

# *Analytical Performance Specifications: the benefit for the laboratory!*



*Haverkate lecture  
10<sup>th</sup> ECAT Participants' meeting  
Sverre Sandberg, Noklus / EFLM  
Bergen, Norway*



# NEWS



## Data on trial of anticoagulant is to be reanalyzed after discovery that investigators used faulty device

Deborah Cohen

The BMJ

Data from the key phase III trial that allowed the oral anticoagulant rivaroxaban on to the market for the prevention of ischemic stroke in non-valvular atrial fibrillation is being reanalyzed because of the use of a faulty device.

The European Medicines Agency has also told *The BMJ* that it has launched an investigation into the trial.

The ROCKET-AF trial, published in the *New England Journal of Medicine* in 2011, compared rivaroxaban, which is

verified by laboratory tests. Alere told the Project on Government Oversight that it was working on a software upgrade to fix the inaccuracies.

Even before the recall, the ROCKET-AF trial had been criticized for the relatively short time that participants taking warfarin were in the correct therapeutic range of the drug. The implication of the criticism was that patients' use of warfarin was not adequately controlled, which, if correct, would have the effect

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### BACKGROUND

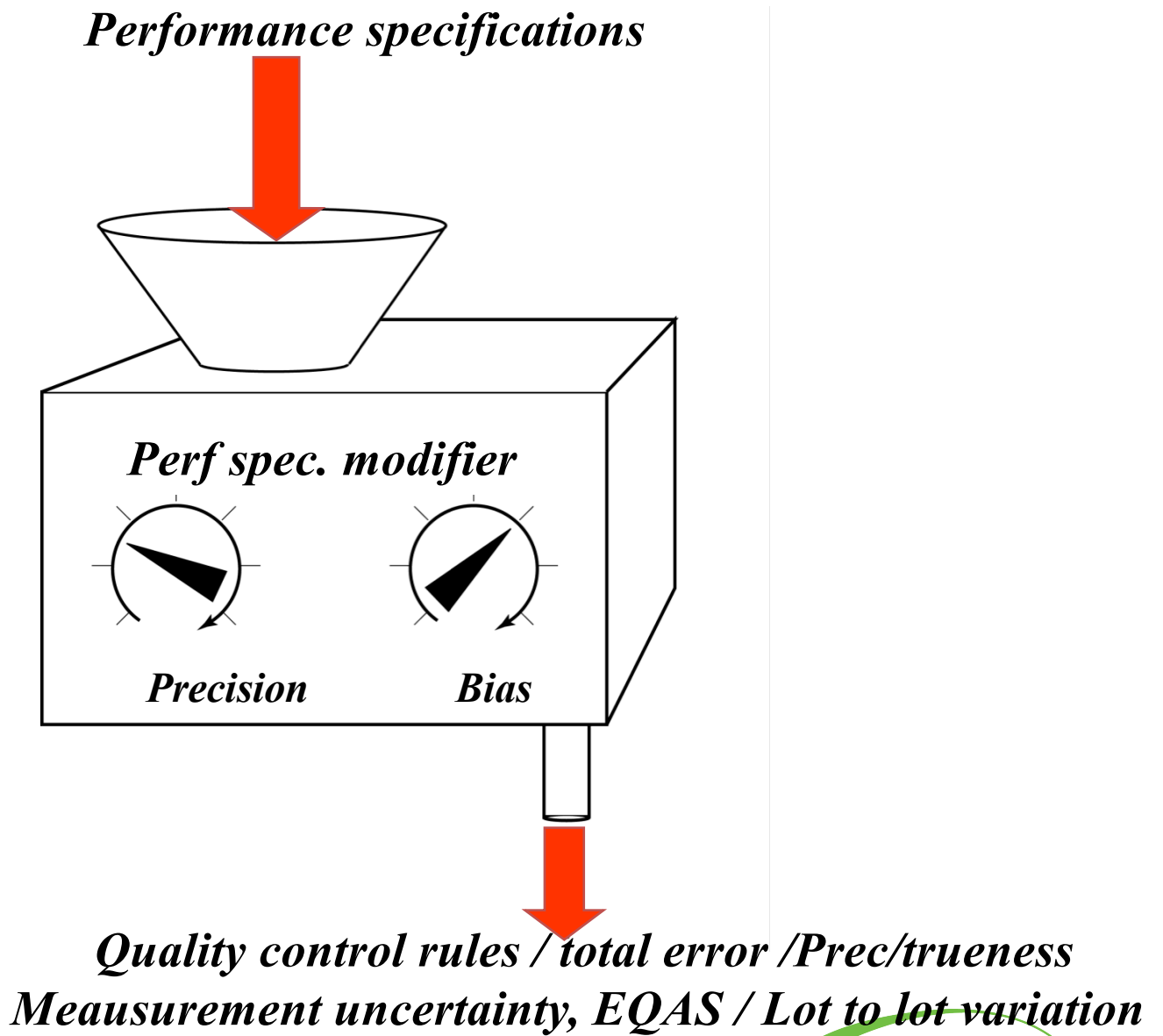
The use of warfarin reduces the rate of ischemic stroke in patients with atrial fibril-

From the Duke Clinical Research Institute



this case. They note that diagnostic devices “rarely require prospective clinical studies for clearance” and that devices that have been recalled can be used as predicates for the approval of new devices. Even the FDA acknowledges that “there are few performance standards... based on clearly defined scientific parameters.”

*The importance  
of  
Performance  
specifications*





European Commission  
Joint Research Centre  
**IRMM**  
Institute for Reference  
Materials and Measurements



# 1<sup>st</sup> EFLM Strategic Conference Defining analytical performance goals 15 years after the Stockholm Conference

8<sup>th</sup> CIRME International Scientific Meeting

Milan (IT)  
24-25 November 2014

with the  
auspices of  **IFCC**  
International Federation  
of Clinical Chemistry  
and Laboratory Medicine



# Consensus statement

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Clin Chem Lab Med 2015; 53(6): 833–835

## Consensus Statement

Sverre Sandberg\*, Callum G. Fraser, Andrea Pignatelli, Per Hyltoft Petersen, Graham Jones, Wytze Oosterhuis, Per Hyltoft Petersen, Heinz G. Löffler, and Mauro Panteghini

**Defining analytical performance specifications:  
Consensus statement from the 1st Strategic  
Conference of the European Federation of Clinical  
Chemistry and Laboratory Medicine**

**CCLM 2015; 53 issue 6**

# Remember that



"All models are wrong, but some are useful."

