

A dark green horizontal banner with white text. The text reads 'SADC Water PT Chemistry workshop 2018 - Part 1: Traceability & Method Validation'. The background of the slide features a large, semi-circular grid pattern in green and orange, resembling a technical drawing or a scale.

# SADC Water PT Chemistry workshop 2018 - Part 1: Traceability & Method Validation

The speaker's name and dates are displayed in a green font. The name 'Maré Linsky' is on the first line, and the dates '26-27 November 2018' are on the second line. The background behind the text is a semi-circular grid pattern in orange and green.

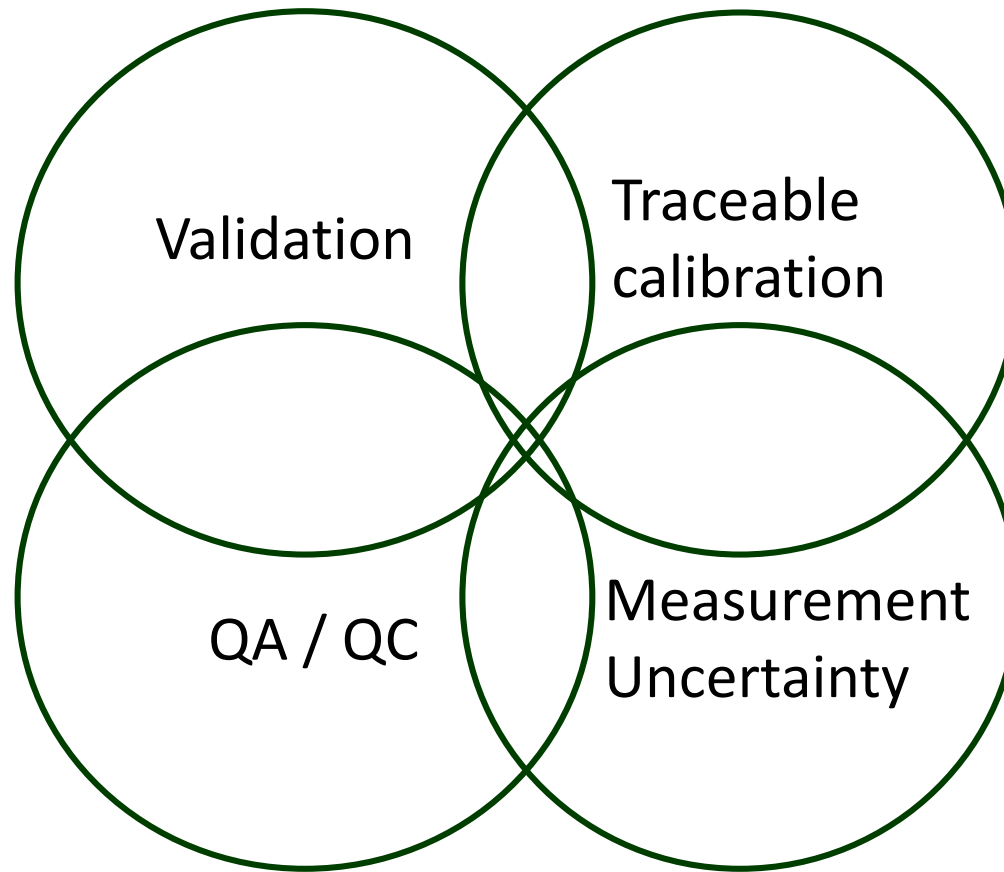
Maré Linsky  
26-27 November 2018

# Importance of Analytical Measurement

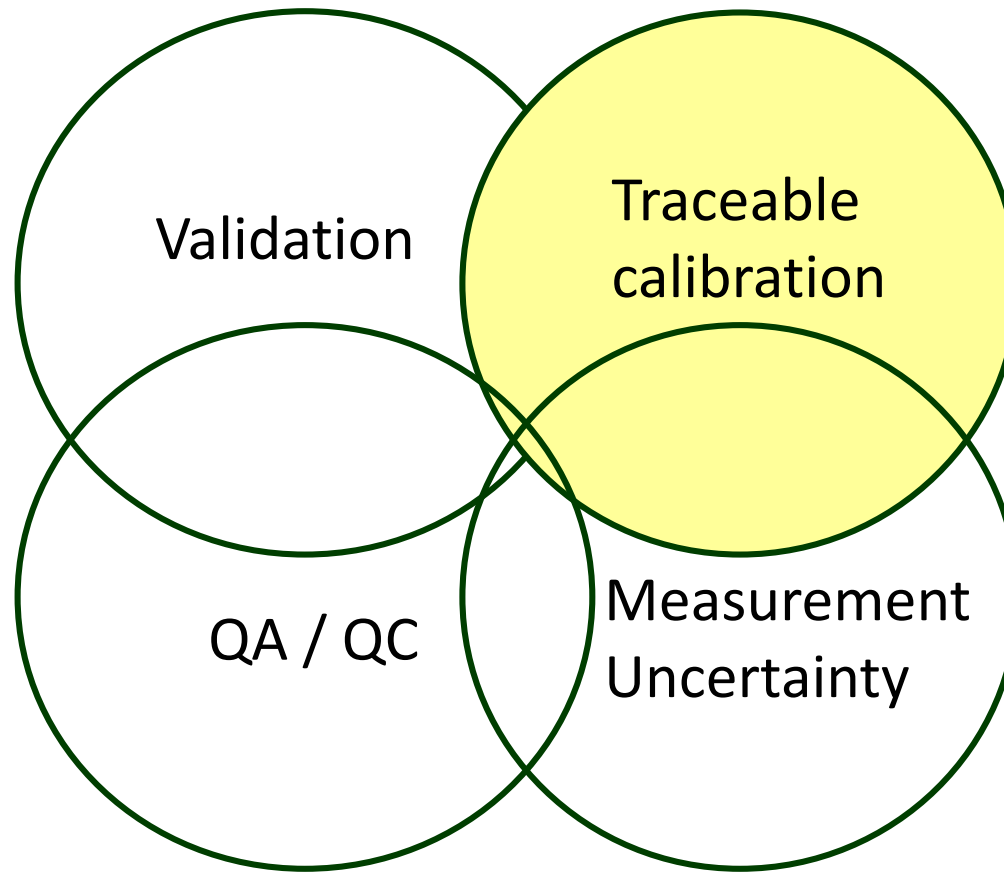
- Everyday millions of tests and measurements performed in thousands of laboratories around the world
  - Trade - Value of product
  - Quality of drinking water, food and feed
  - Healthcare
  - Forensics
  - Environmental analysis
- High costs associated with these measurements
- Impact of these measurements could be far-reaching
  - Health
  - Cost (fines)
  - Legal
- Ensuring the reliability of these measurements is the responsibility of the Analytical Chemist



