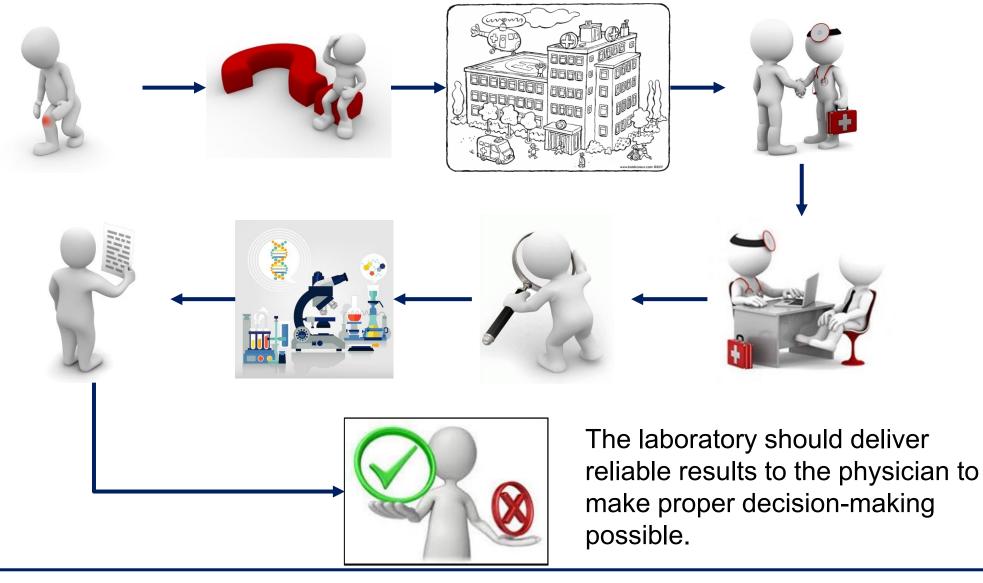
La logique bayésienne peut également optimiser l'analyse longitudinale des résultats d'EEQ.

Piet Meijer ECAT Foundation The Netherlands

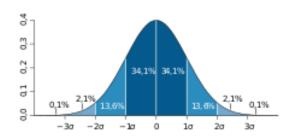


The story of the patient





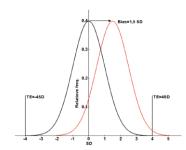


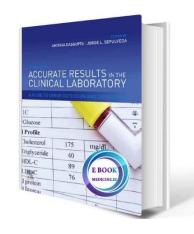


IMPRECISION

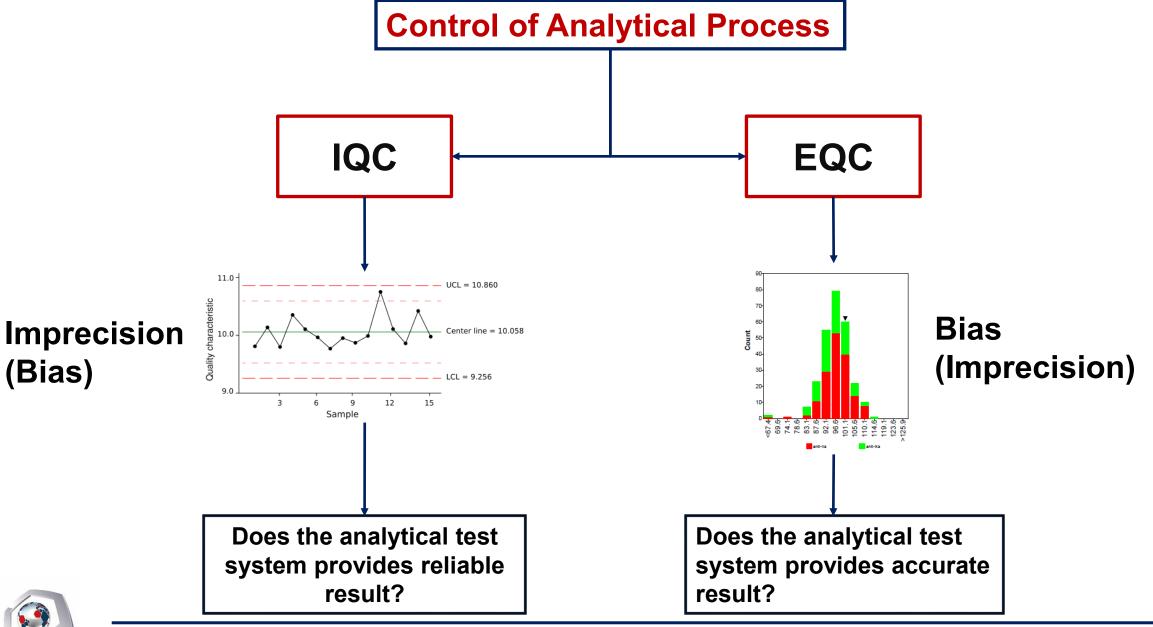
Laboratory Result

INACCURACY -











Control of Analytical Process

	EXTERNAL	INTERNAL		
PERMANENT	Variables of the analytic process that have a permanent character and defined by a third party	Variables of the analytic process that have a permanent character and defined by the laboratory		
VARIABLE	Variables of the analytic process that have a variable character and defined by a third party	Variables of the analytic process that have a variable character and defined by the laboratory		



Control of Analytical Process

	EXTERNAL	INTERNAL
PERMANENT	EQA	EQA
VARIABLE	IQC EQA	IQC EQA



IQC	Primary focus: Random errors (transient deviations), trends in deviation from the internal expected value, shifts
EQA	Primary focus: Systematic deviation from the "true" value = bias, trends and shifts, Random errors.





