



**BIOMATH**

**Department of Applied Mathematics,  
Biometrics and Process Control**

# Optimal Experimental Design in River Water Quality Modelling

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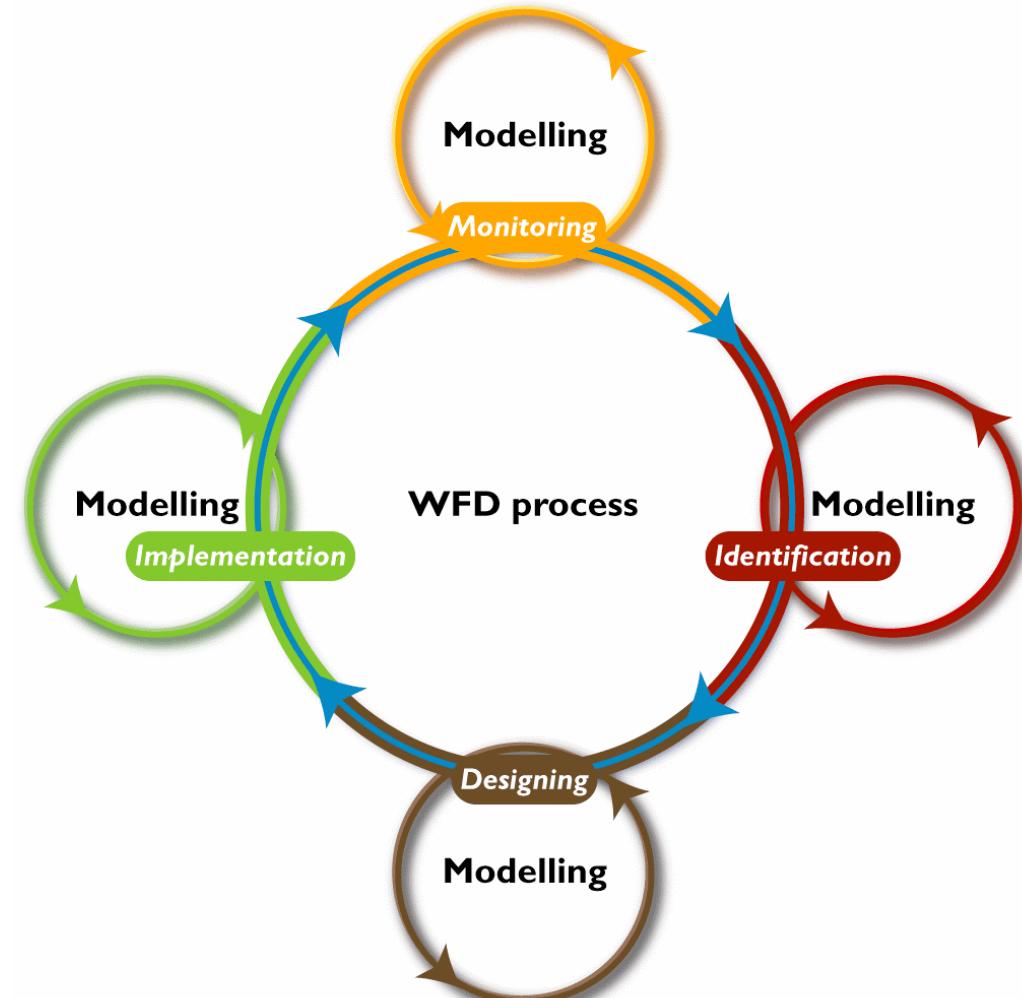
Peter Vanrolleghem

# Overview

- Introduction
- Aim
- ESWAT
- Dender case
- Methodology
- Results
- Conclusion

## Monitoring

- For the implementation of the WFD is monitoring an important step / for identification/evaluation: mostly ok.
- Models can play a role in every step. So also monitoring needed for building models for evaluation of future pollution abatement scenarios



# Monitoring

## Problems

- Costly measurements:  
eg. bod, suspended  
solids, micro-pollutants
- measurements with  
wrong frequency, in  
non-sensitive periods,  
on wrong places
- lot of effort to maintain  
large databanks
- Intensive measurement  
campaigns



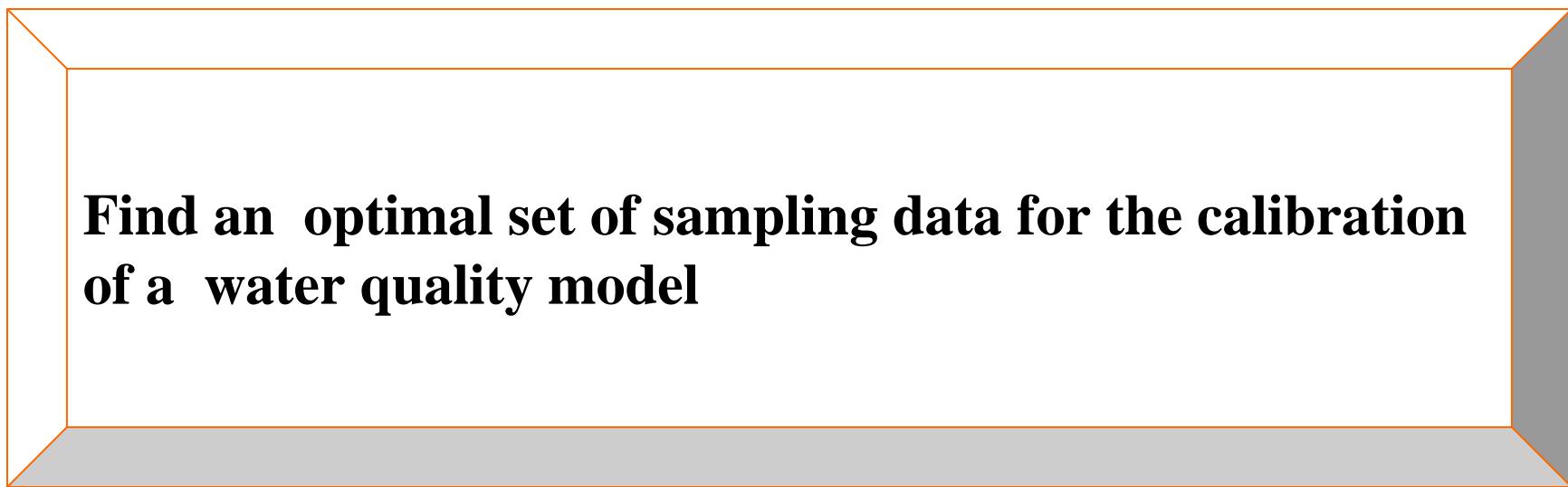
510000 12/18 BZV5 = 5,1 mgO2/L  
510000 12/18 Cd t < 0,2 µg/L  
510000 12/18 A.v.t. < 0 mg/L