# Ensuring valid Analytical Measurements



# Ensuring valid Analytical Measurements



# Metrological Traceability

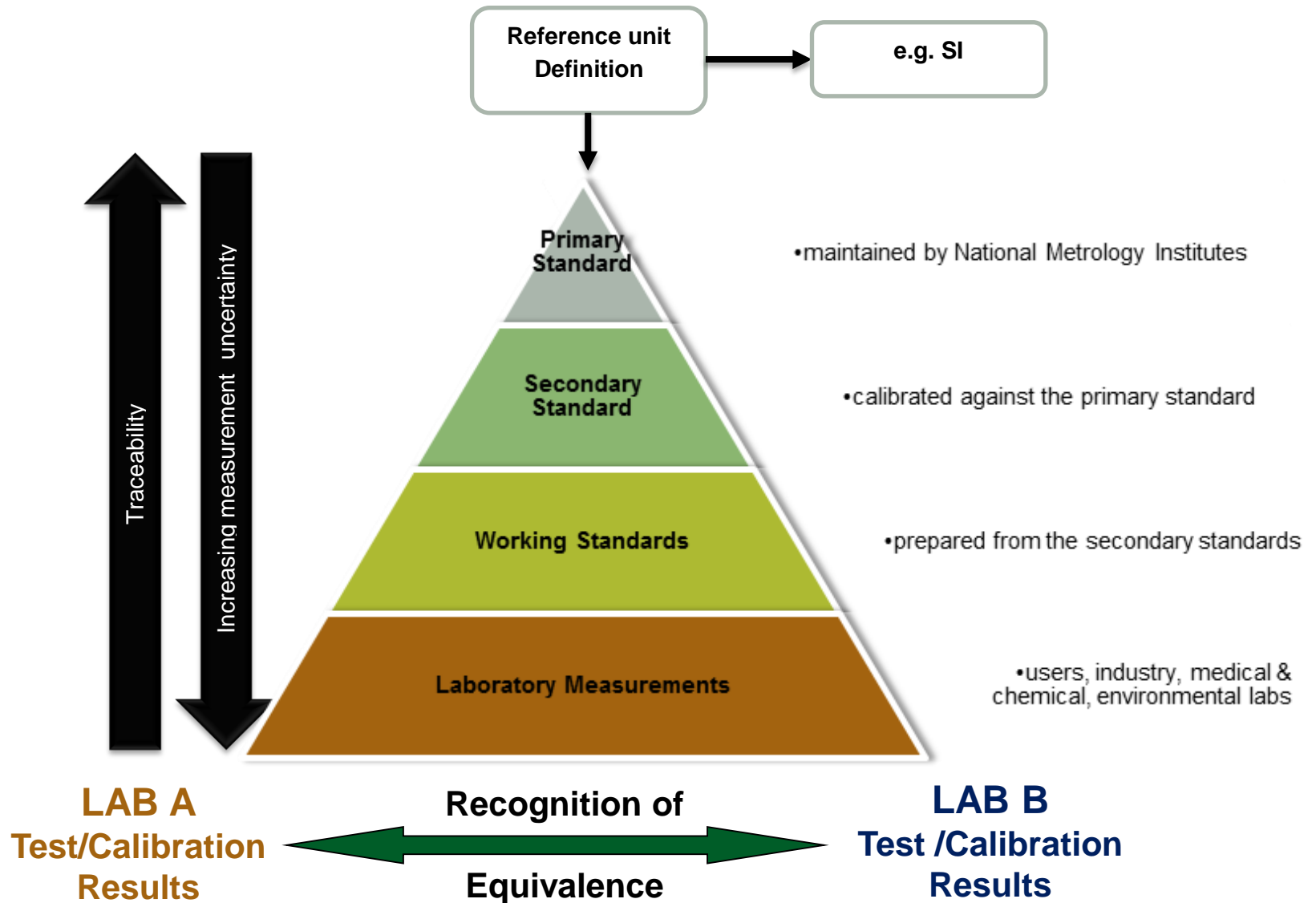
- Good analytical results are essential to ensure reliable decisions
- Result can be related to a reference through a documented unbroken chain of calibrations, each contributing to the measurement uncertainty
- Comparability through traceability to consistent and agreed set of measurement units and scales, i.e. SI



OR



# Establishing measurement traceability



# Metrological Traceability

- For every step in the traceability chain:
  - Documented, unbroken chain of calibrations
    - Traceable to appropriate references
  - Performed according to appropriate method
  - Measurements by technically competent laboratories
  - Measurement uncertainty determined according to agreed methods
- Establish traceability for:
  - Each parameter in mathematical model
  - Each of the specified conditions
    - *NOTE: Essential for critical values in measurement*

# Establishing Metrological Traceability in Analytical Chemistry



- Physical measurements, e.g. mass, volume, temperature
  - Uncertainties are typically not significant compared to those in analytical measurements
- Confirmation of Identity
  - Certified pure material
  - Authentic samples from reputable source
  - Reference data, e.g. reference wavelength spectra
- Amount of substance / Concentration
  - Pure Materials
  - Other reference materials, e.g. single- or multi-element standards, etc.
  - Certified reference materials



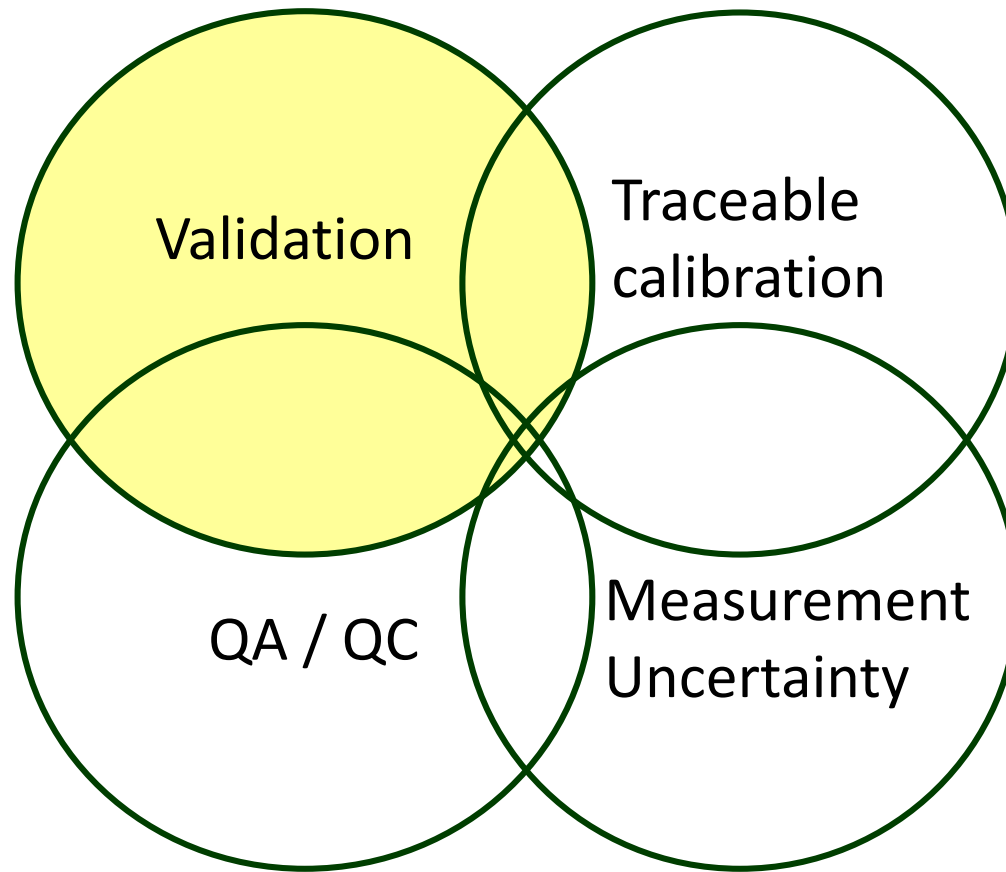
# Metrological Traceability

- Confirming metrological traceability of calibration standards, CRMs, etc.:
  - Accreditation of manufacturer to ISO 17034 and ISO 17025
  - Useful information stated on Certificate:
    - Specification of measurand
    - Measurement unit
    - Characterisation methods
    - Specifications for sample handling
    - Measurement uncertainty
- Confirming metrological traceability of calibration certificates, e.g. for mass balance
  - Accreditation of calibration laboratory to ISO 17025
    - Evaluation of trueness - Correction factors
    - Measurement Uncertainty

# Traceability in Analytical Chemistry

- To conclude:
  - Metrological Traceability – basis for establishing comparability of measurement results
    - Calibrated equipment, e.g. mass balance
    - Certified calibration standards
    - Validated methods
  - Uncertainty is part of the definition
    - Uncertainty of a traceable result = Uncertainty (reference) + Uncertainty (measurement)

# Ensuring valid Analytical Measurements



# Method validation

- Method validation is required to establish the fitness for purpose of a method for the specific requirements of customers when applied to a specific laboratory
- Method validation studies produce data on the overall performance or individual influence quantities associated with the results of a method in normal use in the laboratory

# Method Validation Approaches

- Data on overall method performance parameters are obtained from:
  - Interlaboratory studies
  - Single laboratory: In-house validation protocols
    - Validation
    - Verification

