

Nondestructive Evaluation of
Aerospace Materials
and Structures

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New tools in CIVA for Model Assisted Probability of Detection (MAPOD) to support NDE reliability studies

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EXTEN·D·E
| CIVA |



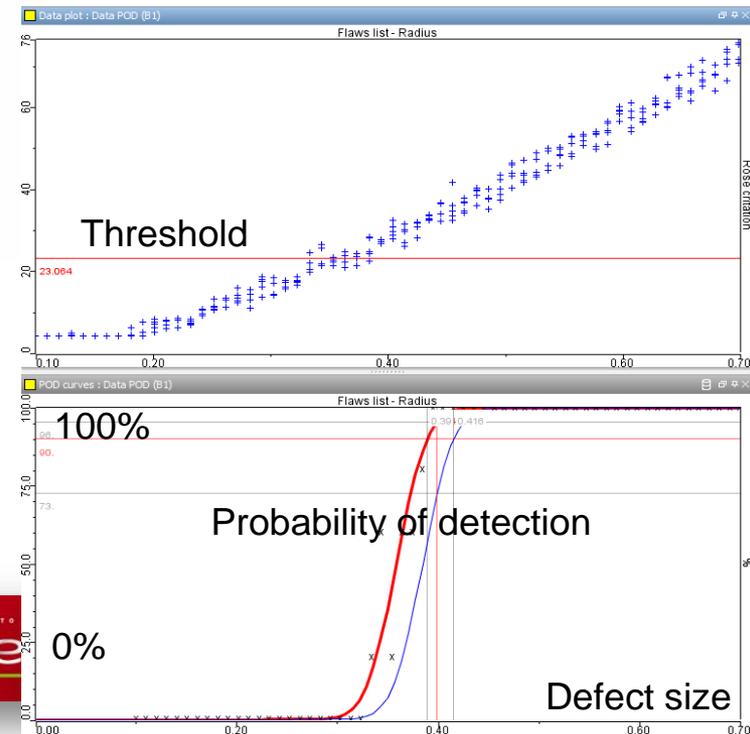
Outline

- | Introduction
- | MAPOD methodology & benefits of modeling
- | CIVA features in this context
- | Examples: Illustrative cases
 - HF ET
 - UT inspection of engine disks
 - UT inspection of titanium billet
- | Conclusion

Reliability in aerospace

Context:

- **Damage tolerance** rules: Aircraft maintenance intervals driven by the knowledge of detectable flaw sizes
- **Probability Of Detections** campaigns are required in order to evaluate statistically the maximum flaw size that can be missed by a given inspection procedure:
 - Military Handbook 1823-A methodology:
Parametric approach of POD (Berens models) and PFA
 - Involves knowledge of **influential parameters** to define relevant **Design Of Experiments**
 - Quite **large amount of data** necessary to provide reliable POD curves
- Quite long & costly: Why not doing a part of this work with simulation?
MAPOD approach (Model Assisted POD)



CIVA

- Leading industrial software dedicated to NDE Simulation & Analysis
(more than 270 customers in 42 countries)
- Multi-techniques:

✓ UT :

- Ultrasounds Testing modelling
- UT Acquisition Data Analysis tools

✓ GWT: Guided Waves

✓ ET : Eddy Current

✓ RT : Radiography

✓ CT: Computed Tomography

- CIVA Education: For universities and training centers

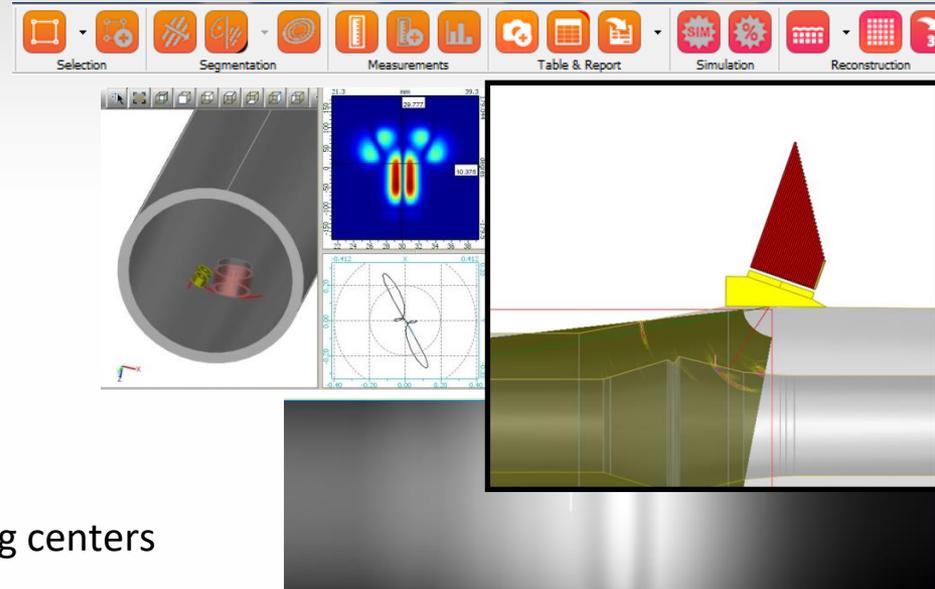
- Help to understand the physics behind NDT

- Mostly based on semi-analytical models (fast), connection with numerical ones (FEM, FDTD,...)

- Developed by CEA (French Atomic Energy commission):
25 years of experience with models & validations



- Distributed by EXTENDE (EXTENDE Inc in USA, VA)



MAPOD methodology

US MAPOD Working Group (2003-2011) driven by USAF has been a pioneer in developing a MAPOD approach

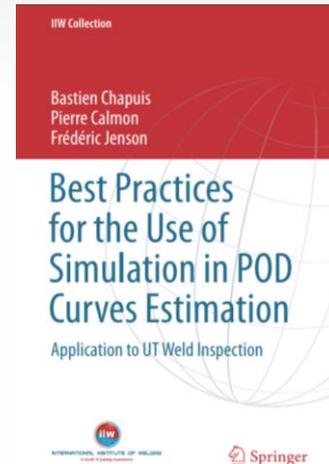


Efforts have been done to fix an **accepted methodology**:

- 2016 : IIW publication « Best practices for the use of simulation in POD Curves Estimation »

Different stages in a MAPOD process:

- Define a nominal configuration
- Identify and characterize the sources of variability which will be accounted for by the POD:
 - Select “aleatory parameters” among the input parameters in the model
 - Assign a statistical distributions to them
- Sample the statistical distributions of aleatory parameters (MC) and run the corresponding simulations.
- Compute POD curve from obtained results with relevant statistical models
- Evaluate the reliability of the POD curve



Benefits of Using Simulation

| In the context of reliability studies:

- Easy and precise **mastering of parameters variation**:
Not always the case experimentally
- Easy and fast to generate **large amount of data** (required for POD analysis)
- Less mock up & less trials : **Lower cost**
- **Insights** for physical understanding

| Modelling also useful also in other contexts

- Inspection method design, expertise, training....