EXTERNAL QUALITY CONTROL TOOLBOX



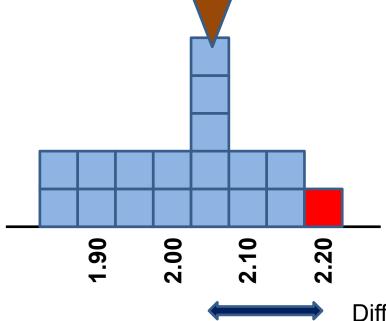
How can Bayesian statistics advance the longitudinal evaluation of laboratory performance?



The primary aim of EQA is to focus on the laboratory's analytical performance in comparison to its peers or to an accuracy-based reference system.

Miller, W.G. Clin Biochem, 2009; 42: 232-5.

Example:



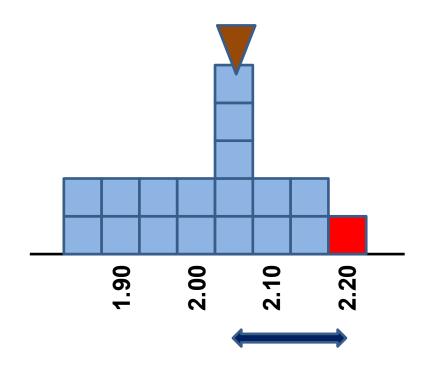


Difference between target and result participant

Performance evaluation

<u>Individual performance indicator</u>:

A numerical indicator representing the position of the individual lab result with respect to the consensus value – measure for accuracy



Examples:

Absolute difference

Relative difference

Z-score



Performance evaluation

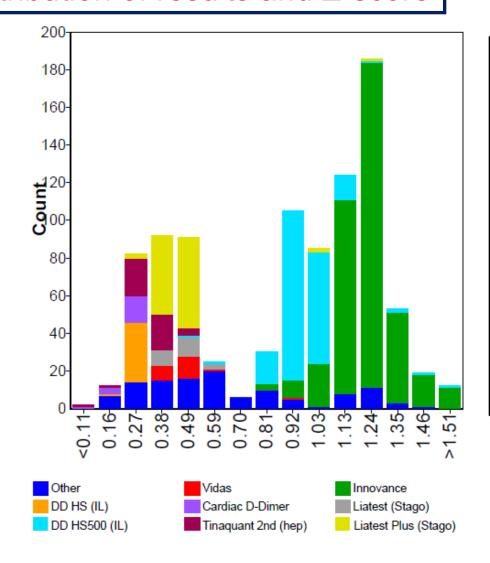
Z-score = Standard Deviation Index

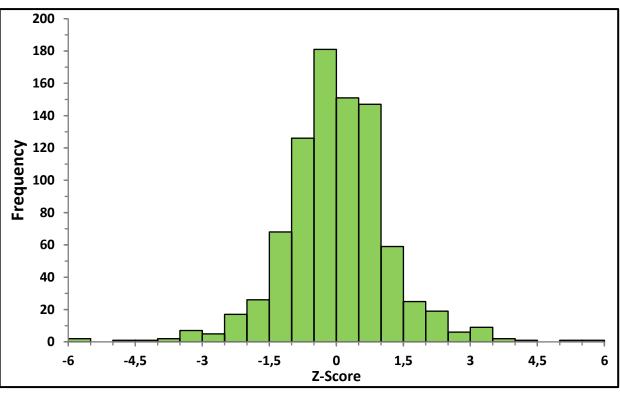
It describes the systematic error or bias of the individual result from the consensus value as a multiplier of the standard deviation.

The z-score is especially helpful for comparing participants' performance of peer groups with consensus values which are not standardised nor harmonised.



