



Statistical Analysis of NPP transients: LBLOCA in a PWR; Parametric, non-parametric methods and EBEPU

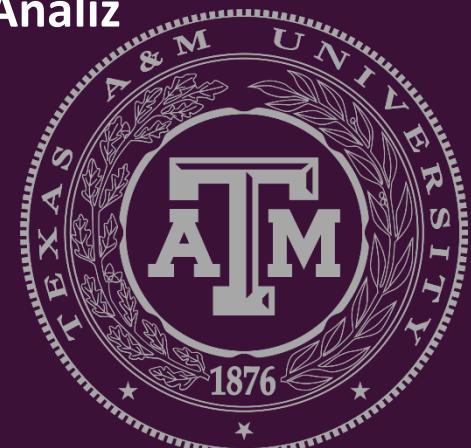
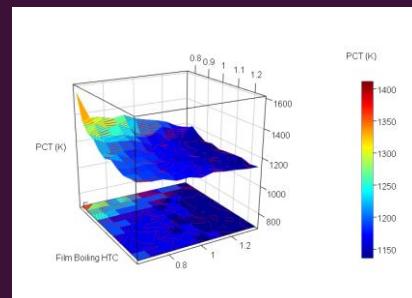
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Seminarium Zakładu Energetyki Jądrowej i Analiz Środowiska, Jan 2022





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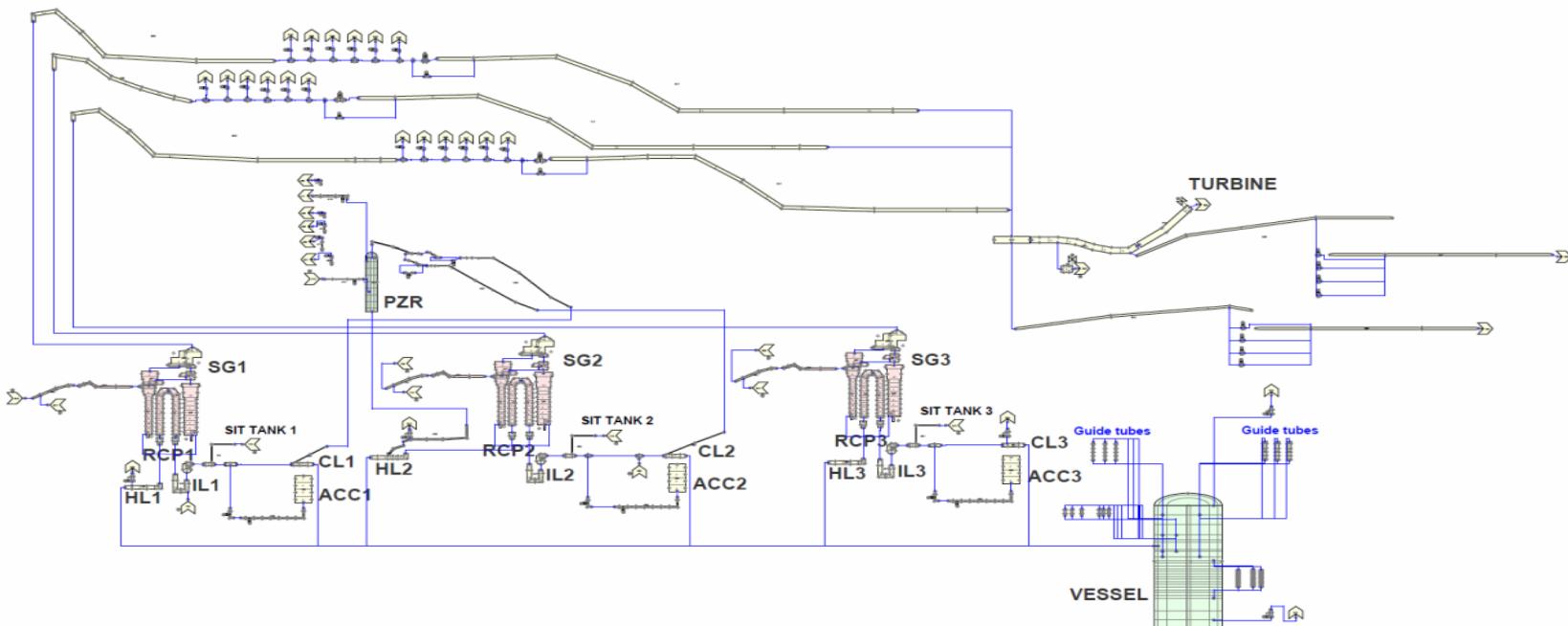


1. Plant Model

Reference NPP TRACE model:

