

30 years of Measurement Uncertainty in Chemistry

A retrospective and a forward look

Steve Ellison, LGC Limited, Guildford, UK

Prehistory – before the GUM

Accreditation

Interlaboratory study

CRMs

PT

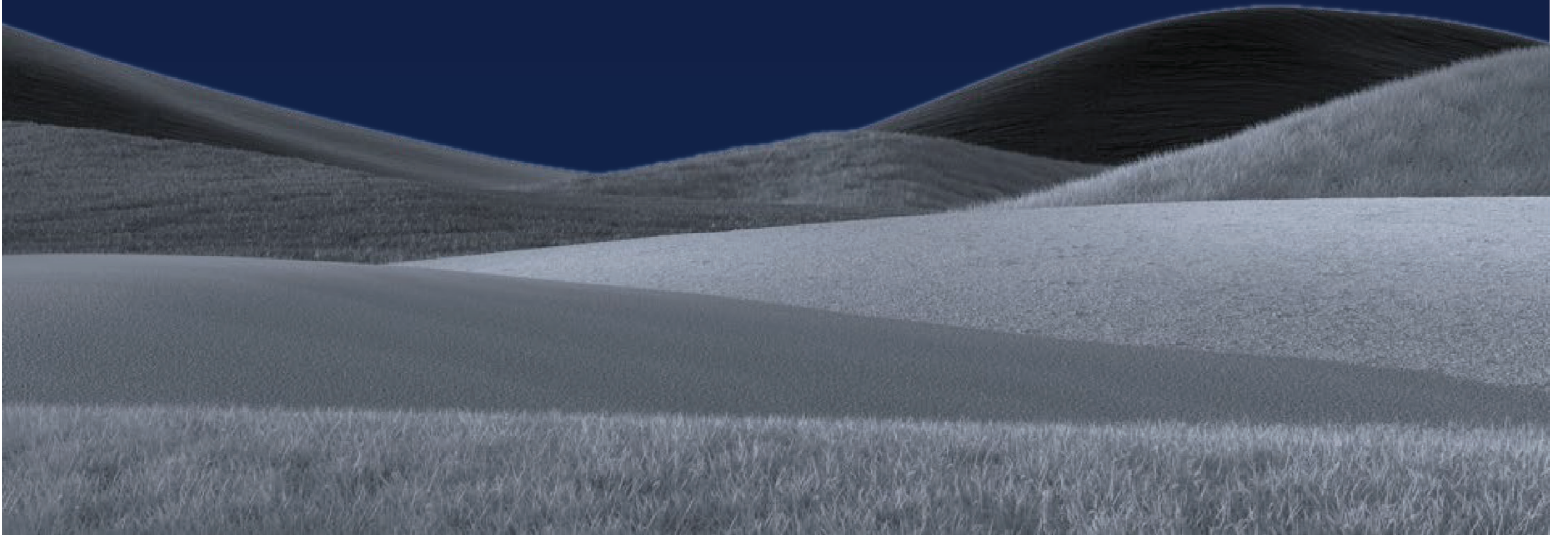
Internal quality control

Method validation

Instrument calibration and qualification

Standardised test methods

1993 – Modern history begins



1993 – Modern history begins

Measurement Uncertainty



0 AG – A new challenge

Measurement Uncertainty



The GUM - Measurement models and 'propagation of uncertainty'

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Mathematical form of uncertainty



$$y = f(x_1, x_2, \dots, x_n)$$

y measurement result
 x_i parameter affecting analytical result y

Sometimes called a
“**measurement model**” or
“**measurement equation**”



Mathematical form of uncertainty



$$y = f(x_1, x_2, \dots, x_n)$$

y measurement result
 x_i parameter affecting analytical result y

$$u(y) = \sqrt{\sum_i^n \left(\frac{\partial y}{\partial x_i} \right)^2 u(x_i)^2}$$

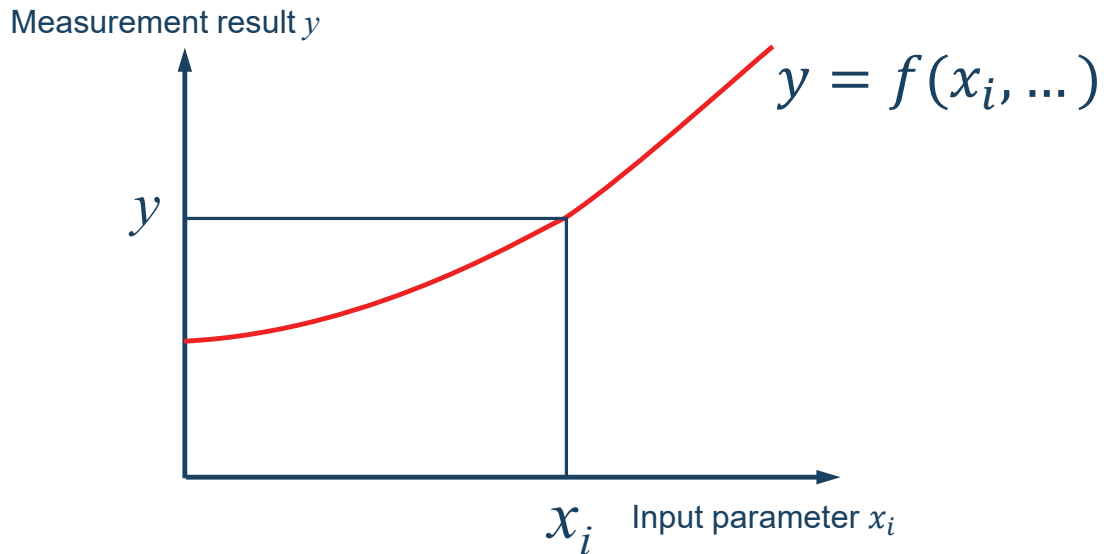
sensitivity
coefficient

$u(x_i)$ uncertainty in x_i
 $u_i(y)$ uncertainty in y due to uncertainty in x_i
 $\partial y / \partial x_i$ Partial differential – a gradient