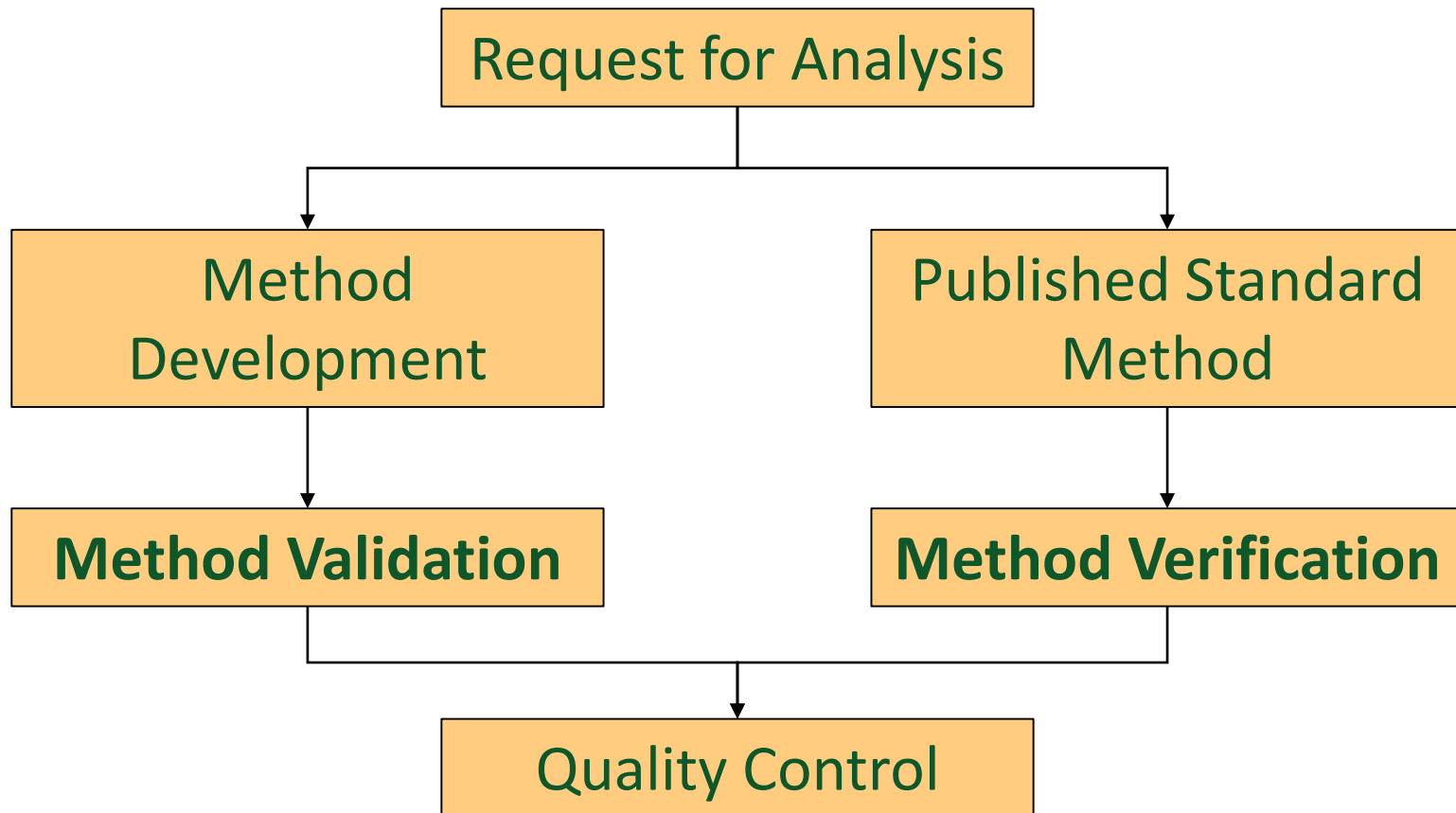
# Method Validation Approaches

- Interlaboratory
  - Published standardised procedure, e.g. ISO, ASTM
  - Validated, employing interlaboratory comparisons, according to international protocols (e.g. ISO 5725 standards)
  - Laboratory's responsibility to confirm that analytical performance can be matched. Typically only for:
    - Precision
    - Bias
    - Note: Robustness, Selectivity covered in Standard

# Method Validation Approaches

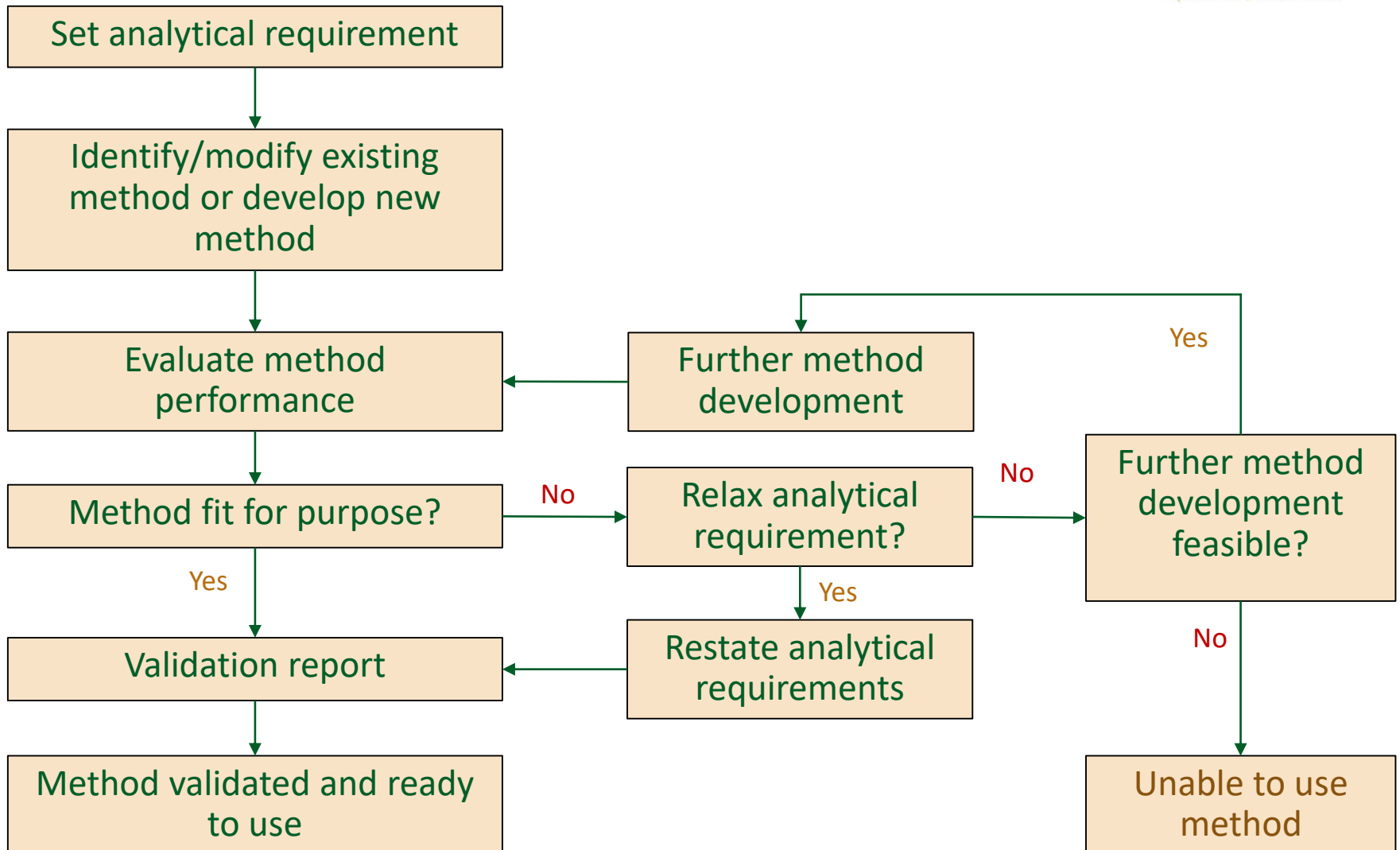
- Single laboratory
  - Method developed in-house
  - Standard method used outside it's scope
  - Evaluate all relevant performance criteria, e.g.
    - Selectivity; LOD, LOQ; Working Range; Trueness; Precision; Ruggedness; Uncertainty
  - Validation is a balance between costs, risks and technical possibilities (routine vs ad hoc)

# Approaches: Validation vs. Verification





# Method Validation Process



# Method Validation Approaches

- Emphasis is on identifying and removing or reducing significant effects, i.e. continue with method development if method performance is not satisfactory
- It is the responsibility of the laboratory to ensure that a method is fit for its intended use.