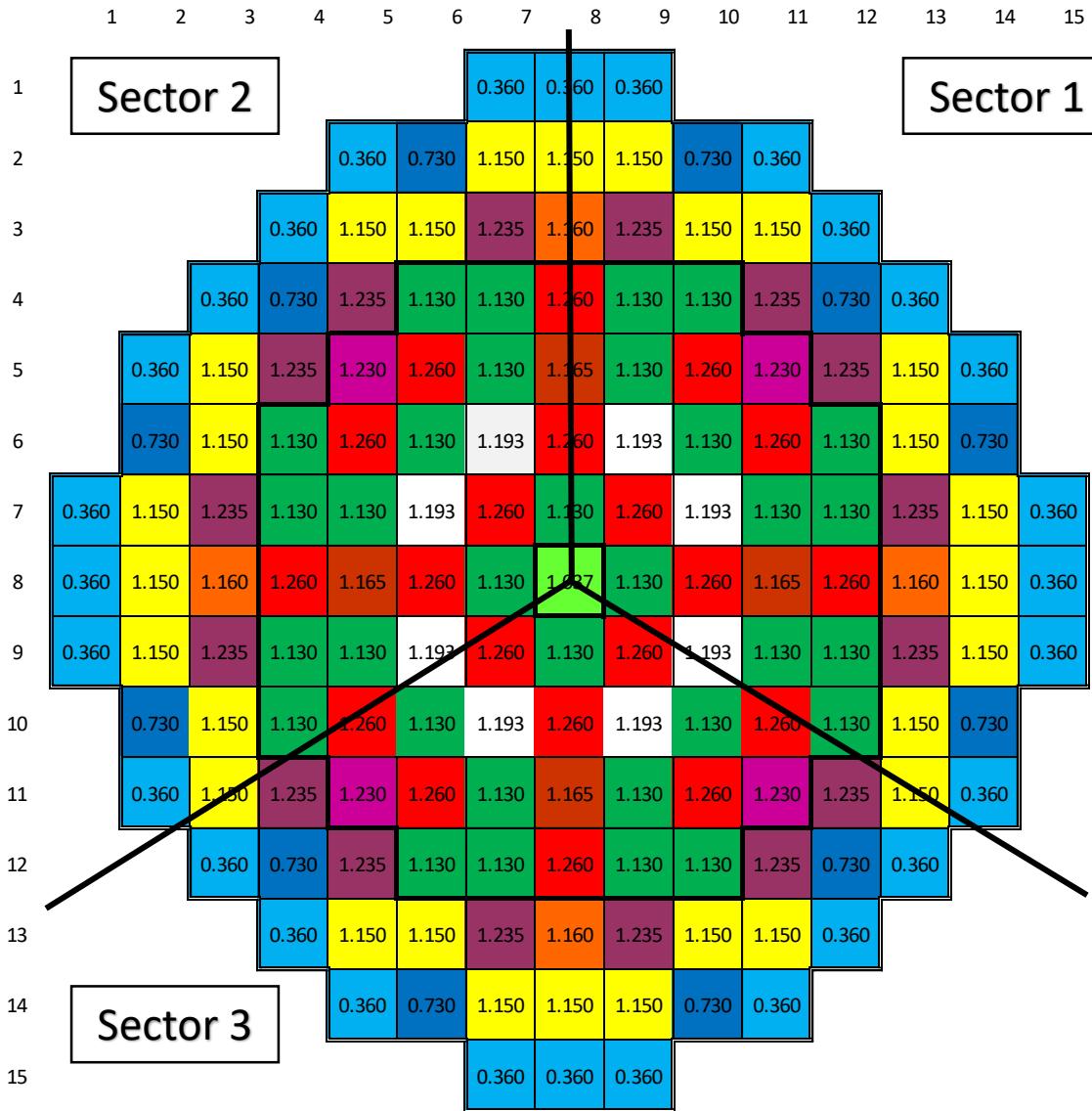
3-LOOP PWR WESTINGHOUSE, 3000 MWT

- 279 Thermal Hydraulic components:
 - 2 VESSELS
 - 73 PIPEs
 - 43 TEEs
 - 54 VALVEs
 - 3 PUMPs
 - 12 FILL
 - 33 BREAK
 - 56 HS
 - 3 POWER
- Control Systems:
 - 740 SIGNAL VARIABLES
 - 1671 CONTROL BLOCK
 - 58 TRIP





1. BE core model



2. Base Case: CL-LBLOCA

2/2 effective ACC - 2/2 LPSI

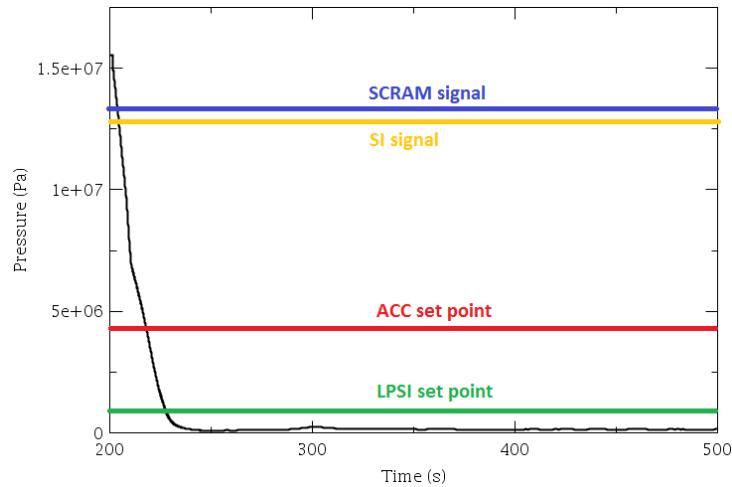


Figure: RCS pressure.