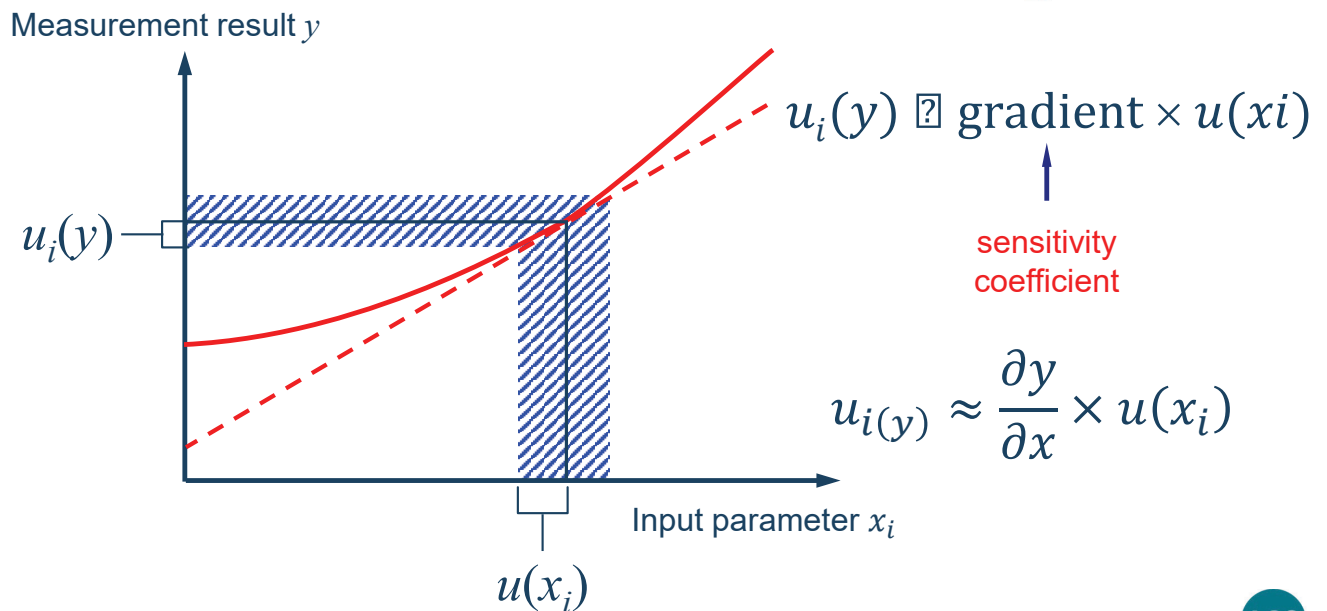
The “**law of propagation of uncertainty**”



Uncertainty propagation



Uncertainty propagation



Other key features of the GUM



- Adoption of INC-1 1980 Recommendations

- The uncertainty in the result of a measurement generally consists of several components which may be grouped into two categories according to the way in which their numerical value is estimated:
 - A. those which are evaluated by statistical methods,
 - B. those which are evaluated by other

No simple correspondence between categories A or B and ... “random” and “systematic”



Other key features of the GUM (cont)



- Adoption of INC-1 1980 Recommendations

- Components in category A are characterized by estimated variances and degrees of freedom
- The components in category B should be characterized by quantities u_j^2 **which may be considered as approximations to the corresponding variances**

Type A and Type B are treated in the same way

- The combined uncertainty should be characterized by the numerical value obtained by applying the usual method for the combination of variances.



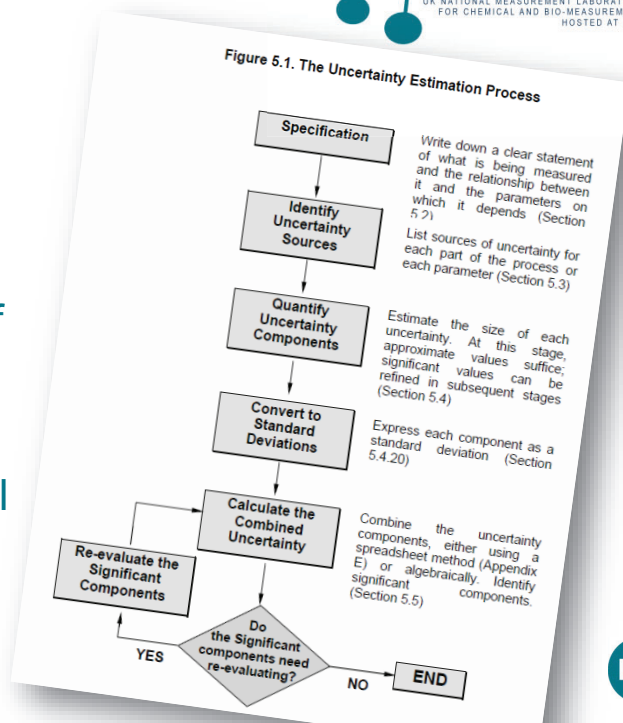
The first Eurachem guide



Quantifying Uncertainty in Analytical Measurement

Development of the first Eurachem guide

- Begun c. 1992-3 in the new MU working group of a young Eurachem
- Initial draft adopted the principles of the (then) draft ISO TAG 4 document
- Closely followed the 'law of propagation of uncertainty'
- Provided a process
- Provided worked examples from analytical chemistry
- Published in 1995
 - with intent to gather experience and review