510000 11/20 O2 verz = 79 %  
510000 11/20 CZV = 33,55 mgO2/L

# AIM

- Calibration of water quality models
  - identifiability of the model parameters
  - reliability of the model

=> depending on good measurements



Find an optimal set of sampling data for the calibration of a water quality model

## Integrated modelling tool

### SWAT98

- Catchment hydrology
- Agricultural pollution
- Constant point sources

### ESWAT (van Griensven)

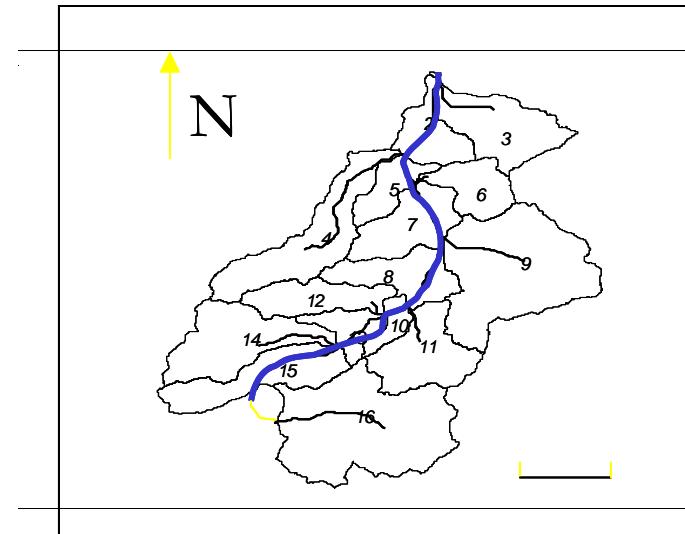
- Hourly time step (land and river hydrology)
- River water quality processes
- Dynamic point sources
- Urban drainage system



## Dender basin model

**MODEL: 700 km<sup>2</sup>**

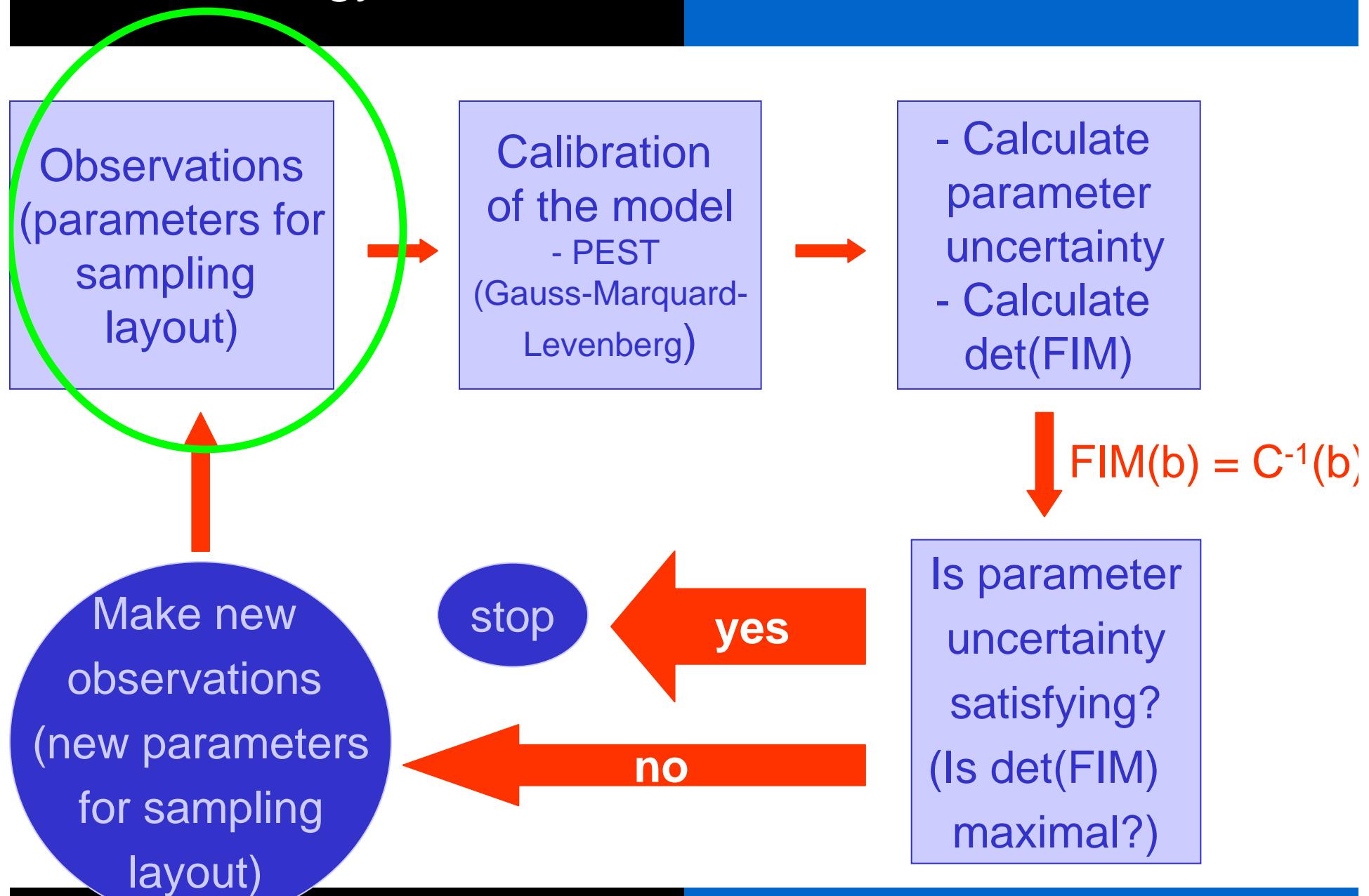
- 15 subbasins / 8 tributaries
- 80 HRU's (=combination land use and soil type)
- 10 point source locations
- 8 sluices



