

The Role of QC in Sample Analysis

ECMC QATS Group - GCP Training Day

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Chapter 13: Clinical trials samples – analysis and evaluation

Throughout this chapter, specific terminology - 'must', 'required' or 'requirement' – has been used to interpret activities that are legislative requirements. These terms have been number-coded in the text where used, and the corresponding reference in the legislation can be found below.

Legislative References

- 1. Regulation 28 (1) of SI 2004/1031
- 2. Regulation 29 of SI 2004/1031
- 3. Schedule 1, Part 2 (4) of SI 2004/1031
- 4. Schedule 1, Part 2 (2) of SI 2004/1031
- 5. Schedule 1, Part 2 (9) of SI 2004/1031
- 6. Regulation 31A (4) of SI 2004/1031
- 7. Regulation 31A (7) of SI 2004/1031
- 8. Schedule 1, Part 2 (13) of SI 2004/1031

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13.6 Maintaining quality within the laboratory

The quality of laboratory work is of utmost importance......Quality requirements.....Quality into two functions: quality control (QC) and QA. Both are equally important but have a different focus and purpose.

13.6.1 Quality control

The accuracy of all laboratory processes should be subject to **<u>some</u>** level of QC checks

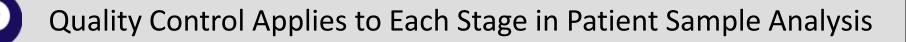


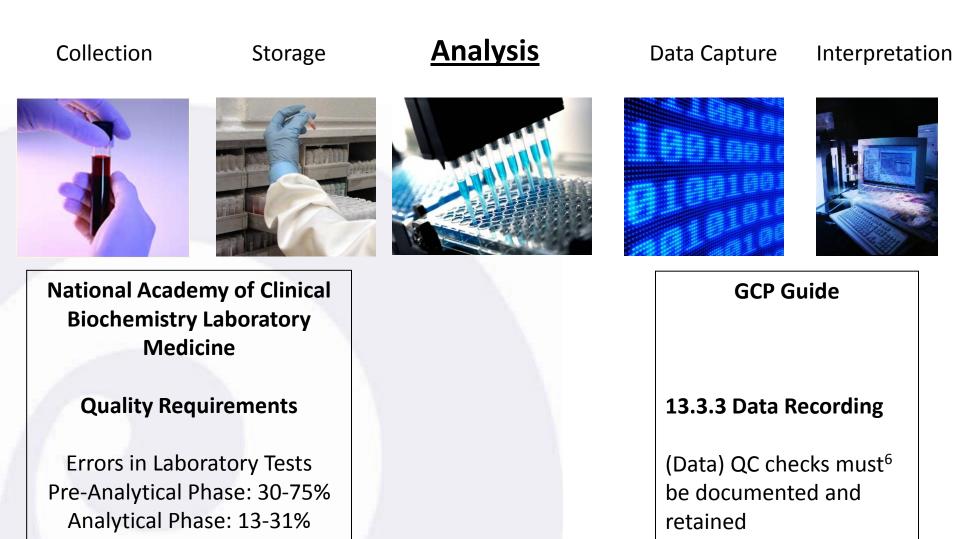
14.3 Quality control

QC is a fundamental approach to verify compliance with the Clinical Trials Regulations (Schedule 1, Part 2 (4) and (9) of SI 2004/1031) and should be implemented.

QC is an activity that involves the review of factors in a process as the process is occurring.

