

Evolution of Philosophy and Description of Measurement (Preliminary Rationale for VIM3)

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Abstract

Different approaches to the philosophy and description of measurement have evolved over time, and are still evolving. There is not always a clear demarcation between approaches, but rather a blending of concepts and terminologies from one approach to another. This sometimes causes confusion when trying to ascertain the objective of measurement in the different approaches, since the same term may be used to describe different concepts in the different approaches. Important examples include the concepts and terms “value,” “true value,” “error,” “probability” and “uncertainty.” Constructing a single vocabulary of metrology that is able to unambiguously encompass and harmonize all of the approaches is therefore difficult, if not impossible. This paper examines the evolution of common philosophies and ways of describing measurement, highlighting some of the differences and providing some of the rationale for the entries and structure of the March 2006 draft of the 3rd Edition of the *International Vocabulary of Metrology, Basic and General Concepts and Associated Terms*, or VIM3.

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Introduction

The concept of measurement covers a wide range of activities and purposes. Different approaches to describing and characterizing measurement have been developed and have evolved to address the various types and uses of measurements, and are still evolving. Many terms have been used over time in the context of describing measurement, and the evolution of the different approaches to measurement has led to sometimes subtle but undoubtedly different uses of some terms.

A “vocabulary” is defined (e.g., ISO 1087-1) as “a terminological dictionary that contains designations and definitions from one or more specific subject fields.” Ideally, every term in a vocabulary should designate only one concept, in order to minimize confusion. However, because of the different concepts that are sometimes associated with the same term in the different approaches to measurement, it is virtually impossible to create a vocabulary of measurement that designates only one concept with each term in the vocabulary. This is a major difficulty that has been encountered in developing the 3rd Edition of the *International Vocabulary of Metrology, Basic and General Concepts and Associated Terms* (VIM3), where “metrology” is defined as “field of knowledge concerned with measurement”.

This paper examines the evolution of the more common approaches to describing measurement, highlighting some of the differences in the use of terms and providing some of the rationale for how several of the terms are likely to be treated in the final version of VIM3.

Common Elements of All Approaches to Measurement

There are a few fundamental concepts in most, if not all, approaches to describing measurement. Probably the most fundamental concept pertains to the kinds of things that can be measured (quantities). Another fundamental concept is the means used to express the magnitude of that which has been measured (in terms of values). Just as fundamental is the concept of measurement itself. The following definitions are taken from the March, 2006, draft of the VIM3:

A **quantity** (1.1) is a “property of a phenomenon, body or substance to which a number can be assigned with respect to a reference” (which allows comparison with other quantities of the same kind). A value of a quantity (**quantity value**, 1.10) is a “number and reference together expressing magnitude of a quantity”. **Measurement** (2.1) is the “process of obtaining one or more quantity values that can reasonably be attributed to a quantity”, usually through some type of experiment.

In VIM3 the term **measurand** (2.3) is defined as “quantity intended to be measured.” This term has ‘evolved’ from the definition in the *International Vocabulary of Basic and General Terms in Metrology*, 2nd Edition [1], VIM2, which is “particular quantity subject to measurement,” that could be different than the quantity intended to be measured. This

distinction must be kept in mind when considering the objective of measurement in the different approaches, and will be discussed further later on.

Figure 1 demonstrates some simple common elements of all approaches to describing measurement. The rectangular box represents the measurand, and the horizontal scale represents the entire set of values that could possibly be attributed to that type of measurand. Note that there is no measurement unit associated with the horizontal line, because the quantity is an **ordinal quantity** (1.27), which is a “quantity, defined by a conventional measurement procedure, for which a total ordering relation with other quantities of the same kind is defined, but for which no algebraic operations among those quantities are defined.” Also indicated in Figure 1 is a value (y) being attributed to the measurand on the basis of a set of replicated measurements, illustrated schematically by a histogram.

For those quantities where there are meaningful algebraic operations among the quantities, a **measurement unit** (1.9) can be defined, which is a “scalar quantity, defined and adopted by convention, with which any other quantity of the same kind can be compared to express the ratio of the two quantities as a number.” This is indicated in Figure 2, where the measurement unit is the reference to be associated with the numerical value in the measured quantity value. The concept of a measurement unit is common to all approaches to describing measurement (for other than ordinal quantities). The bell curve in the figure illustrates a ‘gaussian’ fit to the histogram data. The curve is dashed to indicate that replicate measurements are not always performed in a measurement (that is, sometimes only a single measurement is performed), as will be elaborated below in the discussion of the International Electrotechnical Commission (IEC) Approach.

The two main approaches to describing measurement that will be discussed in this paper are sometimes called the ‘classical’ approach and the ‘uncertainty’ approach. Within each of these approaches are sub-approaches. While the two main approaches are given discrete names, there has in actuality been an evolution of these approaches that makes it difficult to ascribe certain concepts to one approach or another. This evolution of concepts will be discussed below. Also, since probability and statistics usually play an important role in most aspects of measurement evaluation, both the ‘frequentist’ and ‘Bayesian’ theories of inference as used in measurement will be discussed, as appropriate.

Classical Approach (CA) to Measurement

It is generally accepted that the key distinguishing premise of the classical approach to measurement is that, for a specified measurand, there exists a unique value, called the **true value** (1.11), that is consistent with the definition of the measurand. This is shown schematically in Figure 3, where it is indicated that, in the general case, the value being attributed to the measurand based on measurement is different from the true value. This difference could be due to a variety of reasons, including mistakes in formulating the

measurement model (such as not taking into consideration all significant factors and influences), and blunders in carrying out the measurement procedure.

Another premise of the classical approach is that it is possible to determine the true value of a measurand through measurement, at least in principle, if a ‘perfect’ measurement were performed. The objective of measurement in the classical approach is then usually considered to be to determine the true value of the measurand as ‘closely’ as possible, or at least as closely as necessary, by eliminating or correcting for all (known) systematic effects and mistakes, and also by performing enough repeated measurements to adequately minimize effects and mistakes due to random causes.

In the classical approach it is recognized that it is not possible to perform a perfect measurement and so there will be “errors”, both systematic and random, in the value ultimately being attributed to the measurand based on measurement. This value is frequently referred to as the ‘measurement result’ in the classical approach, and in other approaches as well. Figure 4 illustrates the concept of an individual measurement error, defined in the classical approach as the difference between an individual measurement result and the true value. The individual measurement result (denoted by y_i in the figure) is illustrated with respect to the bell-curve, which is now solid to indicate that multiple individual measurements are being considered. Also indicated in Figure 4 are “systematic error,” defined as the difference between the unknown mean of the uncorrected measurement result and the true value, and “random error,” defined as the difference between an individual measurement result and the unknown mean of the uncorrected measurement result. Note that the “mean of the uncorrected measurement result” here is meant to be that of an infinite distribution, and so cannot be known exactly. This is illustrated schematically in Figure 5, where two ‘systematic errors’ are shown, the lower one (systematic error_b) with respect to the average of the histogram data, and the upper one (systematic error_a) with respect to the mean of the theoretical frequency distribution for an ‘infinite’ set of data. The bell curve of the theoretical frequency distribution is dashed to indicate that it is not knowable. The systematic error_a line is also dashed to indicate that its length cannot be known, since the mean of the theoretical frequency distribution cannot be known. The question of whether or not the length of the systematic error_b line can be known, as well as the lengths of the three ‘error lines’ in Figure 4, will be discussed next.

Knowable Error?

Two important and related questions that arise in the classical approach are, first, whether it is possible, in principle, to go about identifying and eliminating, or correcting for, all of the errors in a measurement, and, second, if so, how? One possible way of addressing these questions is to hypothesize that it is possible, at least in principle, to determine the true value by carrying out a very large number of different types of measurements of the same measurand, using different measurement procedures, measurement methods or even measurement principles, a large number of times (so that various systematic errors will ‘average out’). This would require that a lot of information be obtained through measurement (which may not always be practical, even if the philosophy is sound).

Figure 6 illustrates this idea for just two different measurement principles, and Figure 7 is meant to illustrate the advantage of using multiple measurement principles (indicated by the four different curves). Using this idea in the classical approach, a probability is usually assessed that the true value lies within a stated interval, as could be characterized by the ‘widths’ of the large bell-shaped curves associated with the true value in both of the Figures 6 and 7. Since this idea requires that an essentially infinite amount of information be obtained in order to know the true value, it is recognized that, in practice, a true value can never be known exactly using this idea. This is represented schematically in Figure 7, where \bar{y} represents the average of the averages of the four curves.

The questions then remain whether it is possible, in principle, in a different way, to identify and correct for all of the errors in a measurement, and, if so, how?

Error Analysis, Frequentist Theory in CA

One different way of trying to answer these questions is through the application of error analysis, which is based on the frequentist theory of inference as used in measurement. Error analysis is the attempt to estimate the total error using frequency-based statistics. However, the systematic error cannot be estimated in a statistical way, since it is neither observable nor behaves randomly in a measurement series under repeatability conditions. Therefore error analysis, which includes statistical and nonstatistical procedures, leads to inconsistencies in data analysis, especially in error propagation.

Bayesian Theory in CA

Another way of trying to answer these questions is to apply the Bayesian theory of inference to data analysis. Here systematic and random errors are treated on the same probabilistic basis, where probability is no longer understood as a relative frequency of the occurrence of events, but as an information-based degree of belief about the truth of a proposition, for example, about the true value. Using the Bayesian theory, it is still not possible to determine a true value unless an essentially infinite amount of information is obtained, so that it is again recognized that, in practice, a true value cannot be known.

Difficulties with the CA

So far no satisfactory way has been found to identify, let alone correct for, all of the errors in a measurement. The implications are significant, as illustrated in Figure 8, where a hypothetical three ‘known’ components of systematic error are shown (usually estimated as ‘worst-cases’). Since it is virtually impossible to know for sure if there is another component (say, due to a blunder, as indicated by the dashed line), the ‘total’ systematic error is unknown, as also indicated by a dashed line. If the total systematic error is unknown, then the true value cannot be known. If the true value is not known, then the error cannot be known (as again indicated by a dashed line). The random error, when defined with respect to the average of the histogram data, is calculable, as indicated by the solid line in the figure. However, when random error is defined with respect to the

mean of the theoretical frequency distribution, it also becomes unknowable, as illustrated by the dashed line for 'random error' in Figure 9.