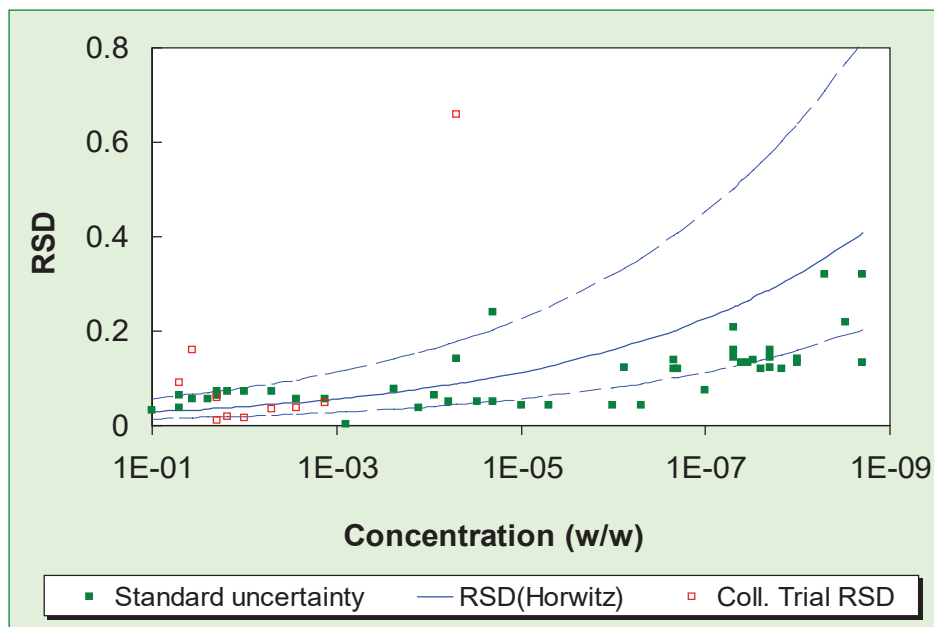


Emerging problems

Problems implementing the ISO Guide approach

- Difficult to write an equation that includes all influence factors
 - what about sample clean-up conditions, recovery of analyte from matrix, instrument conditions, interferences....
- Challenging to evaluate individual uncertainty components
- Process is too time consuming and unworkable in routine testing laboratories
 - a ‘reasonable estimation’ is required

Comparing u with s_R



Laboratory uncertainties using GUM tended to be smaller than reproducibility SD at lower concentrations

Additional problems

- Uncertainties dependent on level
 - No guidance on how to handle ‘top down’ uncertainties expressed as RSD
- Uncertainties near detection limits
 - Should results and uncertainties be reported below LOD?

Developments in MU evaluation 1995-2000

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Cause-and-effect analysis

Accred Qual Assur (1998) 3:101-105
© Springer-Verlag 1998

GENERAL PAPER

S. L. R. Ellison
V. J. Barwick

Estimating measurement uncertainty: reconciliation using a cause and effect approach

Received: 28 October 1997
Accepted: 17 November 1997
Presented at: 2nd EURACHEM
Workshop on Measurement Uncertainty
in Chemical Analysis, Berlin,
29-30 September 1997

Abstract A strategy is presented for applying existing data and planning necessary additional experiments for uncertainty estimation. The strategy has two stages: identifying and structuring the input effects, followed by an explicit reconciliation stage to assess the degree to which information available meets the requirement and thus identify factors requiring further

promotes consistent identification of important effects, and permits effective application of prior data with minimal risk of duplication or omission. The results of applying the methodology are discussed, with particular reference to the use of planned recovery and precision studies.

Accreditation
and Quality
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1998

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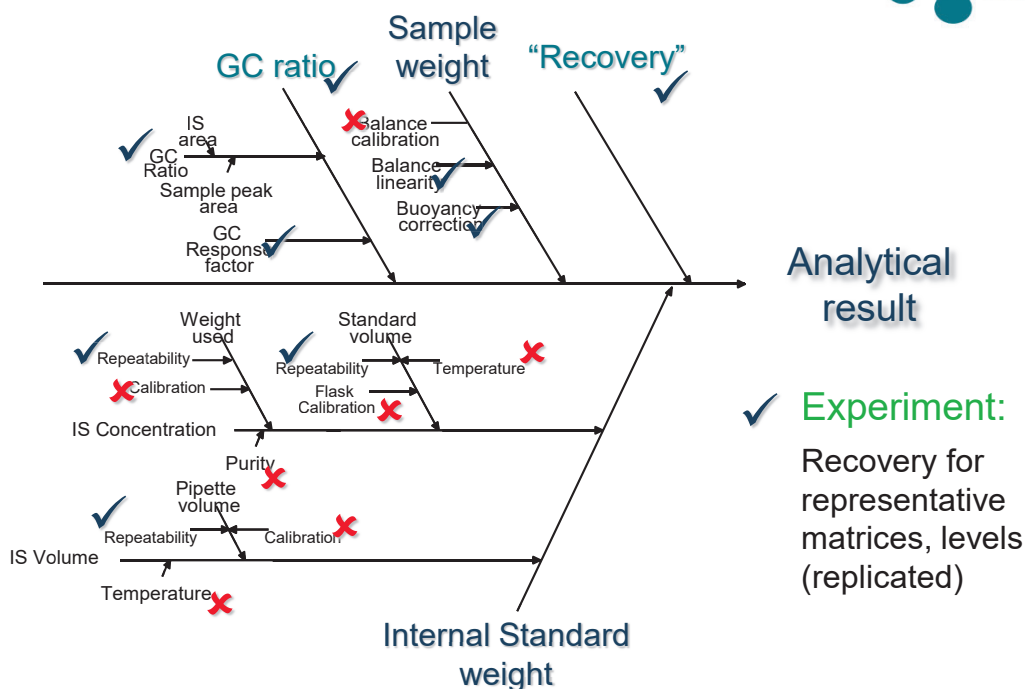
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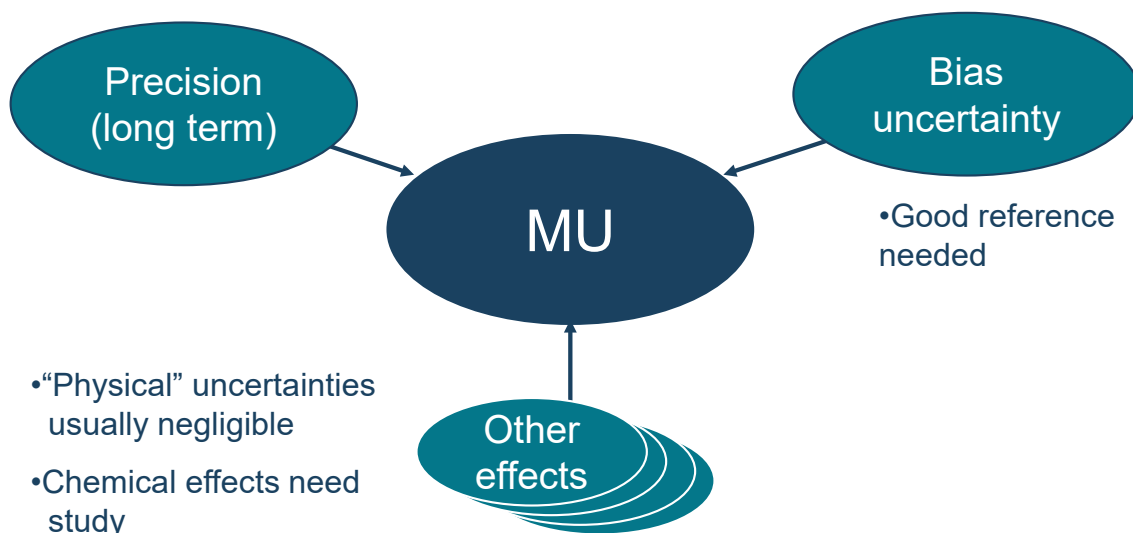