Expectations relating to acceptable standards of QC should be documented; this would usually define **acceptable error rates** and is recommended that the process defines the **actions to be take**n where the QC checks show a failure to meet the acceptable predefined standard







The confirmation by the provision of objective evidence that the particular requirements for a specific intended use are fulfilled*

*ISO 17025 and 9000

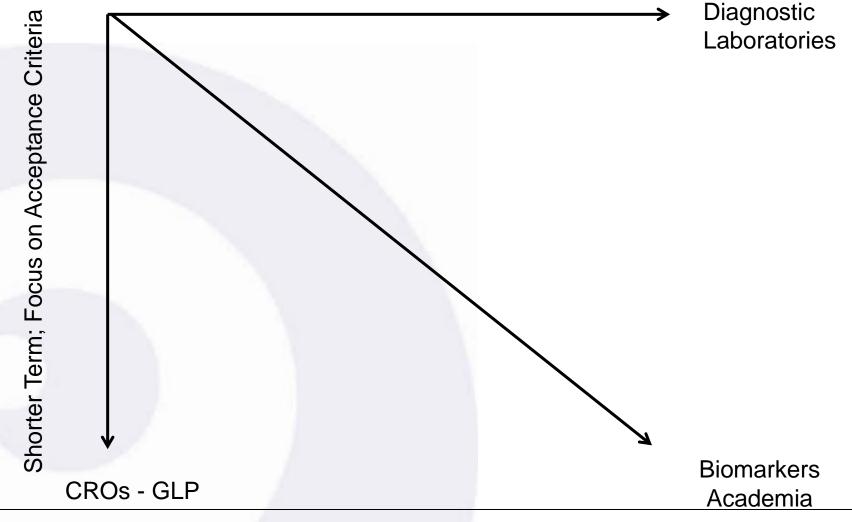
Analytical QC is Conducted Utilising Quality Control Samples (QCs)



- QCs are samples of known concentration/number/staining pattern/Δct/etc run together with the patient samples to objectively evaluate assay performance
- They should resemble as closely as possible the test samples and are therefore meant to replicate the behaviour of patient samples in the assay
- Effectively, QCs function to measure the level of error/uncertainty present in an assay
- Acceptance Criteria are the control limits that are set for the expected performance of QCs based on intended purpose and the operate in conjunction with a Decision Rule

AAPS and FDA Have Identified Two Contrasting Approaches to QC

QC Monitoring Over Time: Focus on Control Charts and Decision Rules

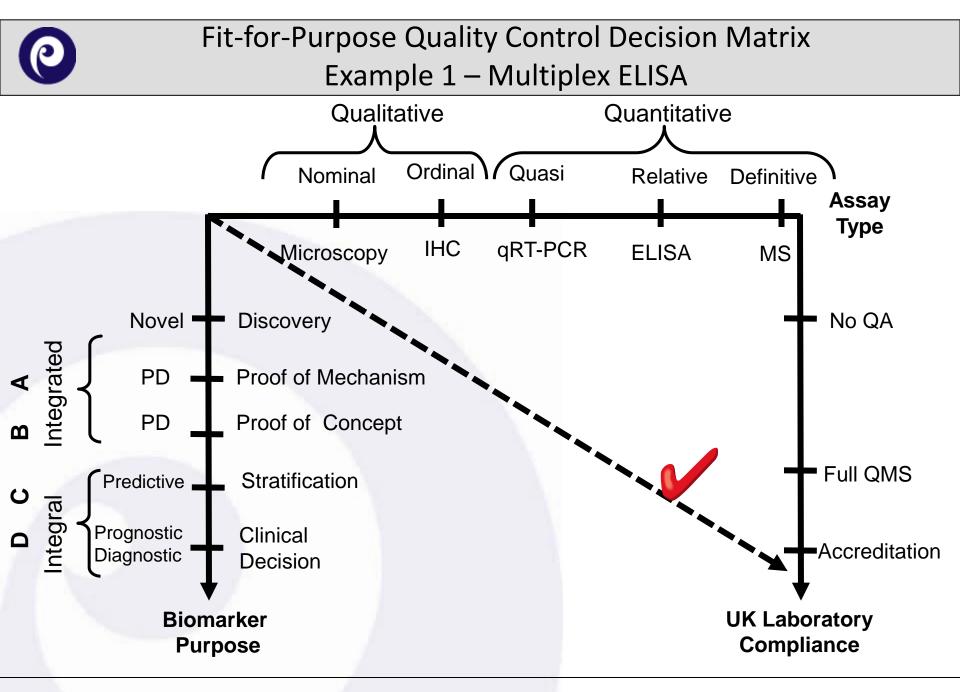


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- 1. Qualification of Multiplex ELISA Determination of Angiogenesis Analytes as Predictive Biomarkers of Response to Anti-Vascular Drugs
- 2. Enumeration of Circulating Tumour Cells (CTC) in Patient Stratification in a Proposed new Trial in Colorectal Cancer
- 3. Immunohistochemistry to Determine Pharmacodynamic Changes in the Expression of the Androgen Receptor in CTC after Treatment of Prostate Cancer Patient with a Phase I Drug