"The best models are not necessarily the most useful models".

*George Box 1919-2013*

**Model 1. Based on the effect of analytical performance on clinical outcomes**

*1a. Direct outcome studies*

*1b. Indirect outcome studies*

**Model 2. Based on components of biological variation of the measurand**

**Model 3. Based on state of the art**



# Model 1. Based on the effect of analytical performance on clinical outcomes

This can, in principle, be done using different types of studies:

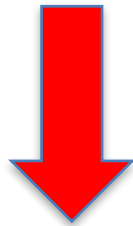
*Direct outcome studies* – investigating the impact of analytical performance of the test on clinical outcomes;

*Indirect outcome studies* – investigating the impact of analytical performance of the test on clinical classifications or decisions and thereby on the probability of patient outcomes, e.g., by simulation or decision analysis.

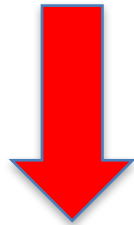
The advantage of this approach is that it addresses the influence of analytical performance on clinical outcomes that are relevant to patients and society. The primary disadvantage is that it is only useful for examinations where the links between the test, clinical decision making and clinical outcomes are straightforward and strong.

# Clinical outcome

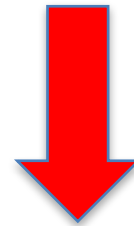
*Patient with a disease*