# Overview: Method Validation

- Performance Criteria

- Precision
- Trueness
- Limit of Detection (LOD) & Limit of Quantification (LOQ)
- Working range (Linearity)
- Selectivity / Specificity
- Robustness (ruggedness)
- Measurement Uncertainty
- Metrological Traceability

# General method validation requirement

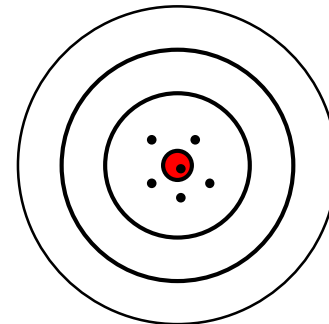
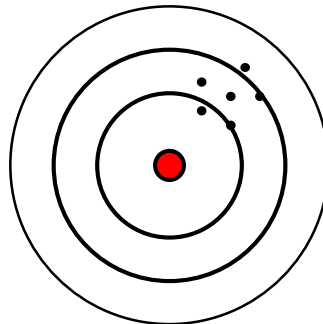
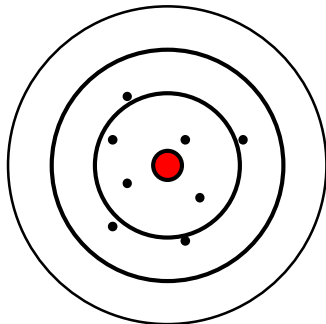
- Representativeness
- Representative variation

Realistic survey of the number and range of effects during normal use of the method, especially concentration ranges and sample types

# Precision

- How close independent results are to each other under specified conditions
- Determine typical variability, not minimum variability, i.e. ensure all operational conditions that would typically vary during routine operation are varied

$$s(x_k) = \sqrt{\frac{1}{n-1} \sum_{k=1}^n (x_k - \bar{x})^2}$$



# Precision

- **Repeatability standard deviation  $s_r$** 
  - Smallest variation in results
  - Single analyst performing analysis on the same equipment in 1 laboratory, over a short timescale (e.g. 1 day), using a single set of standards and reagents
- **Reproducibility standard deviation  $s_R$** 
  - Largest variation in results
  - Variability associated with different laboratories employing the same method



# Precision

- **Within-laboratory reproducibility,  $s_{RW}$** 
  - Largest variation that can be associated with results obtained in a single laboratory, i.e. *within-laboratory reproducibility*
  - Should represent typical variation that may be expected under routine operating conditions, e.g. different analysts performing analysis on the different equipment on different days, using independent sets of standards and reagents
- **Intermediate precision  $s_i$** 
  - Variation of specific variables

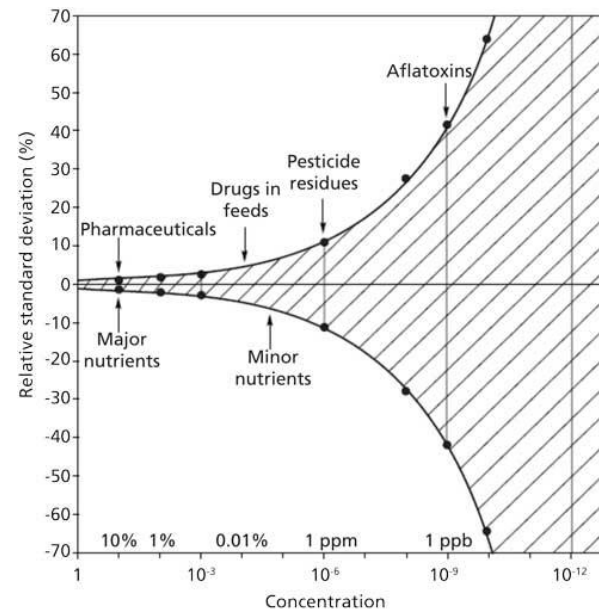
***Repeatability < Intermediate Precision < Reproducibility***



# Precision

- Experimental determination of precision
  - Repeated analysis of test samples (or CRM) at concentration levels covering the working range of the method – precision is generally dependent on analyte concentration
  - n= 6-15 recommended

$$s(x_k) = \sqrt{\frac{1}{n-1} \sum_{k=1}^n (x_k - \bar{x})^2}$$





# Example:

- A concentration of calibration standard for Cd (mg/L) was analysed 6 times on a single day to determine the instrument precision:

0.231

0.235

0.236

0.224

0.230

0.229

## Repeatability:

$$s(x_k) = \sqrt{\frac{1}{n-1} \sum_{k=1}^n (x_k - \bar{x})^2}$$

$$= 0.004 \text{ mg/L}$$

$$\%s_x = \frac{s_x}{\bar{x}} \times 100$$

$$= \frac{0.004}{0.231} \times 100$$

$$= 1.9\%$$

# Example:

- The following data was collected from a control chart for Ca-concentration (mg/L) in a water control sample over a period of 3 months. Calculate the % within-laboratory reproducibility of the method.

55.4	54.7	54.8	55.2
53.1	52.0	56.1	55.1

## Intermediate precision

$$s(x_k) = \sqrt{\frac{1}{n-1} \sum_{k=1}^n (x_k - \bar{x})^2}$$

$$= 1.34 \text{ mg/L}$$

$$\%S_x = \frac{S_x}{\bar{x}} \times 100$$

$$= \frac{1.34}{54.55} \times 100$$

$$= 2.5\%$$

# Example:

- In an interlaboratory comparison for Pb in drinking water, 23 laboratories participated employing the same ISO standard method. The consensus value was 0.0461 mg/L, with the standard deviation for all the participants being 0.0027 mg/L.

## % Reproducibility

$$\begin{aligned}\%s_x &= \frac{s_x}{\bar{x}} \times 100 \\ &= \frac{0.0027}{0.0461} \times 100 \\ &= 5.9\%\end{aligned}$$

# Comparing precision

- F-test:

- Comparison of methods' precision (standard deviation)

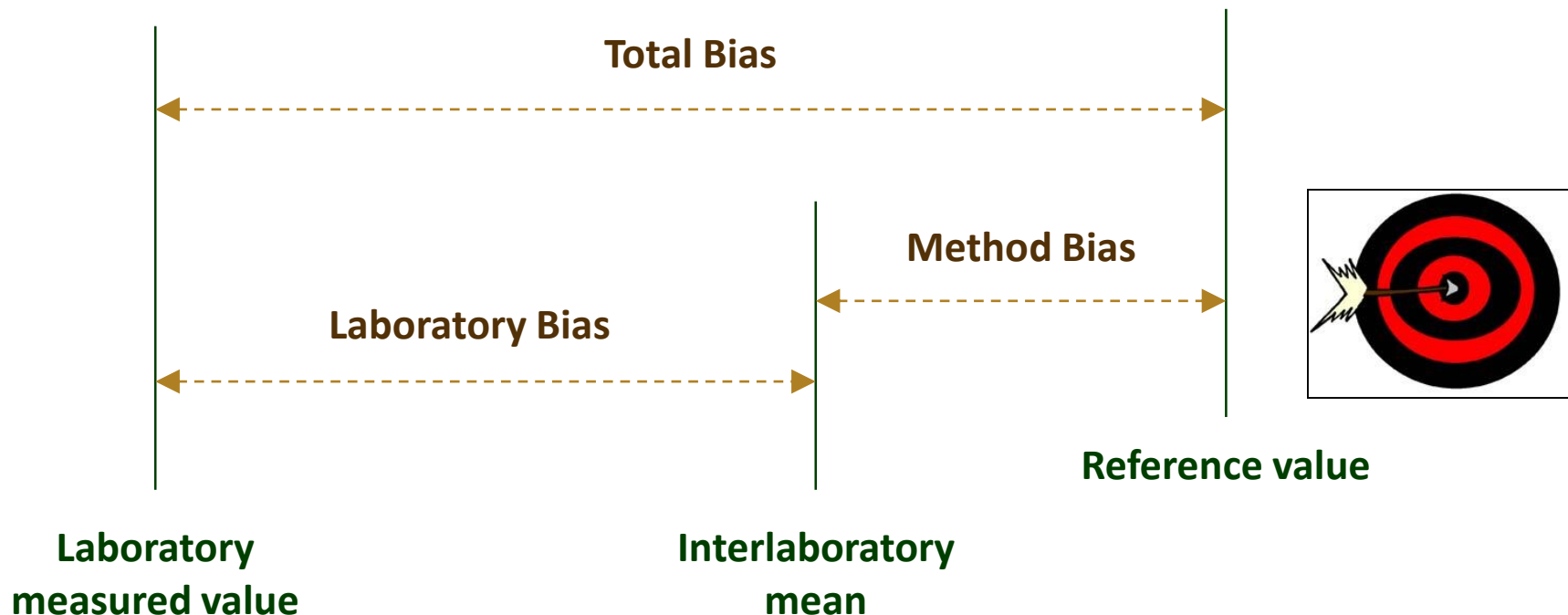
$$F = \frac{s_1^2}{s_2^2}$$

- Where  $F \geq 1$

- $F_{\text{calc}} < F_{\text{crit}}$ : No significant difference between two methods' variances

# Trueness

- Closeness of a number of measurements to the “true” value, i.e. evaluation of potential systematic error
  - Compare measurement mean to reference value



# Trueness

- Establish a reference / trueness through e.g.:
  - The analysis of Certified reference material
  - The analysis of Spiked samples
  - Employing an Alternative method
  - Participation in an Interlaboratory intercomparison / Proficiency Testing scheme
- Criteria for selection of reference material / samples
  - Representative
    - Sample types/matrices
    - Concentration levels of measurand
  - Independent from calibration standards



# Trueness

- Considerations when selecting an approach to evaluate trueness:
  - Certified matrix reference material
    - Ideal
    - Exact matching of matrix and analyte concentration may be difficult to achieve
    - Expensive
  - Spiking
    - Behaviour of added measurand is probably different from naturally incurred measurand (e.g. bound to matrix)
    - Unrealistically high recoveries can be expected