To be Followed Up in More Detail at the Data Quality Workshop*



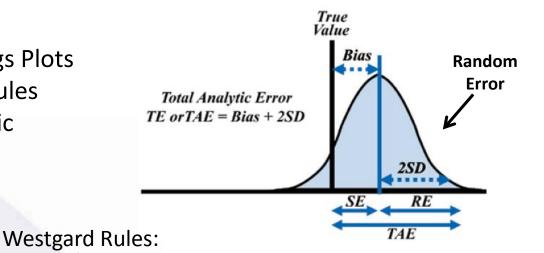
Based on ECMC BAQA Group Publication , BJC, 2010, Volume 103, 1313-1317



QC Approach to a Relative Quantitative Biomarker Assay

Control Limits: Decision: Assessment:

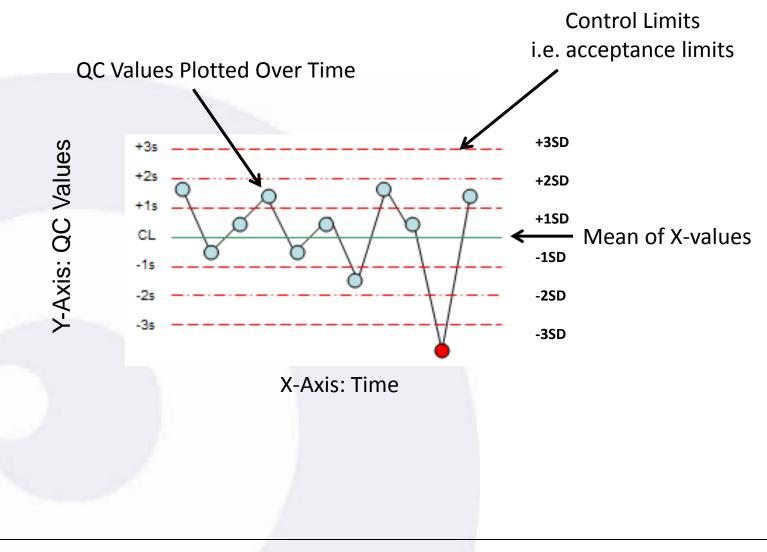
Levi-Jennings Plots Westgard Rules Sigma Metric



- ✓ Based on the concept of Total Error (TAE): the combination of Systematic Error (SE, i.e. Bias) and Random Error (Imprecision) – Therefore ideal for a Relative Quantitative Assay
- ✓ Consists of a series of Decision rules applied in a sequence of increasing analytical rigor
- ✓ The rules have been developed based on power calculations, computer simulations and actual practice and optimised to maximise detection of true assay failures and minimise rejection of valid assays
- \checkmark The rules work in conjunction with the Sigma Metric to optimise QC
- \checkmark Can be customised to meet the requirements of fit-for-purpose approach to QC
- ✓ Live monitoring of assay performance
- ✓ Identifies the nature of analytical error and thus potentially facilitates correction



Levi-Jennings Control Plots for Monitoring the Progress of Biomarker Assays Over Time



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A Selection of the Most Frequently Adopted Westgard Rules

 $\mathbf{1}_{3s}$ refers to a control rule that is commonly used with a Levey-Jennings chart when the control limits are set as the mean plus 3s and the mean minus 3s. A run is rejected when a single control measurement exceeds the mean plus 3s or the mean minus 3s control limit.