Systematic and random errors can therefore typically only be estimated or guessed. No generally-accepted means for combining them into an 'overall error' exists that would provide some overall indication of how well it is thought that a measurement result corresponds to the true value of the measurand (i.e., to give some indication of how 'accurate' the measurement result is thought to be, or how 'close' the measurement result is thought to be to the true value of the measurand). The difficulty in the CA, of the lack of a generally-accepted, good procedure for describing the perceived 'quality' of the measurement result is one important reason that 'modern' metrology is moving away from the philosophy and language of the CA. A solution to this difficulty is addressed in the uncertainty approach (UA) to measurement (as will be described shortly). There are also other reasons, but they will not be discussed here.

VIM3 RATIONALE: There are many measurement situations, typically of a relatively simple nature, where it is likely possible to be able to identify and correct for all of the significant systematic errors, as well as to obtain a sufficient number of replicate measurements for the purpose, such that description of the measurement result using the language and philosophy of the classical approach is a seemingly reasonable thing to do, and many people still do it. This is one of the main reasons that it was decided to keep many of the terms and concepts from the classical approach in the main body of VIM3, and not relegate them to an Annex. Another reason, as mentioned earlier and that will be elaborated further below, is that there is not always a clear demarcation between approaches. As an example, it is not clear to which measurement approach to ascribe the premise of a lack of uniqueness of a true value of a measurand.

Uniqueness(?) of True Value

Obviously no measurand can be completely specified, meaning that there will always be a set of true values that are consistent with the definition of the measurand. The important question is whether the range (defined as the difference between the upper and lower limit) of the set of true values is significant when compared with the range of measured quantity values. This is illustrated schematically in Figure 10, where the interval of the set of true values consistent with the definition of the measurand is indicated by a pair of vertical dotted lines. The corresponding range is shown bracketing the measured quantity value to indicate that, even if a series of replicated, 'perfect' measurements of the measurand were possible, there would still be a set of measured values having that same range. The dotted bell curve illustrates a situation where the range of the set of actual measured quantity values is broader than the range of the set of true values.

It is in general desirable to have a measurement situation where the measurand can be progressively better defined such that the range of the set of true values is relatively insignificant with respect to the range of measured quantity values that can be obtained when using the (best) available measuring system, as illustrated in Figure 11. Under these conditions the measurand can be regarded as having an 'essentially unique' true value

(i.e., ‘the’ true value), and the ‘customary’ language and mathematics of measurement can be employed.

However, this situation is not always found, either because the measurand cannot, or needs not, be specified very specifically, or because the measurement system is so ‘good’ that it is always capable of producing measured quantity values within the range of true values for that type of measurand, as illustrated in Figure 12. Under these conditions it is necessary to think differently about the way of describing measurement, irrespective of the measurement approach. For example, in the classical approach it would no longer be possible to talk about ‘the true value’ of a measurand, or ‘the systematic error’ associated with a measurement result, since such unique values would no longer have meaning. This measurement situation will also be addressed further in the discussion about the uncertainty approach.

Before leaving the discussion of the classical approach, it is worth noting that the classical approach is also sometimes called the ‘traditional approach’ or ‘true value approach.’ However, the latter is a misnomer, since the concept of true value is actually also used in ‘modern’ approaches, such as the ‘uncertainty approach,’ as will be discussed next.

Uncertainty Approach (UA) to Measurement

The concept of measurement uncertainty had its beginnings in addressing the difficulties described above with the CA, namely the questions of 1) whether it is possible, both in principle and in practice, to know the true value and error, 2) whether or not the true value is unique, and 3) how to combine information about random error and systematic error in a generally accepted way that gives information about the overall perceived ‘quality’ of the measurement. Further, if the true value, or set of true values, is not knowable in principle, then the questions arise of whether the concept of true value is necessary, useful or even harmful! All of these issues and perspectives will be addressed below.

While different approaches exist within the UA, the two most prominent approaches are those put forward in the *Guide to the Expression of Uncertainty in Measurement* (GUM, 1993 and 1995) [2] and in IEC 60359 (*Electrical and Electronic Measurement Equipment – Expression of Performance*) [3]. The IEC approach is parallel and complementary to the GUM, but uses a more operational or pragmatic philosophy, focusing primarily on single measurements made with measuring instruments. Both of these approaches, along with their impact on VIM3, will be described.

GUM Approach to UA

The GUM approach to the UA provides a more refined means than the CA for describing the perceived quality of a measurement. One of the main premises of the GUM approach

is that it is possible to characterize the quality of a measurement by accounting for both random and systematic ‘effects’ on an equal footing, and a means for doing this is provided. Another basic premise of the GUM approach is that it is not possible to know the true value of a measurand (GUM 3.3.1): “The result of a measurement after correction for recognized systematic effects is still only an *estimate* of the value of the measurand because of the uncertainty arising from random effects and from imperfect correction of the result for systematic effects.” A third basic premise of the GUM approach is that it is not possible to know the error of a measurement result (GUM 3.2.1, Note): “Error is an idealized concept and errors cannot be known exactly.”

In the GUM approach it is explicitly recognized that it is not possible to know, for sure, how ‘close’ a value obtained through measurement is to the true value of a measurand (i.e., to know the error). Instead a methodology for constructing a quantity, called the standard measurement uncertainty, is established that can be used to characterize a set of values that are thought, on a probabilistic basis, to correspond to the true value, based on the information obtained from the measurement. *The objective of measurement in the GUM approach then becomes to establish a probability density function, usually Gaussian (normal) in shape, that can be used to calculate probabilities, based on the belief that no mistakes have been made, that various values obtained through measurement actually correspond to the ‘essentially unique’ (true) value of the measurand.* Note that the GUM does not explicitly state the objective of measurement this way, but it can be inferred through its description of standard uncertainty (see, e.g., GUM 6.1.2). *Another way of viewing the objective of measurement in the GUM approach is that it is to establish an interval within which the ‘essentially unique’ (true) value of the measurand is thought to lie, with a given level of confidence (probability as degree of belief), based on the information used from the measurement.* The term “true” has been put in parenthesis here as an alert that the GUM discourages use of the term (but not of the concept) “true value,” and instead treats “true value” and “value” as equivalent, and thus omits the term “true”. This, however, causes terminological difficulties that are treated in VIM3, and are discussed below.

VIM3 RATIONALE (measurement uncertainty): The concept of measurement uncertainty is defined in VIM3 (2.12) as “parameter characterizing the dispersion of the quantity values being attributed to a measurand, based on the information used.” As stated above, this important concept is introduced in the UA to provide a quantitative means of combining information arising from both random and systematic effects (if they can be distinguished at all!) in measurement into a single parameter that can be used to characterize the dispersion of the values being attributed to a measurand, based on the information used from the measurement. The VIM3 definition is modified from the VIM2 [1] (and GUM [2]) definition because of the way that the term “measurement result” has been redefined in VIM3 (see next rationale below).

VIM3 RATIONALE (measurement result): The GUM uses the VIM2 definition of “measurement result” (value attributed to a measurand, obtained by measurement), which is the same as the *estimate* mentioned above. However, it was decided by the developers of VIM3 to emphasize the importance of including measurement uncertainty in reporting

the outcome of a measurement by incorporating into the definition of measurement result the notion that “a complete statement of a measurement result includes information about the uncertainty of measurement,” as stated in Note 2 of the VIM2 definition of measurement result. Accordingly, measurement result is defined in VIM3 as “set of quantity values being attributed to a measurand together with any other available relevant information,” which implicitly includes information not about just a single value, but rather about the measurement uncertainty as well. The “other available relevant information,” when available, pertains to being able to state probabilities.

VIM3 RATIONALE (measured value): Since the term “measurement result” is defined in VIM3 in the more general sense given above, it was decided to introduce a separate term in VIM3 that could be associated with the concept of each of the individual values of the set of values being attributed to the measurand based on measurement. Any individual value representing (or that belongs to) the measurement result is called a **measured quantity value** (2.9) in VIM3.

VIM3 RATIONALE (definitional uncertainty): Another basic premise of the GUM approach is that no measurand can be completely specified, as has already been discussed earlier in the context of lack of uniqueness of the true value. In the GUM approach this premise is implemented such that, at some level, there is always an ‘intrinsic’ uncertainty that is the minimum uncertainty with which an incompletely defined measurand can be determined (GUM D.3.4). An explicit term covering this concept (**definitional uncertainty**, 2.13) has been introduced into VIM3. The implication of this concept, as discussed above, is that there is no single true value for an incompletely defined measurand. However, a very important point to remember concerning the GUM approach is that it “is primarily concerned with the expression of uncertainty in the measurement of a well-defined physical quantity – the measurand – that can be characterized by an essentially unique value.” (GUM 1.2) ‘Essentially unique’ means that the definitional uncertainty can be regarded as negligible when compared with the range of the interval given by the measurement uncertainty. Therefore, when using the GUM ‘mathematical machinery’ and language, it is important to make sure that this ‘negligibility’ condition applies. If it does not, then use of different approximations and language might be required. This will be elaborated further below.

VIM3 RATIONALE (value, true value): As already noted, in the GUM approach the term “true” in “true value” is considered to be redundant (GUM D.3.5), and so a “true value” is just called a “value”. *It is important to recognize that this does not mean that the concept of true value is discouraged or ignored in the GUM.* Rather, the concept of “true value” (defined in VIM3 as “value consistent with the definition of a quantity”) has only been renamed “value”, or “the value,” in the GUM. This sometimes causes serious confusion, especially since the same term “value” is also frequently used in the GUM in the more general, super-ordinate sense of “number and reference together expressing magnitude of a quantity.” Another reason for potential confusion is that, if true value is unknowable, then the need for the concept can be questioned (this will also be discussed later in connection with the IEC approach). However, as discussed earlier, *in the GUM approach, the concept of true value is necessary for describing the objective of*

measurement. The concept of true value is also necessary for formulating a measurement model.

The GUM Approach is illustrated schematically in Figure 13, where the objective(s) of measurement are given at the top. Note that the vertical axis is no longer the number of times that a possible quantity value that could be attributed to the measurand is obtained by replicated measurements. Rather, the vertical axis is now the probability that individual ‘estimates’ of the value of the measurand actually correspond to the (essentially unique true) value of the measurand, where probability here means degree of belief under the assumption that no mistakes have occurred. The curve is now a probability density function (PDF) that is constructed on the basis of both replicate measurements (using so-called Type A evaluation) and other information obtained from measurement, such as values obtained from reference data tables and professional experience (using so-called Type B evaluation).