Distribution of results and Z-score





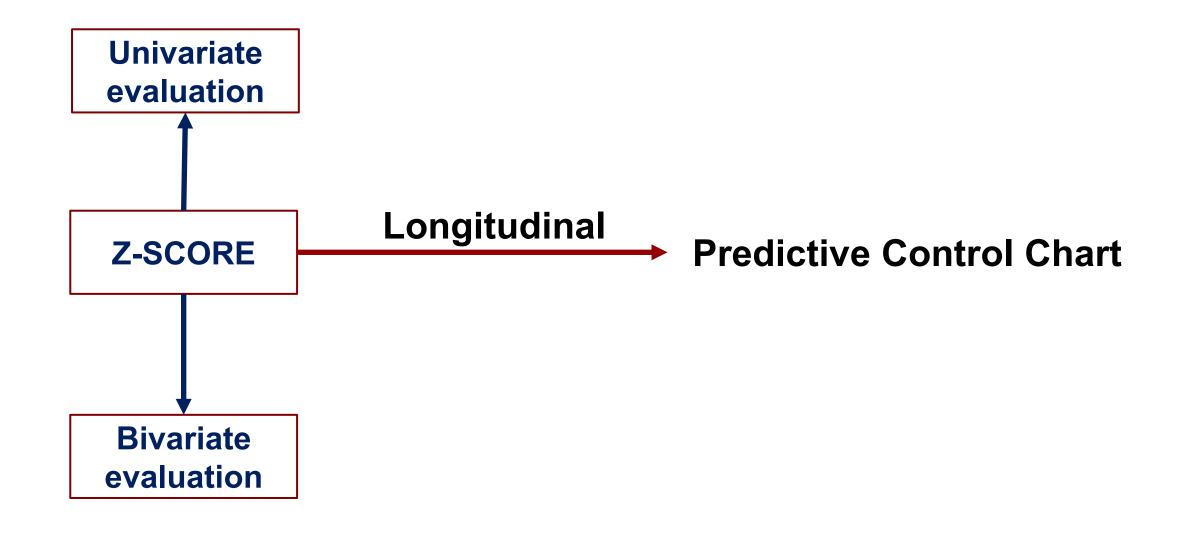




EXTERNAL QUALITY CONTROL TOOLBOX

- Univariate Z-score analysis (per survey)
- Bivariate Z-score analysis (per survey)
- Predictive Control Chart (PCC) on longitudinal Zscores (multiple surveys)



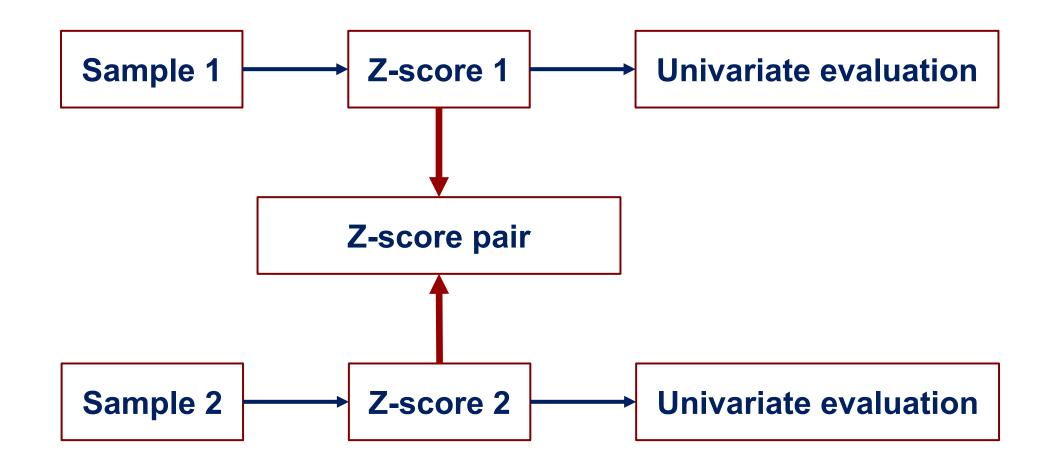






Univariate Z-score analysis









INTERNATIONAL STANDARD

17043

First edition 2010-02-01

Conformity assessment — General requirements for proficiency testing

Évaluation de la conformité — Exigences générales concernant les essais d'aptitude



Satisfactory No signal



-3 ≤ **Z**-score < -2

2 < Z-score ≤ 3





Z-score < -3

Z-score > 3
Unsatisfactory
Action signal





Single Survey

- Univariate analysis.
- Assessment of deviation from target value per sample.

Module	Parameter	Sample	Test System	Total	Assay	Method
Coagulation Factor - I	Factor XI	23.146	3	-0.52	-0.52	-0.10
		23.147	3	1.98	1.97	1.39
Coagulation Factor - I	Factor XII	23.146	3	-0.80	-0.79	-0.26
		23.147	3	-1.09	-1.08	-0.36
Coagulation Factor - II	Factor II	23.148	3	3.37	3.37	2.87
		23.149	3	1.10	1.10	-0.33
Coagulation Factor - II	Factor V	23.148	3	-2.14	-2.14	-3.24
		23.149	3	-0.36	-0.36	-1.97
Coagulation Factor - II	Factor VII	23.148	3	0.31	0.31	0.31
		23.149	3	-1.30	-1.30	-2.24
Coagulation Factor - II	Factor X	23.148	3	-0.87	-0.87	-0.27