Cause-and-effect analysis

“Reconciliation” – what have we covered?



Top-down evaluation with additional effects



A simple spreadsheet method

Analyst, October 1994, Vol. 119

2161

Tutorial Review

Calculating Standard Deviations and Confidence Intervals with a Universally Applicable Spreadsheet Technique

J. Kragten

Laboratory of Analytical Chemistry, University of Amsterdam, Nieuwe Achtergracht 166, 1018 WV Amsterdam, The Netherlands

A quick and universally applicable spreadsheet method is outlined for the calculation of standard deviations based on the general formula for error propagation:

$$s_R^2 = \left(\frac{\partial R}{\partial x}\right)^2 s_x^2 + \left(\frac{\partial R}{\partial y}\right)^2 s_y^2 + \left(\frac{\partial R}{\partial z}\right)^2 s_z^2 + \dots$$

With this method, standard deviations are calculated numerically without violating the condition of mutual independence, with a substantial time gain and with no risk of calculating errors. Satterthwaite's approximation of the degrees of freedom is a logical extension of the technique with which confidence intervals can be easily established. Direct insight is obtained about the separate contributions of the different error sources.

of x, y, \dots , the simple rules lead to erroneous results. This will be shown with the calculation of the surface of a block: $R = 2(lb + bh + hl)$. Most workers will split R into the parts lb , bh and hl . The rules are applied to these separate parts and the standard deviations of these separate parts are obtained. Eventually the separate parts are summed to obtain R and the simple error propagation rules are applied again to find s_R . At this point the error is made: commonly the separate parts of R have some variables in common and hence are mutually dependent. (Use of the word *correlation* is restricted to covariance between measured quantities. Terms containing the same variable in a mathematical relationship will be called *dependent*.) The block-surface $R = 2(lb + bh + hl)$ is a good example with the product terms lb , bh and hl sharing b , h and

Kragten,
Analyst
1994

Kragten's method Spreadsheet implementation



	x	u						
m	100	2		102	100	100	100	100
V_T	100	0.1		100	100.1	100	100	100
α	0.001			0.001	0.001	0.001	0.001	0.001
T	25	1.15		25	25	25	26.15	25
T_0	25			25	25	25	25	25

$x+u(x)$

C	1	0.020		1.02	0.999001	1	1.00115	1	Recalculation
				0.02	-0.001	0	0.00115	0	Differences
				0.0004	9.98E-07	0	1.32E-06	0	Diff ²

Combined
uncertainty

Details in QUAM



Quam:2000



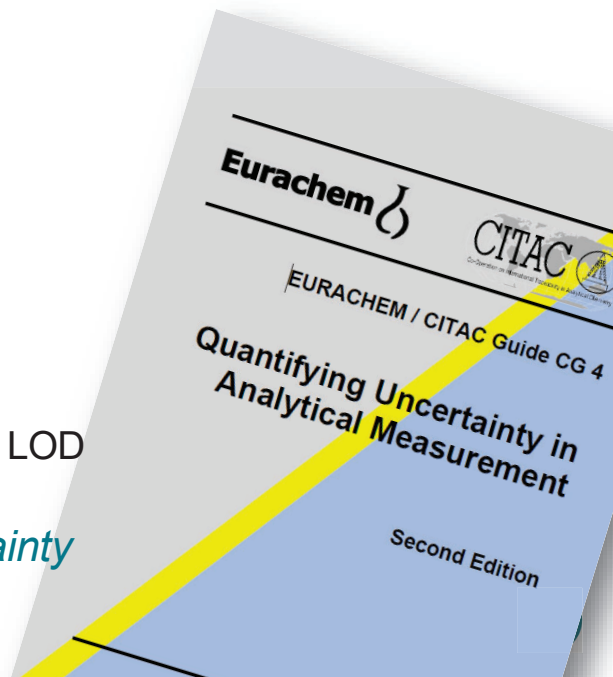
Quantifying Uncertainty in Analytical Measurement