The combined standard uncertainty, expanded uncertainty and coverage interval are also illustrated in Figure 13, where the **coverage interval** is defined in VIM3 (2.19) as “interval containing the set of true quantity values of a measurand with a stated probability, based on the information available.” As indicated above, the GUM does not use the word “true” in connection with the concept of true value, and so “value” is indicated in the figure. Also indicated is the ‘intrinsic’ uncertainty associated with the fact that the (true) value is not unique (but only ‘essentially unique’) in the GUM Approach.

Incorporation of the terminology associated with the VIM3 rationales discussed above is illustrated schematically in Figure 14. The objective(s) of measurement are again given at the top of the figure, where the new terminology has also been incorporated. It is important to notice that nothing has changed in going from Figure 13 to Figure 14 other than the terminology, which is meant to emphasize that VIM3 is not intended to change the philosophy of the GUM approach, but only to clarify and possibly harmonize some of the terminology.

Figure 15 demonstrates the situation where the definitional uncertainty is not small compared to the rest of the measurement uncertainty, in which case the objective(s) of measurement are stated differently in recognition that probabilities must now be stated with respect to a set of true values, and not to an essentially unique true value. This measurement regime, and use of probability, is not treated in the GUM. However, the GUM indicates (e.g., Figure D.2) that definitional uncertainty is to be included in the calculation of measurement uncertainty.

The PDF from Figure 14 (solid curve) is reproduced as the solid curve in Figure 15. A broadened PDF (dashed curve) and larger coverage interval are presented in Figure 15 in order to emphasize the necessity of now incorporating the definitional uncertainty into the probability considerations. Because of the new definition of measurand in VIM3, as “quantity intended to be measured,” if it is thought (but not known) that the quantity actually being measured is different from the measurand, then, using the GUM approach,

the corresponding uncertainty is a part of the measurement uncertainty, and similar considerations concerning use of ‘probability’ would apply.

Since they were discussed earlier in connection with the CA, it is interesting to consider how the Bayesian and frequentist theories of inference relate to the GUM approach. In a sense, it can be said that the GUM approach, and in fact the UA in general, are consequences of the Bayesian theory of describing one’s state of knowledge about a measurand. Using the Bayesian theory in the GUM approach, measurement can be thought to consist of incrementally improving one’s state of knowledge and belief about a true value based on all of the accumulated information that is available through measurement. Using the Bayesian theory, the measurement uncertainty based on probability density functions associated with a particular measurand will continually change according to additional information obtained through measurement. The frequentist theory of inference can be useful for determining certain Type A components of measurement uncertainty, but is not capable of treating most Type B components. An example of the difficulty of the frequentist theory of inference within the GUM approach is that the frequentist theory is not able to be used to assess the uncertainty of a single measured value when using a measuring instrument, such as a voltmeter. The reason is that the uncertainty here derives from ‘nonstatistical’ information obtained from the instrument’s calibration certificate. This type of single measurement comprises a large fraction of the types of measurements routinely made daily throughout the world.

IEC Approach to the UA

The other major approach to describing and characterizing measurement that will be discussed here is that used by the International Electrotechnical Commission (IEC), as presented primarily through their IEC 60359 (*Electrical and Electronic Measurement Equipment – Expression of Performance*) [3]. The IEC philosophy questions the existence, in principle, of a true value of a quantity. The objective of measurement in this view is not to determine a true value of a measurand with a given probability, but concentrates instead on compatibility of measurement results.

The IEC approach uses a more operational or pragmatic philosophy than the GUM approach. Most notably, *the IEC approach treats the concept of true value as both unknowable and unnecessary, discouraging and in fact eliminating at least explicit use of the concept of true value, even in stating the objective of measurement.* In the IEC approach, as presented in the Introduction and Annex A of IEC 60359 [3], the stated objective of measurement is to obtain measurement results that are compatible with each other, within their respective measurement uncertainties. The philosophy is that, from an operational perspective, this is all that can really be done in measurement. This is illustrated schematically in Figure 16, where the four horizontal lines represent sets of measured values (central values, along with their measurement uncertainties) for four separate measurements of the same specified quantity being measured (which might be different than the measurand). From the IEC perspective, it can even be argued that the

concept of true value is potentially harmful, since it leads to thinking about something that is not relevant!

VIM3 RATIONALE: As a result of this key difference in philosophy between the IEC approach and the GUM approach to the UA, it is necessary to generalize several of the key concepts and definitions in VIM3 to accommodate both approaches whenever possible. For reasons discussed earlier, the important concept of **true value** (VIM3 1.11) is kept in VIM3, but is not explicitly used in the context of definitions that also apply to IEC. For example, the definition of **measured value** (VIM3 2.9) has been generalized to “quantity value representing a measurement result,” instead of “quantity value representing the set of true values of a quantity ...” so that true value does not need to be explicitly mentioned, but can still be inferred for the classical and GUM approaches. Similarly, **measurement result** (VIM3 2.10) has been defined in VIM3 as “set of quantity values being attributed to a measurand together with any other available relevant information”, rather than as “information about the set of quantity values being attributed to a measurand,” in order to accommodate the IEC view that a measurement result is just a set of values, with every element of the set having equal status. The probabilistic aspect of the GUM approach is left to the end of the definition as “any other available relevant information,” which can be ignored for the IEC approach. One last example that will be given is **definitional uncertainty** (VIM3 2.13), now defined in VIM3 as “minimum measurement uncertainty resulting from the inherently finite amount of detail in the definition of the measurand,” rather than “parameter characterizing the estimated dispersion of the true values of a quantity ...,” in order to remove explicit reference to true value.

Another key aspect of the IEC approach is that it focuses very heavily on providing guidance for estimating measurement uncertainty for situations where single measurements are made using measuring instruments, and where the measuring instrument is operating not only under reference conditions, but anywhere within its rated operating conditions or even extreme operating conditions. The key aspect of the IEC approach in this regard, as described in IEC 60359 [3], is to construct a calibration diagram for establishing the measurement uncertainty that can be ascribed to a single indication of a measuring instrument under various operating conditions. A calibration diagram is illustrated in Figure 17, where the horizontal axis (or ‘reading axis’) corresponds to the indications or ‘reading’ of a measuring instrument (in ‘units of output’), and the vertical axis (or ‘measurement axis’) corresponds to measured values (in ‘units of measurement’) as obtained using measurement standards. The ‘boundary of measured values’ around the calibration curve is obtained during the course of calibration of the measuring instrument, and is used to assess the measurement uncertainty to associate with a given subsequent indicated value for an unknown measurand (‘reading’ of the measuring instrument), as illustrated in the figure.

Returning to the fundamental IEC philosophy that the concept of true value is unnecessary, and that all that really matters is that measurement results are compatible with each other, one might ask what to do when measurement results are not compatible with each other, as illustrated schematically by ‘measurement number 5’ in Figure 18? In

this case it is necessary to investigate whether any mistakes have been made in performing all of the measurements. If no mistakes can be found, then it is assumed that the quantity that was measured was different for some of the measurements. In this case it becomes necessary to somehow ‘average all of the measurements’ and create an uncertainty that encompasses all of the measurement results.

Conventional Value Hybrid Approach; Knowable Error!

Before concluding, it is useful here to discuss a hybrid approach to the CA and UA that is frequently employed as a practical solution for handling the conceptual and terminological problems described earlier concerning the inability to know error, yet not abandon use of the concept and term, since they are still so widely used. This hybrid-approach, which will be called here the ‘Conventional Value Hybrid Approach’, or CVHA, is typically used in measurement situations where a decision must be made concerning whether a measured quantity conforms to a particular requirement, such as a specified machine tolerance or a legal regulation. The ‘hybrid’ aspect of the CVHA is that, while error is used, measurement uncertainty is also taken into account, as will be explained.

The CVHA is a two-step approach, where in the first step a measurement standard is calibrated and assigned an (essentially-unique) ‘conventional’ quantity value, and then, in the second step, a second measurement is performed on the calibrated measurement standard. Error is assessed in the second step with respect to the conventional value that was assigned to the measurement standard in the first step. This error can be expressed as a rational quantity since it is defined with respect to the conventional value, and not the true value, of the measurement standard.

Figures 19 and 20 schematically illustrate the two step process of the CVHA. Figure 19 shows the conventional value being assigned (through measurement, using a “high-level” measuring system) to the measurement standard. In this first step the systematic error, and hence the error, as defined with respect to the true value, cannot be known (the systematic error is set to zero by convention). The curve represents a fit to a set of histogram data that are obtained when using the measuring system to measure the measurement standard. Note that a measurement uncertainty of the conventional value can be determined, but this is not illustrated in the figure.

Figure 20 illustrates the second step of the process, where the quantity associated with the measurement standard (to which a conventional value has been assigned) is now measured with a “lower-level” measuring system. The measured values obtained when using this system are denoted schematically by the “fit to histogram data₂” on the right side of the figure, and an individual measured value (y_{i2}) is also indicated. Note that the measurement scale has been shifted in Figure 20, such that the difference between the conventional value and true value is meant to be the same in the two figures, and the “fit to histogram data₁” in the two figures is also meant to be the same. Figure 20 illustrates that, typically in the CVHA, the measured value using the “lower-level” measuring

system is not expected to be as “close” to the true value as the conventional value is and, further, the width of the “fit to histogram data₂” is not expected to be as small as that of the “fit to histogram data₁”. More importantly in Figure 20, however, is the illustration that systematic error and error can be defined in the second step of the CVHA both with respect to true value (in which case they are unknowable) or with respect to conventional value (in which case they are knowable). Note that systematic error here is also defined with respect to the mean of the histogram data and not the mean of the theoretical frequency distribution, as discussed earlier.

The advantage of the CVHA is that it can be used in measurement situations where the uncertainty associated with the conventional value is small with respect to the typical “knowable error,” and so it is possible to perform relatively straightforward measurements using the lower-level systems, and make equally straightforward conformity assessment decisions, without having to perform a possibly complicated uncertainty analysis. This approach has been used for many years and covers many types of measurement situations where, in fact, the “knowable error” is frequently treated as the measurand.

An example of the CVHA is the use of a standard weight to verify the performance of a balance. The weight is the (calibrated) measurement standard, and the balance is the lower-level instrument used to obtain the measured value in Figure 20. The knowable error is the difference between the conventional value of the weight and the indication when the weight is placed on the balance. This measured knowable error is then compared to a maximum permissible error (MPE) quoted in a regulation for that type of balance in order to make a decision about whether the balance conforms to the MPE requirement.