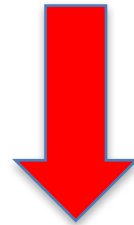
*Drug 1*



*Outcome 1*



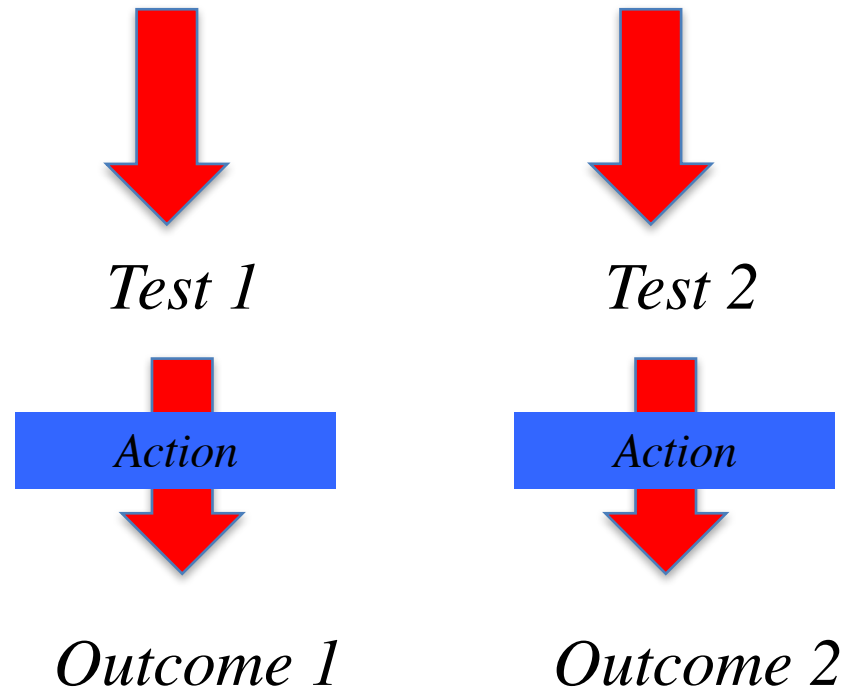
*Drug 2*



*Outcome 2*

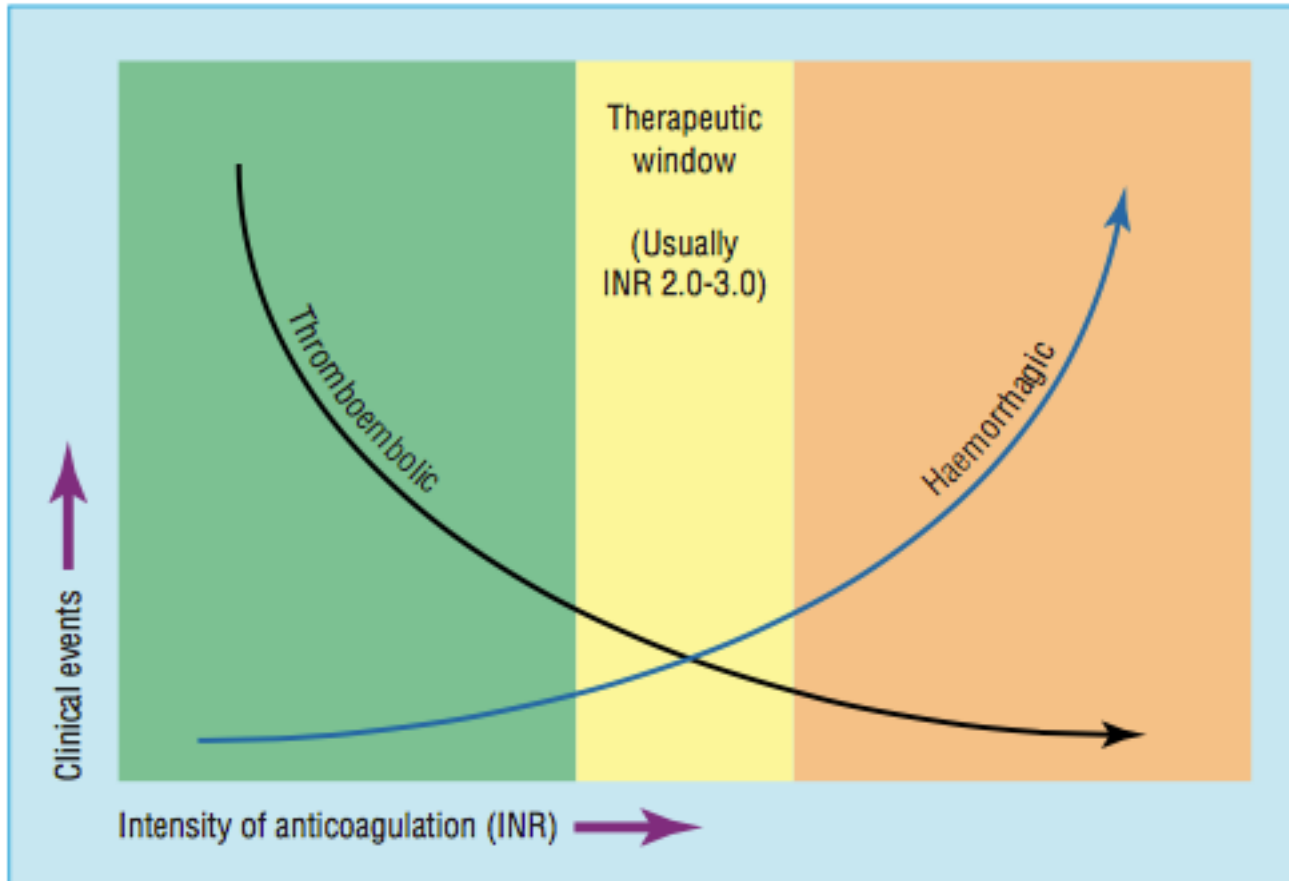
# Clinical outcome

*Patient with a disease*



# 1b:

## INR – balance between complications



So Model, type 1 is very difficult to perform and will take a lot of resources.

However type 1b is not that difficult – but we still have to concentrate on one clinical condition – for example to see how many false positive or false negative we can accept

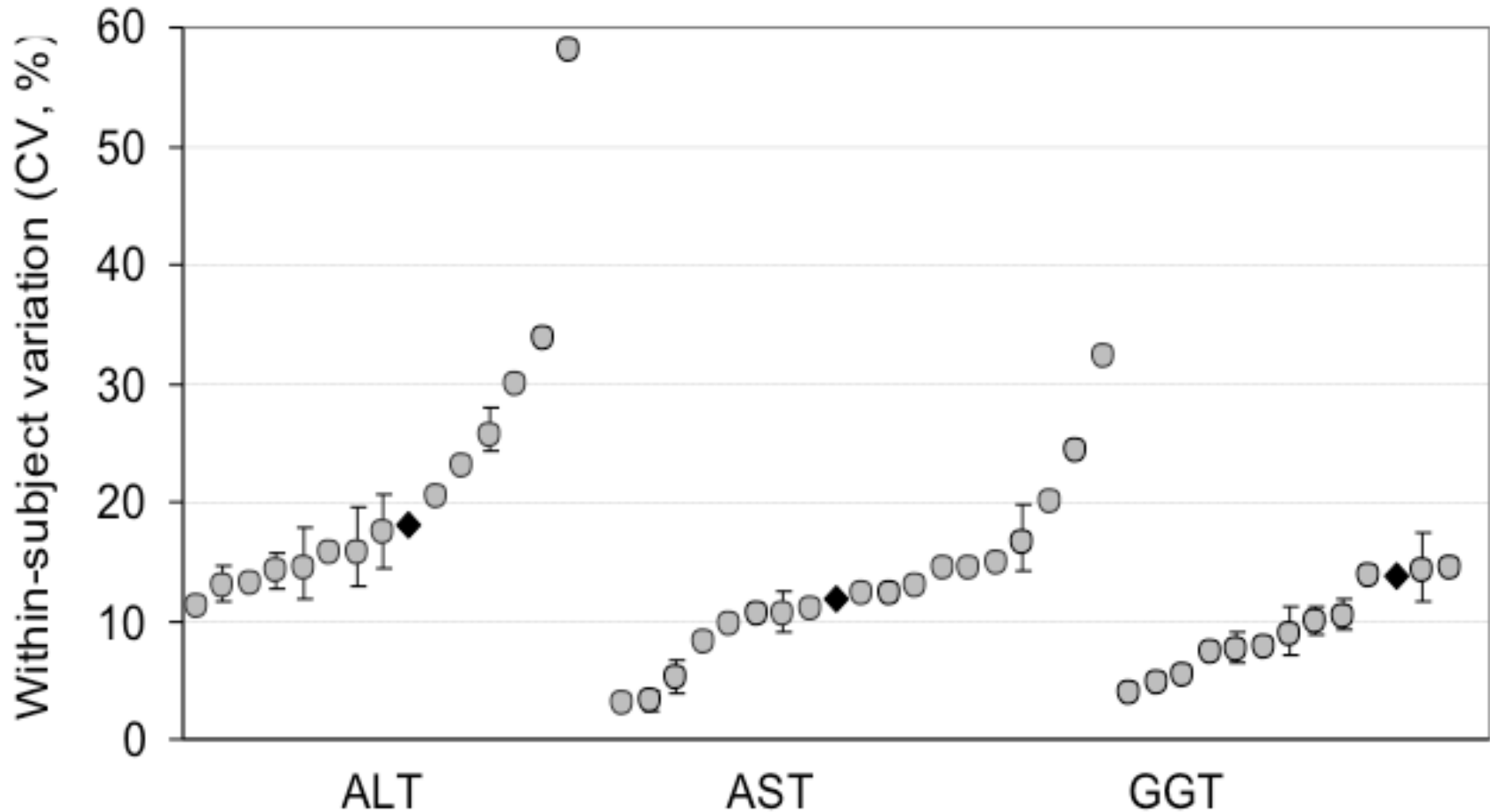
## *Model 2. Based on components of biological variation of the measurand*

This attempts to minimize the ratio of 'analytical noise' to the biological signal.

The *advantage* is that it can be applied to most measurands for which population based or subject-specific biological variation data can be established.

The *limitations* are that much of the current data/studies have not been carried out in a proper scientific way and therefore contains flaws.

# Within-subject variation



*Carobene A et al Clin Chem Lab Med. 2013;51:1997–2007.*



# *Model 3. Based on state-of-the-art*

This relates to the highest level of analytical performance technically achievable. Alternatively, it could be defined as the analytical performance achieved by a certain percentage of laboratories.