# OPTIMAL EXPERIMENTAL DESIGN

- Start with a calibrated model but with high uncertainty bounds caused by a lack of good measurements for a better calibration.
- Generate time series with the model for water quality variables
- See where, when, how many measurements/samples are practically feasible and define parameters of sampling layout

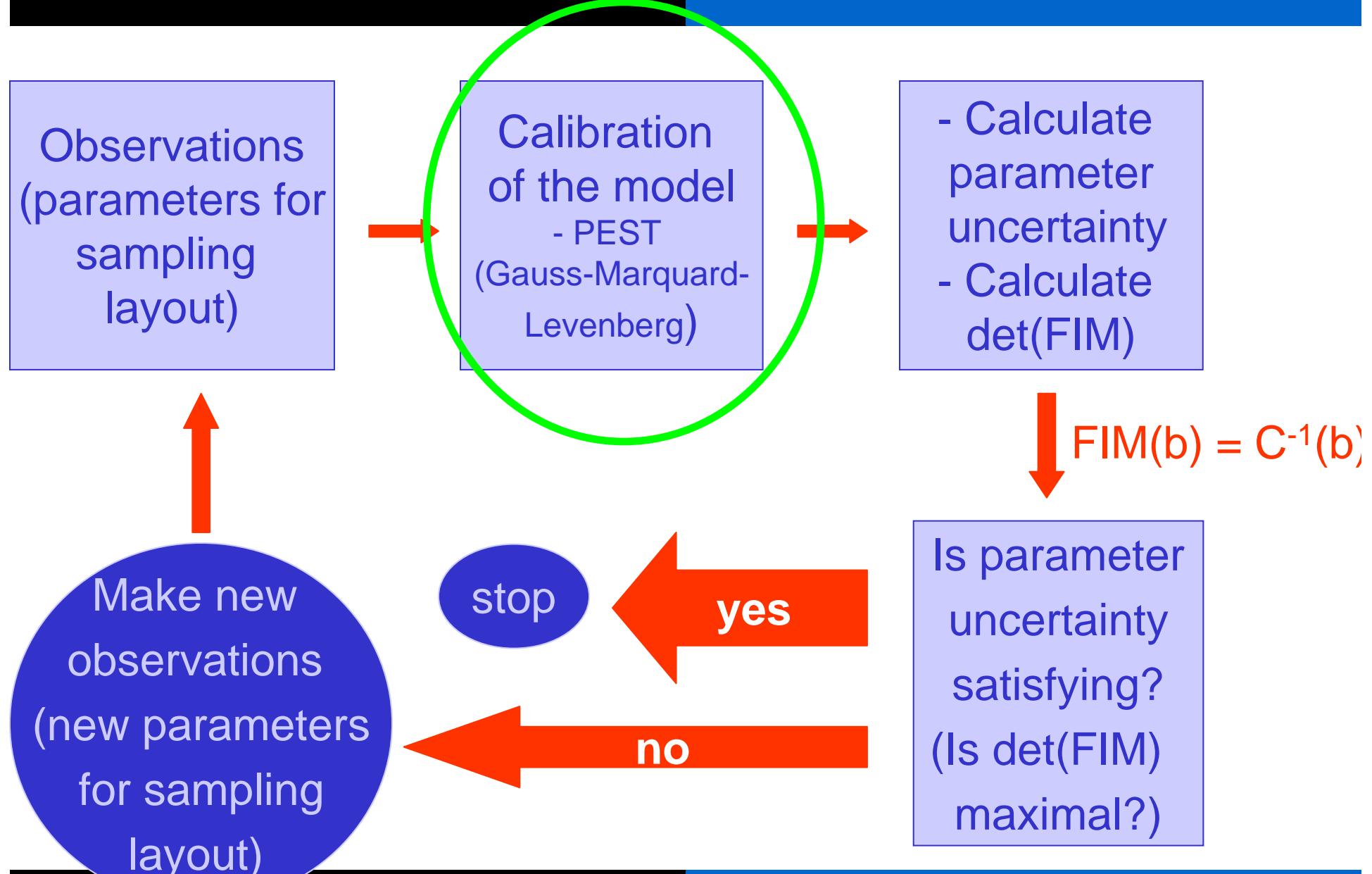
# Methodology



## observations

- No historic series of high frequency water quality data series available
- => Synthetic observation series generated by ESWAT + pseudo-random noise terms
- Noise terms: consistent with the accuracy of the measurement devices
- Parameters for sampling design: frequency, measurement place, period of the year  
e.g. every 12 hours, at the mouth + 5km more upstream, from 01/04 till 31/08