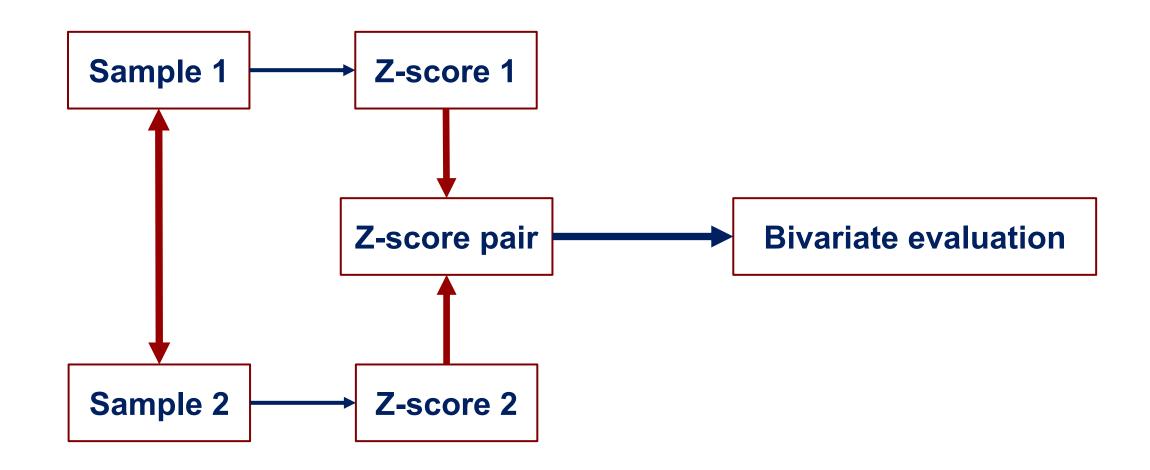
• Investigate potential causes for significant deviation.





Bivariate Z-score analysis



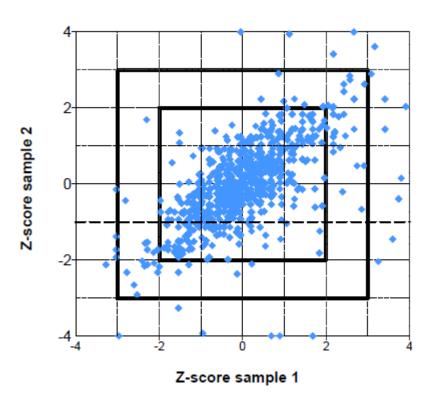




Classical Youden plot

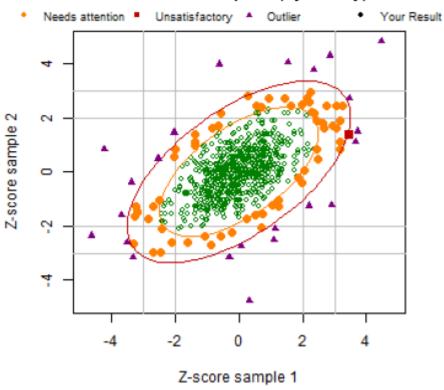


Z-score plot



Bivariate Z-score analysis

Bivariate Z-score plots (by survey)





Piet Meijer*, Frederic Sobas and Panagiotis Tsiamyrtzis

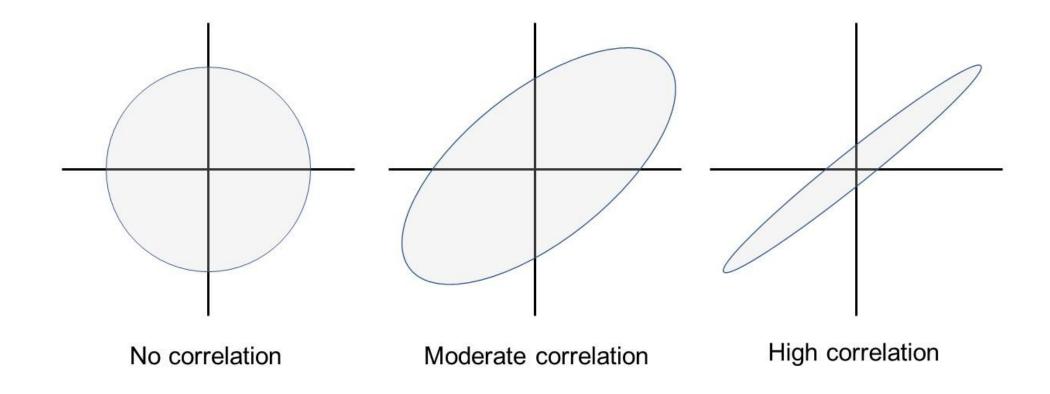
Assessment of accuracy of laboratory testing results, relative to peer group consensus values in external quality control, by bivariate z-score analysis: the example of D-Dimer

Conclusions: The use of the bivariate z-score analysis is of major importance when multiple samples are distributed around in the same survey and dependency of the results is likely. Important lessons can be drawn from the shape of the ellipse with respect to random and systematic deviations, while individual laboratories have been informed about their position in the state-of-the-art distribution and whether they have to deal with systematic and/or random deviations.