As modern measuring equipment used for even routine measurements becomes more sophisticated, it is not always possible to find a measurement standard or measuring instrument that is significantly better than the lower-level system (instrument), and so the knowable error is not always significantly larger than the uncertainty associated with the conventional value of the measurement standard. Further, as the pressure to become more efficient in every phase of business, including that concerning measurement, increases, there is a need to make better conformity assessment decisions. The irony is that it is then becoming increasingly important, when using the CVHA, to consider the uncertainty of the (knowable) error. It therefore becomes necessary to consider whether there is less terminological and conceptual confusion by calculating the uncertainty of the measured value itself (and specifying a maximum permissible uncertainty), than by estimating the knowable error.[4]

VIM3 RATIONALE: This dual usage of the term “error”, both in an unknowable sense when a measured quantity value is compared with a true value, and in a knowable (calculable) sense when that same measured quantity value is compared with a conventional value, is another dilemma faced in the development of VIM3, since two different concepts are being designated by the same term. The solution presented in VIM3 is to slightly re-define **error** (of measurement, 2.24) in a more general sense, as

“difference of measured quantity value and reference quantity value,” where the reference quantity value may or may not be the true value (e.g., it could be a conventional value). This new definition then encompasses both the unknowable and knowable usages of the term ‘error’.

VIM3 RATIONALE (**accuracy**): A concept closely related to “error” is that of “accuracy,” mentioned earlier, which even in the classical approach is in common use and is therefore kept in VIM3. The VIM3 definition (2.11): “<classical approach> closeness of agreement between a measured quantity value and a true quantity value of the measurand” is similar to the VIM2 definition, which also is based on true value. However, since IEC does not use the concept of true value, and also because a somewhat different usage of “accuracy” has developed in connection with the uncertainty approach, it was decided to include a second definition of accuracy (VIM3 2.11 bis): “<uncertainty approach> closeness of agreement between quantity values that are being attributed to the measurand.” This is a situation where a harmonized definition was not considered possible.

Summary

Different approaches and philosophies of measurement still exist and are in common use, most notably the classical approach and the uncertainty approach. Trying to create a vocabulary of metrology that harmonizes the language of measurement among the different approaches, and that keeps one term designating only one concept, has presented tremendous challenges in developing VIM3. While a principle used for VIM3 has been to harmonize terminology to the extent possible (e.g., error), it has in a few cases been necessary to allow two concepts having the same term (e.g., accuracy), or different terms for the same concept (e.g., value/true value), in the different approaches. Several of the decisions and rationales have been presented.

References

- [1] *International Vocabulary of Basic and General Terms in Metrology*, (VIM), 2nd Edition, International Organization for Standardization (ISO), 1993.
- [2] *Guide to the Expression of Uncertainty in Measurement*, (GUM), 1st Edition, International Organization for Standardization (ISO), 1993, Corrected and reprinted, 1995.
- [3] *Electrical and Electronic Measurement Equipment – Expression of Performance*, International Electrotechnical Commission (IEC) International Standard 60359, 3rd Edition, 2001-12
- [4] Role of measurement uncertainty in conformity assessment in legal metrology and trade, Kallgren et. al., *Accred. Qual. Assur.* (2003), 8:541-547.

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Common Elements of Philosophies and Descriptions of Measurement