Second Edition

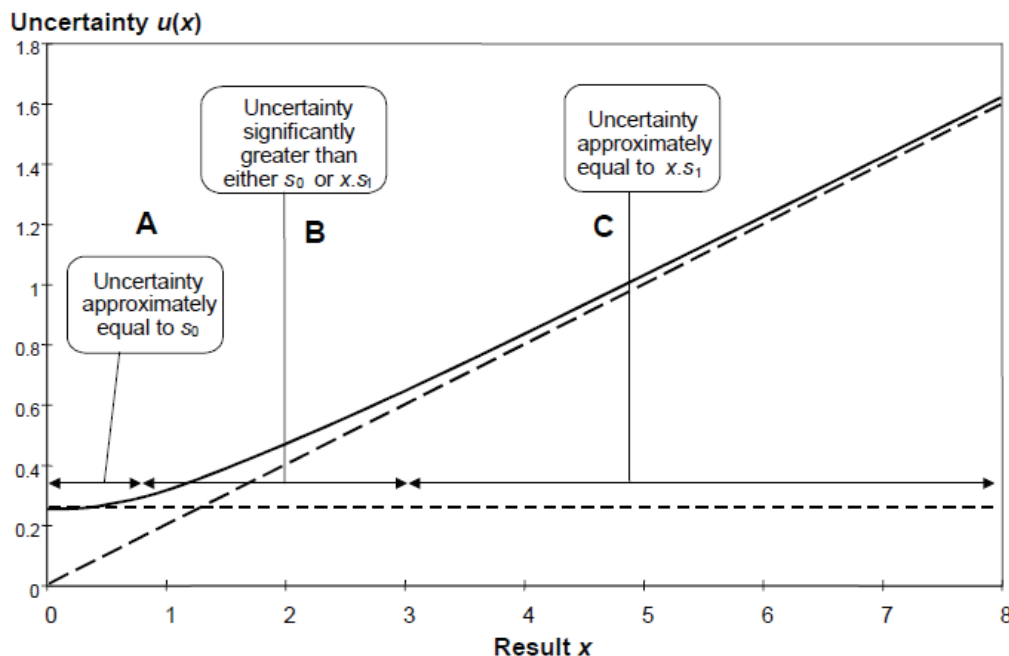
QUAM Second edition

- Based on QUAM:1995
- Included
 - Clear guidance permitting use of validation data
 - Guidance on cause-and-effect analysis
 - Guidance on Kragten's method
 - All examples used cause-and-effect analysis and Kragten calculation
- Basic guidance on results and uncertainties near LOD

“The ideal is accordingly to report valid observations and their associated uncertainty regardless of the values.”



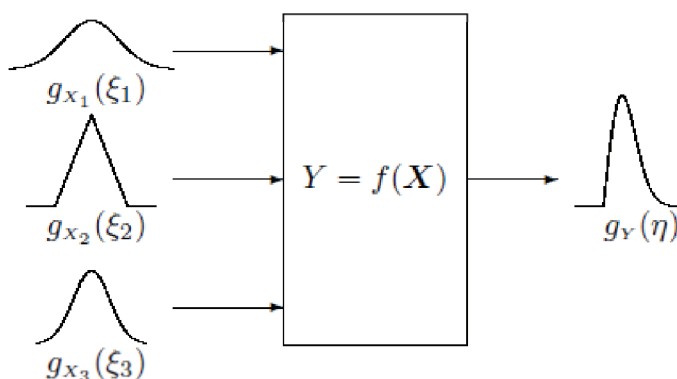
Dealing with uncertainties dependent on level



2000 –2012: Further developments

GUM Supplement 1 (JCGM 101)

- *Evaluation of measurement data — Supplement 1 to the “Guide to the expression of uncertainty in measurement” — Propagation of distributions using a **Monte Carlo method***



*Illustrations from
JCGM:101, Figures
2 & 3*

New approaches to uncertainty near LOD



PAPER
www.rsc.org/analyst | The Analyst
Reporting measurement uncertainty and coverage intervals near natural limits

Simon Cowen and Stephen L. R. Ellison

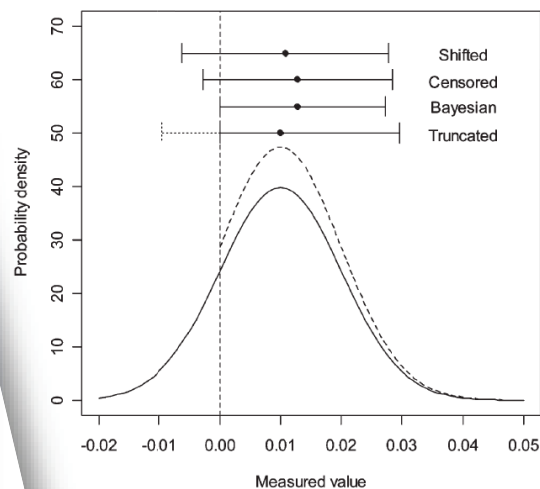
Received 21st December 2005, Accepted 24th April 2006
First published as an Advance Article on the web 11th May 2006
DOI: 10.1039/b518084h

Different methods of treating data which lie close to a natural limit in a feasible range, such as zero or 100% mass or mole fraction, are discussed and recommendations made concerning the most appropriate. The methods considered include discarding observations beyond the limit, shifting observations to the limit if outside the feasible range, truncation and renormalisation of an assumed normal measurement distribution and a uniform prior within the feasible range. Based on consideration of bias and simulation to assess coverage, it is recommended that for most purposes, a confidence interval near a natural limit should be constructed by first calculating the usual confidence interval based on Student's t , then truncating the out-of-range portion to leave an asymmetric interval and adjusting the reported value to within the resulting interval if required. It is suggested that the original standard uncertainty is retained for uncertainty propagation purposes.

Introduction

Analytical measurement is concerned with the quantitation of some or all of the constituents of a sample, together with an assessment of the uncertainty associated with the result. The principles of measurement uncertainty are elaborated in the Guide to the Expression of Uncertainty in Measurement (GUM) and are the basis of the ISO 9000 and ISO 14000 standards.

Establishment of the detection limit for an instrument and method. Assuming that systematic errors are absent or have been accounted for in the instrument or method being used, the remaining uncertainty arises from effects which can be assumed to have random contributions to the observed result. For example, the variability in a set of replicates is the random error due to instrument (or method) repeatability, and the actual deviation from the value of the measurand has (within a specified uncertainty) unknown sign and magnitude. Such uncertainty arises from effects other than sample

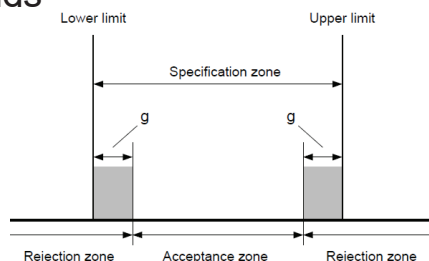


DOI: 10.1039/b518084h



New detailed guidance on conformity assessment

- Published as supplementary guidance
- Introduced 'new' ideas
 - Decision rules
 - Guard bands