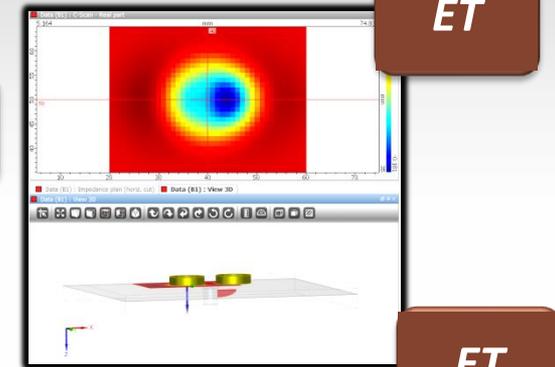
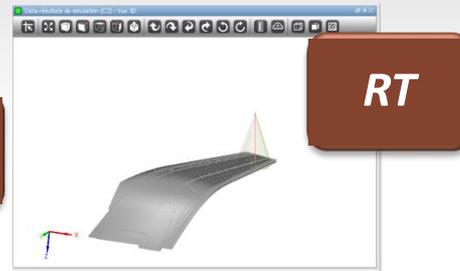
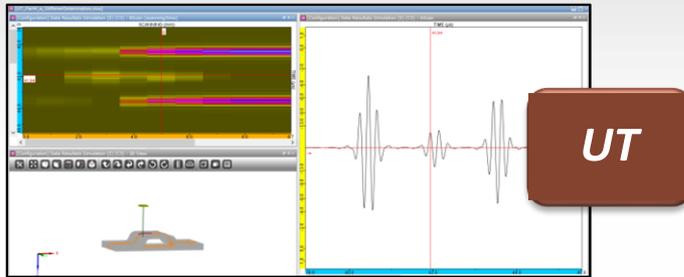
| Some limitations & challenges:

- Models capture a part of the variability but maybe not all (human factor, structural noise, etc.)
- Needs to define a priori in the model the sources of variability: Can be difficult
- Requires sufficient modelling accuracy (needs for validation) and acceptance by stakeholders

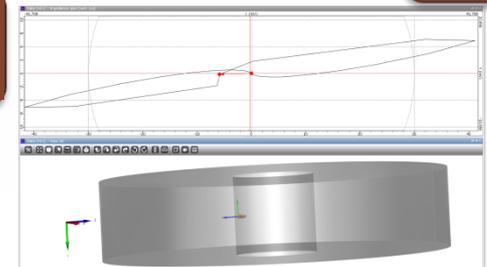
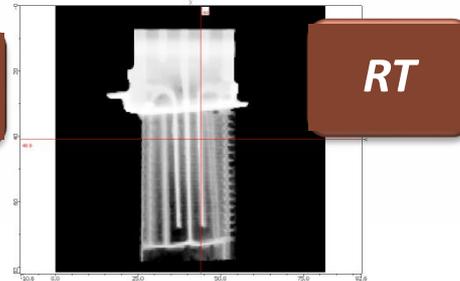
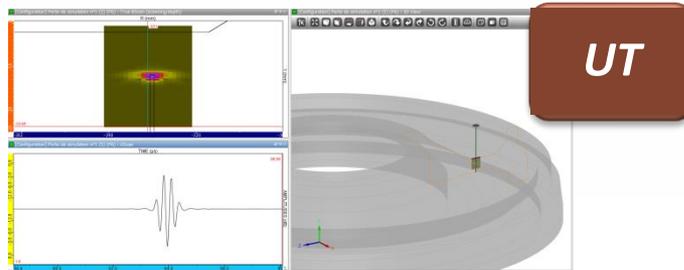
CIVA features in this context

1. Versatile & fast physic-based models, a user-friendly GUI:
Adapted for parametric studies

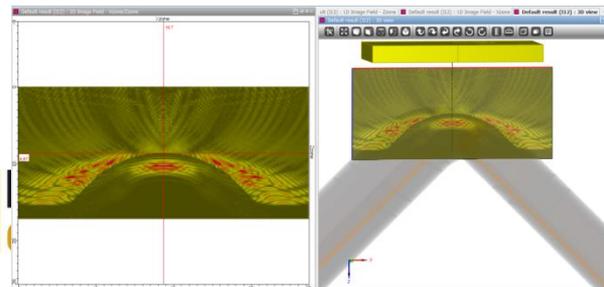
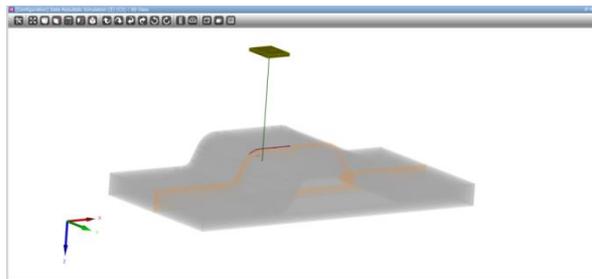
- For aeronautical structures



- For engines manufacturers



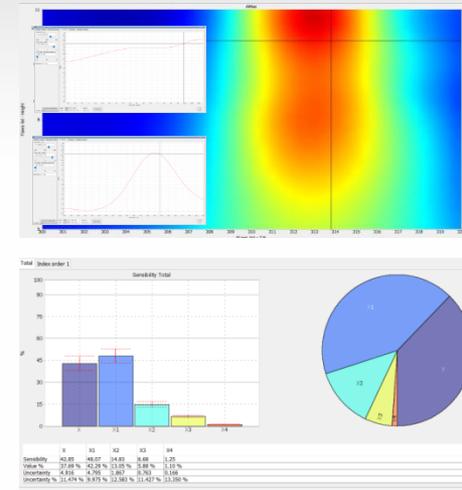
- For metallic or composite parts



CIVA features in this context

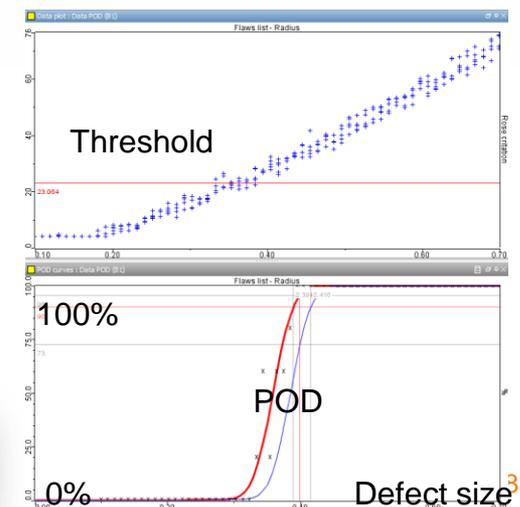
2. Implementation of « **metamodels** » or « Surrogate models:

- Smart Interpolators
- Built from a set of physic-based model results
- Can replace (after validation) the physic-based models:
 - For an **ultra-fast exploration** of the full range of parameters variation and « on demand » resampling
 - Generate even larger amount of data:
 - Makes possible **sensitivity analysis** (Sobol Indices)
 - Can « feed » **POD requirements**



3. Built-in POD Analysis tools:

- Signal Response or Hit Miss Berens models
- Data transform tools (log, lin, box-cox)
- Non parametric curves
- Array of PODs
- Import/Export data
- ...



Model validation

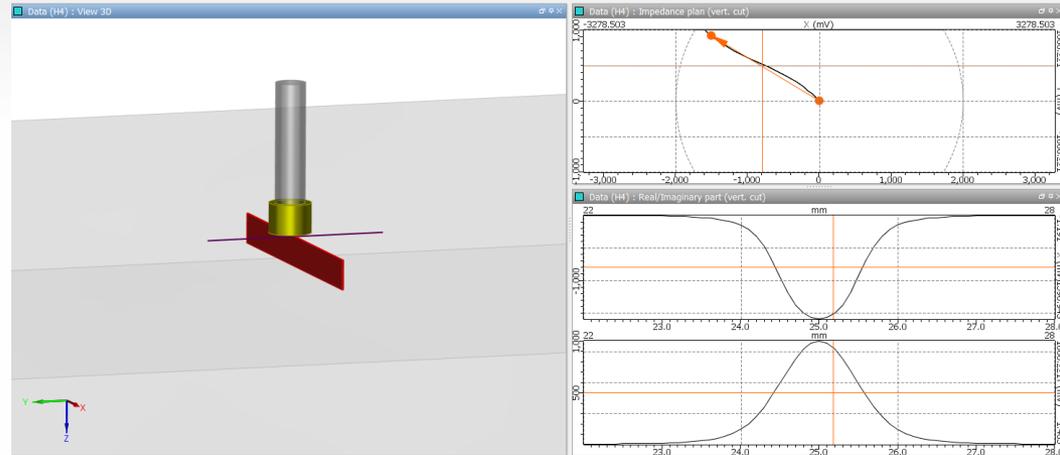
- | To be able to rely quantitatively on simulation, **models reliability and accuracy** is of first importance.
- | CIVA software development goes along with extensive **test & validation** campaigns:
 - To demonstrate applicability of new models when they come out
 - Quality Assurance and Non regression tests between each release
 - Annual participation to WFNDEC benchmarks presented at QNDE conferences
 - Targetted validation works performed in the frame of EXTENDE / CEA a collaboration, then published on EXTENDE website:
www.extende.com
 - Overview of CIVA validation efforts to be presented in the upcoming ECNDT conference (Goteborg, Swe):