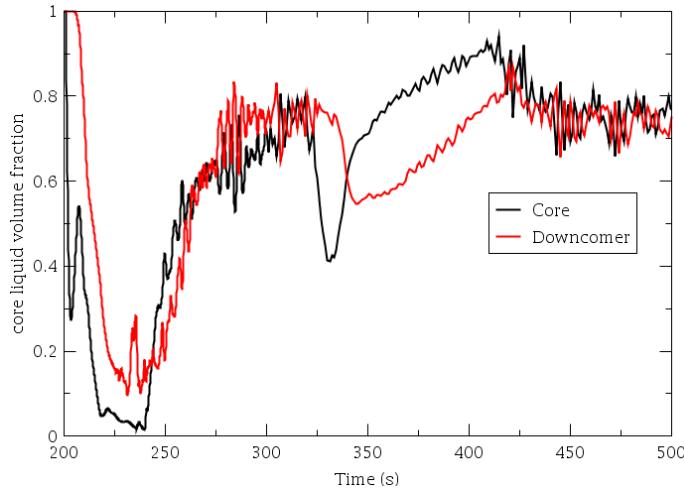


Figure: Core and downcomer coolant level.

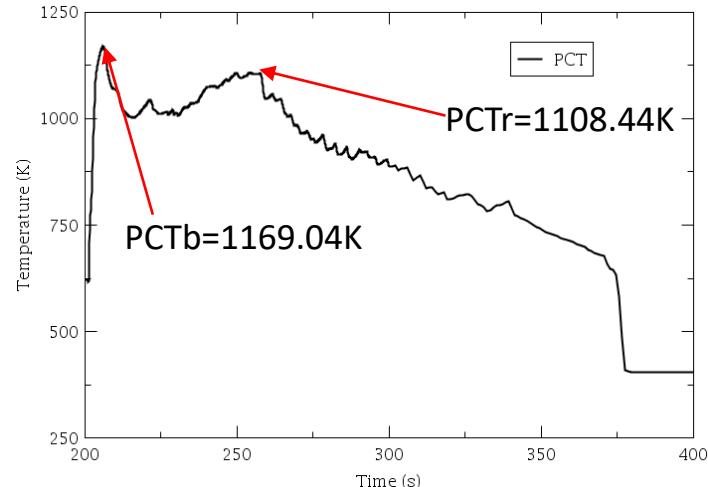


Figure: Peak Cladding Temperature.

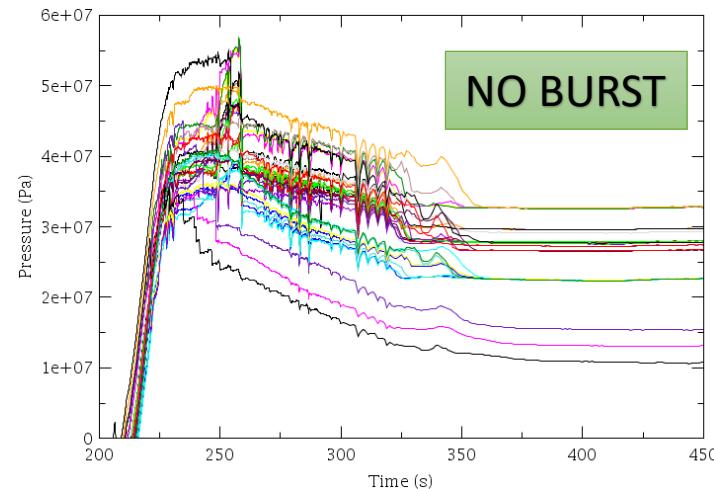
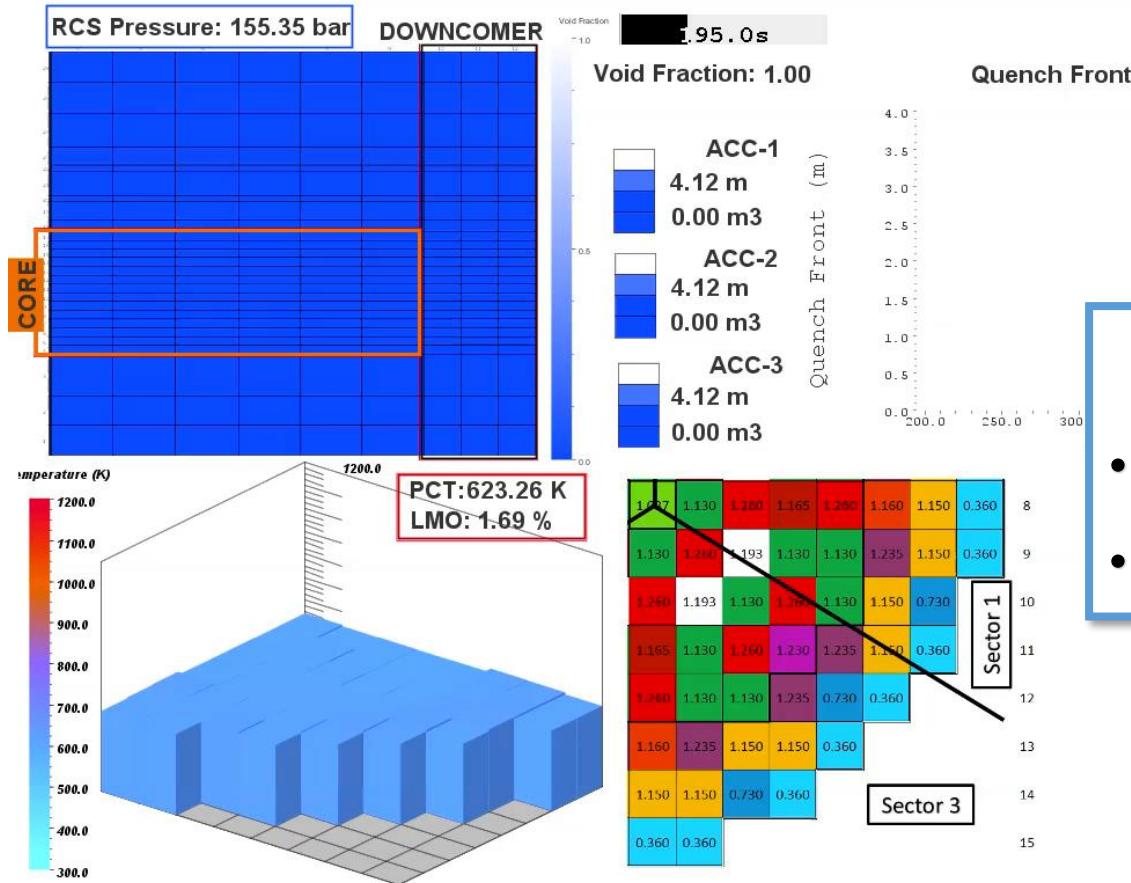