The advantage of this model is that state-of-the-art performance data are readily available. The disadvantage is that there may be no relationship between what is technically achievable and what is needed to minimize the ratio of 'analytical noise' to the biological signal or needed to obtain an improved clinical outcome.

# Explanatory notes

- It should be noted that the three models use different principles.
- The hierarchy assumes that high quality studies or data are available for each model.
- Proposed analytical performance specifications should therefore always be accompanied by a statement of the rationale, the source and the quality of the evidence behind the recommendation.

# Same measurand – different performance specifications?

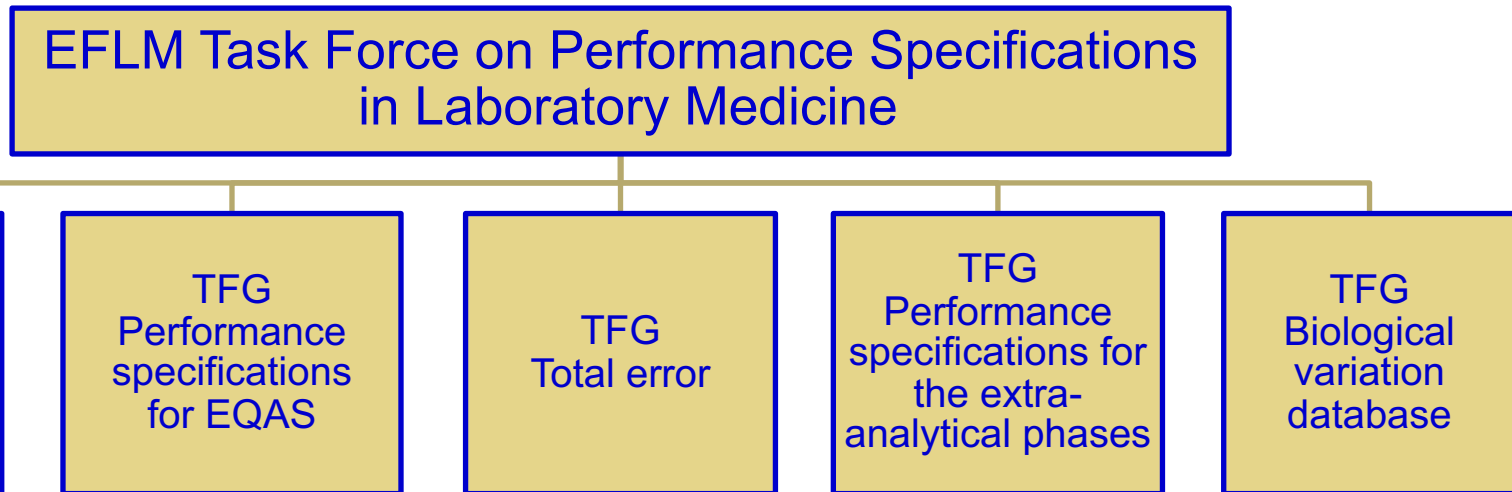
Some measurands can have different performance specifications defined when the test has multiple intended clinical applications.

This is difficult – if not impossible – for the laboratory to handle.

# EFLM Task Force on Performance Specifications in Laboratory Medicine

## An EFLM Task Force on Performance Specifications in Laboratory Medicine (TF-PS)

has been created to coordinate the activities of the Task & Finish  
Groups (TFG) established as outcome of the  
1<sup>st</sup> Strategic Conference



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### EFLM Task Force on Performance Specifications in Laboratory Medicine

TFG  
Allocation of  
laboratory tests to  
different models  
for performance  
specifications

TFG  
Performance  
specifications  
for EQAS

TFG  
Total error

TFG  
Performance  
specifications for  
the extra-  
analytical phases

TFG  
Biological  
variation  
database

# TFG: Allocate tests to different models

Possible principles:

Model 1 - outcome

Measurands that are mainly used for one purpose e.g. HbA1c / INR – can be used as general PS - in EQAS for example

Specific purposes e.g. Selfmonitoring INR, glucose, kreatinin. Can be used in guidelines for specific situations but not as general PS