F. Foucher, S. Lonne, G. Toullelan, S. Mahaut, S. Chatillon, "AN OVERVIEW OF VALIDATION CAMPAIGNS AROUND THE CIVA SIMULATION SOFTWARE", Monday June, 11th 2018, ECNDT conference

Illustrative examples

Sample case: High Frequency Eddy Current Testing model

- Aluminum slab with surface breaking notch
- Pencil ET sensor $\Phi 1,4\text{mm}$ with Ferrite core, common mode function operating at 1MHz
- Simulation results of the calibration case on a reference defect, 10mm long and 1mm high :



- 4 main essential variables kept in the design of experiment:
 - Lift-off: [0,15mm; 0,5mm]
 - Sensor orientation: [-5° ; +5°]
 - Defect Height [0,5mm; 3mm]
 - Defect aperture [0,03mm; 0,07mm]

Defect length considered as defect size parameter

Illustrative examples

Building of the metamodel :

- Built from a database of 500 CIVA simulations
- Sobol sampling schemes to fill the space of parameters variation
- Overview of the DOE and results in a **parallel plot**

Probe Lift-off

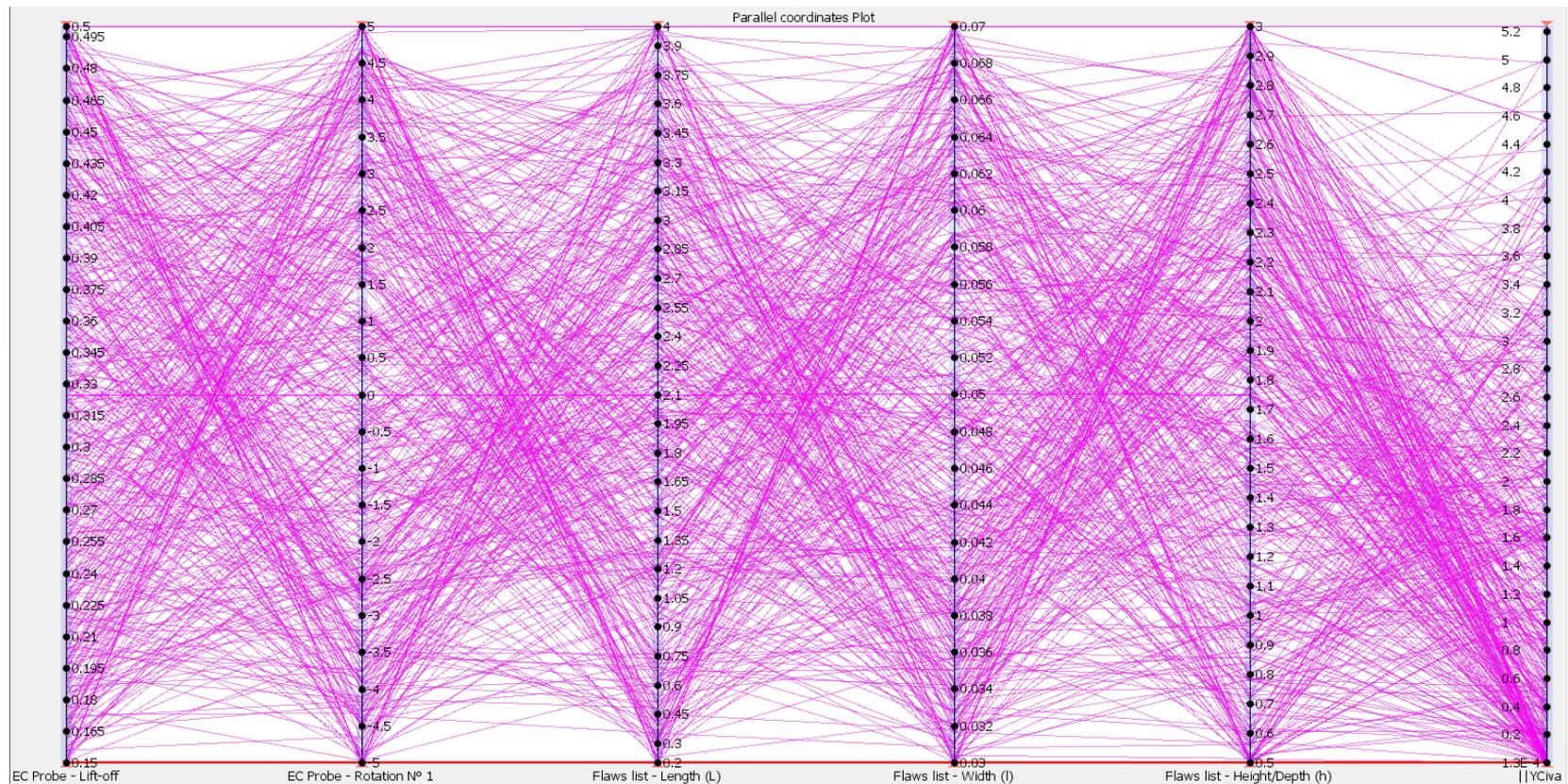
Probe orientation

Flaw length

Flaw width

Flaw height

Signal Amp.



Illustrative examples

Building of the metamodel :

- Built from a database of 500 CIVA simulations
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- Overview of the DOE and results in a **parallel plot**