Figure: Cladding Hoop Stress.

2. Base Case: CL-LBLOCA

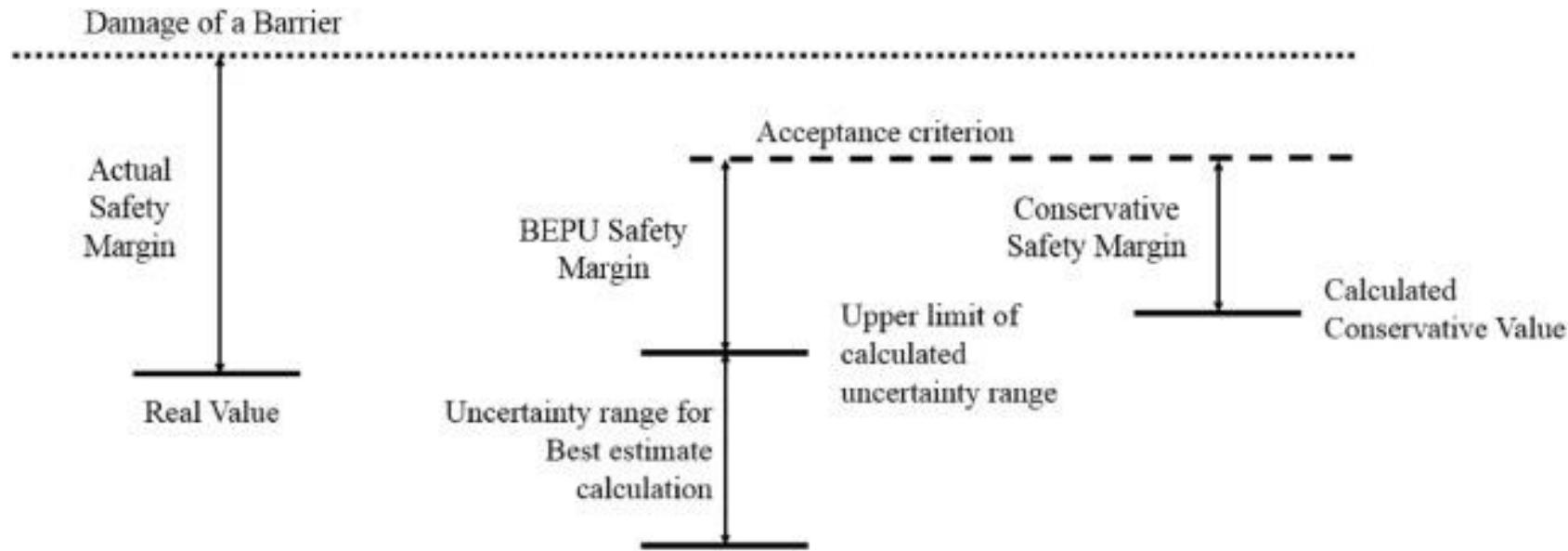
2/2 effective ACC - 2/2 LPSI



Analysis objectives

- PCT = 1169.03K < 1477K
 - LMO = 1.69% < 17%

3. Best Estimate Plus Uncertainties : CL-LBLOCA

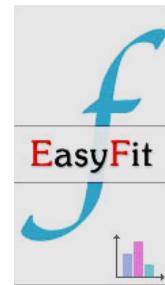


- RG-1.157, Best-Estimate Calculation of Emergency Core Cooling System Performance (1989)
- RG-1.203 Transient and Accident Analysis Methods (2005).