# Methodology



## CALIBRATION

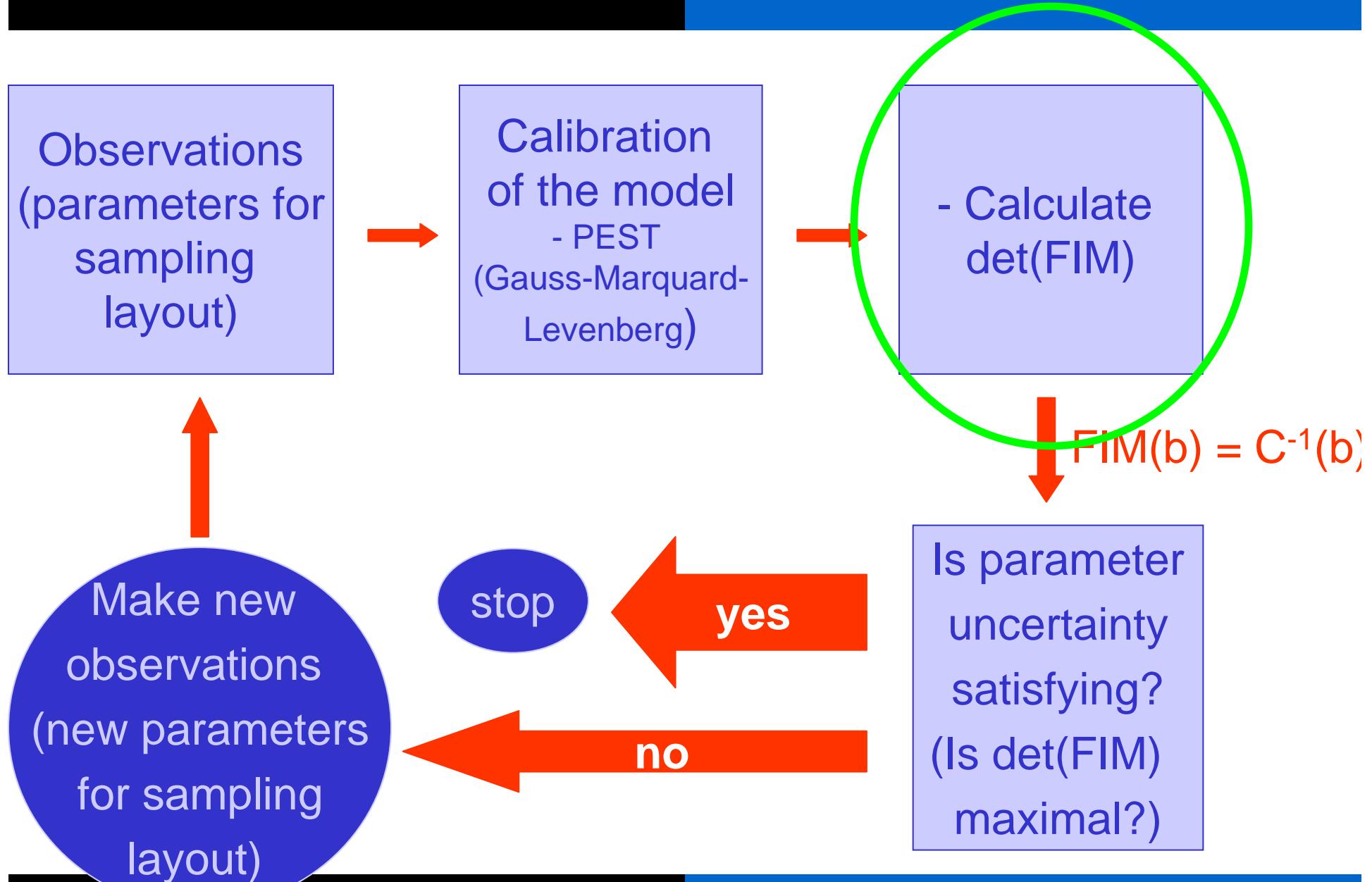
- Start calibration at a point ‘close’ to the former calibration (to avoid local minima)



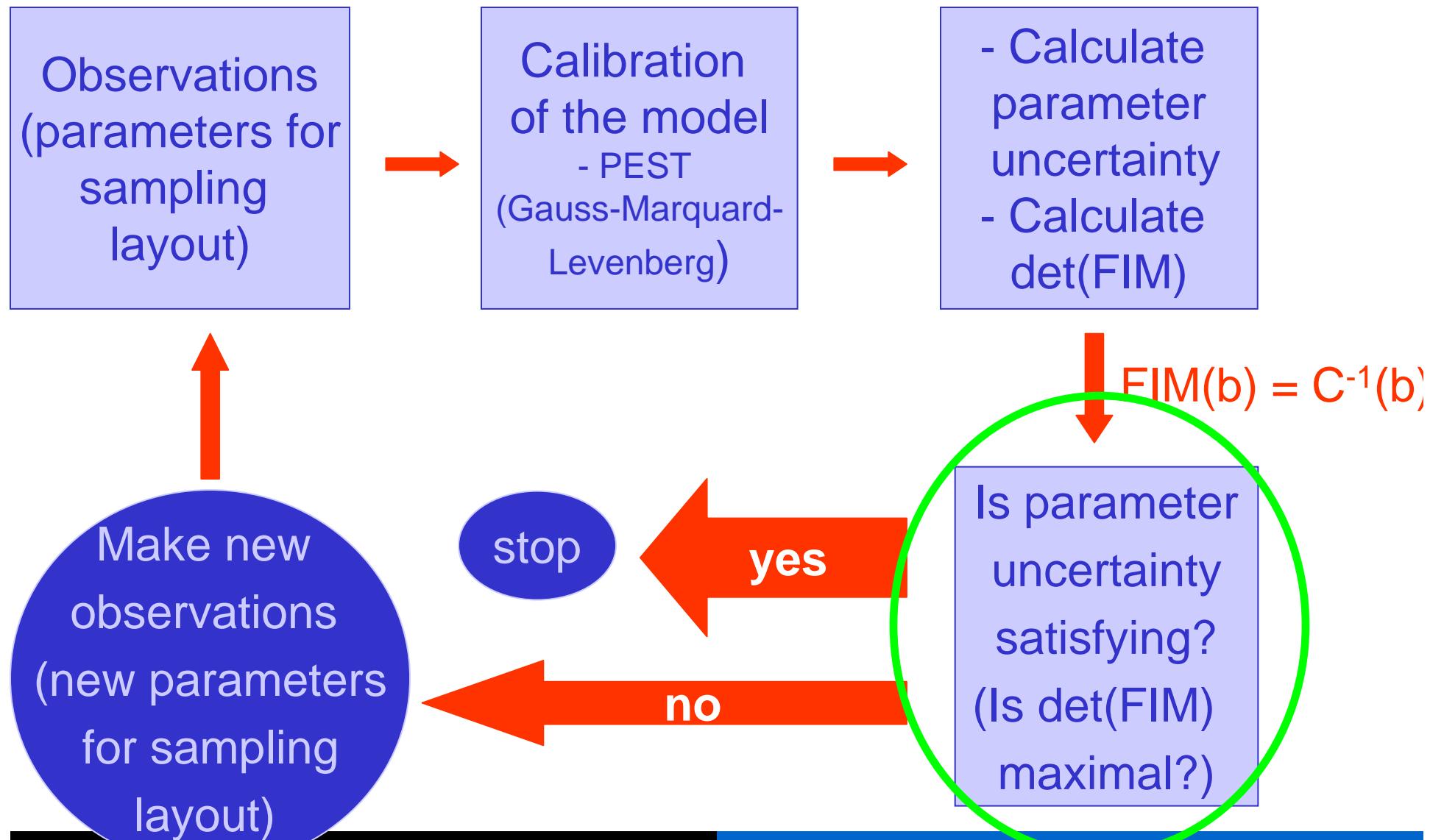
**During the calibration the covariance matrix and the uncertainty bounds on the parameters are calculated**