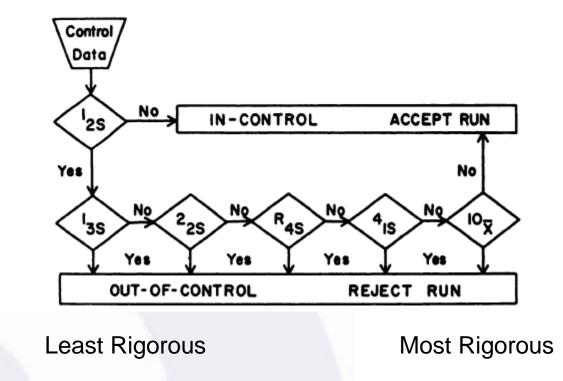
 $\mathbf{2}_{2s}$ - reject when 2 consecutive control measurements exceed the same mean plus 2s or the same mean minus 2s control limit.

4_{1s} - reject when 4 consecutive control measurements exceed the same mean plus 1s or the same mean minus 1s control limit.

 $\mathbf{8}_{\mathbf{x}}$ - reject when 8 consecutive control measurements fall on one side of the mean.

 7_{T} - reject when seven control measurements trend in the same direction, i.e., get progressively higher or progressively lower.

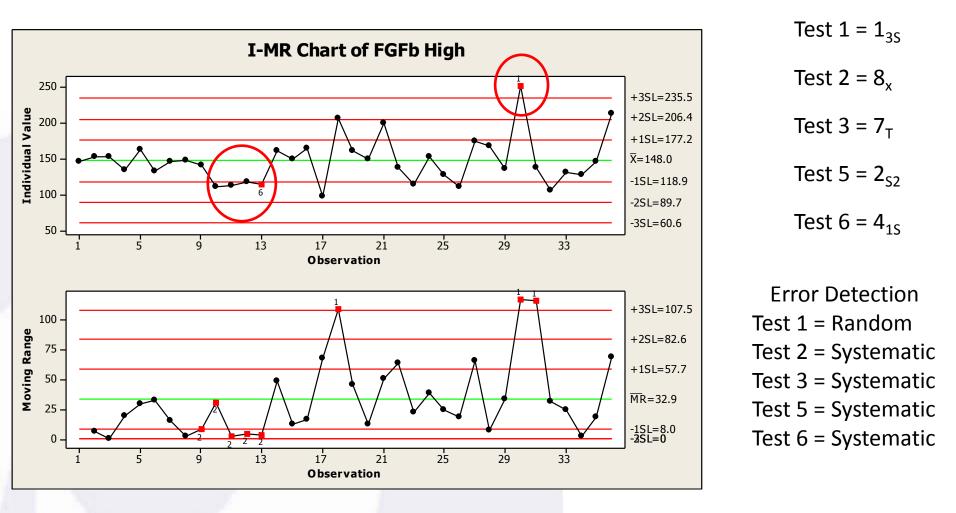
Application of Westgard Rules – Decision Rule Matrix



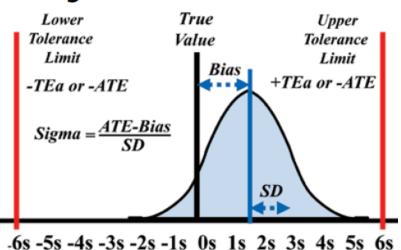
Matrix and Rules Based on An Assay with 2 Different QC Concentrations Run in Duplicate i.e. 4 QC values

A greater number of QCs will affect the Power Calculation and change the rules

Analysis of Levi-Jennings Plots and Westgard Rules of Multiplex QC Data for FGF beta Evaluated Utilising Minitab