3. Best Estimate Plus Uncertainties : CL-LBLOCA



MATLAB®



BEPU
Methodology

Phenomena
and
Parameter
Identification

Uncertain
Parameters

Uncertainty
Analysis

Sensitivity
Analysis

3. Best Estimate Plus Uncertainties : CL-LBLOCA



MATLAB®



BEPU
Methodology

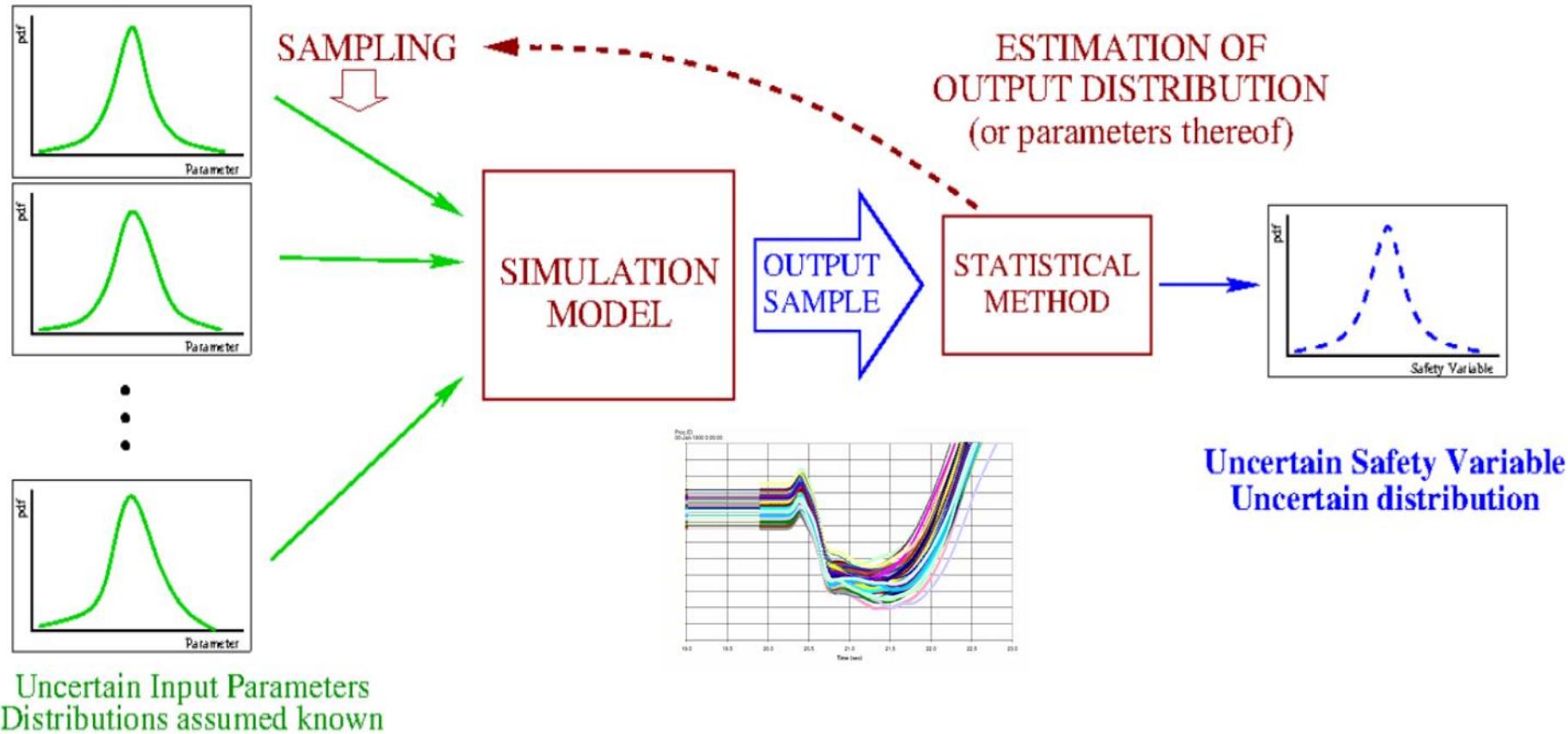
Phenomena
and
Parameter
Identification

Uncertain
Parameters

Uncertainty
Analysis

EBEPU
Remarks

3. BEPU: Methodology



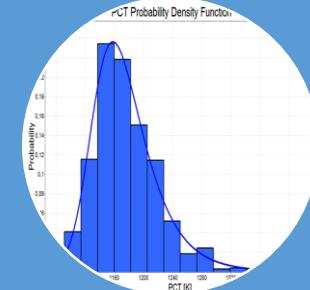
3. Best Estimate Plus Uncertainties : CL-LBLOCA