- Decisions under relative uncertainty



Eurachem

CITAC
Co-Operation in International Technology & Analytical Chemistry

EURACHEM / CITAC Guide
Use of uncertainty information
in compliance assessment

First Edition 2007

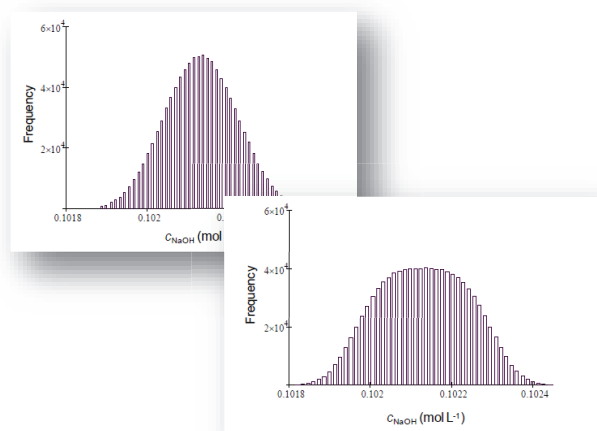


Quantifying Uncertainty in Analytical Measurement

Third Edition

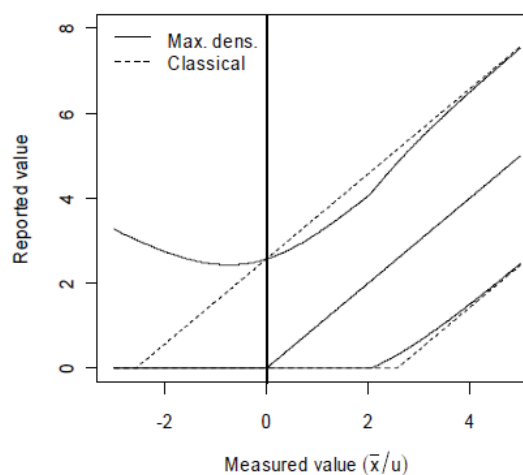
Additional material

Monte Carlo examples



Uncertainties near zero

- Bayesian solution



2013 – 2025: Evolution continues

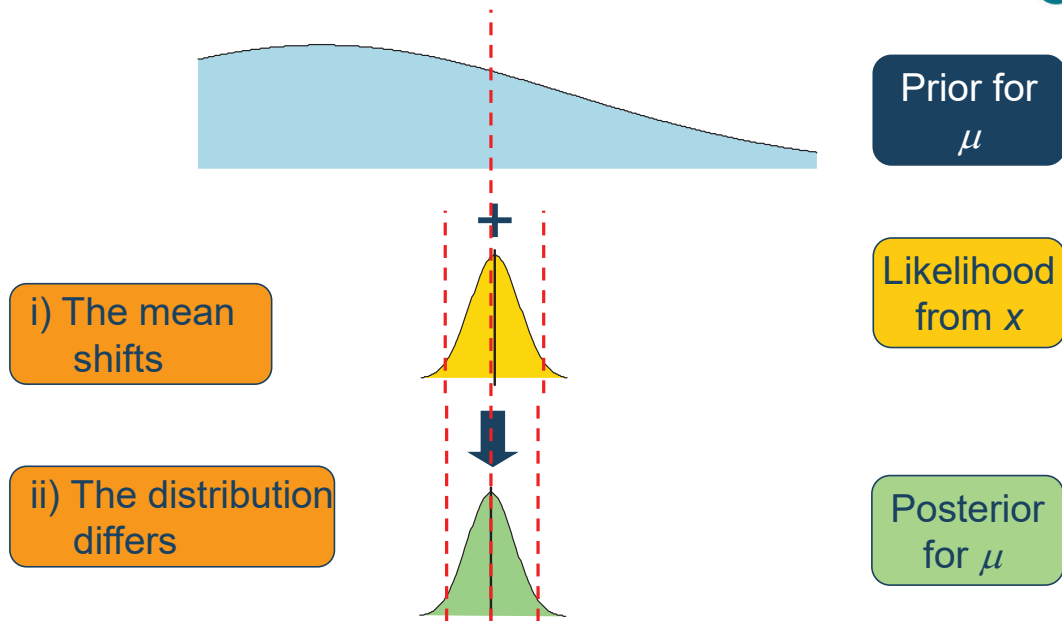
LGC

The Uncertainty Factor

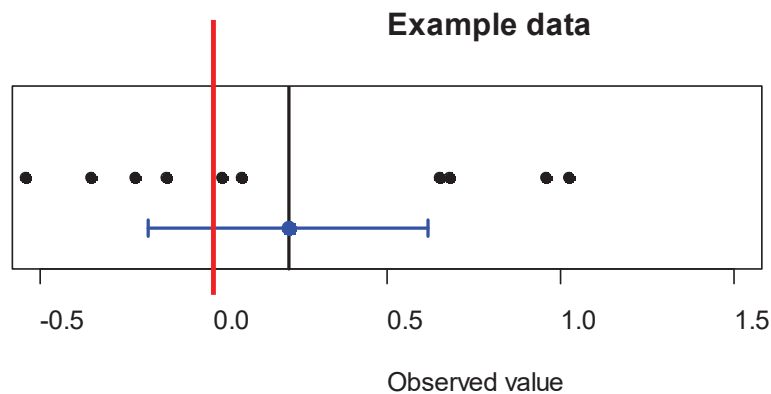
- Introduced in the Eurachem guide on Uncertainty from Sampling (2nd Edition)
- Gives an asymmetric interval
- Useful for large relative uncertainty with approximately lognormally distributed uncertainty



Bayes applied to Measurement Uncertainty



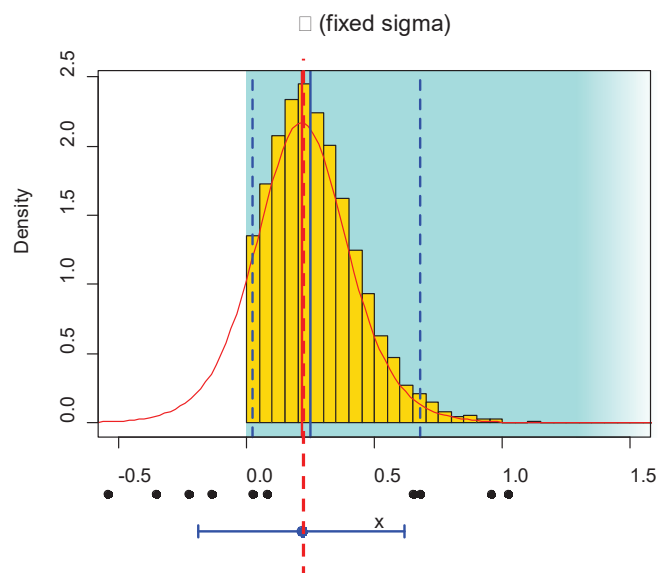
MCMC example 2: Constant RSD - SD proportional to μ



