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ADA screening assay cutpoint *Introduction*

Anti-drug antibody (ADA) assays

- Aim
 - Find a cutpoint that allows to distinguish between ADA *positive* and *negative* samples

Common proceeding

- Define cutpoint by characterization of a *negative* sample population with a 95 % quantile:
 - 5 % of negative cases will be false positive;
 - few positive cases shall be missed.





ADA screening assay cutpoint *Ideal case*

- 1. Data show a **normal distribution**
- 2. Plate normalization leads to equal means and variances





3. Determination of **parametric cutpoint** based on mean and standard deviation





Roche ADA assays *Assay format*



 Diagnostic assay
 Small background aspired
 Highly specific capturing surface (e.g. streptavidin – biotin interaction) & high-quality assay components
 Highly specific binding of unspecific binding of biological matrix components



Roche ADA assays *Assay data*

Aspired low matrix effect leads to new challenges in data analysis

Small ODs close to instrument level Measuring samples that are negative by definition with very low technical noise Standard reader settings can lead to binned data due to number of decimals

- Data not continuous
- → Data show no normal distribution
 - Mostly neither normal nor log normal distribution – even after outlier exclusion
 - Rules out standard parametric and 'robust' methods for cutpoint determination





Roche ADA assays *Cutpoint determination*

Nonparametric cutpoint calculation

- **Due to skewed distribution** of data close to instrument level
 - Screening cutpoint empirical 95 % quantile
 - Confirmatory cutpoint empirical 99 % quantile (or even 99.9 %)



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Challenges with nonparametric cutpoints

- Limited sample size leads to strong influence of maximum value
 - Sample size between 3x15 (preclinical) and 3x100 (clinical)
 - Cutpoint can correspond to the maximum of all observed values potentially compromising robustness as based on only one sample
 - This can result in a deviation of the aspired percentage of false positive samples
 - Report "actual" quantile (e.g. 98 % quantile) otherwise claiming to be more strict than actually the case
- Resulting **cutpoint depends on applied software** as algorithms vary



Case study – mAb XY *Unexpected high amount of positives in study data*

Pre-dose data of 120 healthy volunteers (phase I study)



Screening cutpoint was statistically evaluated to lead to **5% false positives** in validation data.

Study data: **12.5 % screening positive** samples (15/120)

(10.8 % without two borderline cases with only one out of two replicates above cutpoint but mean below)

Percentage of positives unexpectedly high in set of pre-dose samples of healthy volunteers!



Case study – mAb XY *Re-evaluation of validation data*

1. Original approach (CP1)

12 outliers (x) were identified in the validation study data and **excluded** for screening cutpoint calculation.



Validation study data

- 50 samples (25 disease, 25 healthy)
- measured on triplicate plates



Case study – mAb XY *Re-evaluation of validation data*

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12 outliers (x) were identified in the validation study data and **excluded** for screening cutpoint calculation.

- However: biological not technical outliers !
- They reflect part of the negative population that we aim to characterize, and are therefore not to be excluded from screening cutpoint calculation (unless samples assumed to be positive).



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2. Re-evaluated approach (CP2)

No (biological) outlier exclusion



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- 50 samples (25 disease, 25 healthy)
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Case study – mAb XY *Re-evaluated screening cutpoint*

Back to study data



Original approach (CP1):

- 12.5 % screening positive samples (15/120)
- Percentage of positives unexpectedly high in pre-dose samples of healthy volunteers!



Re-evaluated approach (CP2):

- 4.2 % screening positive samples (5/120)
- Percentage of positive samples now in expected range



Case study – mAb XY *Re-evaluated screening cutpoint*

Change of validation parameters after re-evaluation

Validation parameter	Validation result CP1	Validation result CP2
Mean NC signal (OD) during validation runs	0.0374	0.0374
Normalization value (additive normalization)	0.006	0.0339
Assay sensitivity	0.288 ng/mL	1.64 ng/mL
Drug tolerance factor	80	13
 ratio of drug concentration and lowest positive control concentration giving a signal above the cutpoint 	→ 250 ng/mL ADA can still be found with 20 µg/mL drug	→ 250 ng/mL ADA can still be found with 3.25 µg/mL drug



Validation data of **25 healthy samples**

1. Original approach (cCP1)

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2. Re-evaluated approach (cCP2) No outlier exclusion



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Pre-dose data of 120 healthy volunteers

• % confirmed positive samples

		Screening (OD)	
		sCP1 0.043	sCP2 0.071
		12.5 % (15/120) screening positives	4.2 % (5/120) screening positives
Confirmation (% inhibition) (% cCb 22 % 28 %	cCP1 25 %	9.2 % (11/120) confirmed positives	4.2 % (5/120) confirmed positives
	cCP2 78 %	4.2 % (5/120) confirmed positives	4.2 % (5/120) confirmed positives



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positive confirm both cC	erning es are led with CPs.	12.5 % (15/120) screening positives	4.2 % (5/120) screening positives
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'Conservative' approach (cCP1) chosen to mitigate risk of false negatives

Case study – mAb XY '*Real' positive samples*



Patient 1: Expected PK profile Patient 2: PK decrease & ADA increase (patients from same dose group)

Clinical on-treatment study data of patients



"Real" positive samples ADA signal in different range

Challenges for the determination of cutpoints *Summary*



Lack of normal distribution which hinders usage of "standard" methods

Roche

- Imprecise determination of empirical quantile depending on sample size
- Strong influence of outlier treatment/interpretation on result

Statistics can offer only limited support

- Mainly for 'ideal' cases
 - But even then seemingly in irrelevant OD range
- Solution more on biological / experimental level ?
 - Looking for and assessing different new approaches
 - Increased background, ...
 - Potentially go via positive controls
 - As actual positives seem to lie in completely different range anyways



Challenges for the determination of cutpoints *Acknowledgements*

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THANK YOU FOR YOUR ATTENTION! ANY QUESTIONS?



Doing now what patients need next