Selection of **worst cases**

Probe Lift-off

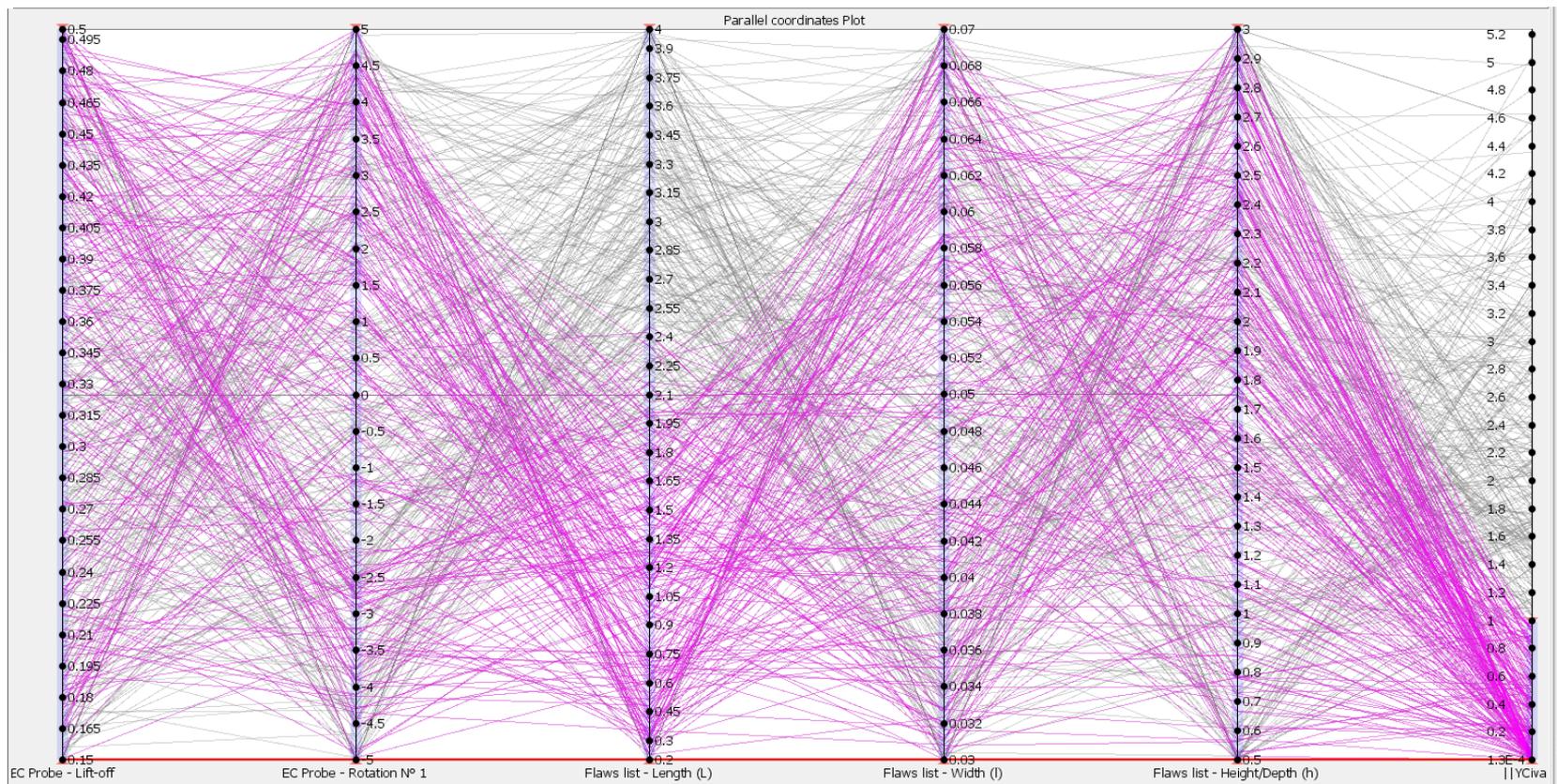
Probe orientation

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Signal Amp.

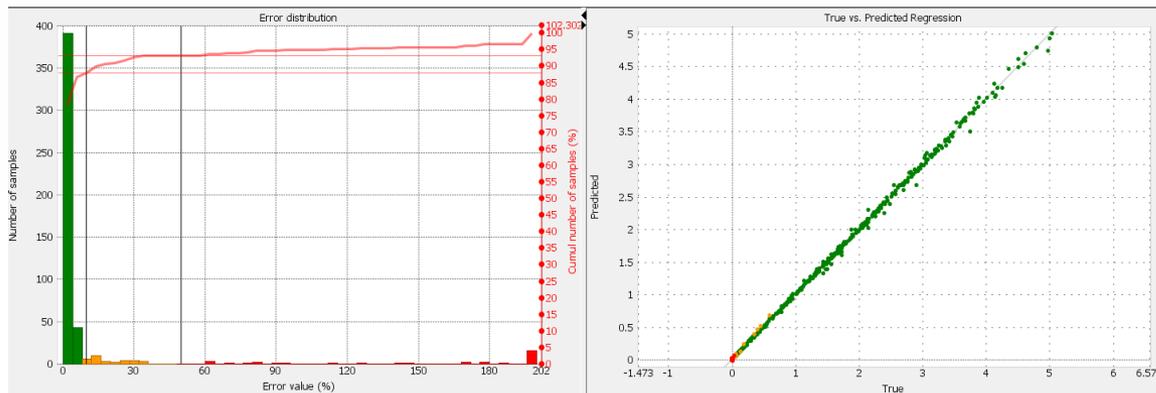


Illustrative examples

Analysis of the metamodel : Validation

Based on **Cross validation** methodology:

- Division of the physic-based samples database in k folds
- Comparison of metamodel results obtained from k-1 folds with the remaining samples.
- « Error measurement »:
 - Histograms of « errors » between metamodels and samples
 - « True vs predicted » plot



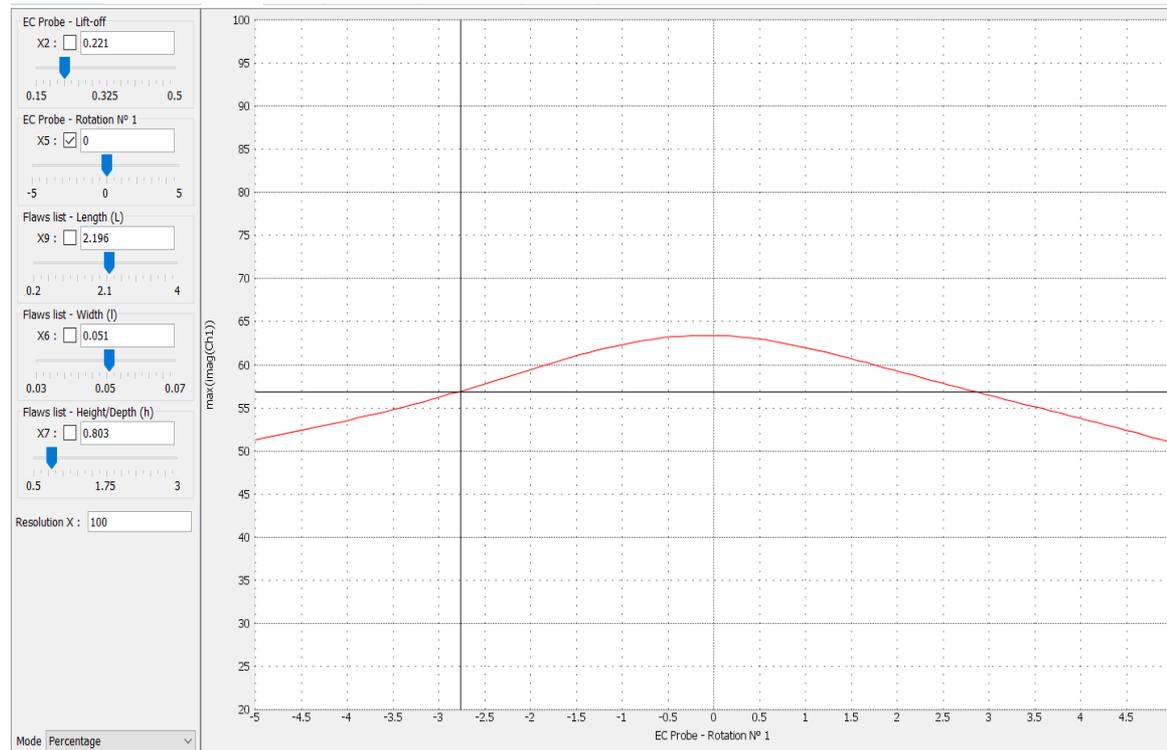
- Several interpolators available to build the metamodel (Kriging, Linear, RBF)

Evaluation of the metamodel “fit” and selection of the best interpolator

Illustrative examples

Parametric analysis from metamodel data:

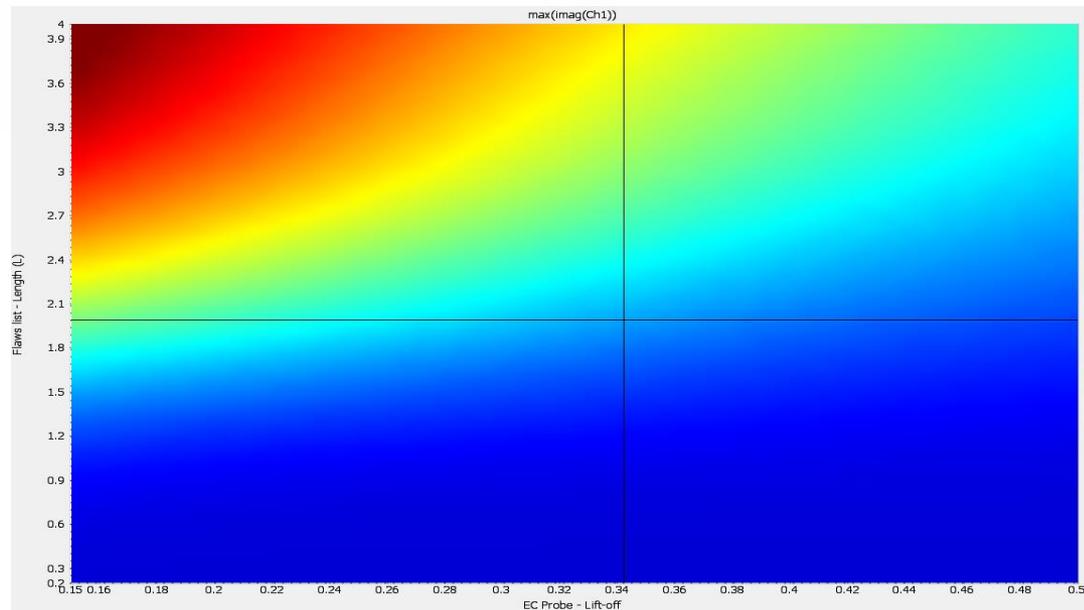
- Access to **1D or 2D plots** built with metamodel data (and not only the 500 results grid)
- Really fine sampling and exploration of the full range of multi-parameters variation:
 - Impact of sensor orientation (-5° ; $+5^\circ$) when other parameters fixed to a selected value