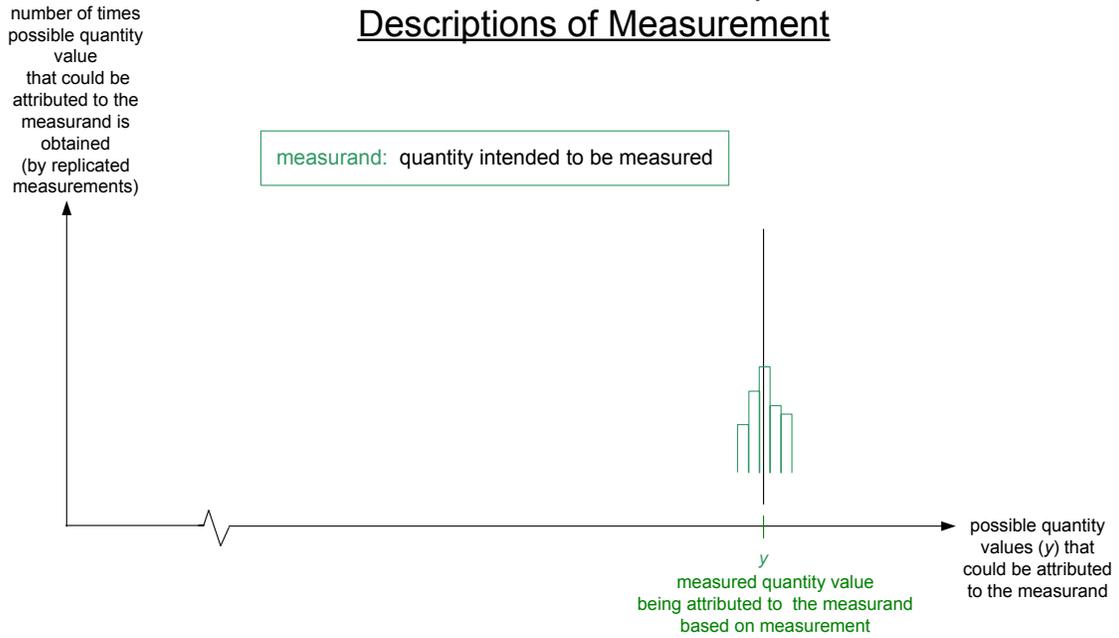


Figure 1

Common Elements of Philosophies and Descriptions of Measurement

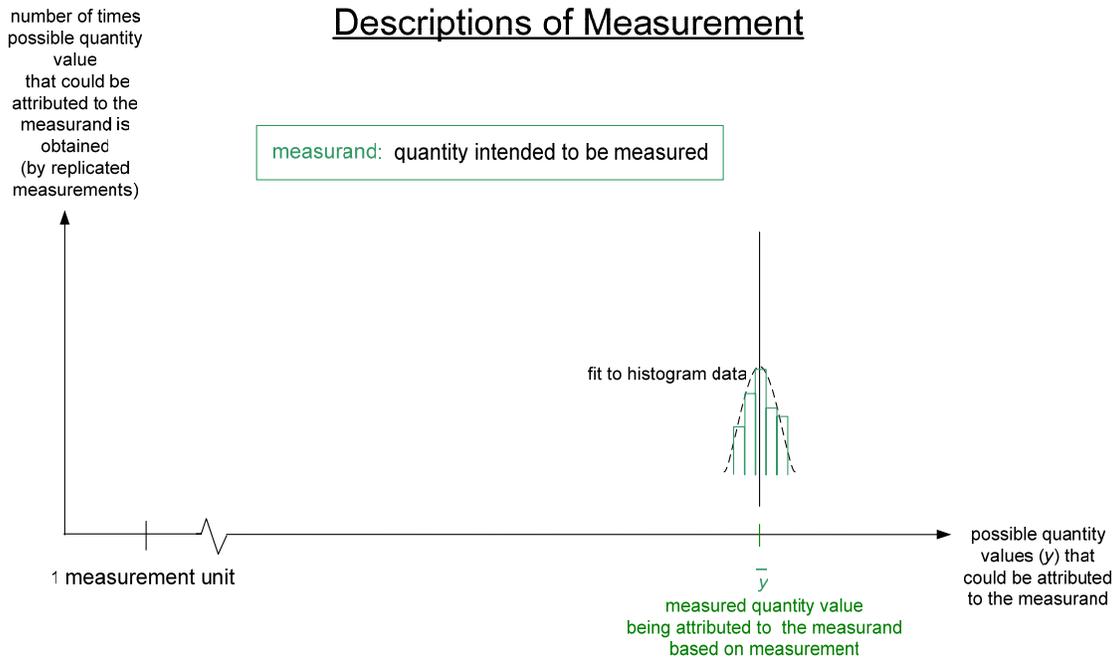


Figure 2

Classical Approach to Measurement

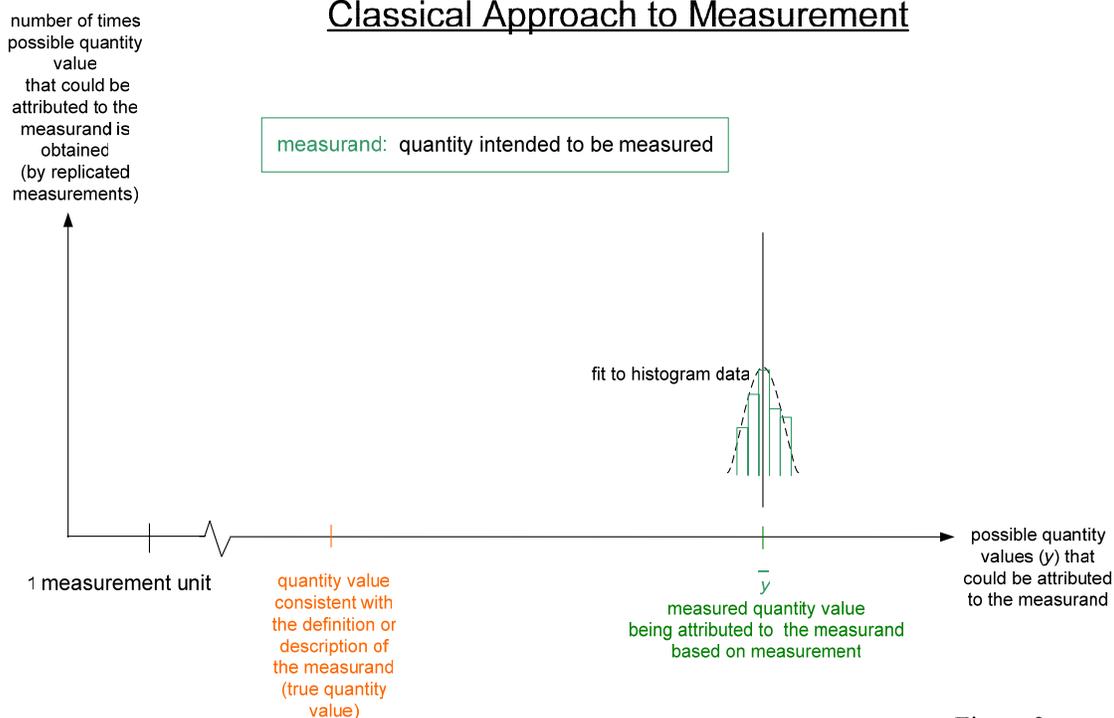


Figure 3

Classical Approach to Measurement

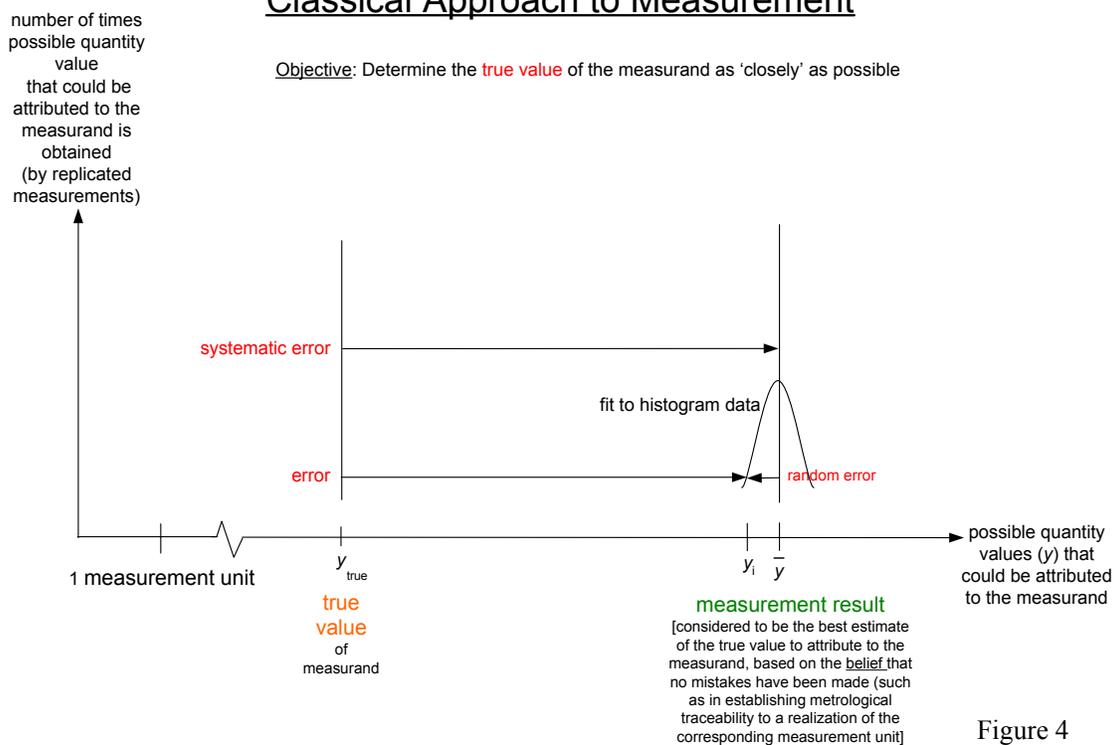


Figure 4

Classical Approach to Measurement

number of times possible quantity value that could be attributed to the measurand is obtained (by replicated measurements)

Objective: Determine the true value of the measurand as 'closely' as possible

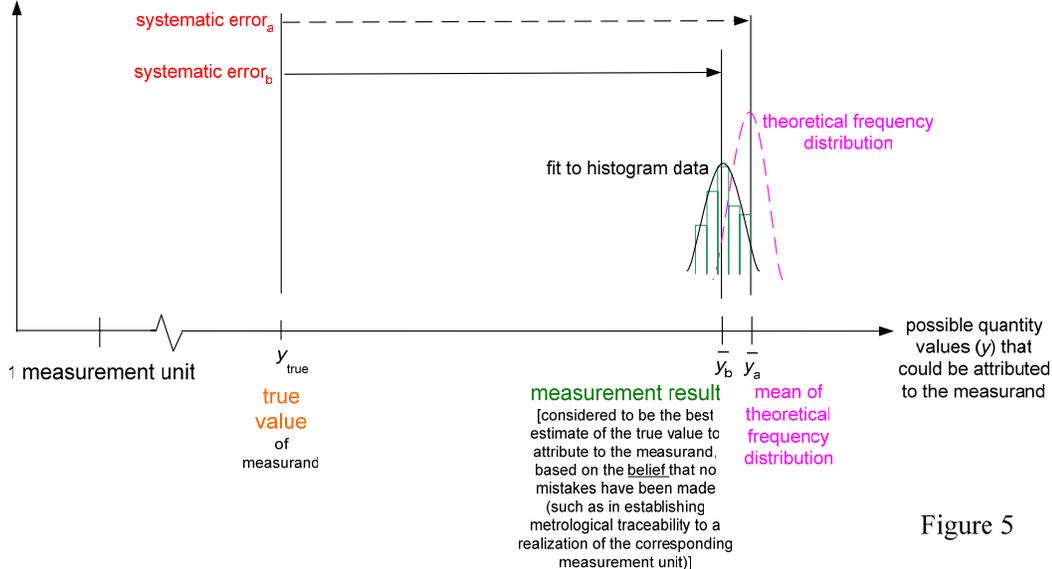


Figure 5

Use of Two Measurement Principles

number of times possible quantity value that could be attributed to the measurand is obtained (by replicated measurements)