- PEST: uses Gauss-Marquardt-Levenberg

# Methodology



# Methodology



# D-optimal design technique

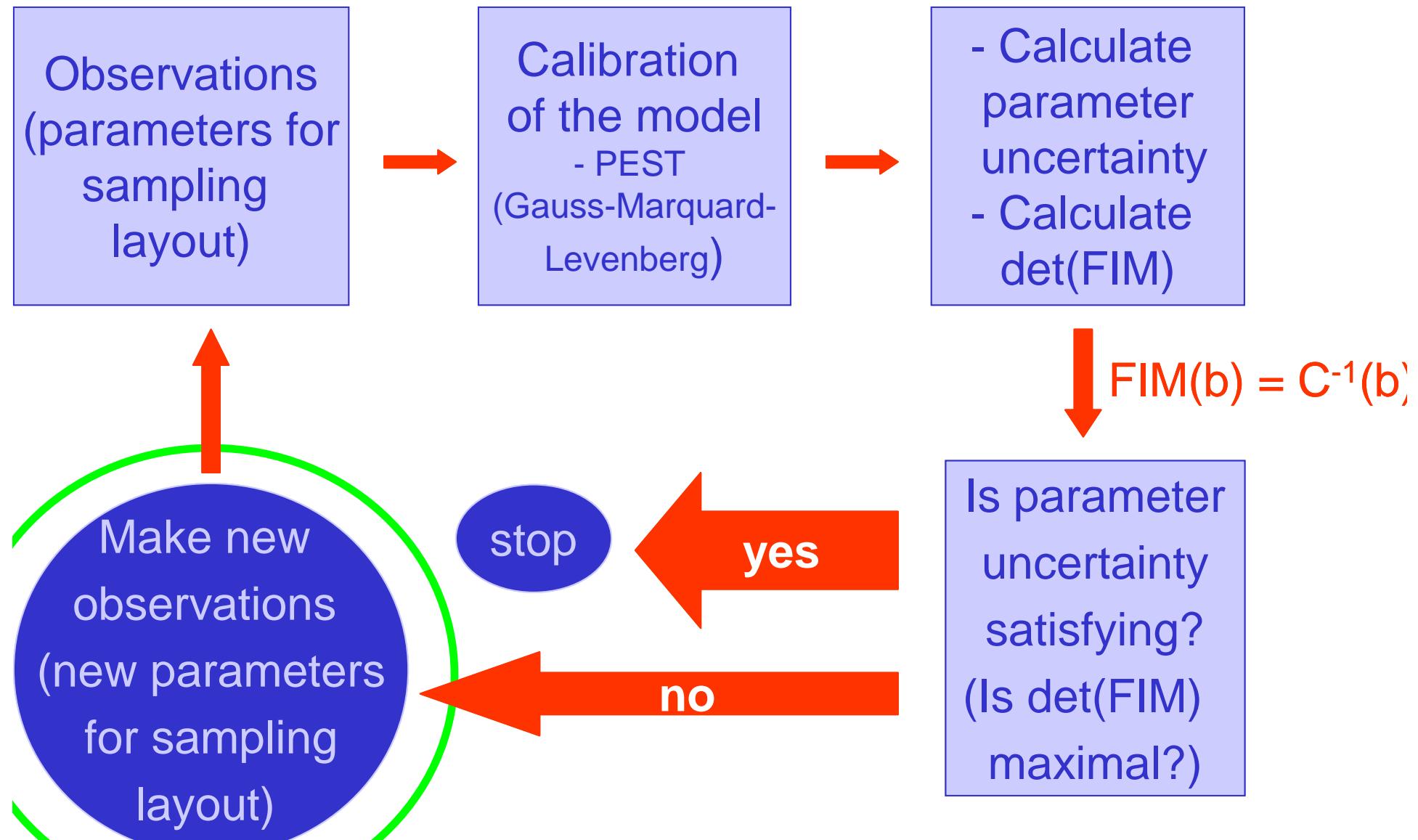
- new measurement layout => with D-optimal design technique (= maximisation of  $\text{Det}(\text{FIM})$ )