Illustrative examples

Parametric analysis from metamodel data:

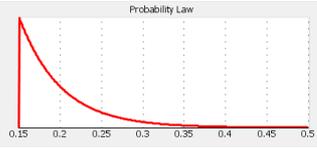
- Access to **1D or 2D plots** built with metamodel data (and not only the 500 results grid)
- Really fine sampling and exploration of the full range of multi-parameters variation:
 - Impact of defect length (ordinate) and lift-off (abscissa) on the output signal (color level)

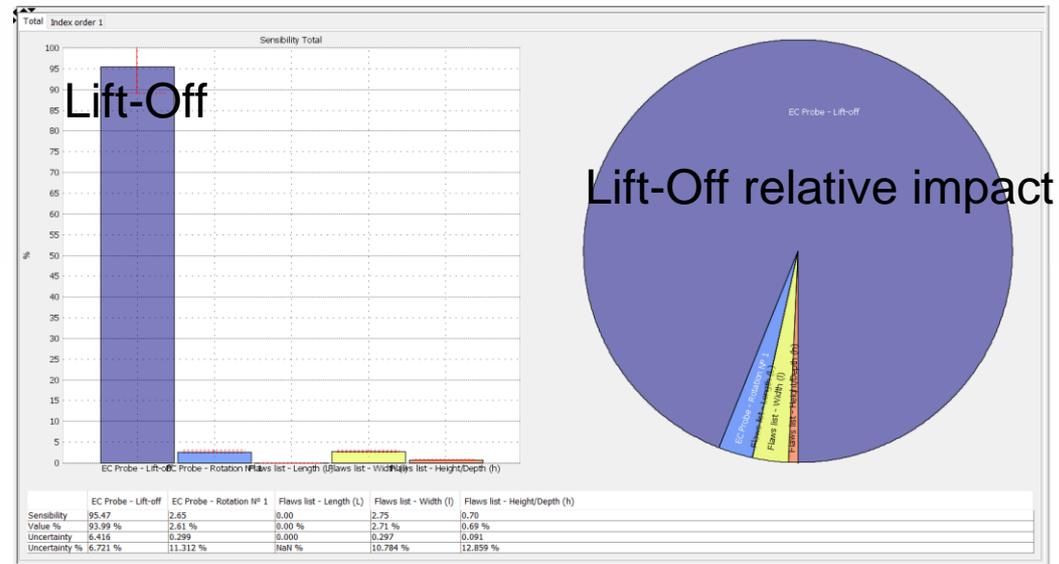


Illustrative examples

I Metamodel makes possible a statistical analysis of the **parameters sensitivity**:

- Computation of Sobol Indices (Total Order, 1st order)
- Obtained from variance decomposition computation
- User defines assumed **statistical distributions** for variables
- Sobol gives the relative influence of each parameters to the output

Lift-off	
Probe Orientation	
Defect Aperture	
Defect Height	
Defect length	Constant at 2 mm

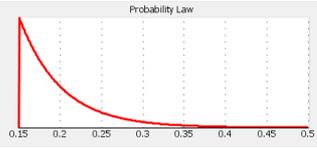
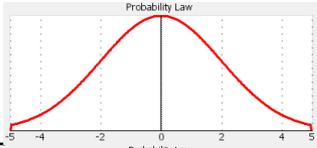


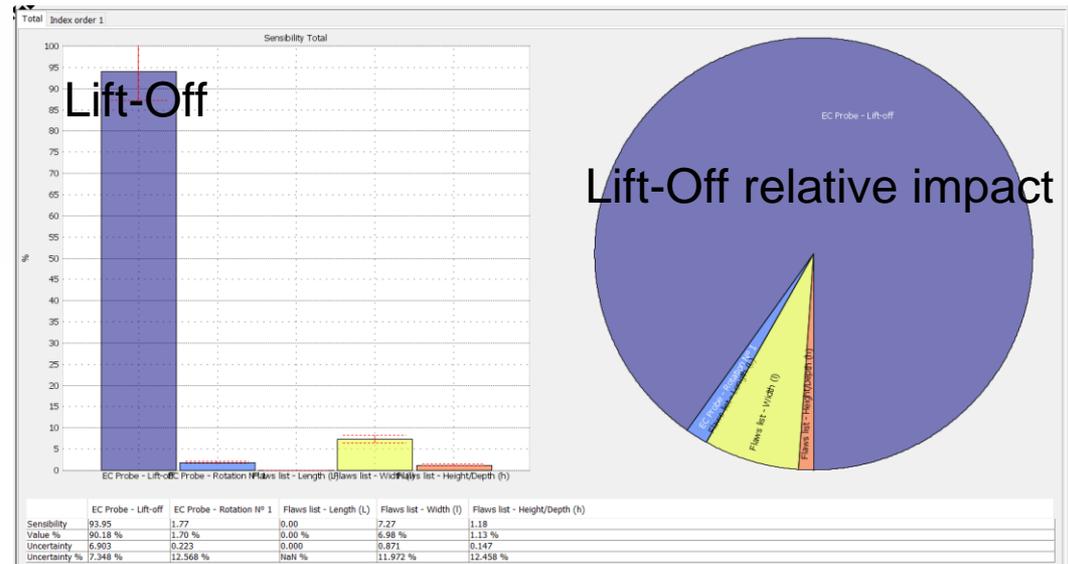
Helps for technical justifications and definition of a relevant Design Of Experiment

Illustrative examples

Metamodel makes possible a statistical analysis of the **parameters sensitivity**:

- Computation of Sobol Indices (Total Order, 1st order)
- Obtained from variance decomposition computation
- User defines assumed **statistical distributions** for variables
- Sobol gives the relative influence of each parameters to the output

Lift-off	
Probe Orientation	
Defect Aperture	
Defect Height	
Defect length	Constant at 1 mm

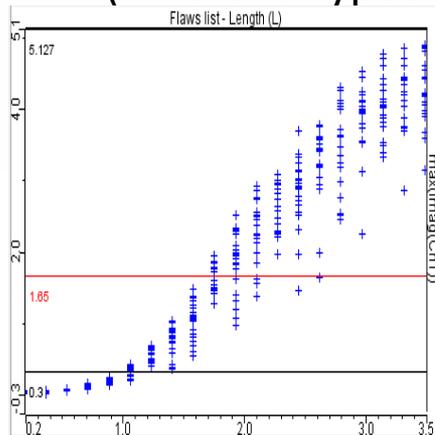


Helps for technical justifications and definition of a relevant Design Of Experiment