Objective: Determine the true value of the measurand as 'closely' as possible

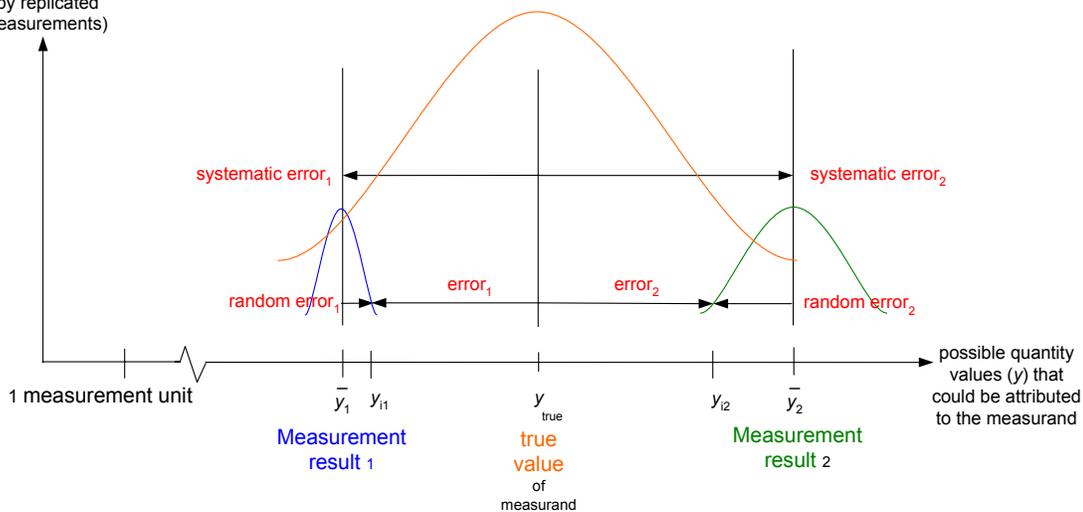


Figure 6

Use of Multiple Measurement Procedures, Methods and Principles

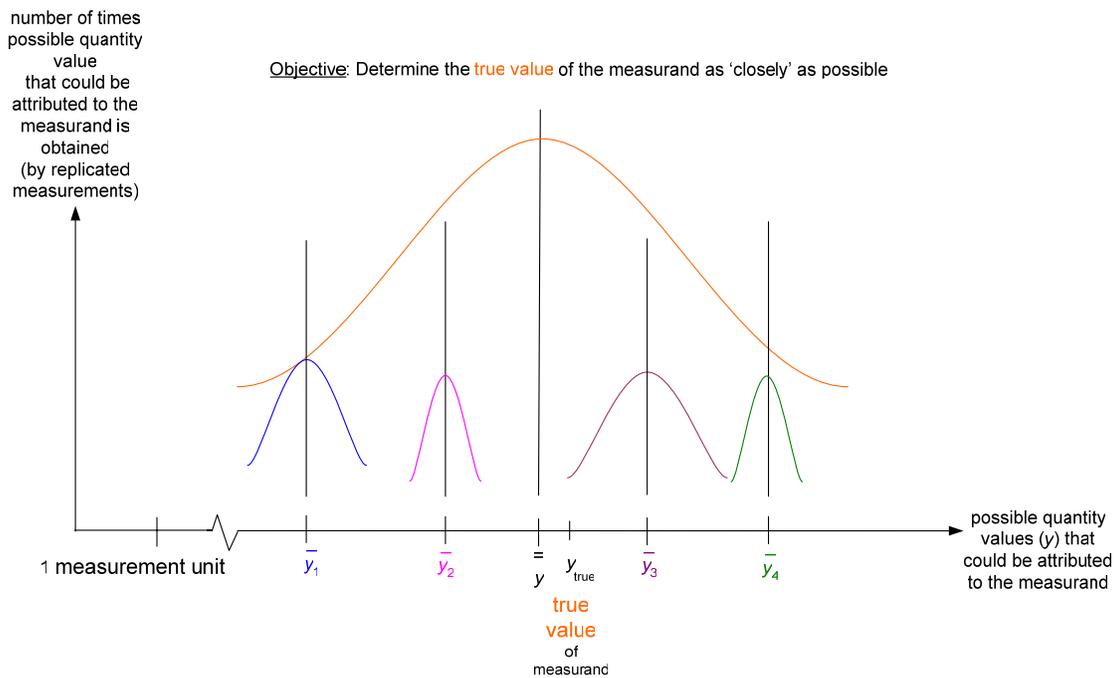


Figure 7

Classical Approach to Measurement

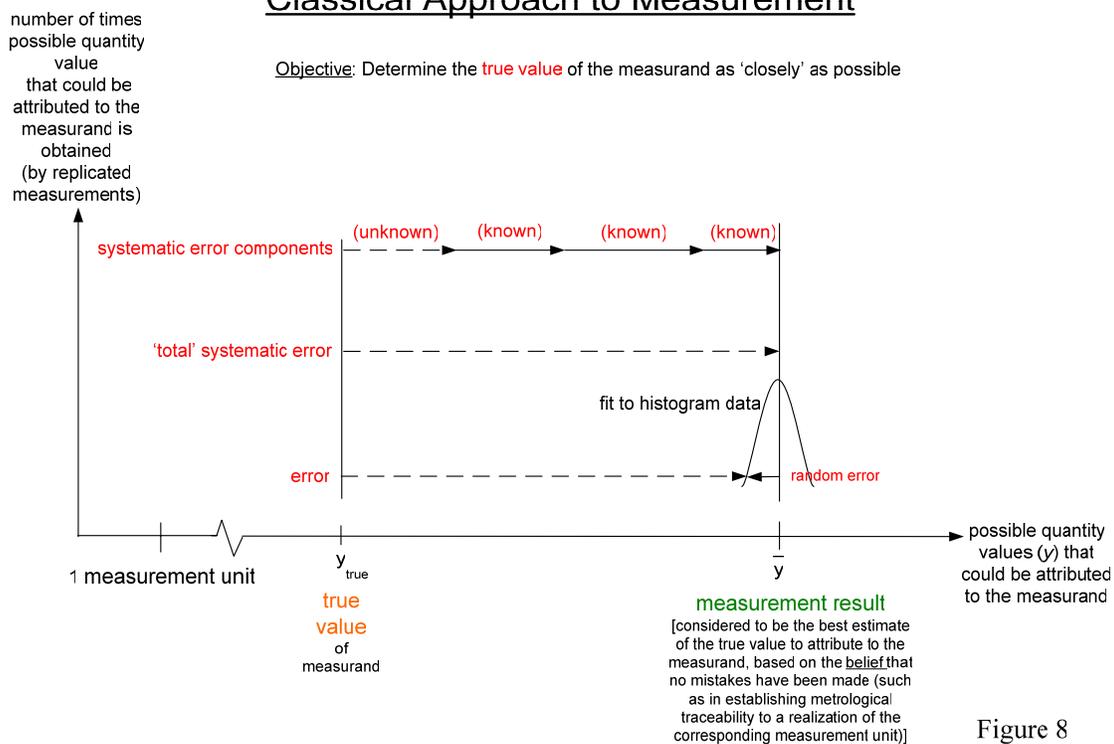


Figure 8

Classical Approach to Measurement

number of times possible quantity value that could be attributed to the measurand is obtained (by replicated measurements)

Objective: Determine the **true value** of the measurand as 'closely' as possible

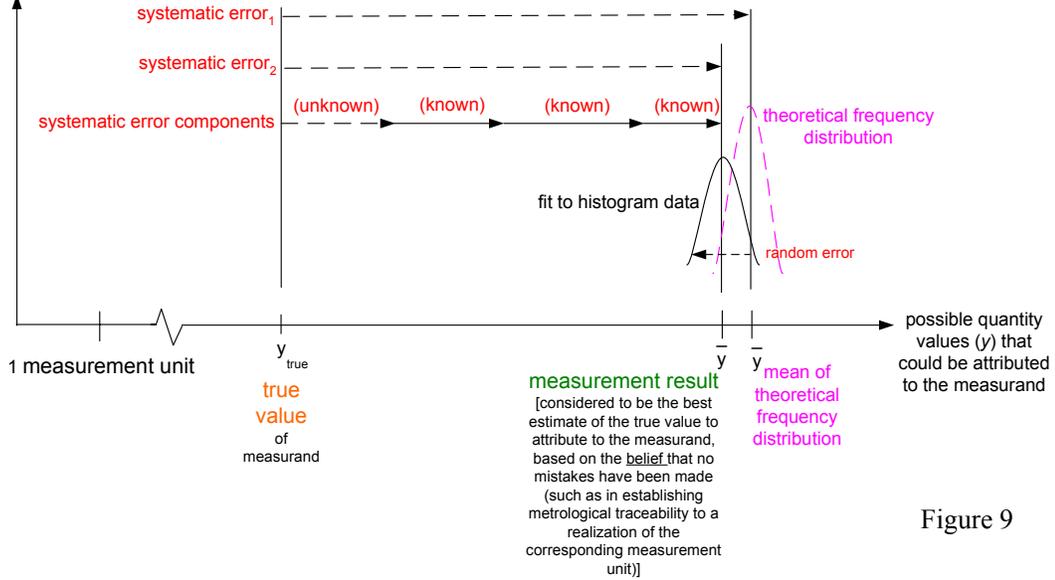


Figure 9

Non-Unique True Value

number of times possible quantity value that could be attributed to the measurand is obtained (by replicated measurements)

measurand: quantity intended to be measured

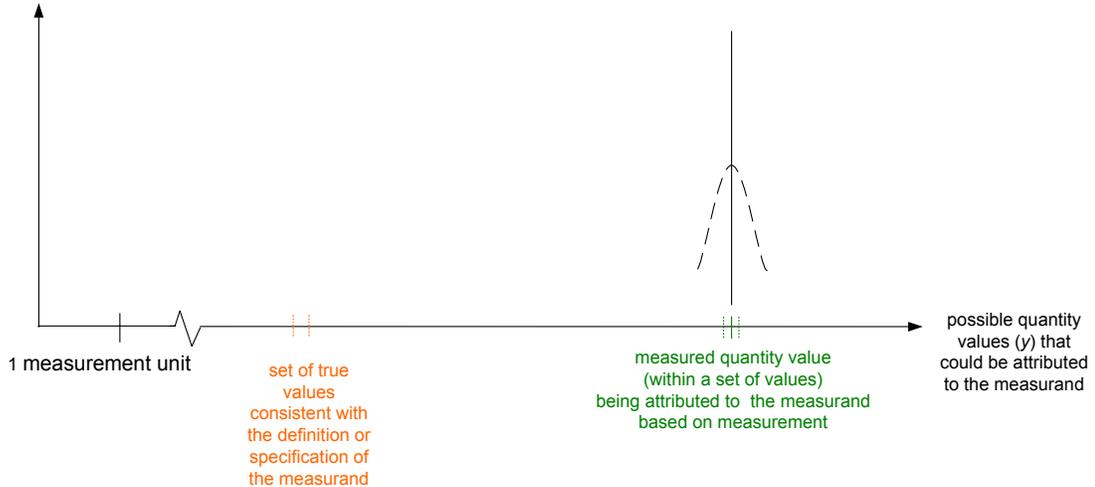


Figure 10

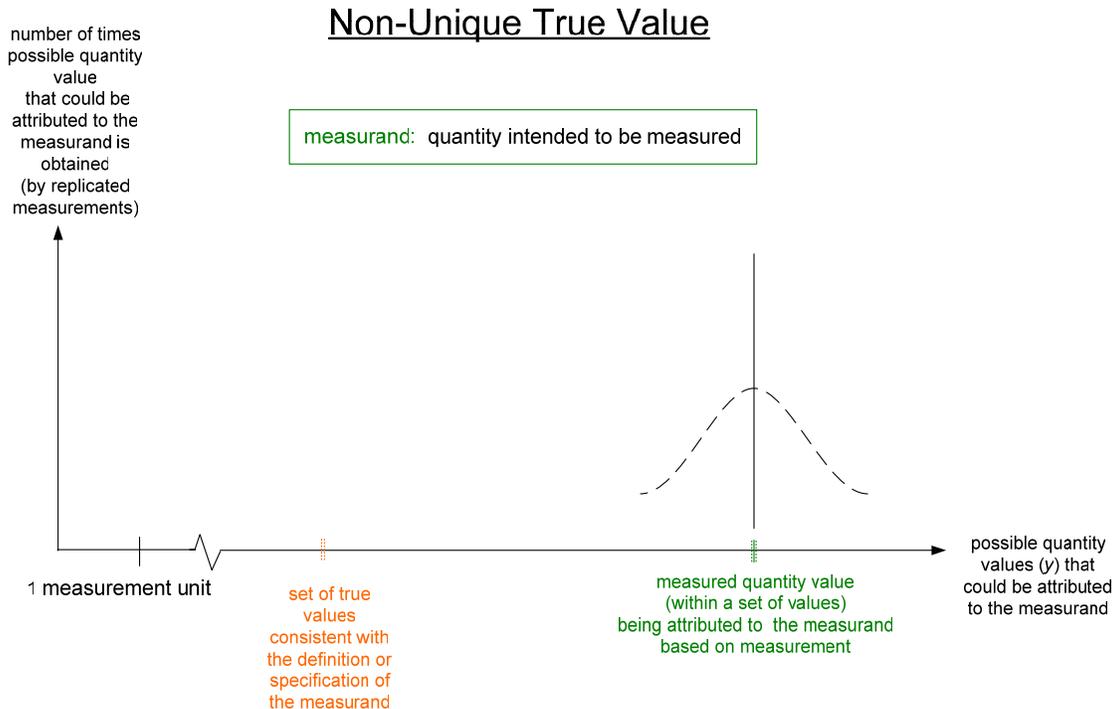


Figure 11

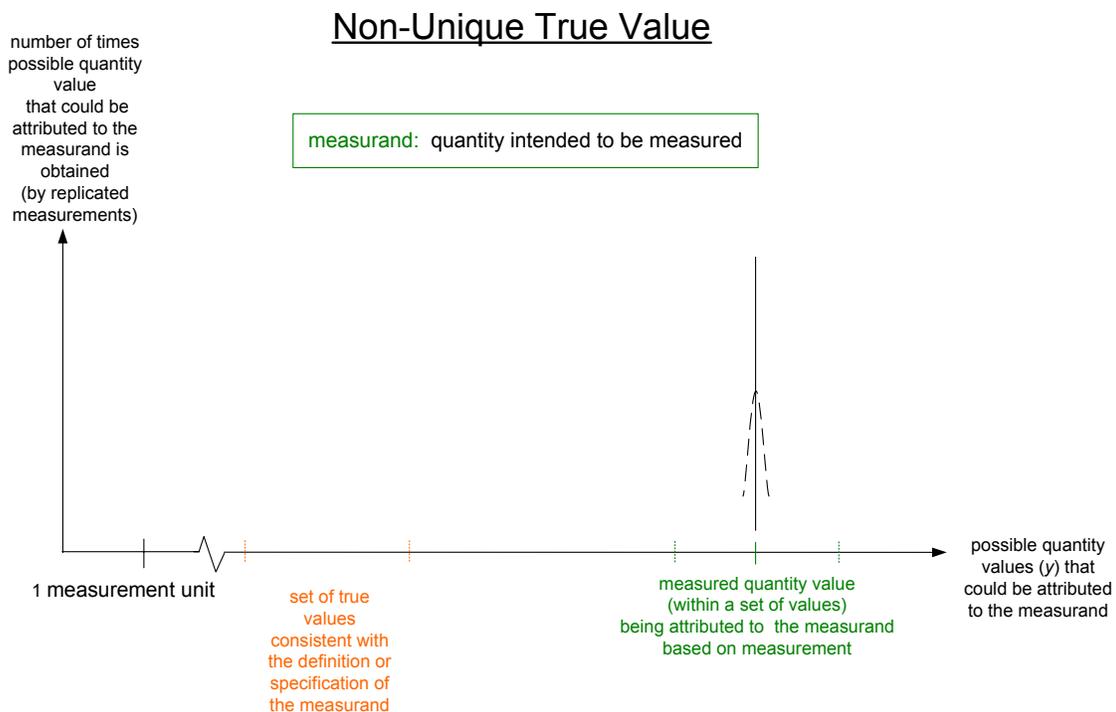


Figure 12

GUM Approach to Measurement

Establish probabilities (expressed as a probability density function, PDF) that individual **estimates** are actually 'the' (essentially unique true) **value** of the measurand, based on the information used from the measurement