https://www.degruyter.com/document/doi/10.1515/cclm-2023-0835/html

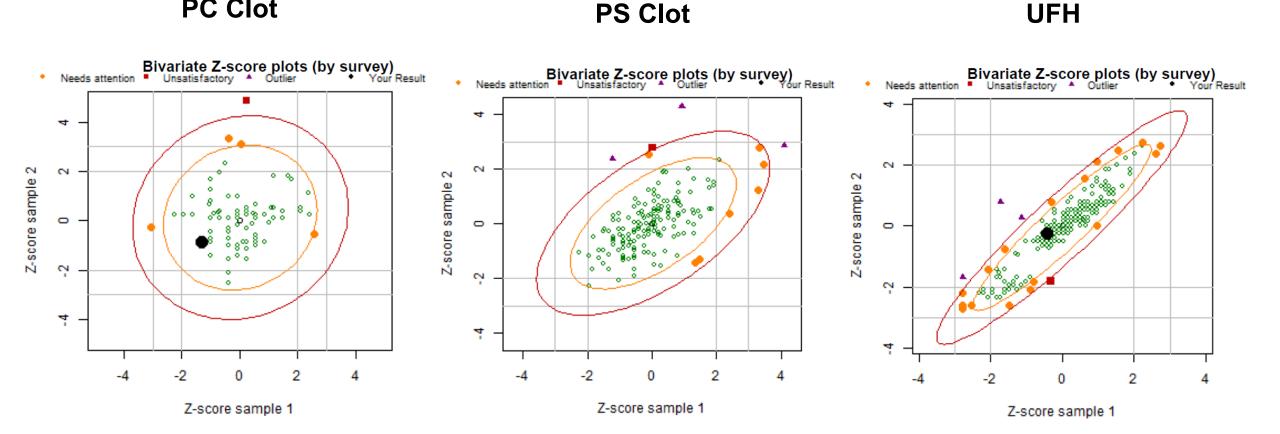
Bivariate Z-score analyse





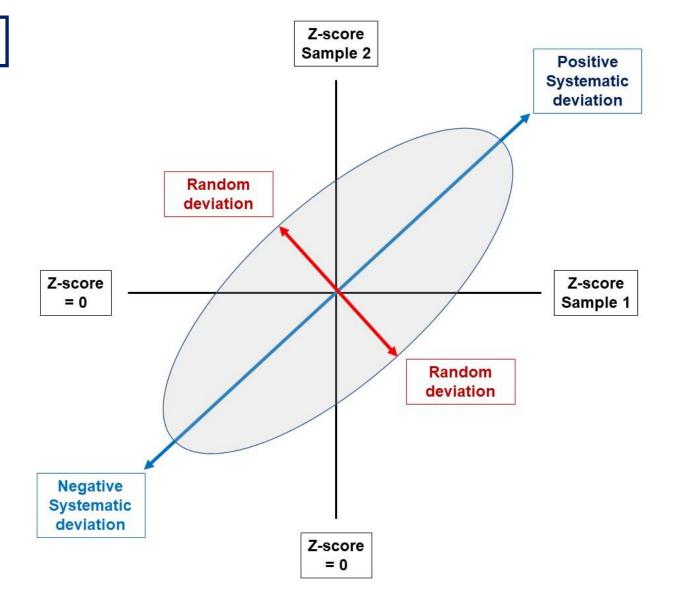
Examples Bivariate Z-score analyse

PC Clot





Bivariate Z-score analyse

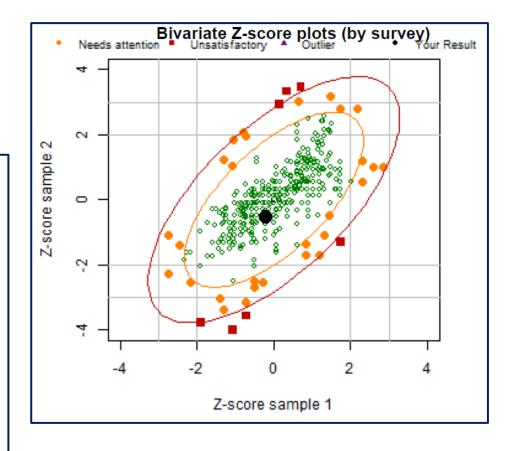




Single Survey

- Bivariate analysis.
- Position of your own Z-score pair in total distribution of Z-score pairs.

The use of the bivariate z-score analysis is of major importance when multiple samples are distributed around in the same survey and dependency of the results is likely. Important lessons can be drawn from the shape of the ellipse with respect to random and systematic deviations, while individual laboratories have been informed about their position in the state-of-the-art distribution and whether they have to deal with systematic and/or random deviations.

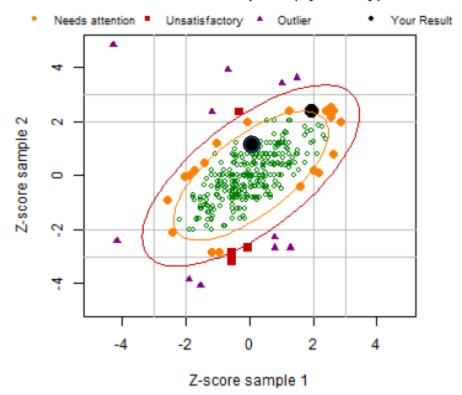




Example 1

Prothrombin Time / INR

Bivariate Z-score plots (by survey)



Z-score pair in the higher right quadrant corresponding to an increase of systematic error.