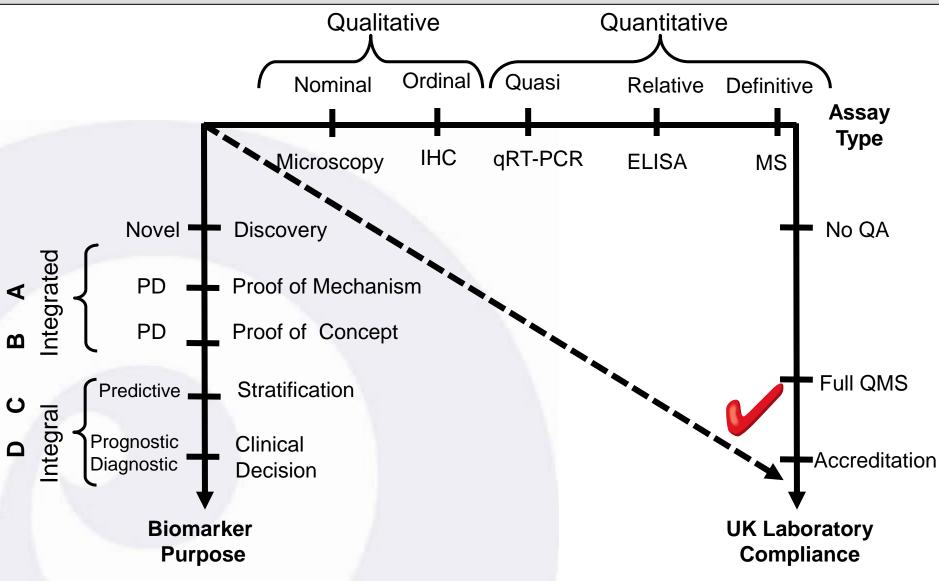
Sigma-metric Calculation

Example Calculation – Taken from Westgard QC Web-Site The CLIA criterion for acceptable performance for cholesterol is 10%. If a laboratory method shows a bias of 2.0% on proficiency testing surveys and a CV of 2.0% on internal QC results, the Sigma-metric is (10-2/2) = 4

> Sigma = 1-2, Poor/Very Poor, high level QC required Sigma = 3-4, Moderate/Good, regular level of QC required Sigma = 5-6, Very Good, less QC required

Fit-for-Purpose Quality Control Decision Matrix

Example 2– CTC Enumeration for Patient Stratification



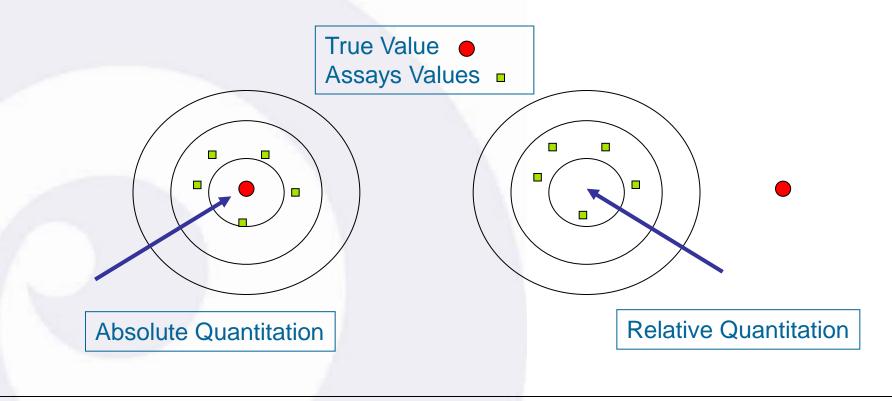
Based on ECMC BAQA Group Publication , BJC, 2010, Volume 103, 1313-1317



Levi-Jennings Plots and Westgard Rules Not Necessarily Sufficient for Absolute Quantitation

The Most Important Parameter of a Definitive Quantitative Assay is: Analytical Accuracy

The Ability to Measure the True Concentration/Amount Present in the Patient Sample