# Trueness

- Considerations when selecting an approach to evaluate trueness:
  - Alternative method
    - Uncertainty (Reference Method) < Uncertainty (Candidate Method)
    - Tested on real samples
  - Interlaboratory Comparison / PT
    - Exact matching of matrix and analyte concentration may be difficult to achieve
    - Reputability of study, e.g. number of participants, reliability of reference value



# Trueness

- General considerations
  - Experimentally
    - At 3 concentration levels
      - Close to limit of detection
      - Mid-range
      - Upper concentration limit
    - n=9 recommended
  - Expected to be negligible or accounted for

# Trueness

- Statistical evaluation of data
  - Outlier testing
    - Grubbs
    - Dixon
- Performance criteria
  - % Recovery
  - % Bias
  - t-test
  - $E_n$  or  $\zeta$ -score



# Trueness: Outlier testing

- View results graphically
- Grubbs' test
  - Small number of samples
  - ISO recommended

$$G = \frac{|SuspectValue - \bar{x}|}{s}$$

- Outlier if  $G_{calc} > G_{crit}$
- Dixon's test (Q-test)
  - Sample sizes:  $n = 3 - 7$

$$Q = \frac{|SuspectValue - NearestValue|}{LargestValue - SmallestValue}$$

- Outlier if  $Q_{calc} > Q_{crit}$

n	$Q_{crit}$ CL at 90%	$Q_{crit}$ CL at 95%	$Q_{crit}$ CL at 99%
3	0.941	0.970	0.994
4	0.765	0.829	0.926
5	0.642	0.710	0.821
6	0.560	0.625	0.740
7	0.507	0.568	0.680
8	0.468	0.526	0.634
9	0.437	0.493	0.598
10	0.412	0.466	0.568

# Dixon Test

The Dixon test can be used to test for a single outlier in a univariate data set. This test is primarily used for small data sets (Dataplot limits the sample to be between 3 and 30).

Specifically, given a set of ordered observations  $Y_1, Y_2, \dots, Y_N$ , the Dixon test is computed as follows:

Sample Size	Test for Minimum	Test for Maximum
$3 \leq N \leq 7$	$(Y_2 - Y_1) / (Y_N - Y_1)$	$(Y_N - Y_{N-1}) / (Y_N - Y_1)$
$8 \leq N \leq 10$	$(Y_2 - Y_1) / (Y_{N-1} - Y_1)$	$(Y_N - Y_{N-1}) / (Y_N - Y_2)$
$11 \leq N \leq 13$	$(Y_3 - Y_1) / (Y_{N-1} - Y_1)$	$(Y_N - Y_{N-2}) / (Y_N - Y_2)$
$14 \leq N \leq 30$	$(Y_3 - Y_1) / (Y_{N-2} - Y_1)$	$(Y_3 - Y_1) / (Y_{N-2} - Y_1)$

# Dixon Test

Dixon's Q test, or just the "Q Test" is a way to find outliers in very small, normally distributed, data sets. Small data sets are usually defined as somewhere between 3 and 7 items

$$Q_{exp} = \frac{x_2 - x_1}{x_n - x_1}$$

Finding the Q statistic for different sample sizes (n) of between 8 and 30 (in Step 2 above):

8 < n < 10: use  $R_{11}$ :

$$r_{11} = \frac{x_2 - x_1}{x_{n-1} - x_1}$$

11 < n < 13: use  $R_{21}$ .

$$r_{21} = \frac{x_3 - x_1}{x_{n-1} - x_1}$$

14 < n < 30: use  $R_{22}$ .

$$r_{22} = \frac{x_3 - x_1}{x_{n-2} - x_1}$$

# Trueness

- % Recovery

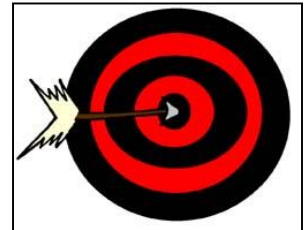
- Expected to be close to 100%
  - Depending on application field and concentration levels

$$\% Recovery = \frac{C_{meas}}{C_{Ref}} \times 100$$

- % Bias

- Expected to be close to 0%
  - Depending on application field and concentration levels

$$\% Bias = \frac{C_{meas} - C_{Ref}}{C_{Ref}} \times 100$$

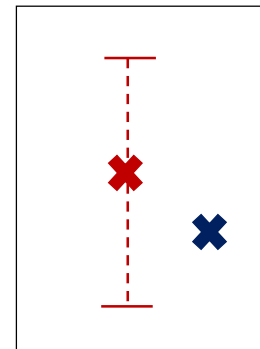


# Trueness

- t-test

- Comparison of mean with reference value
- Uncertainty of reference value not considered
- Null hypothesis ( $H_0$ ): No significant difference between measured and “true” value
  - $H_0: t_{\text{calc}} < t_{\text{crit}}$ : No significant bias

$$t = \frac{(\bar{x} - \mu)\sqrt{n}}{s}$$



# Trueness

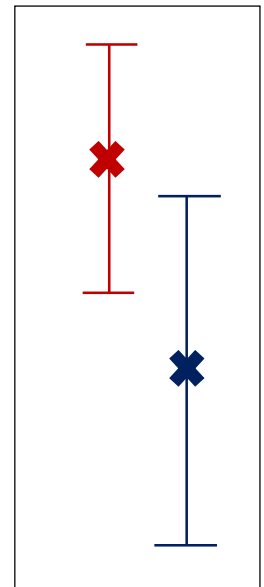
- $E_n$ -score /  $\zeta$ -score

- A measure of agreement between the reference value ( $X$ ) and the method's result ( $x$ ), i.e. are they within their respective uncertainty ranges?

$$E_n = \frac{x - X}{\sqrt{U(x)^2 + U(X)^2}}$$

$$\zeta = \frac{x - X}{\sqrt{u(x)^2 + u(X)^2}}$$

En Score	$\zeta$ -score	
$ E_n  \leq 1$	$ \zeta  \leq 2$	Satisfactory
$ E_n  > 1$	$ \zeta  > 2$	Unsatisfactory





# Trueness

## - Example

- During method validation for analysis of alloy samples a CRM was analysed 10 times for Zn:

$$C_{CRM} = 2.013 \pm 0.034 \text{ g/kg}$$

$$C_{meas} = 2.035 \pm 0.054 \text{ g/kg}$$

- Calculate the % Bias and % Recovery for this method.

$$\%Bias = \frac{C_{meas} - C_{CRM}}{C_{CRM}} \times 100$$

$$= \frac{2.035 - 2.013}{2.013} \times 100$$

$$= 1.09\%$$

$$\%Recovery = \frac{C_{meas}}{C_{CRM}} \times 100$$

$$= \frac{2.035}{2.013} \times 100$$

$$= 101.09\%$$

# Trueness

## - Example

- During method validation for analysis of alloy samples a CRM was analysed 10 times for Zn:

$$C_{\text{CRM}} = 2.013 \pm 0.034 \text{ g/kg}$$

$$C_{\text{meas}} = 2.035 \pm 0.054 \text{ g/kg}$$

- Calculate if there is a significant difference between the mean and the consensus true value, employing the t-test.

$$t_{\text{calc}} = \frac{(\bar{x} - \mu)\sqrt{n}}{s} = \frac{(2.035 - 2.013)\sqrt{10}}{0.054} = 1.29$$

$$t_{\text{crit}} = 2.26$$

$H_0: t_{\text{calc}} < t_{\text{crit}}: \text{ No significant bias}$

# Trueness

## - Example

- During method validation for analysis of alloy samples a CRM was analysed 10 times for Zn:

$$C_{\text{CRM}} = 2.013 \pm 0.034 \text{ g/kg}$$

$$C_{\text{meas}} = 2.035 \pm 0.054 \text{ g/kg}$$

- Calculate if there is a significant difference between the mean and the consensus true value, employing the  $E_n$ -score.

$$\begin{aligned} E_n &= \frac{x - X}{\sqrt{U_x^2 + U_{ref}^2}} = \frac{2.035 - 2.013}{\sqrt{(0.054)^2 + (0.034)^2}} \\ &= 0.34 \end{aligned}$$

$|E_n| \leq 1$ : No significant bias

# Trueness

## - Example

- During method validation for analysis of alloy samples a CRM was analysed 10 times for Zn:

$$C_{\text{CRM}} = 2.013 \pm 0.034 \text{ g/kg}$$

$$C_{\text{meas}} = 2.035 \pm 0.054 \text{ g/kg}$$

- Calculate if there is a significant difference between the mean and the consensus true value, employing the  $\zeta$ -score.