$$C(b) = \partial^2 (J^t Q J)$$

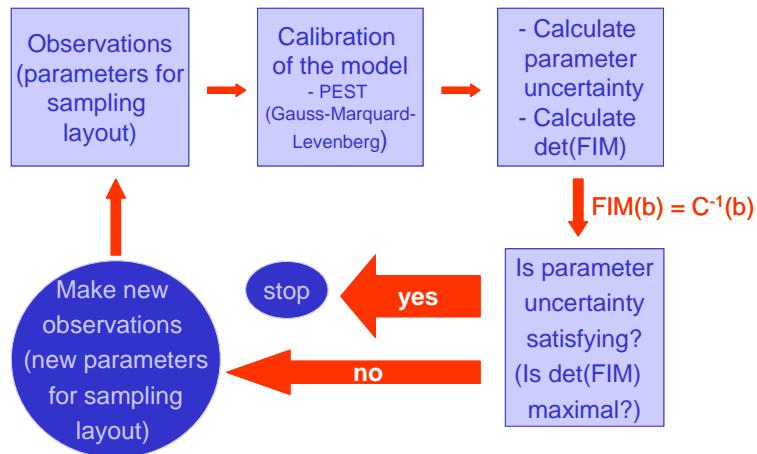
*Fisher Information Matrix =  $C^{-1}(b)$*

- maximisation with shuffled complex technique (SCE-UA method) by changing parameters of sampling layout (frequency, place, period,...)

# Methodology



# Methodology



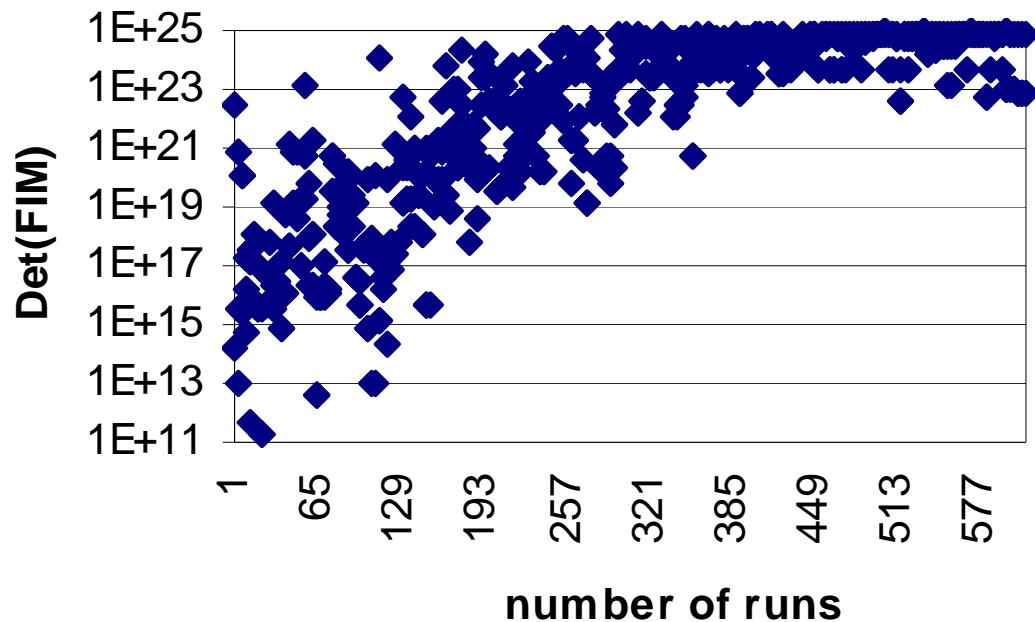
- When 'STOP' = make new measurements following the obtained sampling parameters
- Then start the procedure all over again=>iterative method
- The sampling procedure can change year after year, following natural trends e.g. climate change or human interactions e.g. a dam, a new discharge point

# Results

## Five parameters of the sample layout variable:

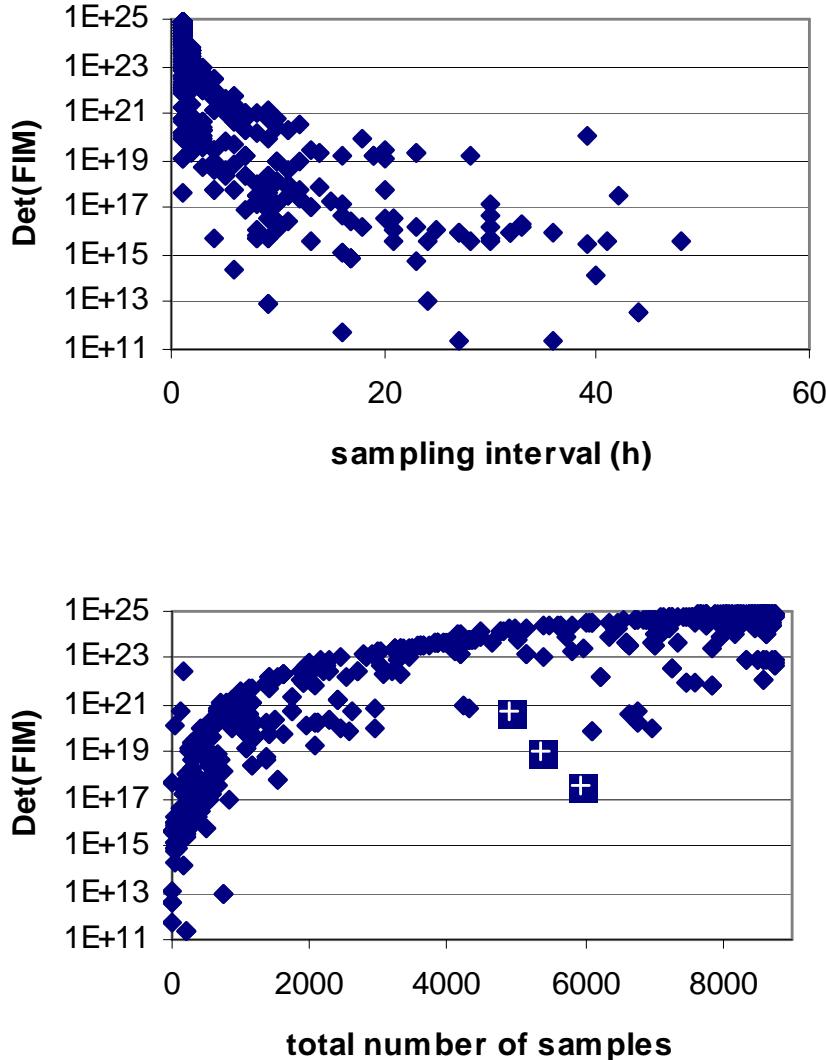
- **frequency (every hour - every two days)**
- **period (summer, winter, mixed summer-winter)**
- **total amount of samples (365\*24)**
- **only DO or combined DO-NO<sub>3</sub>, DO-NO<sub>3</sub>-BOD or DO-NO<sub>3</sub>-BOD-NH<sub>4</sub>)**
- **the sample location (4 possible combinations of 3 possible locations: upstream, halfway, downstream)**