- **Concentration: not below zero**
- **Common observation: standard deviation proportional to true value**

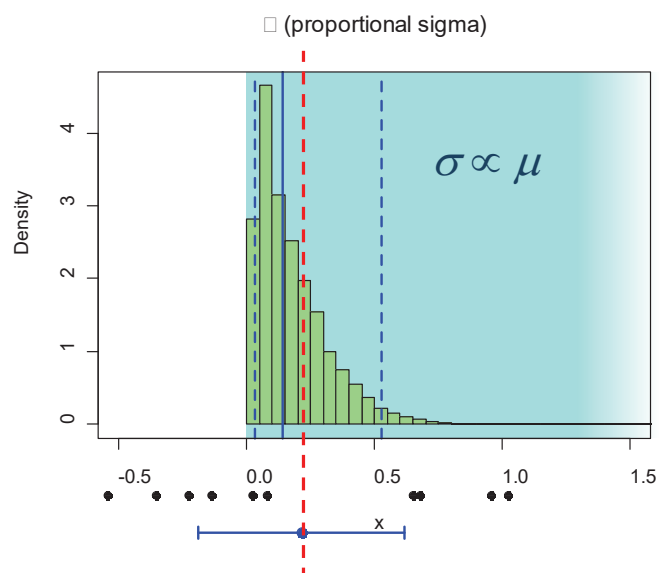
MCMC results:

i) Fixed standard deviation



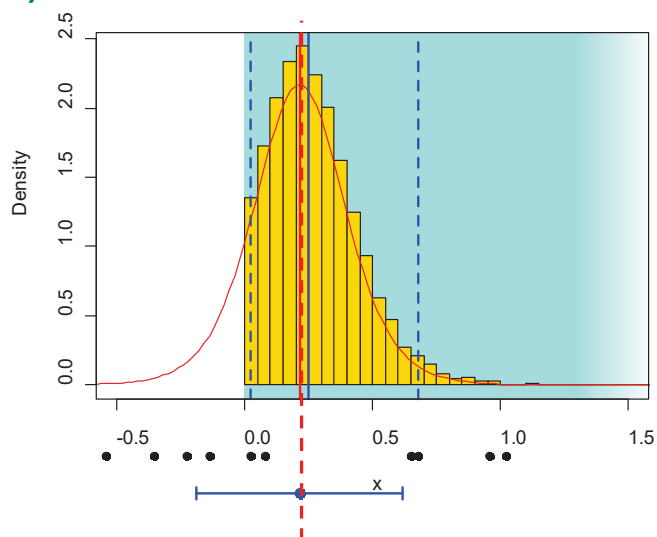
MCMC results

ii) Proportional standard deviation

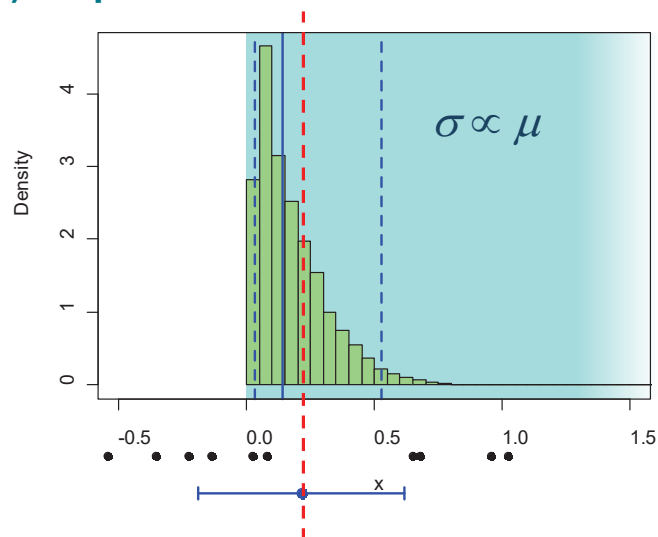


MCMC results:

