

Mechanistic and Statistic Partitioning the Technical Variability of Ligand Binding Assays in Distinct Formats

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ICH M10 LBA Guideline
Distinct LBA assays
LBA (PK, Biomarker, ADA)
Partitioning technical variabilities
Deterministic effect
4PL model
Linear Mixed Models
LLOQ
Gyros-Lab
SIMOA
Zero-inflated models



- ±20% for RE & CV, except for the LLOQ and ULOQ, where it should not exceed 25%.
- For non-accuracy and precision validation runs, at least 2/3 of the total QCs and at least 50% at each concentration level should be within ±20% of the nominal values.
- TE should not exceed 30% (40% at LLOQ and ULOQ).
- Detailed regression models including weighing factors for LBA assays, which are deterministic factors for technical variability for LBAs, are not available in ICH M10.
- Stochastic factors, including batch effects, are not detailed in ICH M10.
- Both distinct formats of LBAs and inherent Ab-Ag affinity and avidity are crucial for technical variability.

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*Committee for Medicinal Products for Human Use* 

ICH Guideline M10 on Bioanalytical Method Validation and Study Sample Analysis

Step 5











# Oistinct Formats for Anti-Drug Antibodies Assays



## Partitioning Technical Variabilities (Stochastic)



## Variability and LLOQ are Not Independent

In-Well (Without X	Deterministic	Stochastic
Dilution Factor)	Factors	Factors

Platform	LLOQ	CV%
ELISA	~1 ng/mL	<15%
MSD	~2 pg/mL	<15%
ELLA	~1 pg/mL	<20%
Gyros-Lab	~1 pg/mL	<25%
SIMOA	~0.2pg/MI	<30%

In general, LLOQ and variability are negative correlated.



$$\frac{[RL]}{[R][L]} = \frac{k_{\text{on}}}{k_{\text{off}}} = K$$

*K*, the equilibrium affinity constant, has dimensions of  $M^{-1}$ . As *K* (analogous to the thirst of the delegates) increases, so the concentration of the receptor–ligand complexes increases at the expense of the free species. Alternatively,

$$\frac{[R][L]}{[RL]} = \frac{k_{\text{off}}}{k_{\text{on}}} = K_{\text{d}}$$



$$Y(t) = A + rac{K-A}{(C+e^{-B(t-M)})^{1/
u}}$$

$$y = d + \frac{a - d}{1 + \left(\frac{x}{c}\right)^b}$$

$$x = c \left(\frac{a-d}{y-d} - 1\right)^{\frac{1}{b}}$$

- Simple linear regression can work well for ELISA, especially for IVD kits;
- Weighing factor for 4/5 PL impact result, including CV% and RE%;
- Log-log transformation may be needed;
- For laboratory developed test (LDT), determined by LBA platforms.

# Standard Curves (Data Transformation)





$$egin{array}{rcl} \mathbf{y}_i &=& \mathbf{X}_i oldsymbol{eta} + \mathbf{Z}_i \mathbf{b}_i + oldsymbol{arepsilon}_i \ \mathbf{b}_i &\sim& \mathbf{N}_q(\mathbf{0}, oldsymbol{\Psi}) \ oldsymbol{arepsilon}_i &\sim& \mathbf{N}_{n_i}(\mathbf{0}, oldsymbol{\sigma}^2 oldsymbol{\Lambda}_i) \end{array}$$

where

- $\mathbf{y}_i$  is the  $n_i \times 1$  response vector for observations in the *i*th group.
- $\mathbf{X}_i$  is the  $n_i \times p$  model matrix for the fixed effects for observations in group *i*.
- $\beta$  is the  $p \times 1$  vector of fixed-effect coefficients.
- $\mathbf{Z}_i$  is the  $n_i \times q$  model matrix for the random effects for observations in group *i*.
- $\mathbf{b}_i$  is the  $q \times 1$  vector of random-effect coefficients for group i.
- $\varepsilon_i$  is the  $n_i \times 1$  vector of errors for observations in group *i*.
- $\Psi$  is the  $q \times q$  covariance matrix for the random effects.
- $\sigma^2 \mathbf{\Lambda}_i$  is the  $n_i \times n_i$  covariance matrix for the errors in group *i*.

Scientist Lot Batch Instrument



High Analyte Concentration



Figure 1: Example of desired binding profile in Gyrolab Viewer



Figure 2: Example of binding profile that indicates saturated detector signal



Figure 3: Example of binding profiles that indicate lower affinity (left) and higher affinity (right) between capture reagent and analyte









$$f(k;\lambda) = \Pr(X{=}k) = rac{\lambda^k e^{-\lambda}}{k!}$$

 $\lambda$ =-ln(N-neg/N-total)

 $\lambda = \mathrm{E}(X) = \mathrm{Var}(X).$  CV = std/mean =  $\sqrt{\lambda}/\lambda$ 

$$f_{Y|\mathbf{X},\mathbf{W}}(y;\mathbf{X},\mathbf{W},\boldsymbol{\beta},\boldsymbol{\vartheta},\boldsymbol{\alpha}) = \begin{cases} \pi + (1-\pi)p_{Y|\mathbf{X}}(0;\mathbf{X},\eta,\boldsymbol{\vartheta}), & \text{for } y = 0; \\ (1-\pi)p_{Y|\mathbf{X}}(y;\mathbf{X},\eta,\boldsymbol{\vartheta}), & \text{for } y \in \mathbb{N}^+, \end{cases}$$

For Fit-for-Purpose Design, Relaxed Criteria Might be Needed.



- Technical variability control (TVC) is the prerequisite for assess biological variabilities;
- A better understanding the deterministic and stochastic partitioning LBA holds the promise to foster validation accomplishment;
- Distinct formats my have specific criteria for validation as the sensitivity might be at the cost of technical variabilities;
- > LBA platforms should be carefully chosen, given the characteristics of analytes;
- Linear mixed model can be employed to dissect technical variability;
- Zero-inflated model can be applied to SIMOA platform;
- > ICH M10 Guideline may require science-based adaptation, namely, fit-for-purpose.



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