There was a delay in preventive maintenance in combination with the end of the current batch of thromboplastin.

After maintenance and also change of reagent batch the z-scores became 0.22 - 0.81. The instrument went back to the green area by eluding the increase of systematic error thanks to the z-scores bivariate analysis!!



Example 2

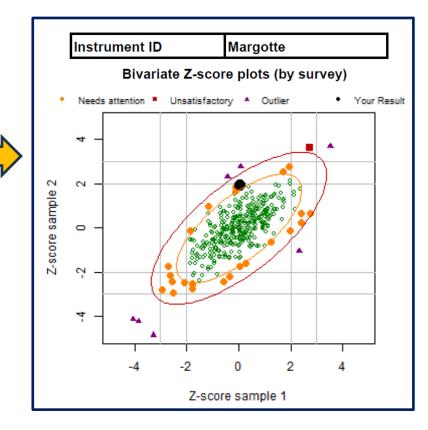
Fibrinogen

Reagent Reagent	Equipment Analysis

Screen - I	Fibrinogen (Clauss)	24.05	Margotte	0.04	0.04	
			Zébulon	0.49	0.49	
		24.06	Margotte	1.94	1.94	
			Zébulon	1.64	1.64	·

Screen - I	Fibrinogen (Clauss)	24.05	Azalée	0.34	0.34	
			Pollux	-0.67	-0.67	
		24.06	Azalée	0.99	0.99	
			Pollux	0.57	0.57	

	Instrument ID	Zébulon		Instrument ID	Azalée		Instrument I	D Pollux
	Bivariate Z-	score plots (by survey)		Bivariate Z-	score plots (by survey)		Biva	riate Z -score plots (by survey)
•	Needs attention Unsat	isfactory * Outlier * Your	Result •	Needs attention Unsat	sfactory * Outlier * '	Your Result .	Needs attention	Unsatisfactory A Outlier
	4 -	•		4 -	•		4 -	
ple 2	~		ple 2	~		le 2	7 -	
Z-score sample	0 -		ore sample	0 -		ore sample	0 -	
Z-sc	7	\$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$ \$	Z-score	7	10 K 20 K	Z-score	7	300 300
	4 - *			4 - **			4 - *	
	-4 -2	0 2 4		-4 -2	0 2 4		-4	-2 0 2 4
	Z	-score sample 1		Z	score sample 1			Z-score sample 1



Instrument	Inter-assay CV
Margotte	± 6.0%
Zebulon	± 4.3%
Azalee	± 4.3%
Pollux	± 4.3%



Z-score analyse per survey

Univariate

Bivariate

Individual evaluation of deviation from target value

Evaluation of position of Z-score pair within the full correlation of Z-scores pairs