i) Fixed standard deviation

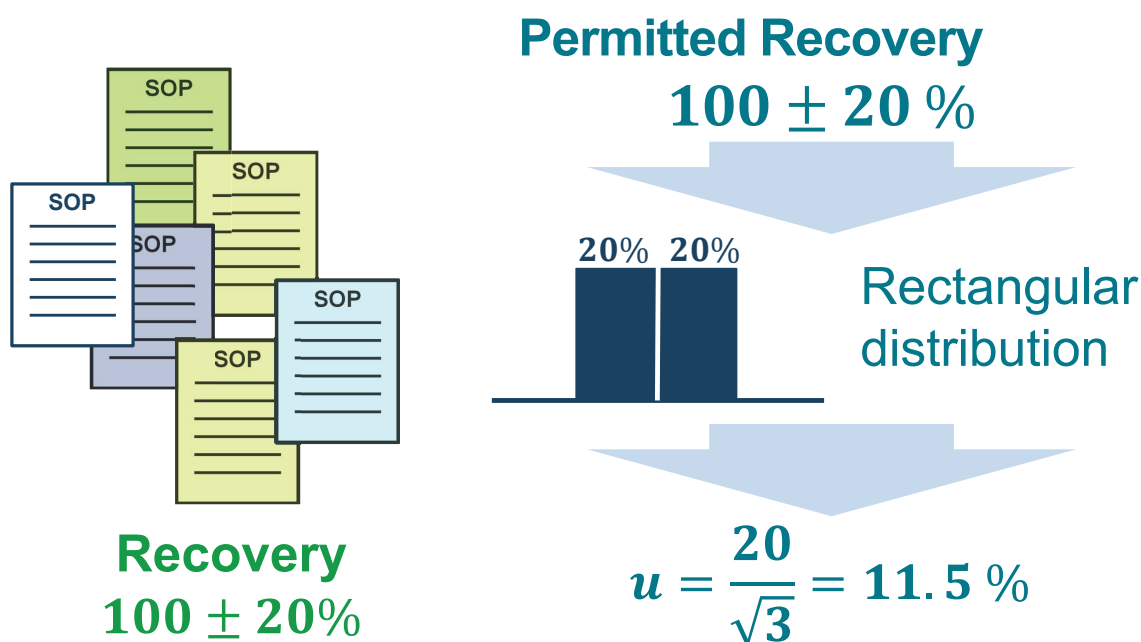


ii) Proportional standard deviation



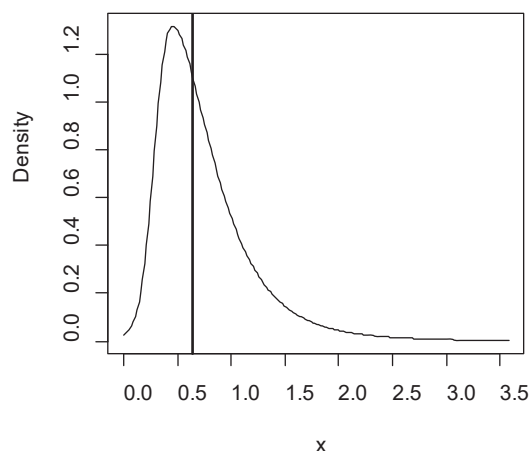
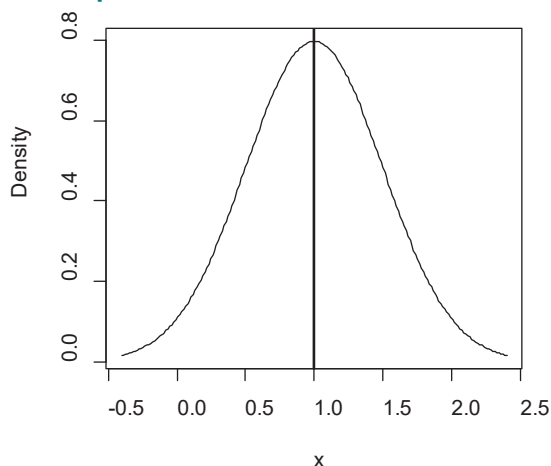
Outstanding problems

Allowable limits and measurement uncertainty – A problem?



Asymmetry – What does it mean for conformity assessment?

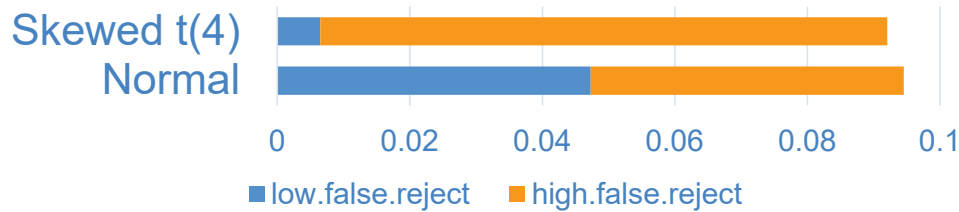
- Real processes can have different distributions



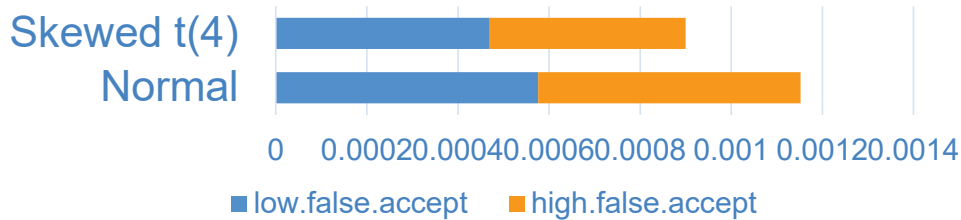
- Does this affect conformity assessment using MU?

Asymmetry – What does it mean for conformity assessment?

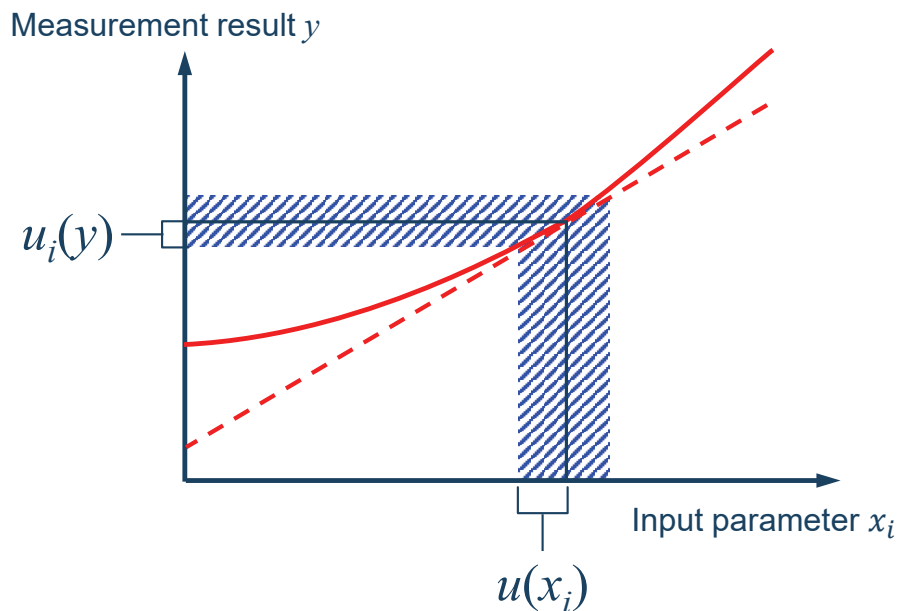
Producer risks



Consumer risks



Linearity – How linear is ‘linear enough’?



The Future



Future MU guidance from Eurachem



- New guidance on MU from validation data
 - Current guide gives general guidance
 - Additional 'how to' guidance needed
 - Draft guide now out for consultation
- Additional guidance in QUAM
 - Uncertainty factors
 - Asymmetry – cautionary guidance
 - Non-linearity – cautionary guidance
 - Bayesian methods ??
 - Effect of permitted limits on MU – Supplementary guidance ?

Conclusion

Measurement Uncertainty



Conclusion

Measurement Uncertainty

- Not the mountain it once was
- ... But work still to do