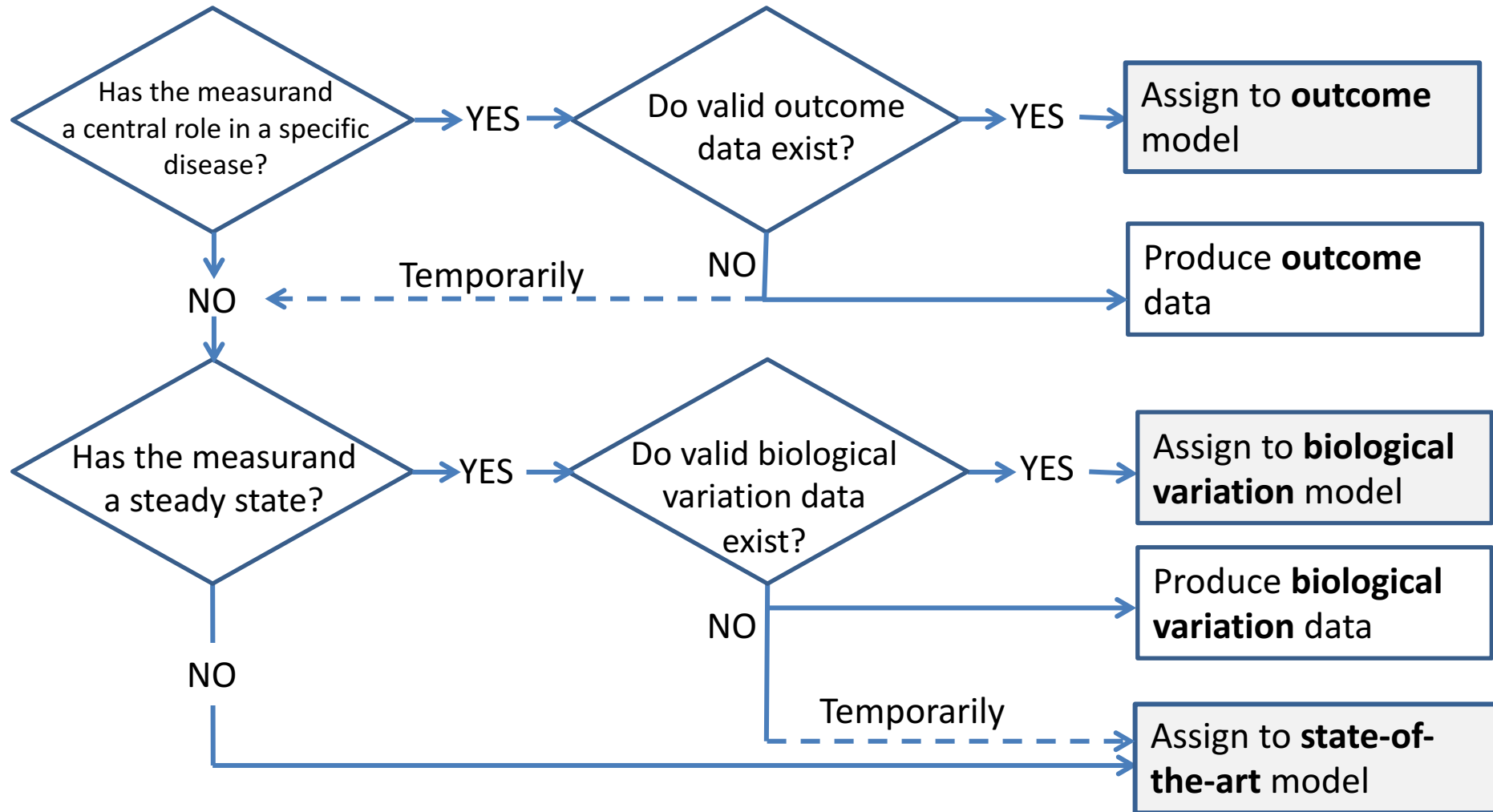
## Model II – biological variation

Measurands in which it is important for the body to keep a constant concentration- for example electrolytes, Hb, some hormones

## Model III – state of the art

Measurands where it is possible to obtain data from model I or model II;

# Model assignment workflow





# Biological variation database - cooperation -

Callum Fraser / Per Hyltoft Petersen

Analytical Quality commission in SEQC

EFLM – TFG

35 persons working with papers for kidney disease,  
cardiac markers, lipids, kidney diseases, diabetes.

# An improved biological variation database

1. To use a critical appraisal check list to evaluate papers on biological variation.
2. To collect samples to analyze most measurands on biological variation
3. To collect data on biological variation in a new database on [www.eflm.eu](http://www.eflm.eu)



## First report from the EFLM Task and Finish Group (TFG) for the Biological Variation Database (BVD)

*By Sverre Sandberg  
Chair of the EFLM TFG on Biological Variation*



# Critical appraisal check list

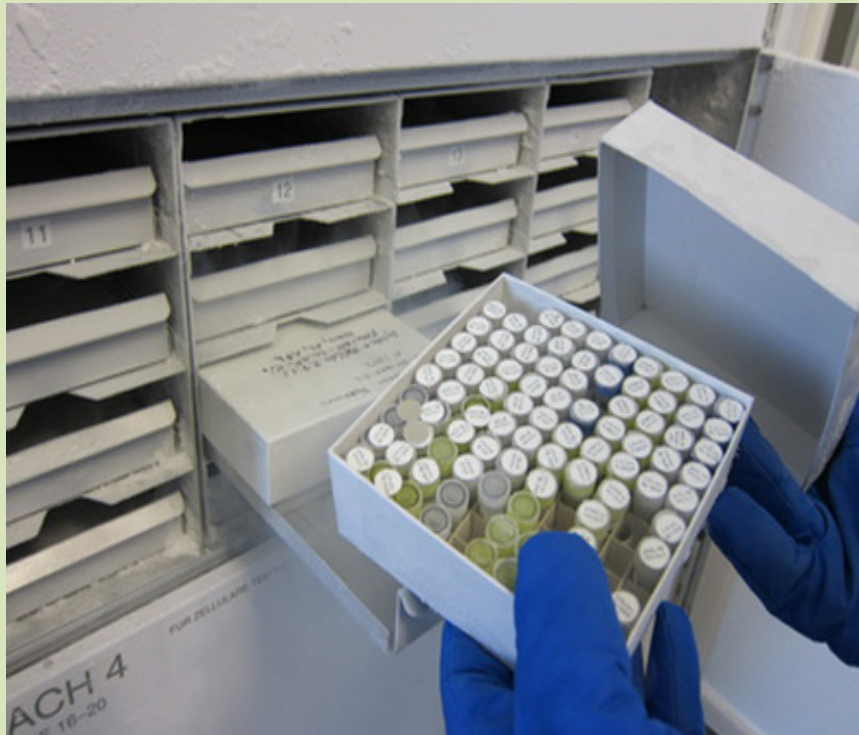
- ✓ Papers are categorized as A, B, C and D depending on their methodological quality, with category A papers indicating high-quality and D poor quality.
- ✓ The checklist contains 14 items, and
- ✓ 22 items will be extracted from each paper and presented in the database.
- ✓ Groups are established for different measurands

# “EFLM BV-WG project: new experimental BV data Samples collection from healthy volunteers for biological variation values update”

A multicenter study  
involving 6  
European  
laboratories



**Methods:** Each lab has processed and frozen the specimens strictly in agreement with the procedure reported in the protocol, in order to avoid pre-analytical variability.