OR

Establish an **interval of possible values** (coverage interval) within which 'the' (essentially unique true) **value** of the measurand is thought to lie, with a given level of confidence, based on the information used from the measurement

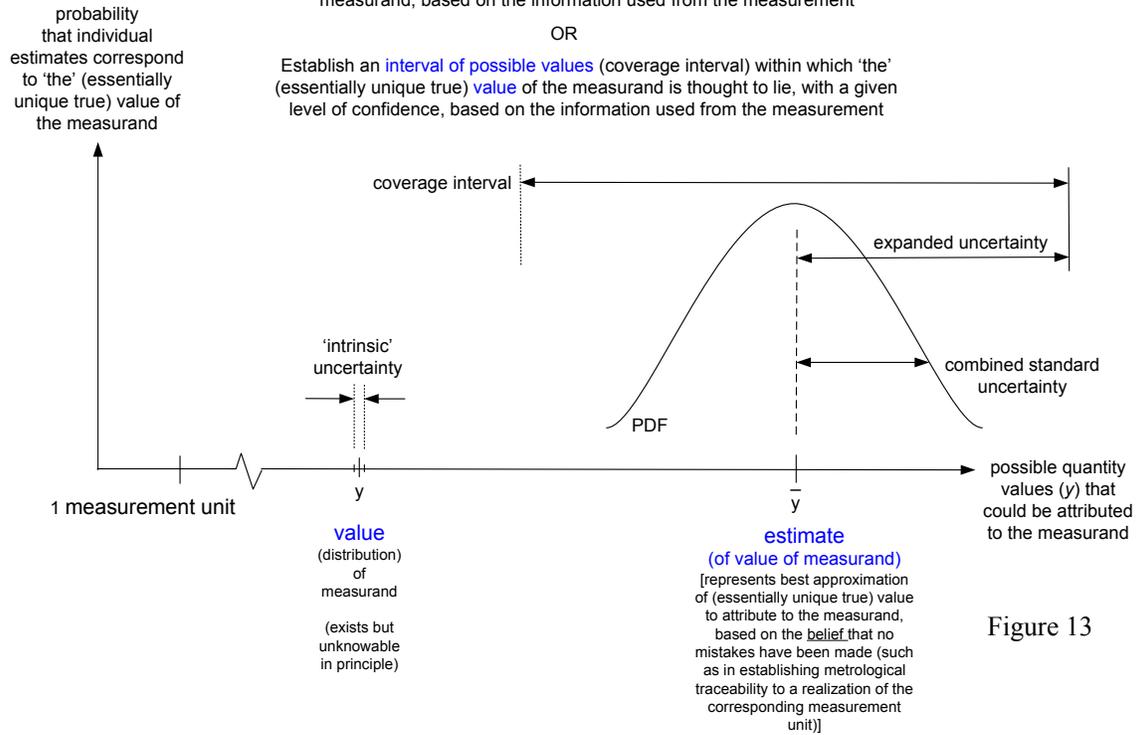


Figure 13

VIM3 Terminology for UA to Measurement

Establish probabilities (expressed as a probability density function, PDF) that individual **measured values** are actually 'the' essentially unique **true value** of the measurand, based on the information used from the measurement

OR

Establish an **interval of possible values** (coverage interval) within which 'the' essentially unique **true value** of the measurand is thought to lie, with a given level of confidence, based on the information used from the measurement