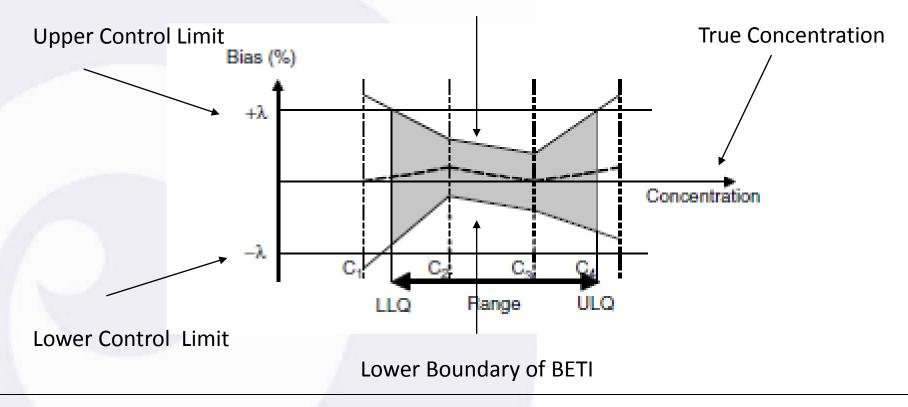




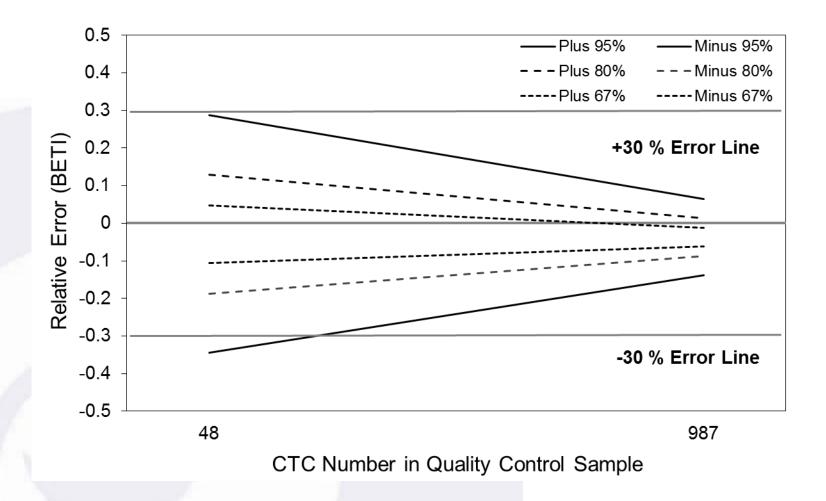
β-Expectation Tolerance Intervals Produce a Plot Called the Accuracy Profile

The β -expectation tolerance interval (BETI) calculates the upper and lower boundaries where each future measurement of patient samples are expected to has a defined probability (β) to fall within and thus informs on analytical accuracy