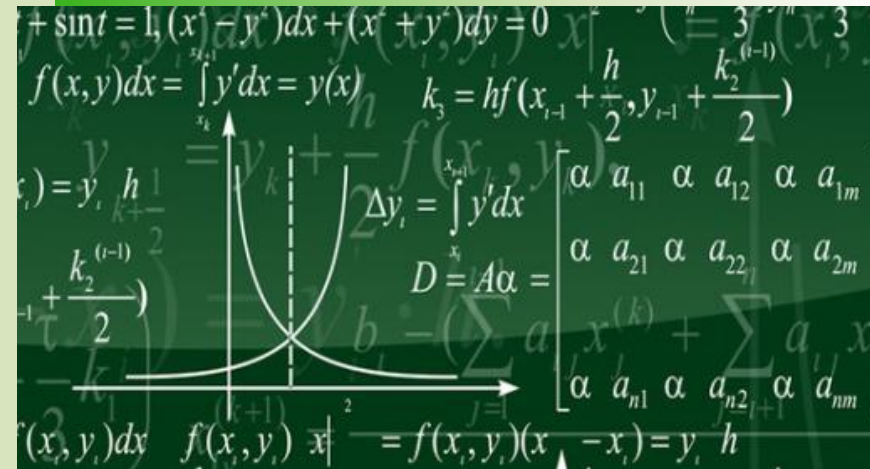
**Results:** A total of about 19.500 aliquots have been collected: 120 aliquots of serum, 40 of plasma EDTA, and 40 of plasma citrate for each subject. The samples, delivered to the coordinator lab (Milan, Italy), are stored at  $-80\text{ }^{\circ}\text{C}$  in a dedicated freezer.



### Future work:

A large number of tests, including enzymes, substrates, proteins, electrolytes, hormones, vitamins, tumor markers and coagulation tests will be performed in duplicate in the same analytical run.

The data will be treated according to recommended procedures for calculating BV including: outliers analysis at three different levels (replicate, samples and subjects), homoscedasticity of variances, normality of the distribution, and a nested ANOVA.



# When we have good studies: How to use the data?

When people say they use "biological variation" "clinical opinion" to set performance specifications, there is a big variation.

Why?

Because they use different (sub)models



# Different levels of models

- Principal models – what principles should we use to set performance specifications
- (The Milan conference)
- Practical models – when we have chosen the principles – which models should we then use to set practical performance specifications
- (A lot of work to be done)

# Second level models

First choose what of the three principal models to use,  
then choose a "submodel"

# Performance specifications

- what are you interested in?

Bias

Imprecision

Total error

Measurement uncertainty

# Outcome based models

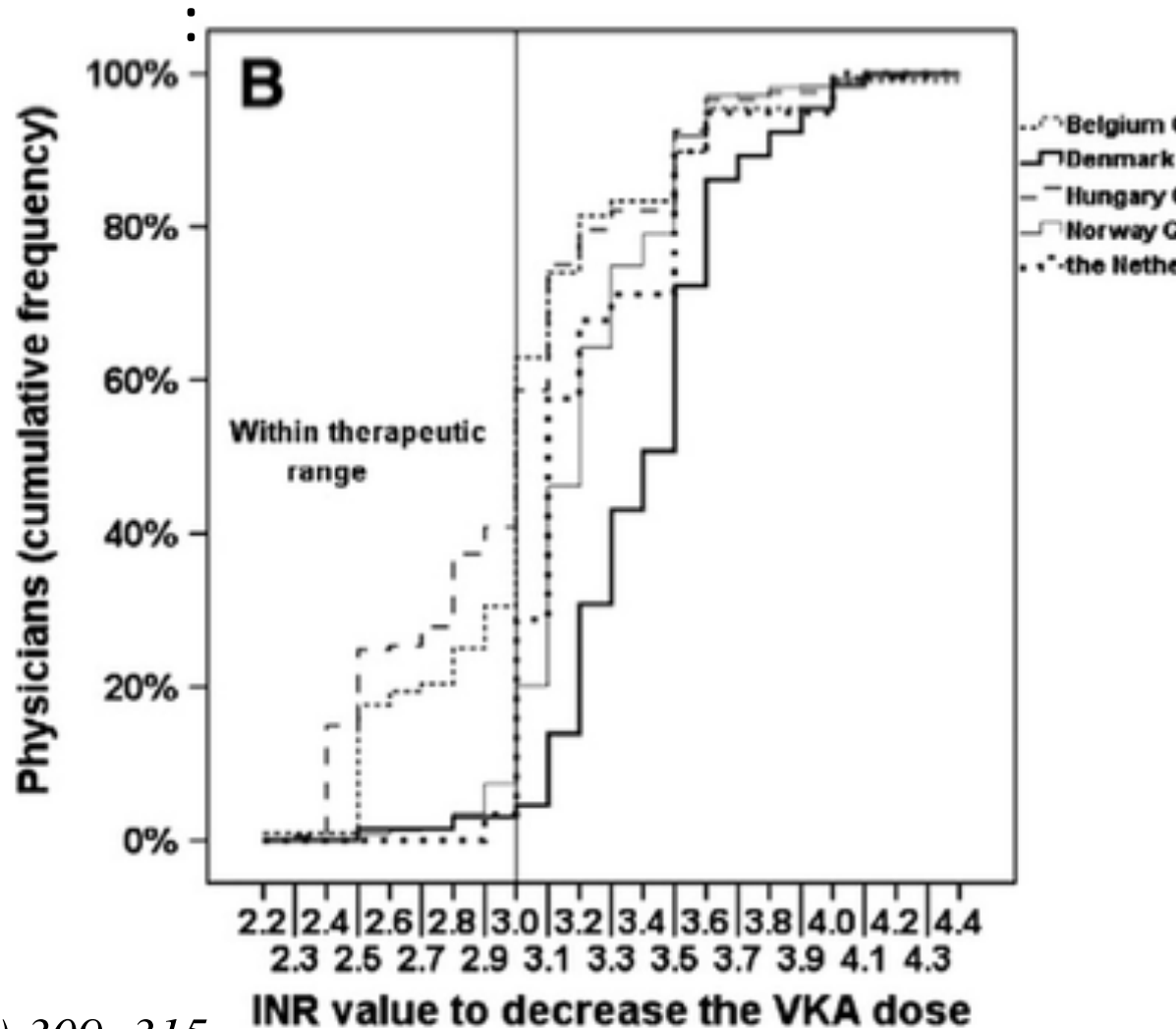
If outcome data should be used – should then the most strict data be used or the less strict?