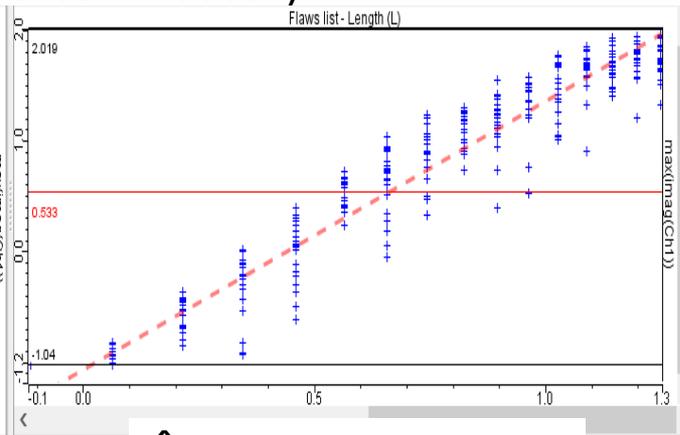
Illustrative examples

| A **POD analysis** can be created from the metamodel in a few seconds:

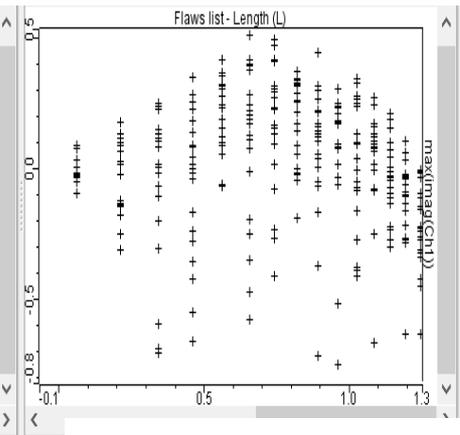
- Selection of the « characteristic value » (e.g. length) for defect size
- Selection of assumed statistical distributions for test variables
- Data sampling definition (# of defect sizes, # of tests) :
No limits thanks to metamodel
- Definition of threshold
- \hat{a} vs a plot can be used to compute POD curve (if Berens hypotheses validated)



\hat{A} vs a plot in linear scale



\hat{A} vs a plot in log/log scale
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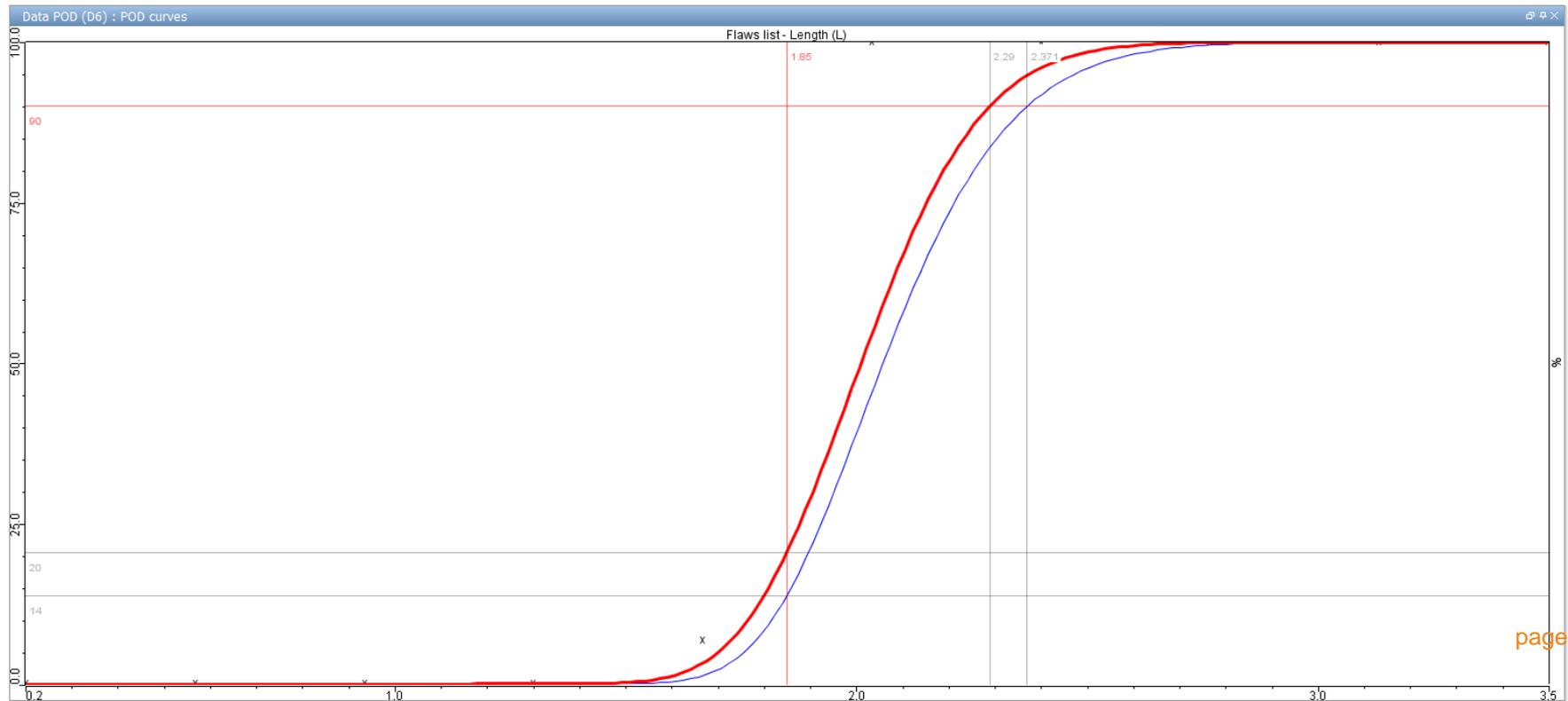


Residuals from the linear regression computed by MLE



Illustrative examples

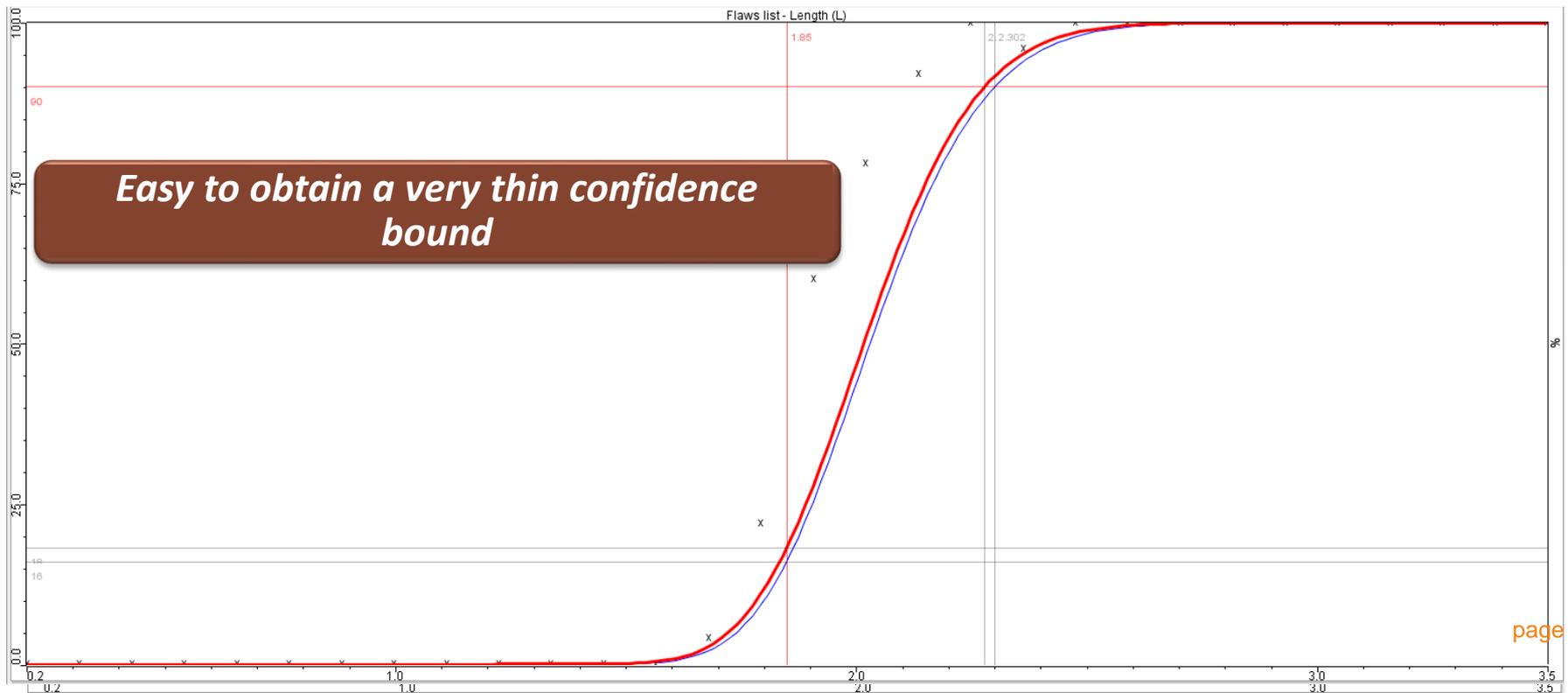
- | A **POD analysis** can be created from the metamodel in a few seconds:
 - POD curve obtained with a quite coarse sampling (150 samples with 10 different defect sizes) : $a_{90} = 2,29\text{mm}$; $a_{90/95} = 2,37\text{mm}$