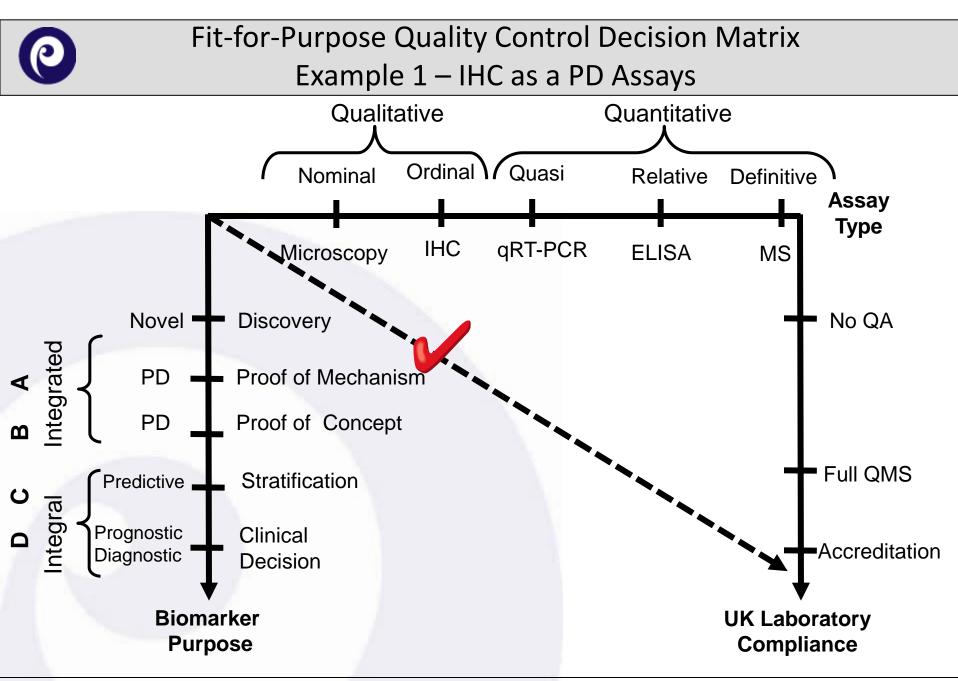
Upper Boundary of BETI



The Accuracy Profile of CTC Enumeration in QC Samples Monitored Over 3 Months and Conducted by 4 Different Analysts*



*Taken From: Cummings et al. BMC Cancer 2013, 13:415; http://www.biomedcentral.com/1471-2407/13/415



Based on ECMC BAQA Group Publication , BJC, 2010, Volume 103, 1313-1317



Quality Control of IHC Focussed on the Reproducibility of the Staining Intensity

Performance Characteristic	Qualitative
Accuracy	
Trueness (Bias)	
Precision	
Reproducibility	✓
Sensitivity	✓
Specificity	✓
Dilution Linearity	
Parallelism	
Assay Range	
Reagent Stability	✓
Sample Stability	✓

Negative	Weak	Moderate	Strong
	0	6	
Negative	1+	2 ++	3 +++



Reproducibility of IHC Staining Intensity Evaluated Using a Modified Accuracy Profile*

Modification of ISR Substitution with 2 Analysts

 Y_i^O = the original measurements (i.e. analyst 1). Y_i^R = the repeat measurements (i.e. analyst 2).

$$\Delta_i = \log(Y_i^R) - \log(Y_i^O)$$

$$\bar{\Delta} = \frac{1}{N} \sum_{i=1}^{N} \Delta_i$$

$$\hat{\sigma}_{\Delta}^2 = \frac{1}{N-1} \sum_{i=1}^{N} \left(\Delta i - \bar{\Delta} \right)^2$$

$\label{eq:stable} \begin{array}{l} \mbox{Calculation of BCTI:} \\ \mbox{β-content$ γ-confidence tolerance intervals} \end{array}$

 $The \ two \ tailed \ \beta \text{-content} \\ \gamma \text{-confidence tolerance interval is therefore defined as:}$

$\bar{\Delta} \pm Z_{(1+\beta)/2} \sqrt{1+N^{-1}} \sqrt{(N\!-\!1)\,\hat{\sigma}_{\Delta}^2/x_{N\!-\!1,1\!-\!\gamma}^2}$

 $Z_{(1+\beta)/2}$ is the upper $(1+\beta)/2$ quantile of the standard distribution and $x_{N-1,1-\gamma}^2$ is the lower γ quantile of the chi-squared distribution (N-1 degrees of freedom). Calculation of BCTI was performed utilising MATLAB (as above) at $\beta = 67\%$ and 95% [26]. A plot of BCTI (y-axis) against the operator pair (x-axis) represents a modified form of the 'accuracy profile'. All code developed in MATLAB was validated against previously published data sets as reported previously [26].

*Taken From: Cummings et al. BMC Cancer 2013, 13:415; http://www.biomedcentral.com/1471-2407/13/415



Summary: QC in Biomarker Analysis of Clinical Trial Samples

- 1. The MHRA place great emphasis in Quality Control (QC) in the analysis of trial samples, at each stage in the analytical cycle
- 2. However, the GCP Guide 2012 gives little indication of the level of QC required
- 3. A fit-for-purpose approach has been presented based on a decision matrix which takes account of the nature of analytical procedure and the purpose of the biomarker
- 4. As examples, QC on three different types of assays (ranging from definitive quantitation to categorical) employed as biomarkers with very different purposes (ranging from PD to stratification) are presented
- 5. These examples were be discussed in more detail at the Workshop on Data Quality