$$\beta = \sum_{j=0}^{N-R^*} \binom{N}{j} \gamma^j \cdot (1-\gamma)^{N-j}$$

$$R^* = p \cdot \left(\sum d_i \right)$$

Wilks/Wald non-parametric method

Non-parametric Binomial Distribution w Clopper-Pearson Interval

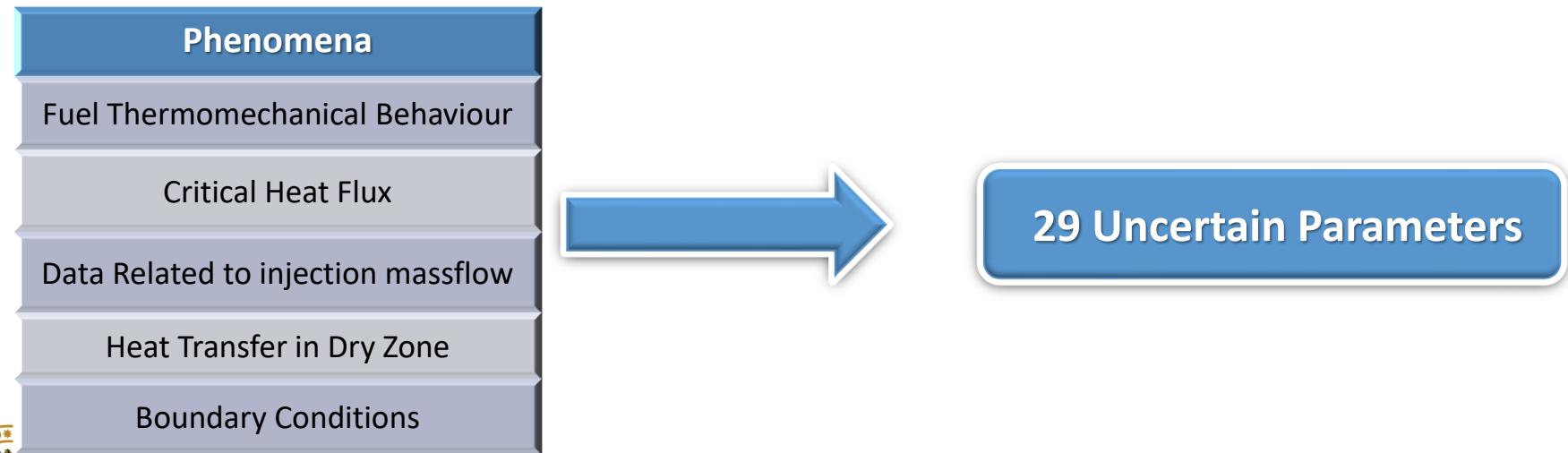


Parametric Adjustments

- ASTRUM (WEC: LBLOCA)
- WILAP verification (IB: ΔCPR, GIRALDA)
- TRACG (GE: AOO, LOCA, ATWS)
- TRACG (GE: AOO, LOCA, ATWS)

3. BEPU: Phenomena and Parameter Identification

- Identification of up to **40 parameters** in previous LBLOCA studies, including:
 - Ranges of uncertainty.
 - Probability Density Functions (PDF).
- Identification of LBLOCA phenomena: LBLOCA-PIRT (**NUREG/CR-6744**).
- Selection of parameters and phenomena by means of previous importance analyses.
- New fuel rod parameters due to the new **TRACE** thermomechanical models.
- 29 uncertain parameters** selection.