Within laboratory

Between laboratories





Predictive Control Chart (PCC) on longitudinal Z-scores



Survey 1

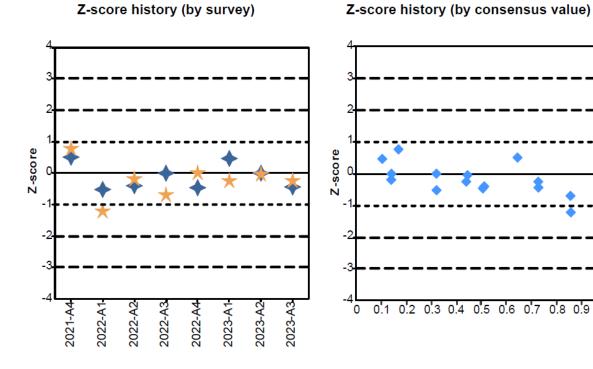
Survey 2

Survey 3

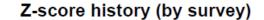
Survey 4

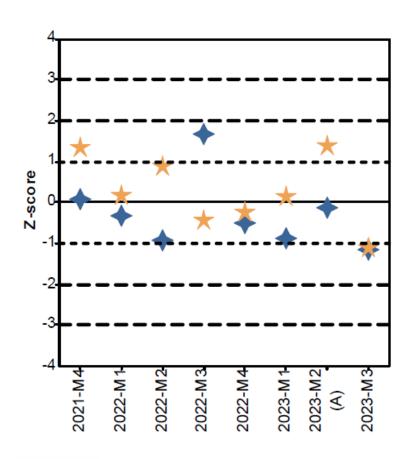
Survey 5

Trend in Z-scores between surveys

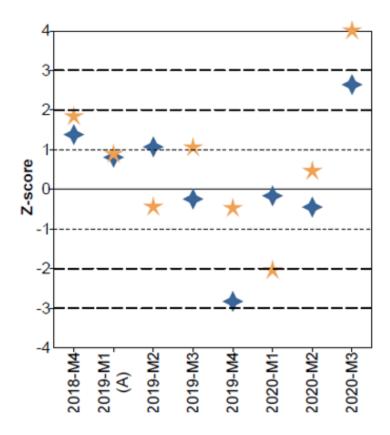




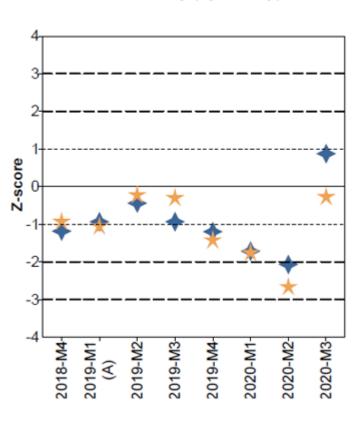




Z-score history (by survey)



Z-score history (by survey)





- Currently only a graphical representation.
- Personal interpretation about trends.
- Valuable for recognising systematic and / or random deviations.

HOWEVER!!!

A sophisticated modern evaluation based on Bayesian Statistical Process Control is lacking!

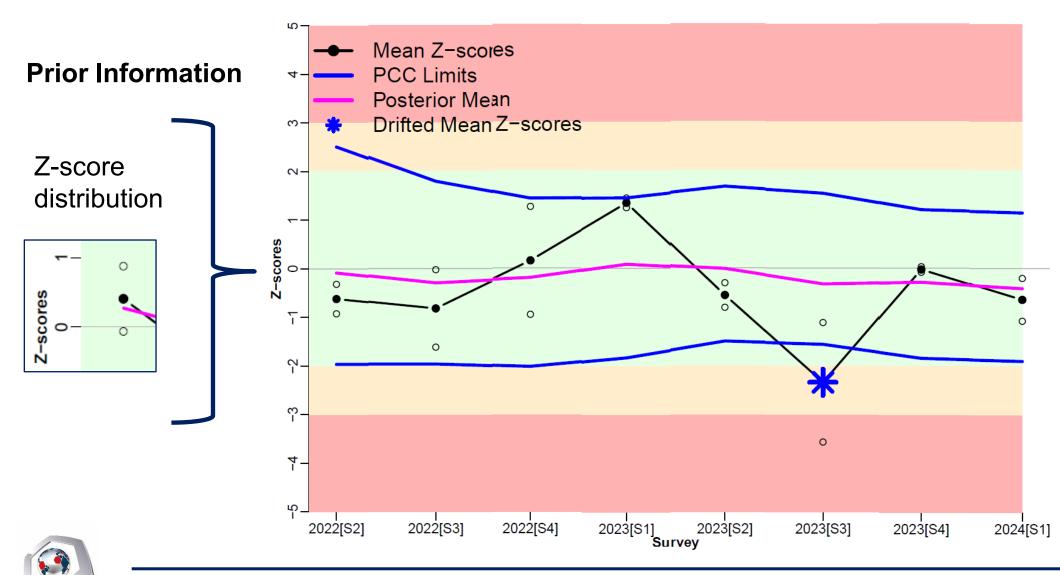
THE KEY OF BAYESIAN STATISTICAL PROCESS CONTROL:

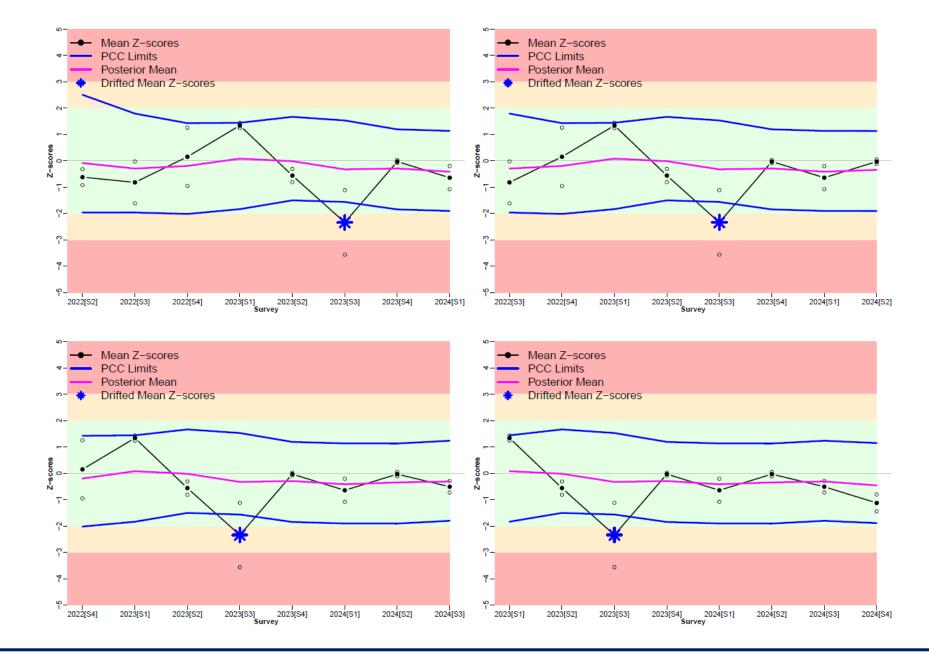
Use of prior information!