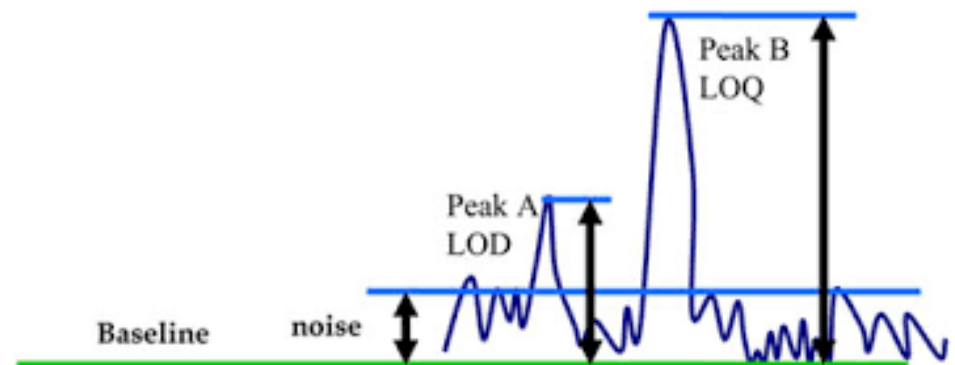
$$\begin{aligned}\zeta &= \frac{x - X}{\sqrt{u(x)^2 + u(X)^2}} = \frac{2.035 - 2.013}{\sqrt{(0.027)^2 + (0.017)^2}} \\ &= 0,69\end{aligned}$$

$|\zeta| \leq 2$ : No significant bias

# Validation parameters

## Limit of Detection & Quantification

- Limit of Detection (LOD):
  - Lowest concentration that can be reliably detected, but not quantified.
- Limit of Quantification (LOQ):
  - Lowest concentration that can be accurately quantified / *at which performance is acceptable for typical application.*



# Validation parameters LOD & LOQ

- Limit Of Detection (LOD)

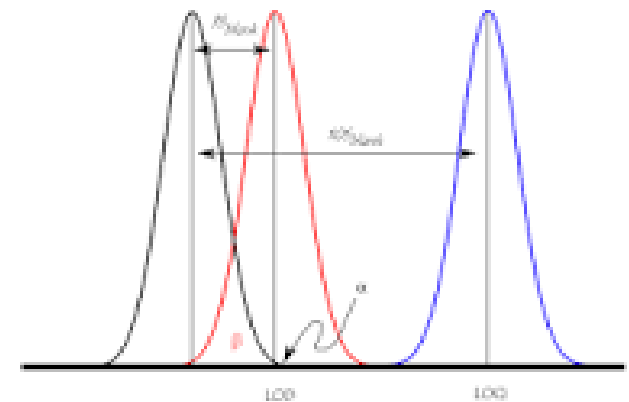
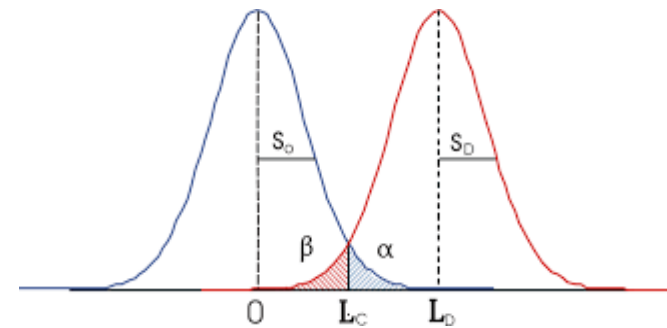
$$LOD = 3 \cdot s'_0$$

- Limit of Quantification (LOQ)

$$LOQ = k_Q \cdot s'_0$$

With:

- $k_Q = 10$  (corresponding to 10% RSD)
- Note:  $s'_0$  should be in concentration units



# Validation parameters

## LOD & LOQ

- Experimentally:
  - Sample:
    - Blank
      - Reagent blank
      - Sample blank - matrix
    - Samples with concentration @ LOD
  - Number of measurements
    - A reliable estimate of standard deviation requires 6-15 measurements.
    - In practice, typically 10
  - Calculate Standard deviation,  $s'_0$

# Validation parameters

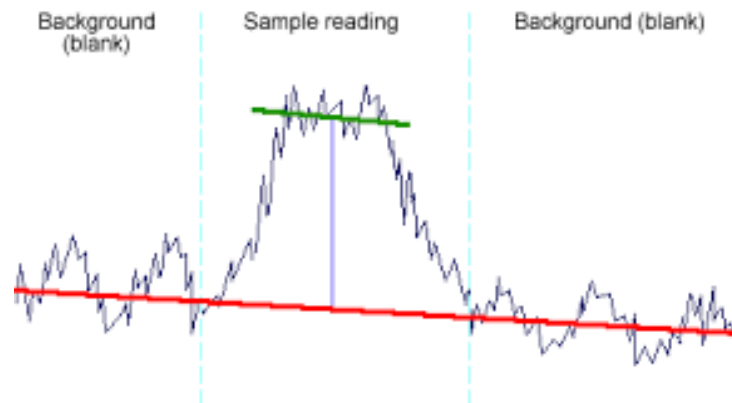
## LOD & LOQ

- **Instrument LOD:**
  - Sample blank / low concentration sample directly analysed on instrument, i.e. no sample preparation.
- **Method LOD:**
  - Sample blank / low concentration sample taken through complete sample preparation procedure.
  - Results calculated as stipulated in measurement procedure (i.e. corrected for dilution effects)

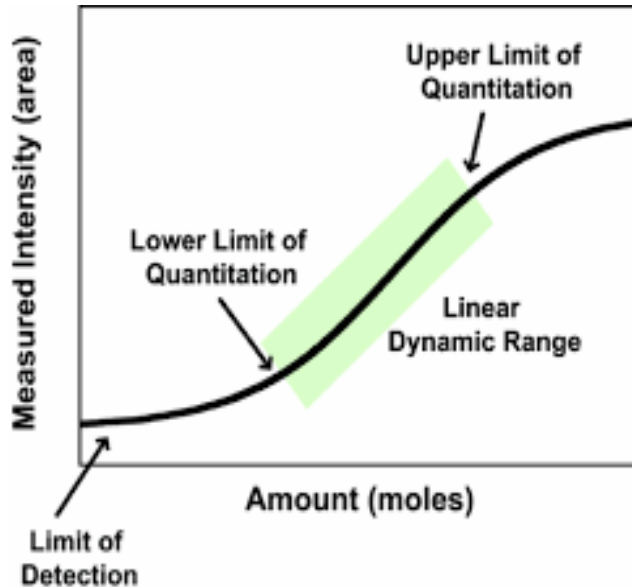


# Validation parameters LOD & LOQ

- Considerations regarding reliability
  - If blank values varies significantly from day-to-day:
    - $s'_0$  should be intermediate precision, rather than repeatability
  - Samples concentrations expected to be close to LOD:
    - LOD/LOQ should be monitored regularly – estimates of standard deviation are inherently variable



# Working range



- Interval over which method provides results with an acceptable uncertainty.
  - Lower end:
    - LOQ, or
    - Minimum expected concentration in samples
  - Upper end:
    - Concentrations at which anomalies in analytical sensitivities become apparent  
\_ *Linearity*

# Working range

## • Linearity Evaluation

- Visually inspect
- Regression statistics
  - e.g.  $r^2 > 0.9995$
- Residuals plot
  - Random distribution about zero: Linear
  - Systematic trends: Non-linear

