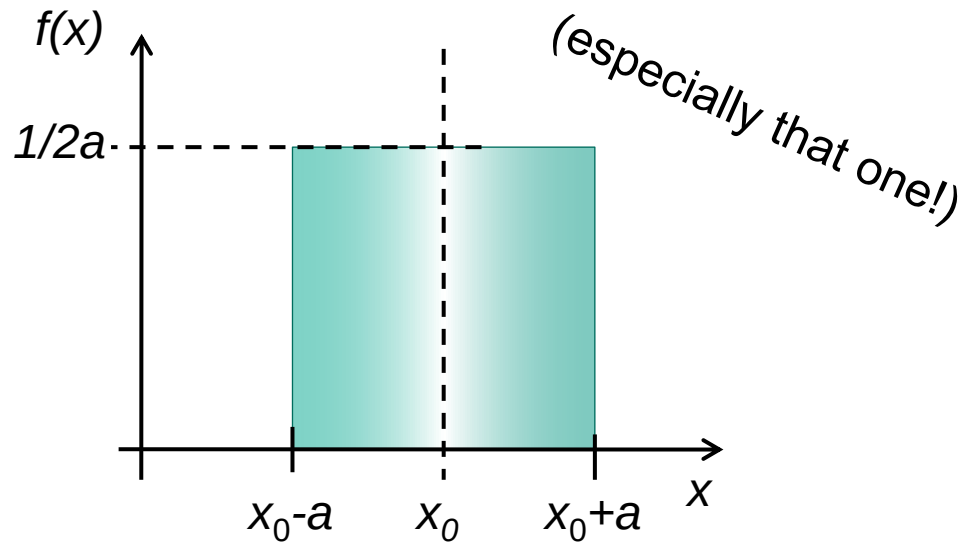
www.bipm.org/utis/common/documents/jcgm/JCGM_100_2008_E.pdf

Student distribution for small sample sizes (< 30)



Take-home-messages

Think of distribution functions!



Thank you!