# Results



- 635 iterations needed

# Results



- **Best : hourly time base, nearly the whole year, on three locations and with the four variables**
- **Other sampling schemes possible that provide a quasi similar accuracy**
- **Some sampling schemes non-optimal**

# Results

## non-optimal sampling schemes

Table 1. Non-optimal sampling designs

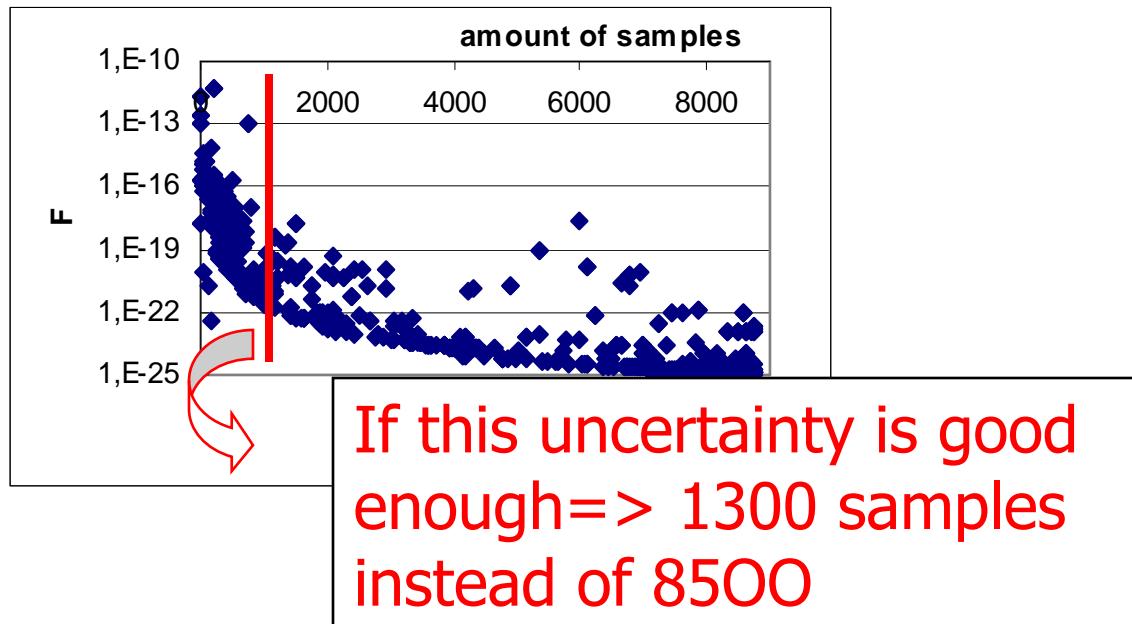
Sampling interval (h)	Number of samples	Period	Location	Observed variables	Det(FIM)
1	5972	16 Apr.-31 Dec.	Geraardsbergen	DO-NO <sub>3</sub>	4,08E+17
1	5340	22 May-15 Nov.	Geraardsbergen	DO-NO <sub>3</sub> -BOD	1,19E+19
1	4902	11 May-31 Dec.	Geraardsbergen	DO-NO <sub>3</sub> -BOD	5,92E+20

- **Factors that are influencing:**
  - **place (upstream)**
  - **period (no measurements in spring)**
  - **no NH<sub>4</sub>**