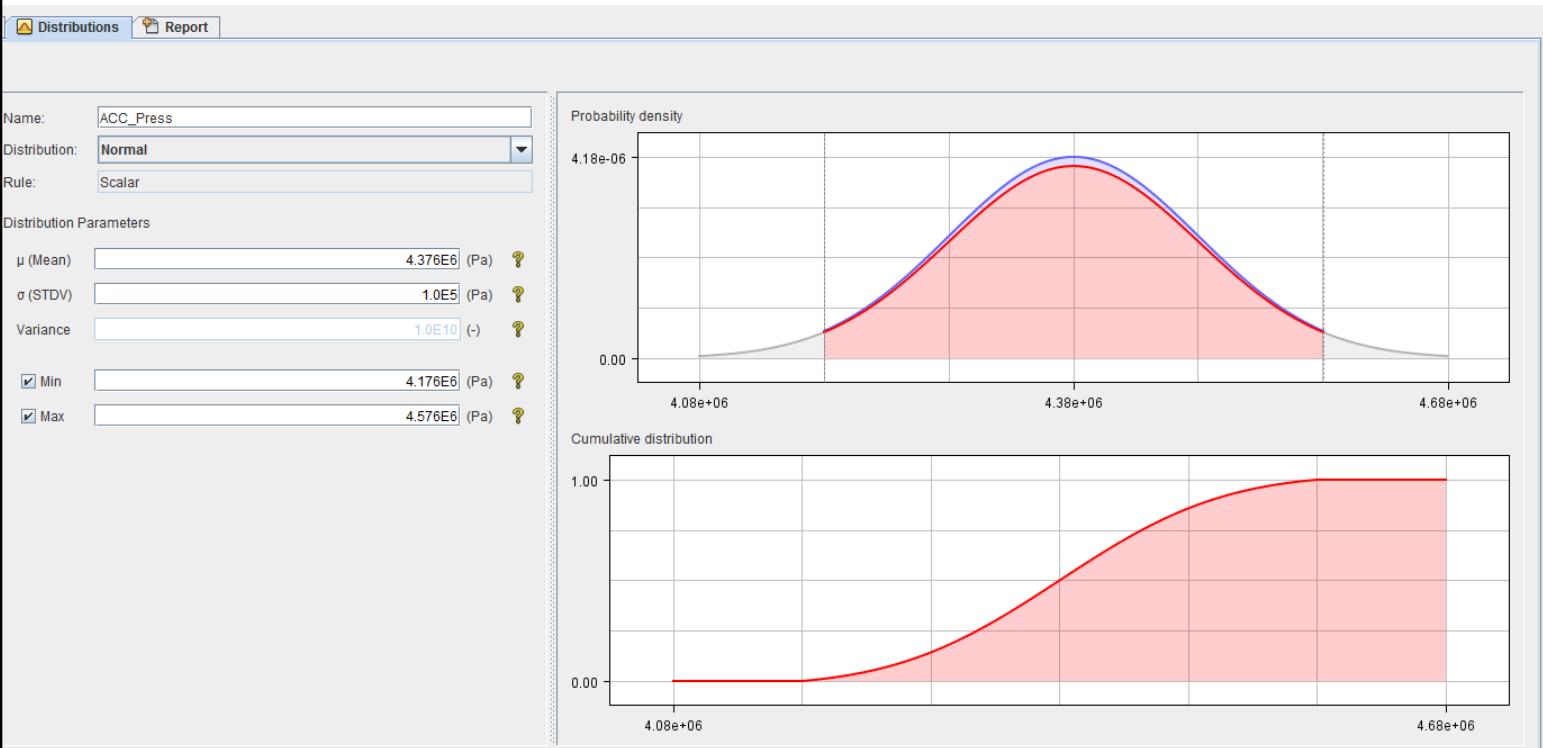




- Break DC
- Power
- DHeat
- PPF
- Forced Convection HTC
- Film Boiling HTC
- Transition Boiling
- CHF
- ACC Pressure
- LPSI injection Factor
- RCP Broken
- RCP Intact
- Gap HTC
- Cladding Inner Radius
- Cladding Thickness
- Pellet Radius
- Pellet Dish Depth
- Pellet shoulder
- Spring Volume
- Plenum Height
- Fuel Density
- KUO2
- Burst Temperature
- M-W Reaction
- Containment Press.
- ACC Temperature
- Gap Pressure
- Burst Strain
- Oxide Layer

3. BEPU: Uncertain Parameters

Characterization of uncertain parameters and Monte-Carlo sampling by means of **DAKOTA** code coupled to **TRACE** code in **SNAP** platform.



3. BEPU: Uncertain Parameters

- Characterization of uncertain parameters in **SNAP** platform:
 - Definition of the variable in **Numerics** option.
 - Model input parameters via “Select a Shared Value”.

The screenshot shows the 'Fuel Properties' table and the 'Select a Shared Real' dialog box. The table lists various fuel properties with their current values and units. The 'Select a Shared Real' dialog box lists 14 shared real variables with their names, current values, and engineering units.

Fuel Properties	
Rod Plenum Height	Rod_Plenum_Height(0.19456) (m)
Pellet Shoulder Width	Shoulder_Width(1.625E-3) (m)
Pellet Dish Depth	Pellet_Dish_Depth(1.2E-4) (m)
Pellet Height	9.83E-3 (m)
Spring Volume Fraction	Spring_Volume_Fraction(0.11) (-)
Sintering Temperature	1872.59 (K)
Maximum Density Change	105.0 (kg/m³)
PuO ₂ Fraction	0.0 (-)
Fuel Density	Fuel_density(0.93) (-)
Gap-gas Definition	Pressure
Average Gap-gas Pressure	Pgap_1x20(2.588E6) (Pa)

Select a Shared Real

Name	Current V...	Engineering Unit
<none>		
ACC_Line_Kfact	1.0E-20	No Unit (-)
BreakCD1	1.0	No Unit (-)
BreakCD2	1.0	No Unit (-)
Decay_Heat	1.0	No Unit (-)
Fuel_density	0.93	No Unit (-)
RPKF1x20	1.043	No Unit (-)
RPKF2x10	1.142	No Unit (-)
RPKF2x20	1.067	No Unit (-)
RPKF2x21	1.045	No Unit (-)
RPKE9v22	1.052	No Unit (-)

- Internal models: 37 Sensitivity coefficients available in **Model Options**.

The screenshot shows the 'Model Options' table. It lists various model options with their current values and validation status. A blue arrow points from the 'Sensitivity Coefficients' row to the right-hand table.