Illustrative examples

I A **POD analysis** can be created from the metamodel in a few seconds:

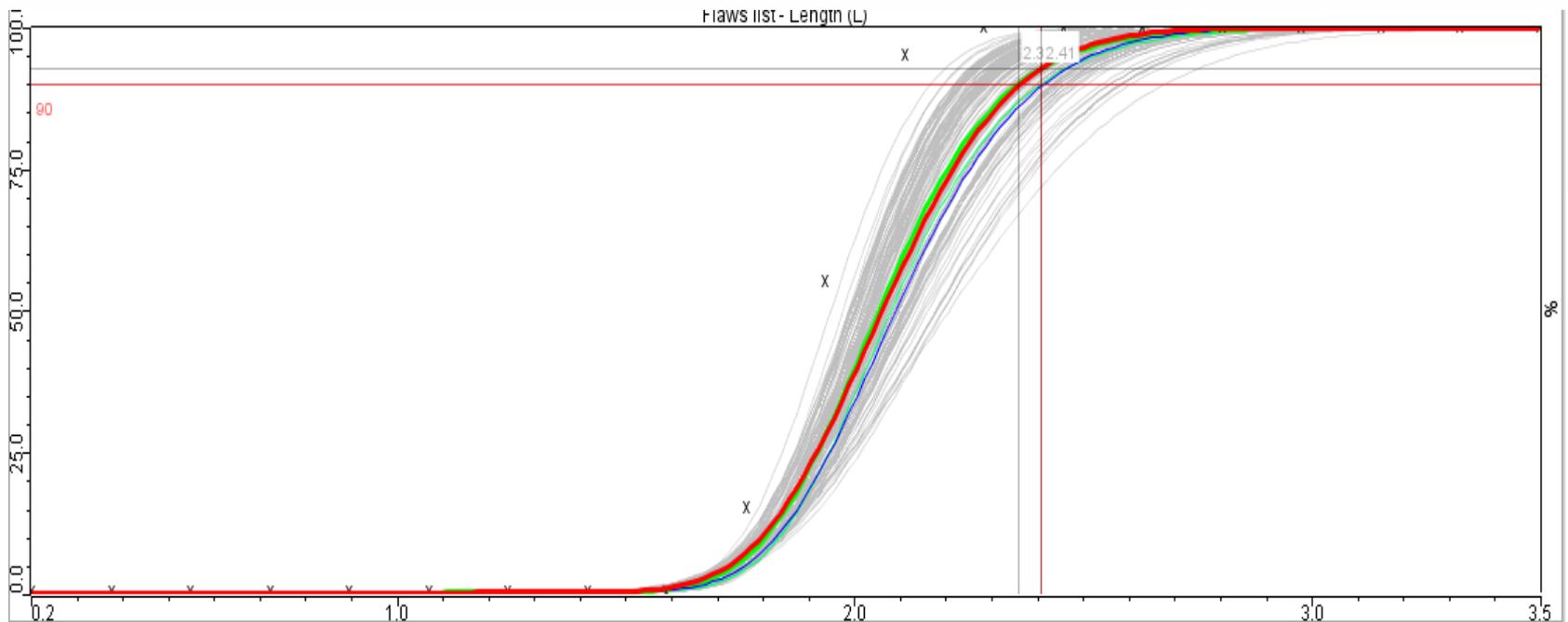
- New POD curve obtained **in a second** with thinner sampling (1500 samples with 30 different defect sizes): $a_{90}=2,28\text{mm}$; $a_{90/95}=2,30\text{mm}$



Illustrative examples

Evaluation of POD curve reliability

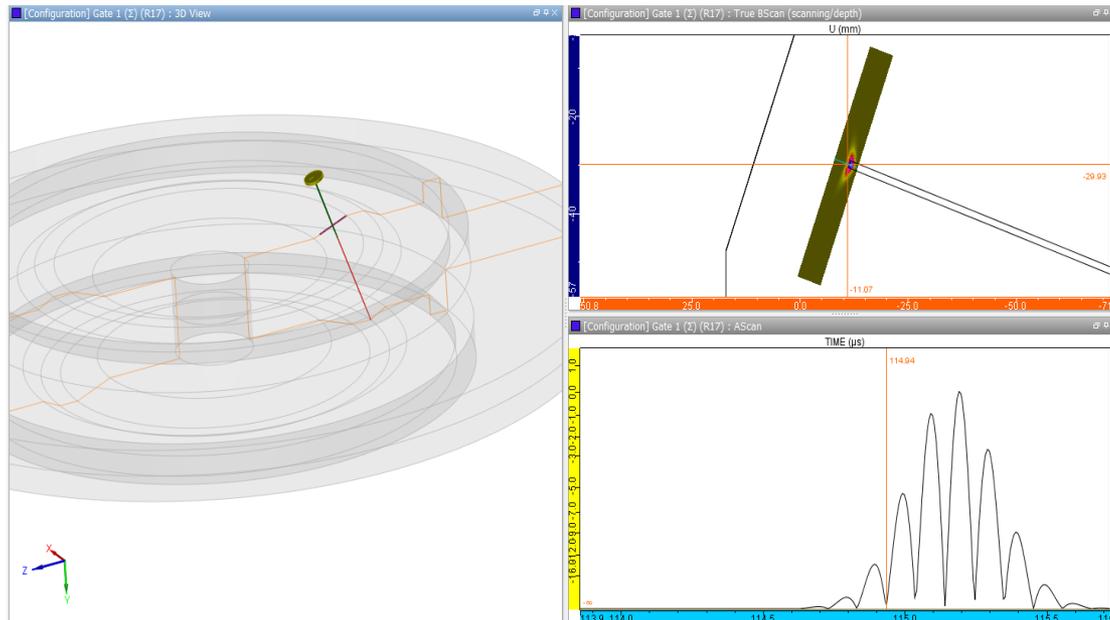
- In MAPOD: Main « error » is not due to a poor sampling but to the necessity to define input parameters value and statistical distribution
- Evaluation of POD curve reliability possible by varying these distributions and assesing the impact. One tool to do this: **Array of POD**
 - A confidence level is given to statistical distribution parameters of the input variables.
 - Monte-Carlo sampling & generation of a set of POD curves instead of only one
 - Evaluation of the scattering of POD curves obtained