# Results

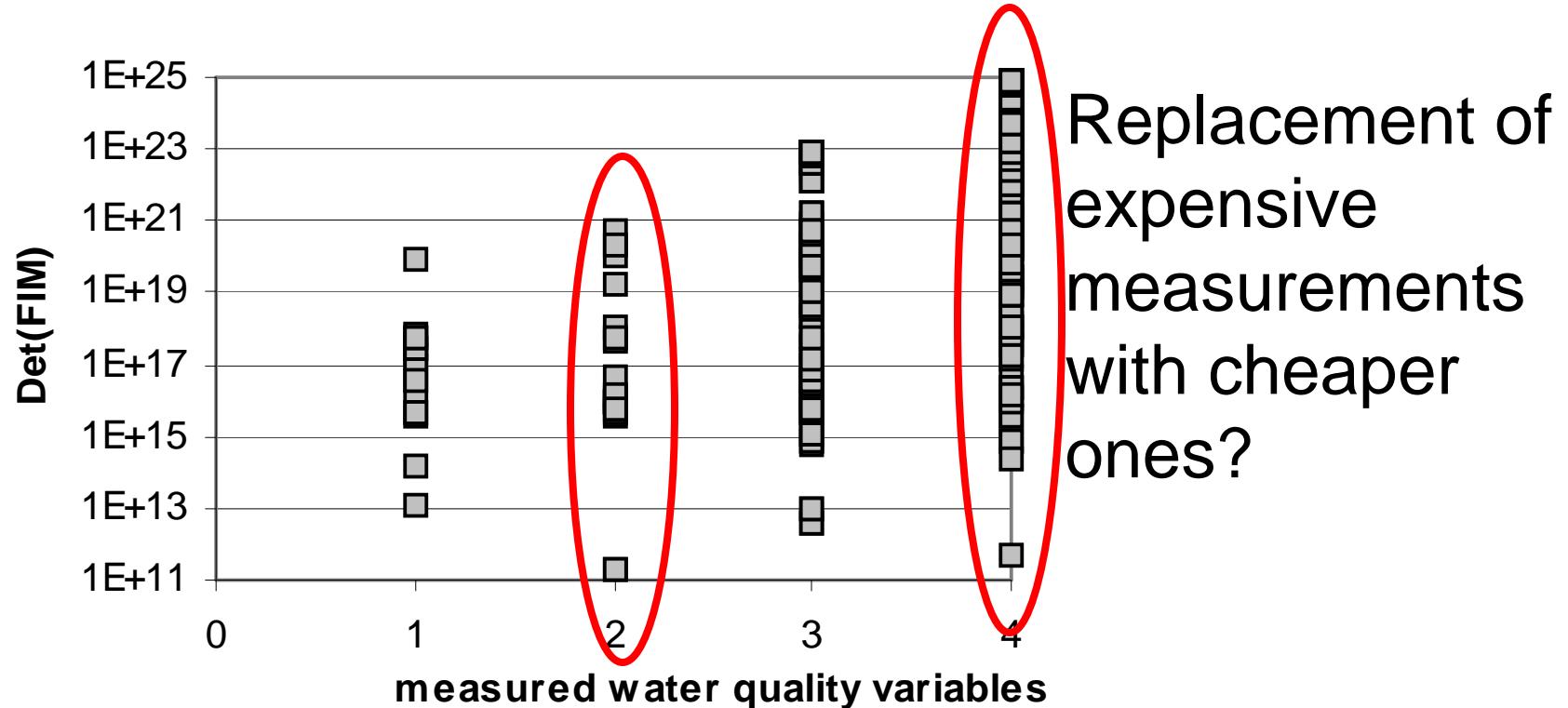
## OED with practical considerations

- Final uncertainty on model results is important => can be related to cost and practical implications



# Results

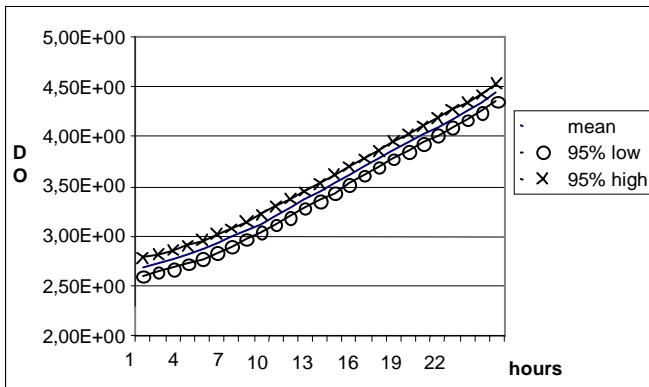
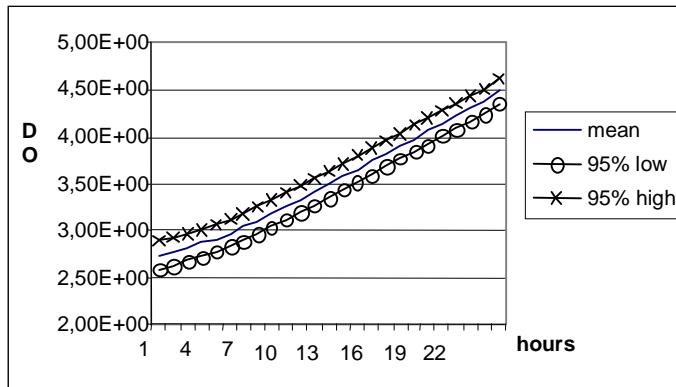
## OED with practical considerations



**DO + NO<sub>3</sub> or DO + NO<sub>3</sub> + BOD + NH<sub>4</sub>**

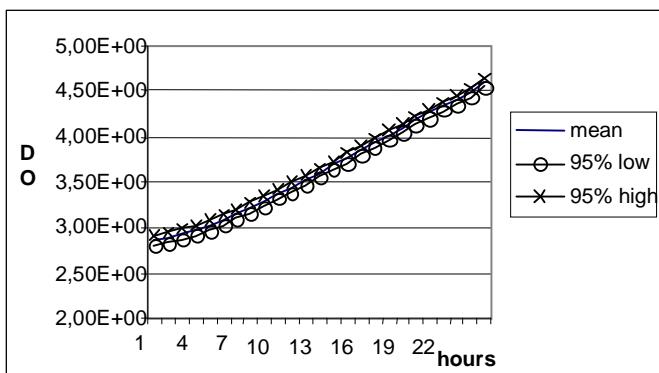
# Results

## Comparison of three sampling schemes



1

2



3

- Average width of the confidence interval on the model results:
  - reduction of 45% (2-1)
  - reduction of 60% (3-1)

# Conclusion

- OED for calibration of water quality models => measurement strategy
- Dender: optimal sampling strategy with highest amount of samples, highest sampling frequency, max. number of locations and max. number of variables measured.
- Usefulness of the method: evaluation of sub-optimal sampling strategies, in view of limitations, costs and practical considerations.