In EQA: prior Z-scores

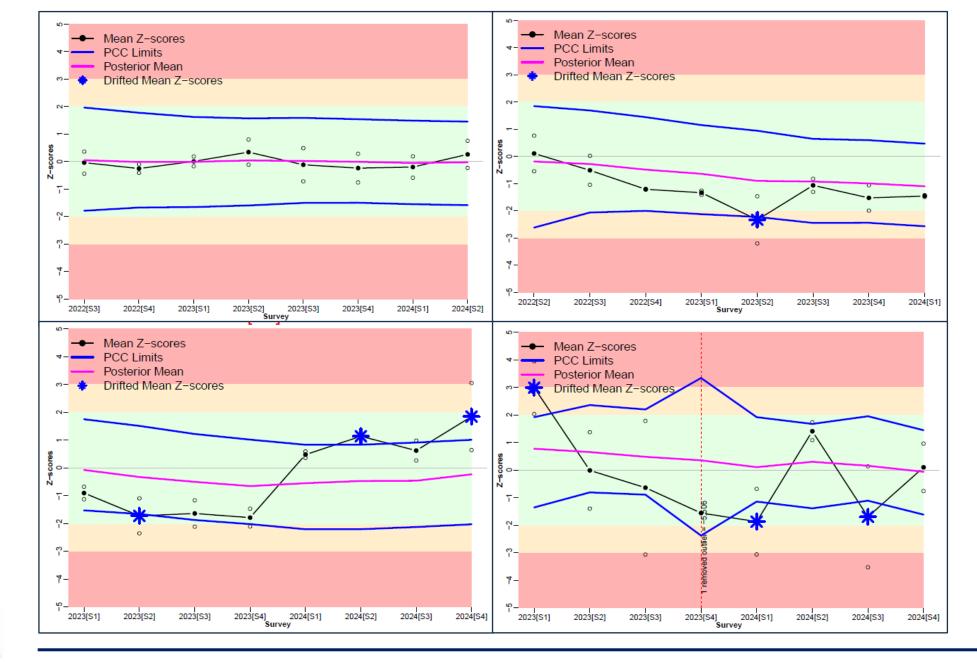


Bayesian-driven Z-score predictive control chart











Bayesian PCC

- Performance over time
- Trends
- Shifts
- Transient deviations



Within laboratory





Univariate

Individual evaluation of deviation from target value



Within laboratory

Bivariate

Evaluation of position of Z-score pair within the full correlation of Z-scores pairs



Between laboratories

Bayesian PCC

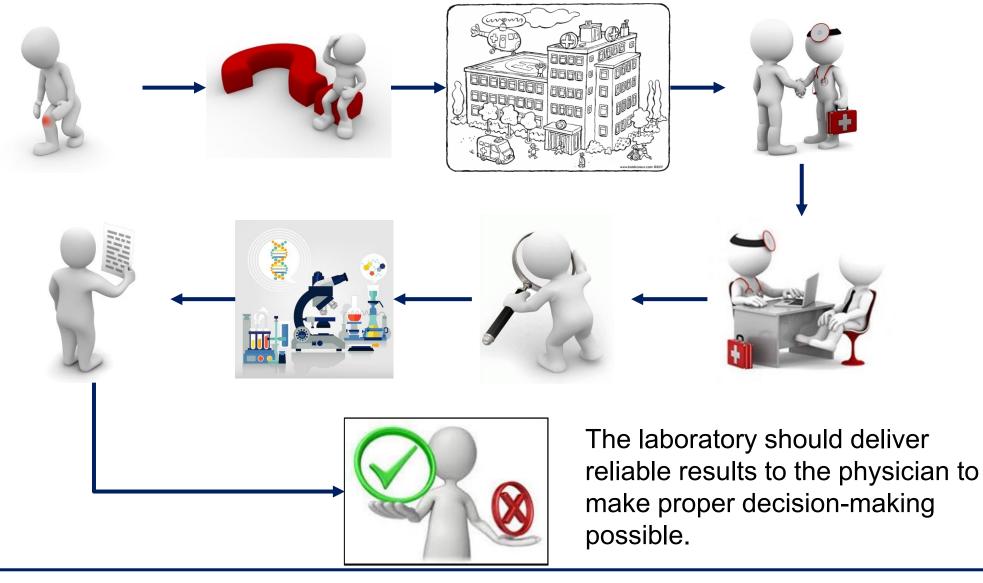
- Performance over time
- Trends
- Shifts
- Transient deviations



Within laboratory



The story of the patient





Acknowledgement

- Werfen
- Frederic Sobas (Lyon, France)
- Panagiotis Tsiamyrtzis (Athens, Greece)
- Participants in ECAT programme



Quality of coagulation testing for the benefit of the patient

Thank you for your attention

Merci beaucoup pour votre attention