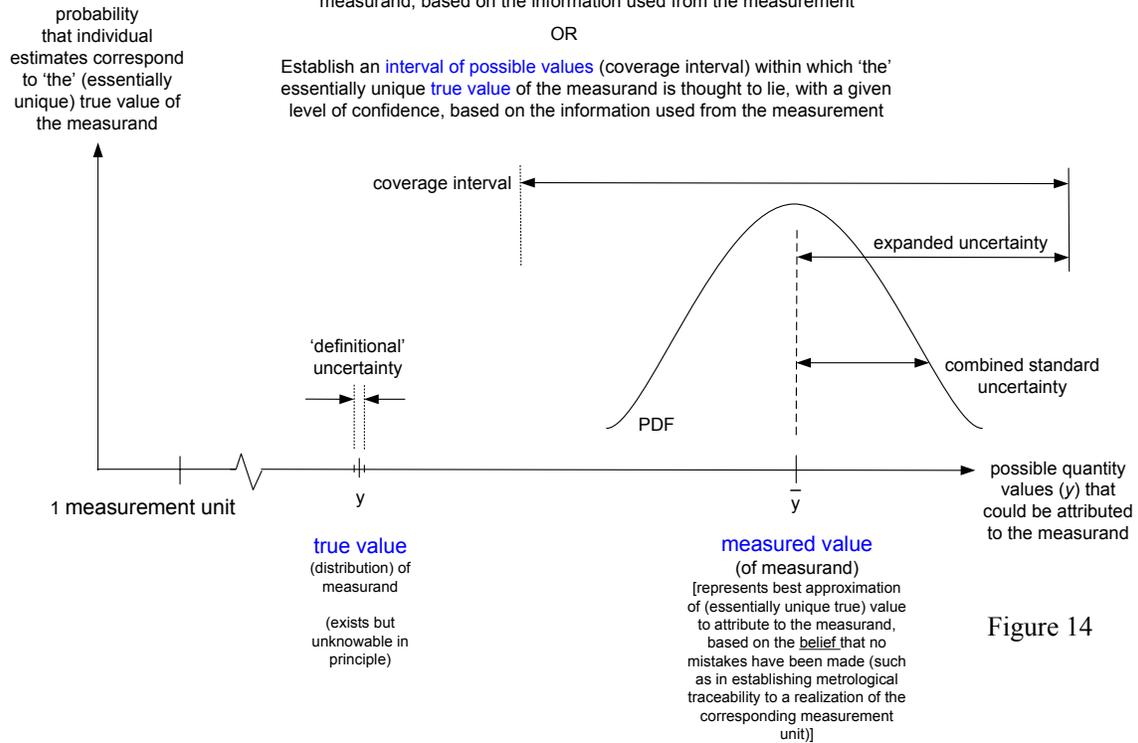


Figure 14

VIM3 Terminology for UA to Measurement

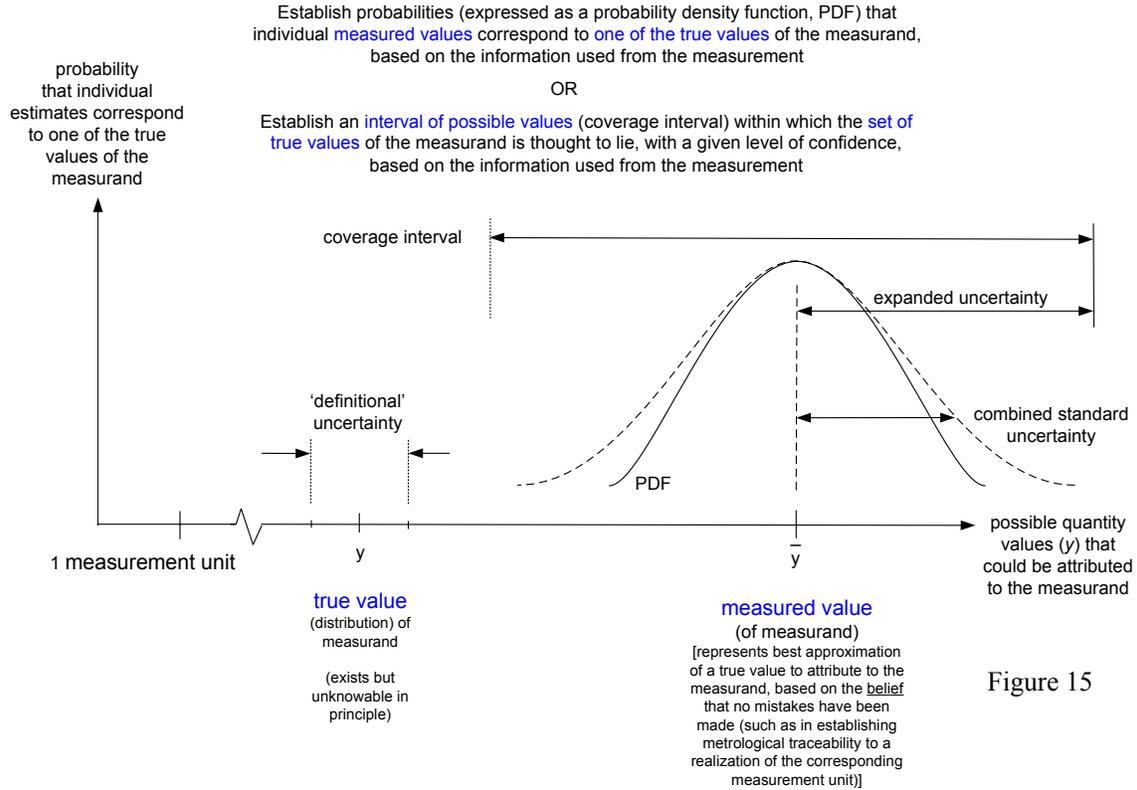


Figure 15

IEC Approach to Measurement

Objective: Assure **compatibility** of measurements

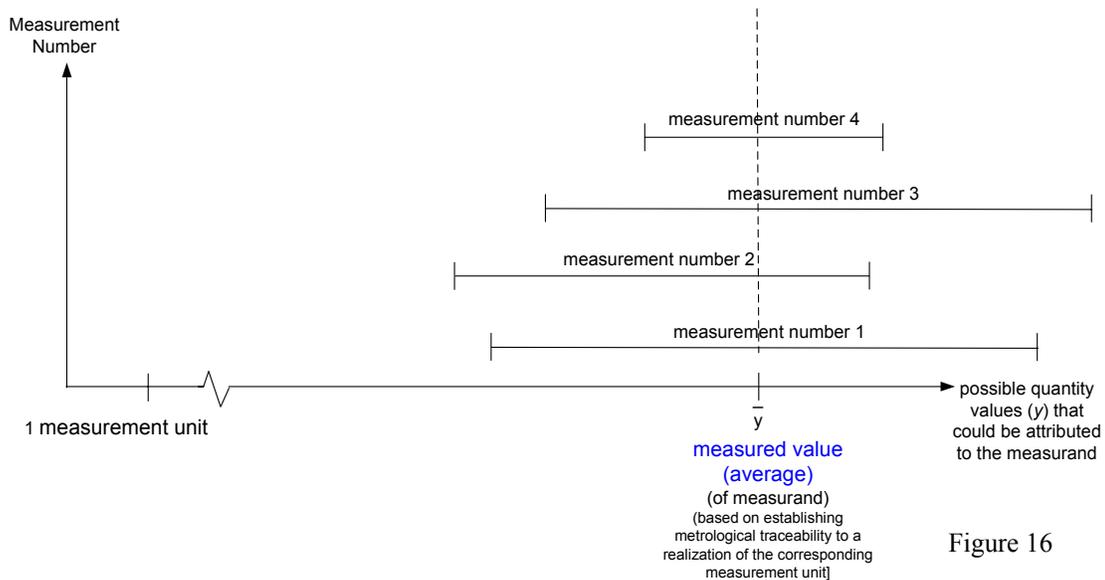


Figure 16

IEC Approach to Measurement

Calibration Diagram

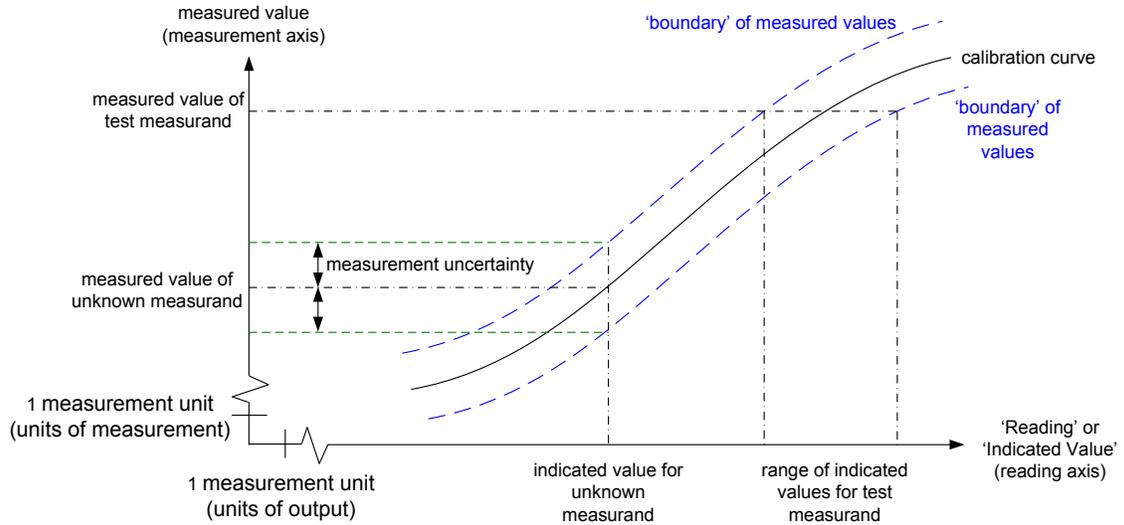


Figure 17

IEC Approach to Measurement

Objective: Assure compatibility of measurements

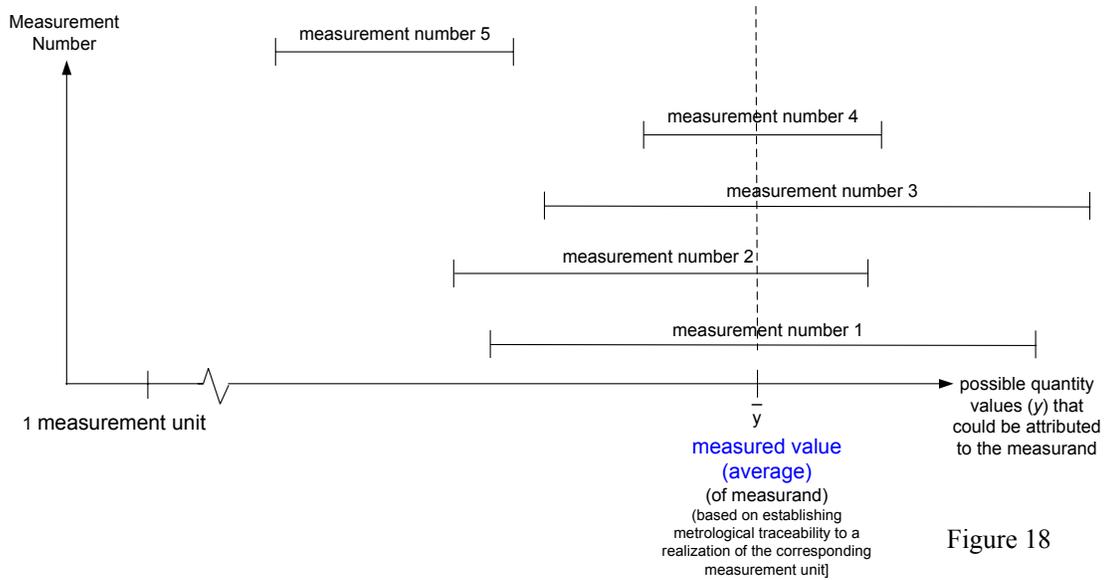


Figure 18

Conventional Value Hybrid Approach to Measurement

number of times possible quantity value that could be attributed to the measurand is obtained (by replicated measurements)

Step 1) Determine the conventional value of the measurement standard as 'closely' as possible

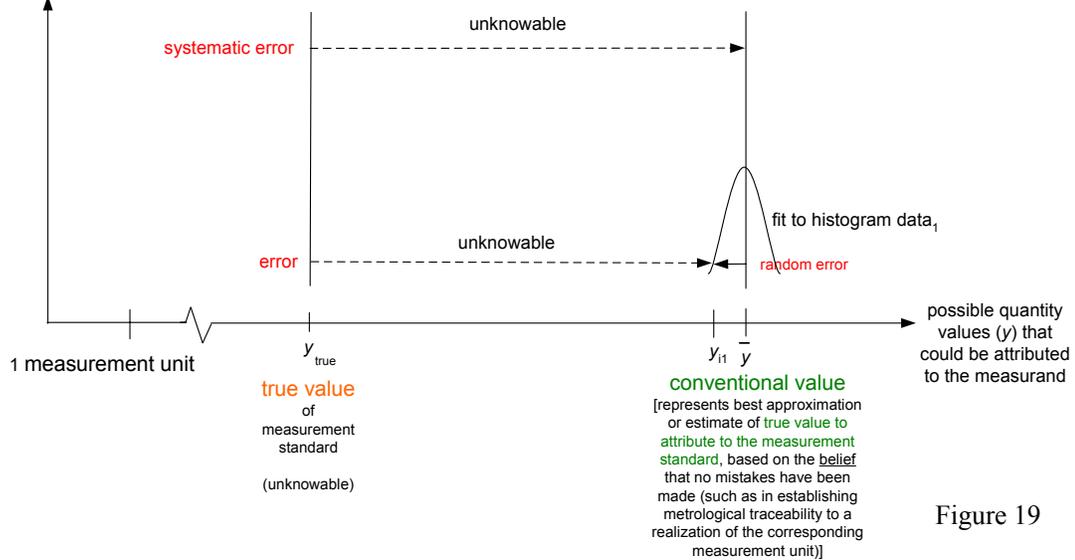


Figure 19

Conventional Value Hybrid Approach to Measurement

number of times possible quantity value that could be attributed to the measurand is obtained (by replicated measurements)

Step 2) Measure the (knowable) error 'sufficiently well'

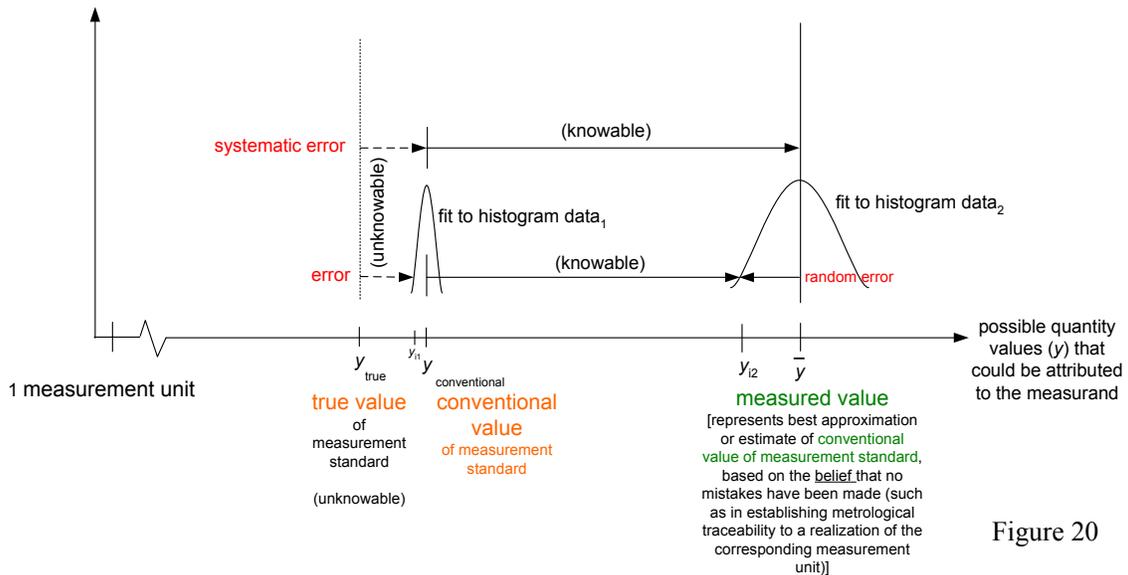


Figure 20