## • Non-linearity corrected for by:

- Restricted operating range
- Non-linear calibration

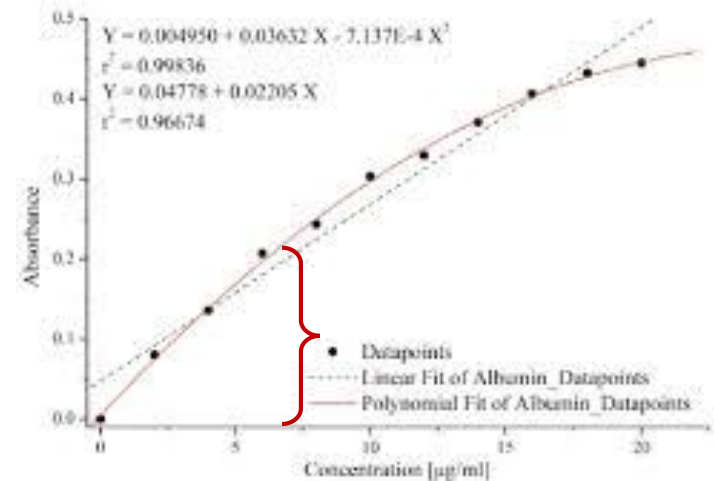
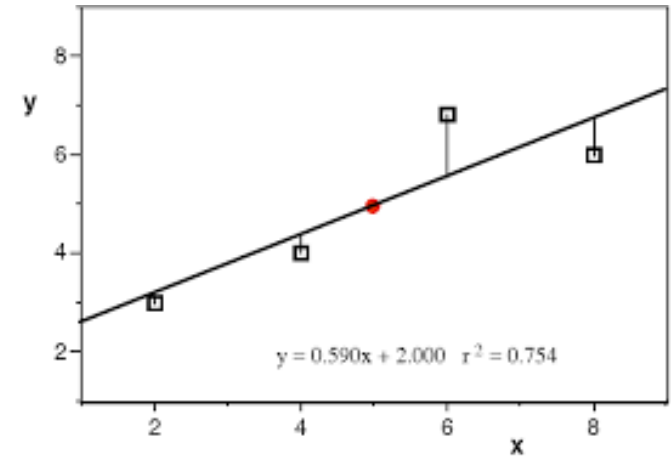
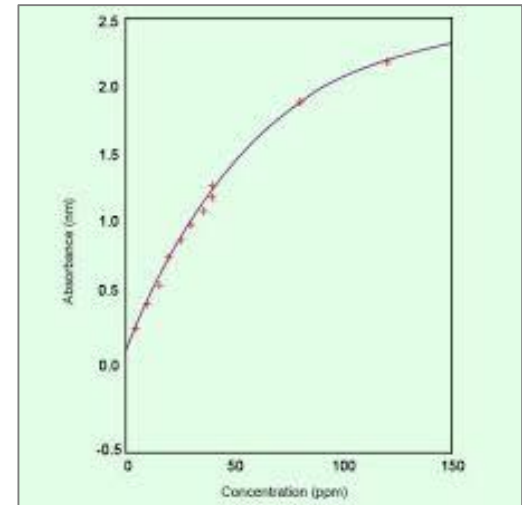
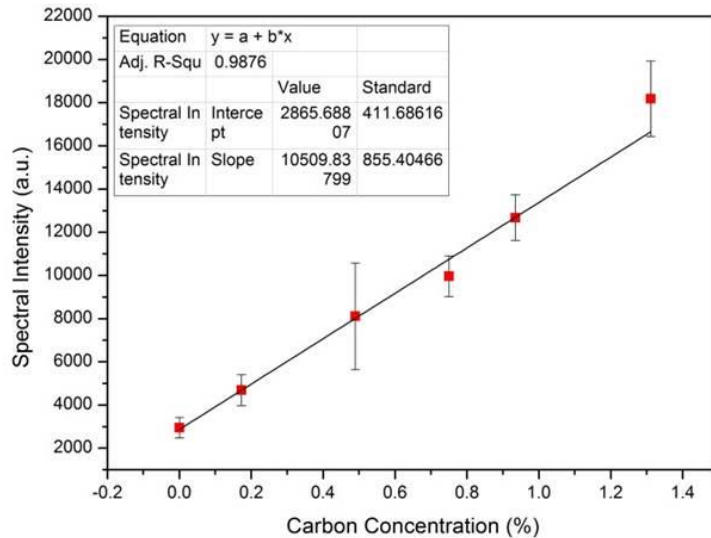


Figure 3 The essence of a linear model optimization for experimental data in Least Squares Method. The model-experiment differences are indicated by solid lines.

# Working range

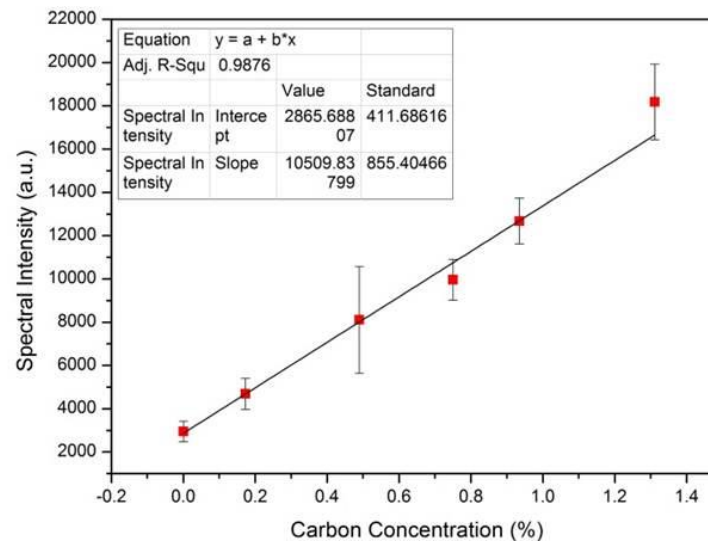
- Instrument calibration

- Linear calibration ( $n \geq 5$ )
- Quadratic calibration ( $n \geq 7$ )
- Higher functions not advisable
- Weighted fit (standard deviation proportional to concentration)
- Transformation of values – e.g. log-normal calibration



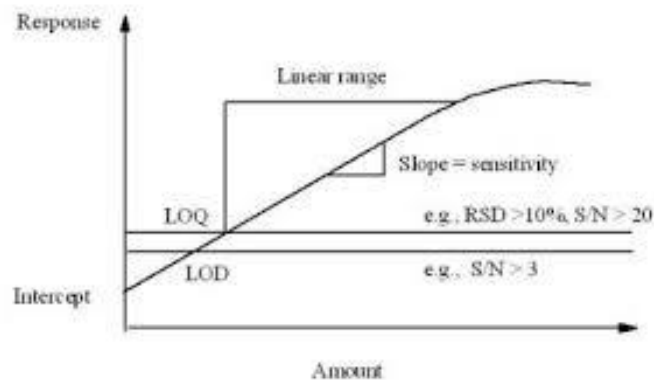
# Working range

- Method working range
  - Instrument working range
  - Sample preparation restrictions
    - Minimum/maximum sample size
    - Dilution factors



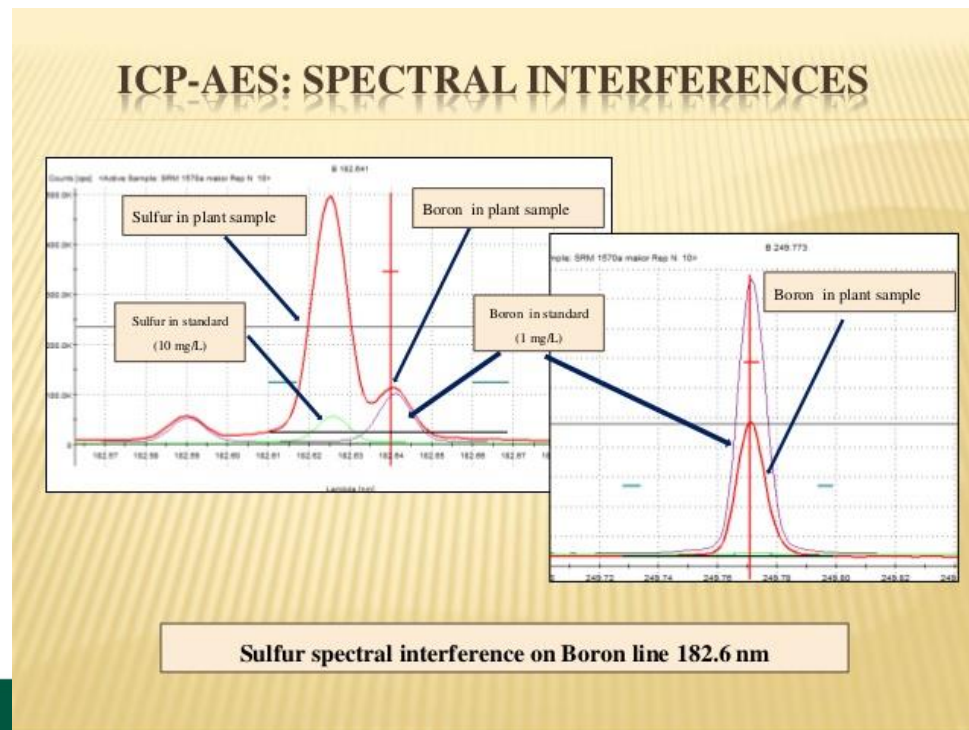
# Sensitivity

- Change in instrument response corresponding to change in measured quantity
- As part of instrument quality assurance the sensitivity can be checked routinely to ensure that it doesn't fall below a minimum level



# Selectivity/specificity

- The degree to which a method responds uniquely to the required analyte
- Interference may increase (enhance) or decrease (suppress) analyte signal