Illustrative examples

Other examples: Engines Turbine Disk UT inspection

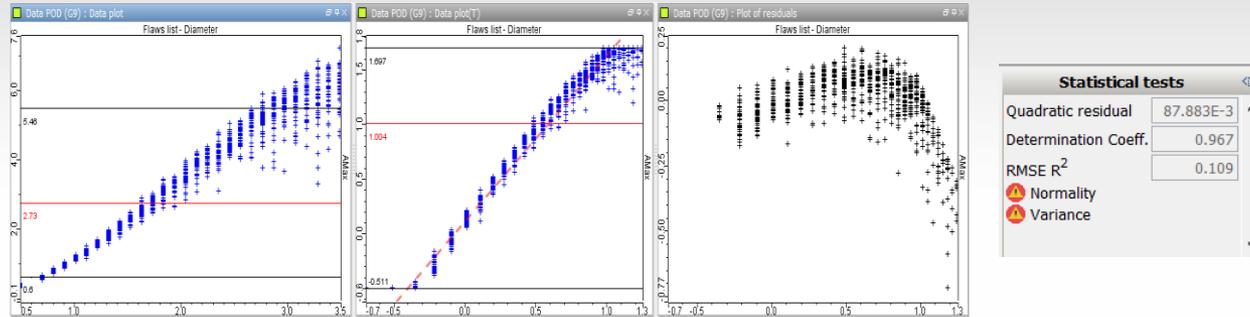
- Nickel Super alloy disk at the mid-manufactured stage
- Immersion, 5 MHz, Focused Single Element transducer
- Flat Bottom Holes defect,
- 4 main essential variables kept in the design of experiment:
 - Incidence Angle $[-3^\circ ; +3^\circ]$
 - Water Path [75mm; 85mm]
 - Attenuation level [40dB/mm; 60dB/mm]
 - Defect orientation $[-5^\circ ; +5^\circ]$



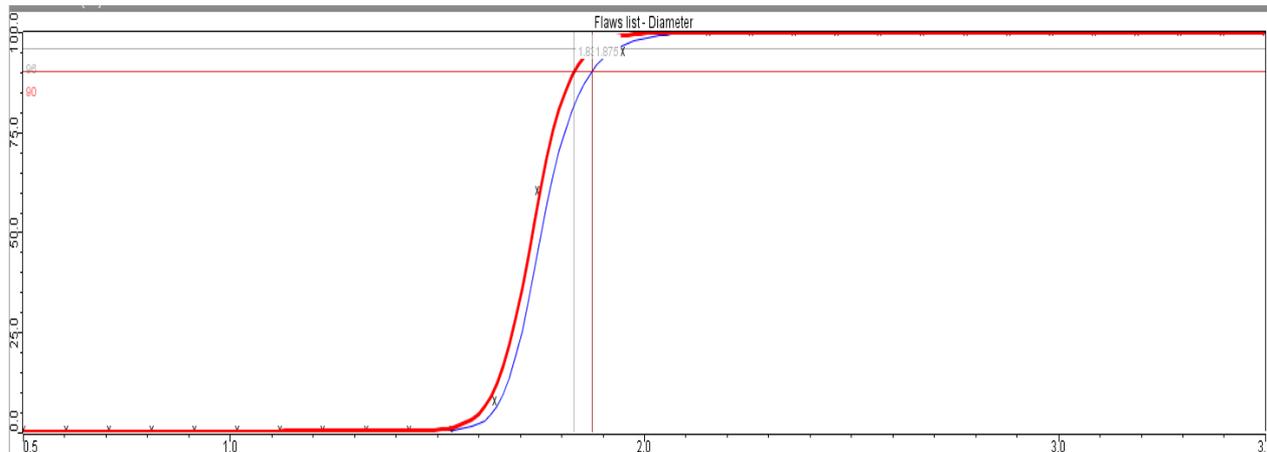
Illustrative examples

POD analysis obtained from metamodels:

- Here, Berens hypotheses not fulfilled for a signal response analysis



- Hit/Miss POD Curve: Obtained from 1200 samples with 30 different FBH diameters : $a_{90} = 1,83\text{mm}$; $a_{90/95} = 1,88\text{mm}$



Illustrative examples

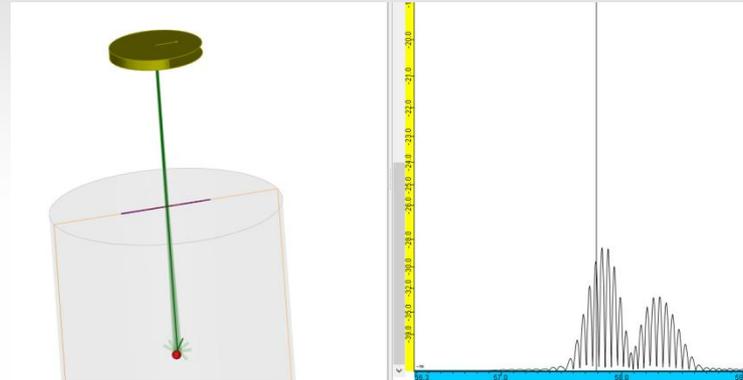
Non parametric POD curves:

- Parametric models (Hit-Miss, Signal Response) have been created to tackle the difficulty to have enough data to generate POD curves with the binomial approach.
- With MAPOD & especially with metamodels, generate a large amount of data is not any more a real problem.
- Possible to directly plot piecewise POD curves, based on **Hit/Miss ratio for each defect size** (just requires to select a sampling with a sufficient amount of results for each defect size):
 - Able to describe **non monotonic POD curves**
 - Can be used to validate (or even replace) a POD curve obtained from with the standard parametric approach

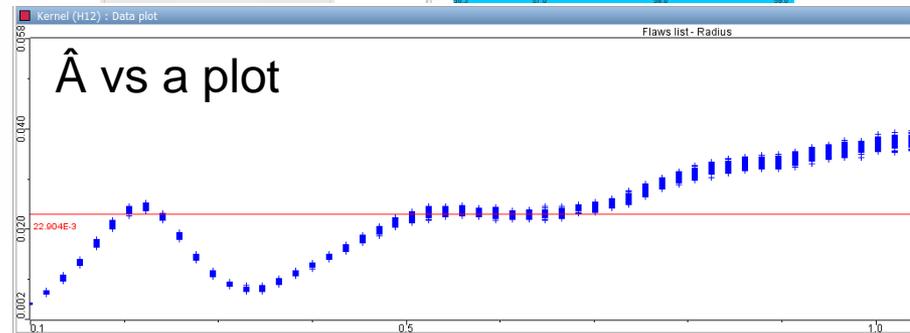
Illustrative examples

Non parametric POD curves:

- Examples of Titanium billet UT inspection simulation
- Alumina Inclusion

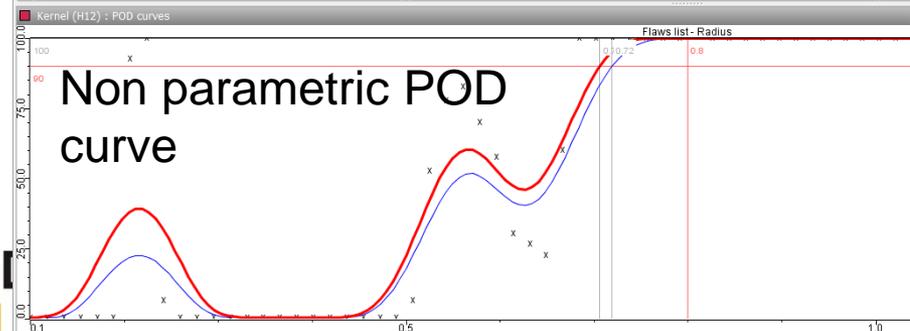


- Resonance phenomena* observed for several defect sizes, non linear \hat{a} vs a plot



- Example of obtained non parametric POD curve

(80 samples for each defect size)



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Conclusion

- | MAPOD approach to support NDE reliability studies in aerospace: **More data, Lower cost.**
- | General acceptance on the **methodology**: Recommended practices published by IIW in 2016.
- | Validation with physical tests remain necessary at some stages.
- | CIVA gives tools to fulfill required main steps of MAPOD method:
 - Versatile & Validated Physic-based models
 - Metamodels to help parameter sensitivity analysis and the definition of relevant Design of Experiment
 - Metamodels to easily generate a large amount of data required to « feed » POD parametric models
 - Embedded POD statistical tools
 - Array of PODs to evaluate POD curves reliability
 - Non parametric POD curves available

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