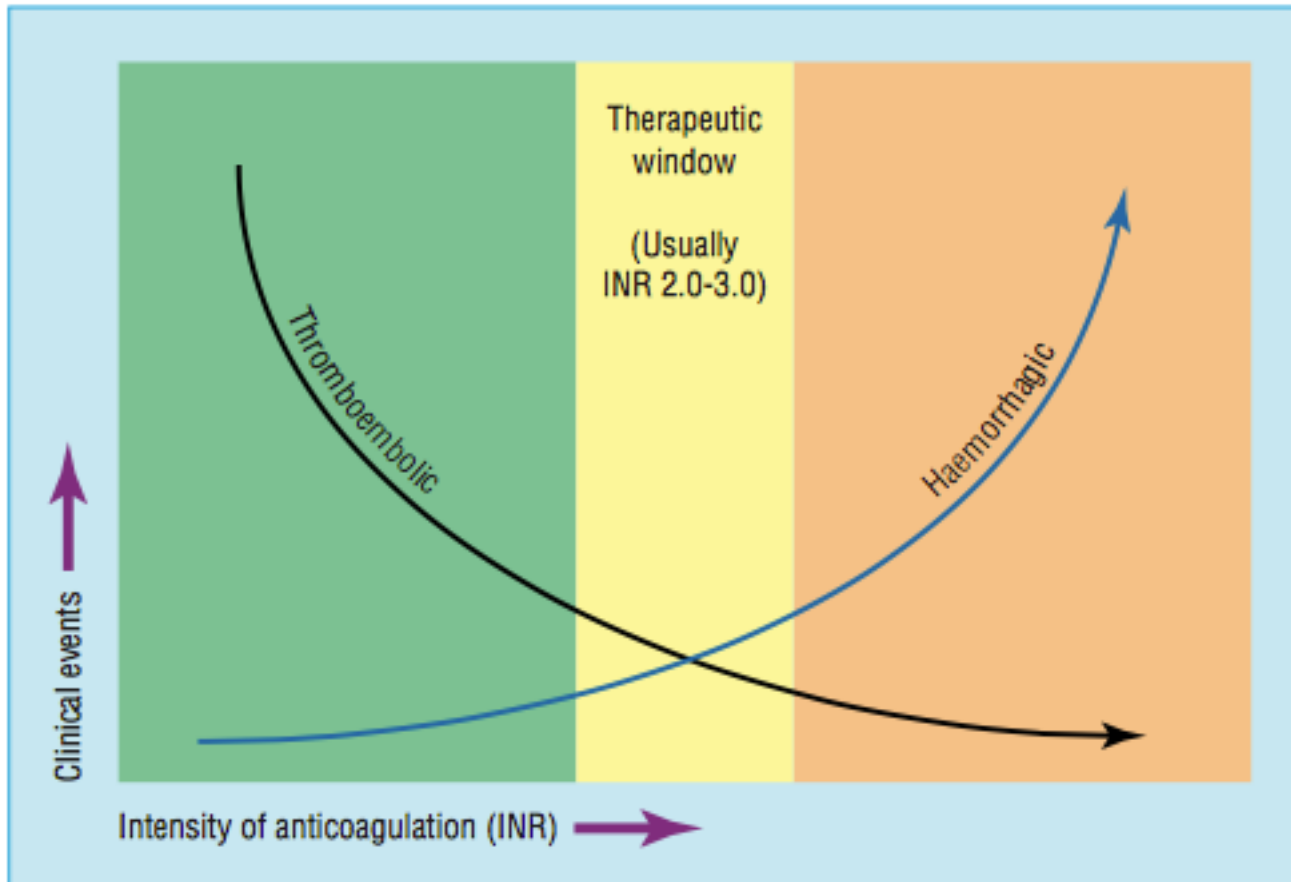
For clinical guidelines it will be easier to use outcome based specifications since they usually deal with one specific clinical indication.

# Example INR: Clinical algorithms for dosing - based on experience

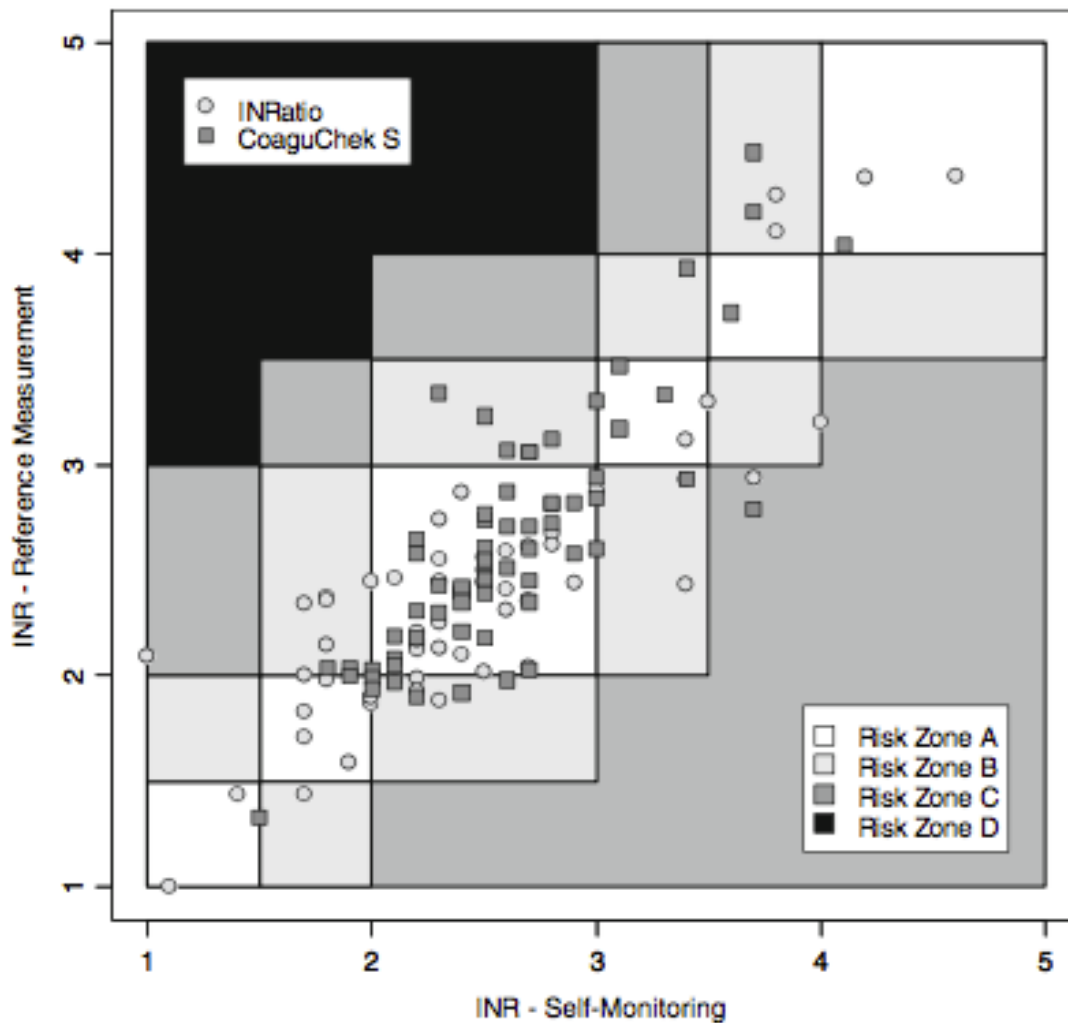
- Indirect experience with analytical quality of measurement
- Frequency of dosing
- Amount of drugs given
- Fixed limits (e.g. target limits)