# Selectivity/specificity

- Experimentally:
- Investigate the effects of potential interferents:
  - Add potential interferent to blank
  - Add potential interferent to samples
  - Independent technique
  - Certified Reference Material
- Determine Trueness
  - Recovery / Bias
  - T-test
- Normally used to demonstrate insignificant effects



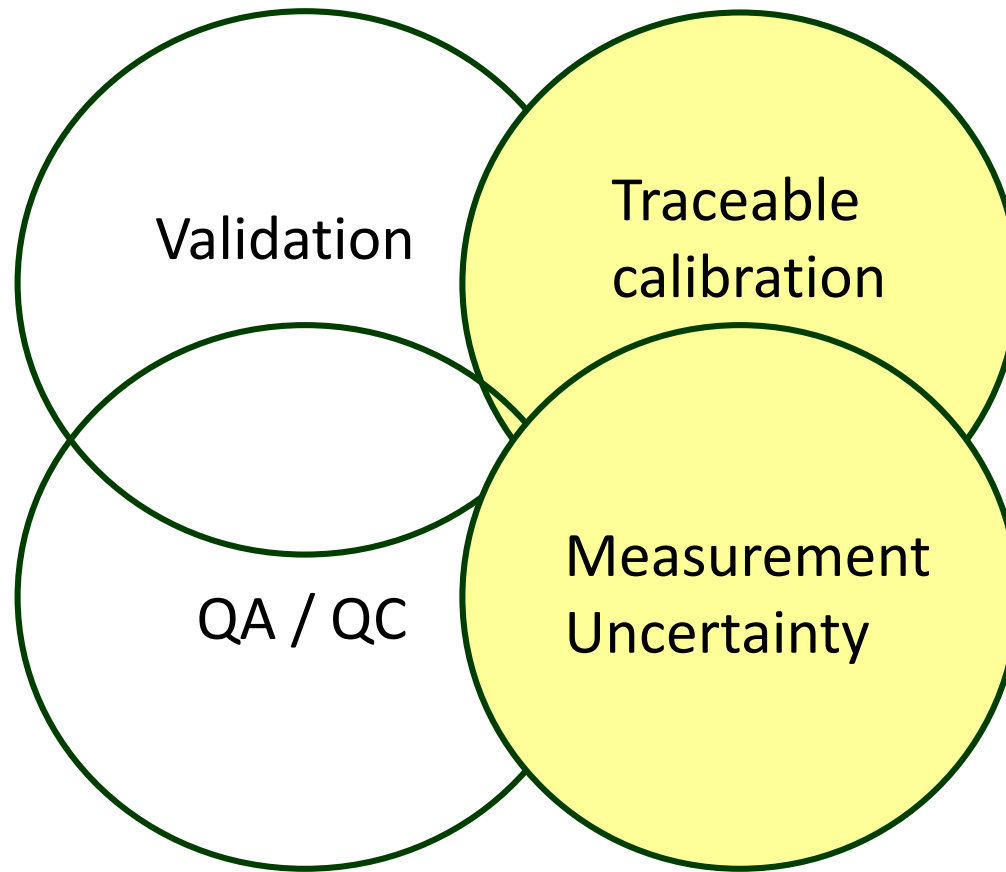
# Robustness (ruggedness)

- Measure of a method's capacity to remain unaffected by small, but deliberate variations in method parameters
- Provides indication of method's reliability during normal use
- Required for:
  - In-house methods
  - Method developed from scientific literature
  - Standard methods used outside the method's scope
- *Not required:*
  - *Standard methods used within method's scope*

# Robustness (ruggedness)

- Evaluation method:
  - Identify variables that could have significant effect on method
  - Make deliberate changes to variables identified to determine the effect of changes on the results
  - Significance testing to establish if statistically relevant
  - If the effect is significant:
    - Ensure that variable is effectively controlled when using the method
    - Improve method

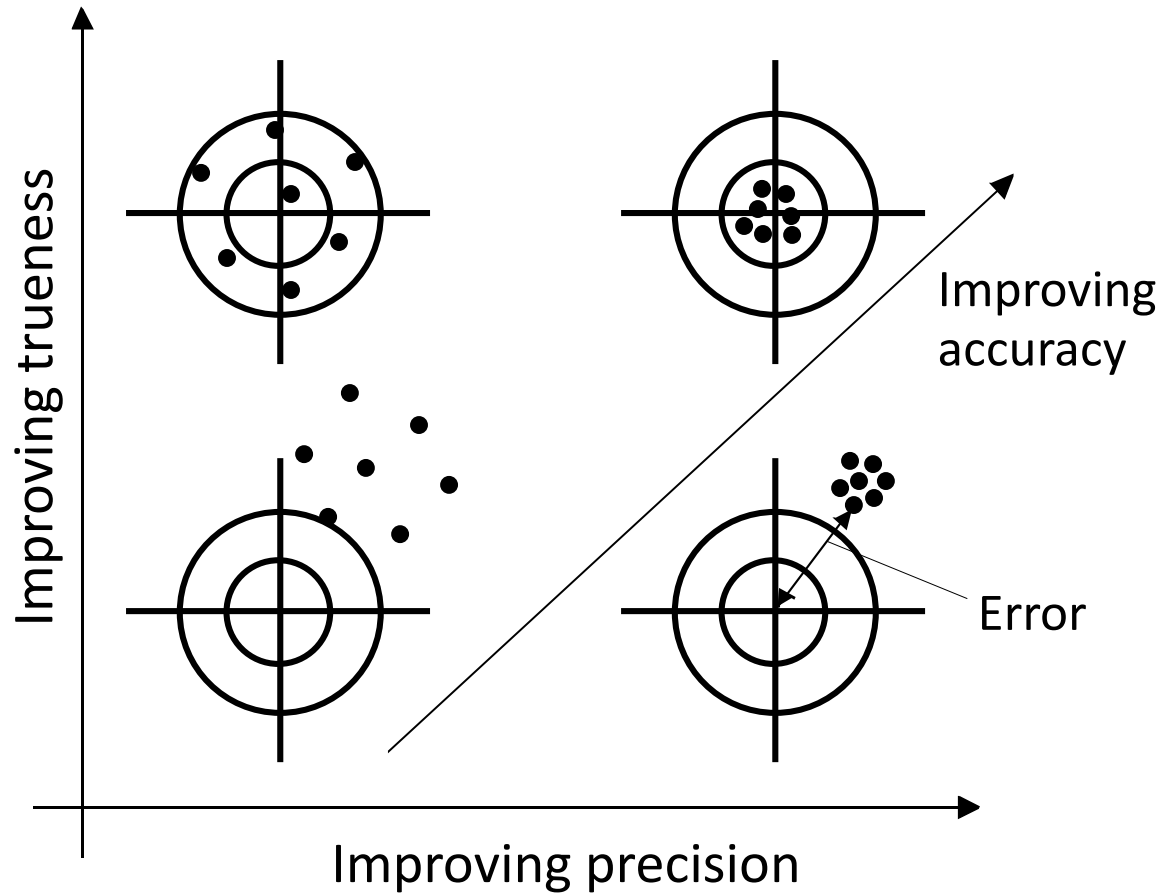
# Ensuring valid Analytical Measurements



# Fit for purpose..

Performance Characteristic	Type of Analytical Application		
	Qualitative	Quantitative: Impurity	Quantitative: Main comp.
Selectivity	✓	✓	✓
LOD			
LOQ		✓	
Working range, incl. Linearity		✓	✓
Trueness		✓	✓
Precision ( $s_r$ , $s_i$ )		✓	✓

# Accuracy: Trueness & Precision



from: LGC, vamstat II

# Conclusion

- For valid analytical measurements, we need
  - Measurement traceability through:
    - Calibrated equipment
    - Certified calibration standards
  - Validated methods that are:
    - Fit for purpose
    - Accuracy (trueness and precision) critical
- Also...
  - Quality control
  - Estimation of uncertainty of measurement

# Sodium in Food: Proficiency testing scheme

<b>Round 1</b>	Savoury Stock Powder (Oct 2018)
<b>Round 2</b>	Bread (Dec 2018)
<b>Round 3</b>	Instant Noodles (Feb 2019)
<b>Round 4</b>	Fat Spread (Apr 2019)
<b>Round 5</b>	Flavoured Snack/Crisp (May 2019)
<b>Round 6</b>	Cured processed meat product (Jul 2019)

# Sodium in Food: Proficiency testing scheme

<b>Main parameter: Salt (sodium) content</b>	
<b>Optional Parameters:</b>	
Iron and Zinc	Moisture
Protein	Fat
Total sugar	Cholesterol
Dietary fibre	Vitamins A, B1, B2, B3, B6, folic acid
Energy	

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*We measure what matters*

**Thank you**

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