Patient with atrial fibrillation and an INR of 2.3.  
INR values (representing critical differences) at which physicians would increase or decrease the VKA dose.





# Model 1b: Error grid





# Submodels for biological variation

- PS for monitoring
- PS for diagnosing
- PS for EQAs
- PS for method evaluation

# Models based on biological variation - relating to the reference interval -

Tonks: “TE in %”  $\leq \pm 100 * (0.25 \text{ times the “normal range”})$

$$\text{Harris: } s_A^2 + B^2 \leq 0.25 * (s_I^2 + s_G^2)$$

Gowans: to keep the common ref intervals within the uncertainty of a 120 person sample. Instead of the usual 2.5% outside a reference limit a maximum of 4.6% (based on the IFCC guide- line on reference values)

$$|B_{MAX}| \leq 0.25 * s_B = 0.25 * (s_I^2 + s_G^2)^{1/2},$$
$$SD_A \text{ max} < 0.52 * (s_I^2 + s_G^2)^{1/2},$$

Pragmatic model where you add analytical CV to within-subject or between subject variation (Oosterhuis et al/ Hoetzel et al – clin chem lab med 2015)

**Table 4.  $CV_A$  as measured in this study in relation to the recommen**

Variable	$CV_A$ : this study	Desirable $CV_A$ : this study		Desirable $CV_A$ : lit	
		Diagnosis	Monitoring	Diagnosis	Monitoring
General assays					
PT	1.3	2.8	1.3	3.0-4.2	1.2-1.3 [C
APTT	1.4	5.5	3.3	5.0-6.4	0.9-2.3 [C (14), Ch et al. (21
Fibrinogen	2.7	12.0	6.0	10.0-13.5	2.6-6.7 [de (13), Ch et al. (16
Thrombosis factors					
AT	1.4	5.2	2.2	4.4-6.1	0.6-2.7 [C et al. (17
PC-clot	5.5	14.0	4.5	7.8-11.1	1.2-4.0 [C (18)]
PC-chrom	1.7	12.1	3.8	–	
PS-act	4.1	14.6	4.0	13.7	3.8 [Wada
		11.1	3.7	5.4	1.5 [Nguye

$CV_A \leq 2\%$	Ricos et al. 1999 (healthy individuals)
$CV_A \leq 1.5\%$	Wada et al 2004 (healthy individuals)
$CV_A \leq 1.65\%$	Rudez et al 2009 (healthy individuals)
$CV_A \leq 4.5\%$	Lassen et al. 1995 (patients on OAT)
$CV_A \leq 4.5\%$	Besselaar et al. 2012 (patients on OAT)
$CV_A \leq 4.6\%$	Geest-Daalderop et al. 2009 (patients on OAT)
$CV_A \leq 5.4\%$	Geest-Daalderop et al. 2010 (patients on OAT)
$CV_A \leq 1,3\%$	De Maat et al 2016 (healthy individuals)

# INR Biological variation

$$CV_A = 0,5 CV_I$$

There is an important relationship  
between

analytical performance (bias and imprecision)

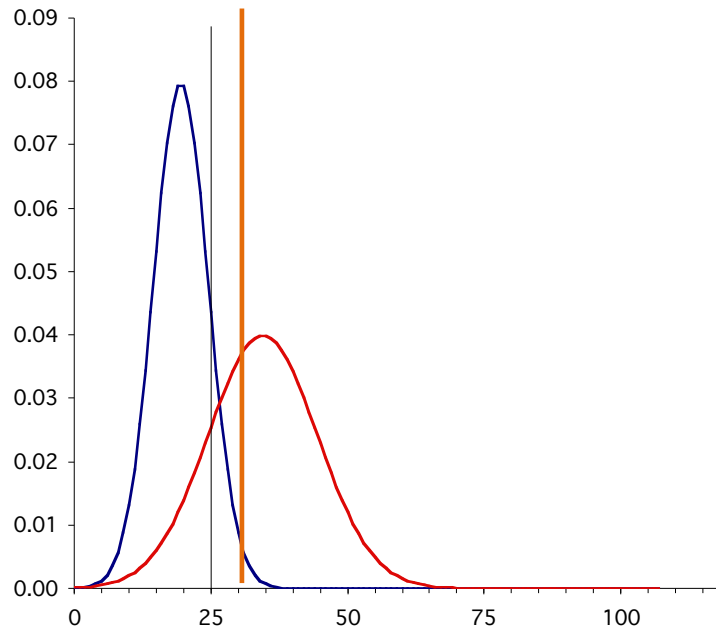
and

clinical performance (sensitivity and specificity /  
predictive values)

# Effect of analytical bias

|Bias 0.0

	ND	D	Cut-off	30
Mean	20.0	35.0	Sens	0.69
SD	5.0	10.0	Spec	0.98



# Conclusions

- Three main models using different principles
- Many "submodels" to calculate performance specifications
- Biological variation most used in laboratories and EQAS
- There is a profound relationship between analytical performance and clinical performance





Thank you