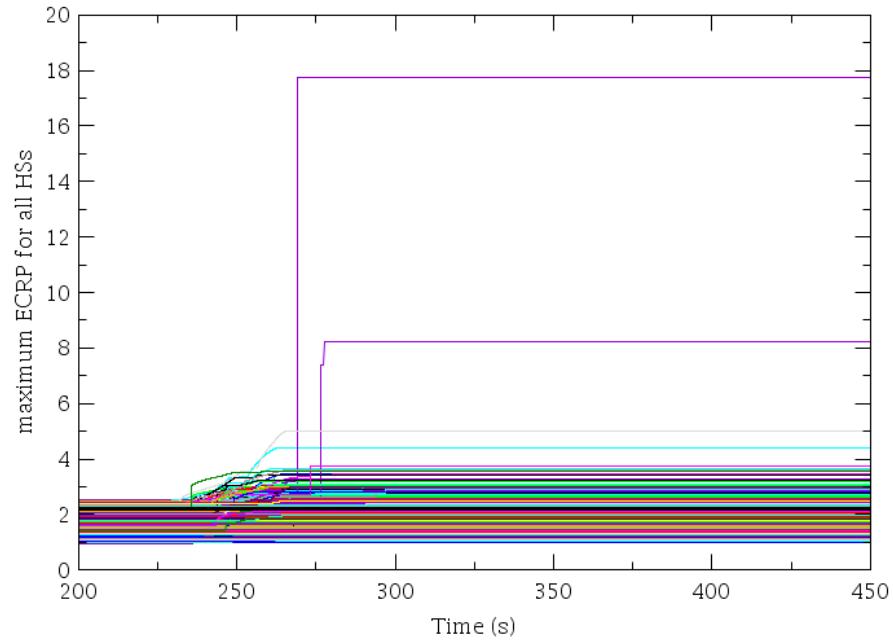
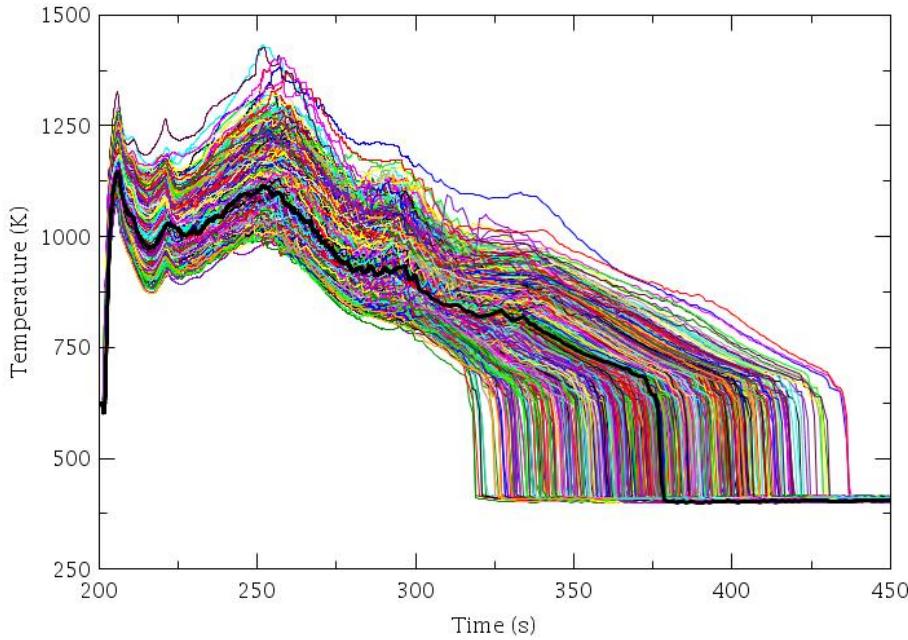
Model Options	
Namelist Variables	Valid values
User Defined Units	< none >
Timestep Data	[4] Timesteps
Trip Initiated Timestep Data	[0] Timesteps
Mixed Numerics	0 of 177 Enabled
System Gas/Liquids	None
Noncond. Gas Option	<input checked="" type="checkbox"/> [1] Air
Model Validation	[10] Active Tests: Loop Check, Noncondensable ...
Initial Condition Sets	[0] Initial Conditions Sets
Sensitivity Coefficients	[37] coefficients available

The screenshot shows the 'Sensitivity Coefficients' table. It lists 37 sensitivity coefficients with their names, types, current values, and descriptions. A blue arrow points from the 'Sensitivity Coefficients' table in the previous screenshot to this one.

[1011] dFFBHTCWallSV	[3] Factor	SV1011(1.0)	W/(m²K)	All	Dispersed Flow Film Boiling heat transfer coefficient
[1012] subchTCWallSV	[0] Off	0.0	W/(m²K)	All	Subcooled boiling heat transfer coefficient
[1013] nuclHTCWallSV	[0] Off	0.0	W/(m²K)	All	Nucleate boiling heat transfer coefficient
[1014] dnbchfWallSV	[3] Factor	SV1014(1.0)	W/m²	All	Departure from nucleate boiling / critical heat flux'
[1015] transHTCWallSV	[3] Factor	SV1015(1.0)	W/(m²K)	All	Transition boiling heat transfer coefficient
[1016] gapCondSV	[3] Factor	SV1016(1.0)	W/(m*K)	All	Gap Conductance coefficient
[1017] fuelCndBBSV	[3] Factor	SV1017(1.0)	W/(m*K)	All	Fuel Thermal Conductivity
[1018] cladMWRxnRteSV	[3] Factor	SV1018(1.0)	m²/s	All	Cladding Metal-Water Reaction Rate coefficient.
[1019] rodIntPressSV	[0] Off	0.0	Pa	All	Rod Internal (Gap) Pressure coefficient
[1020] burstTempSV	[3] Factor	SV1020(1.0)	K	All	Burst Temperature coefficient
[1021] burstStrainSV	[3] Factor	SV1021(1.0)	-	All	Burst Strain coefficient
[1022] wallDissCoeffSV	[0] Off	0.0	-	All	Wall Diss coefficient

3.1 BEPU: Uncertainty Analysis.

- **Monte Carlo sampling: n=1020**

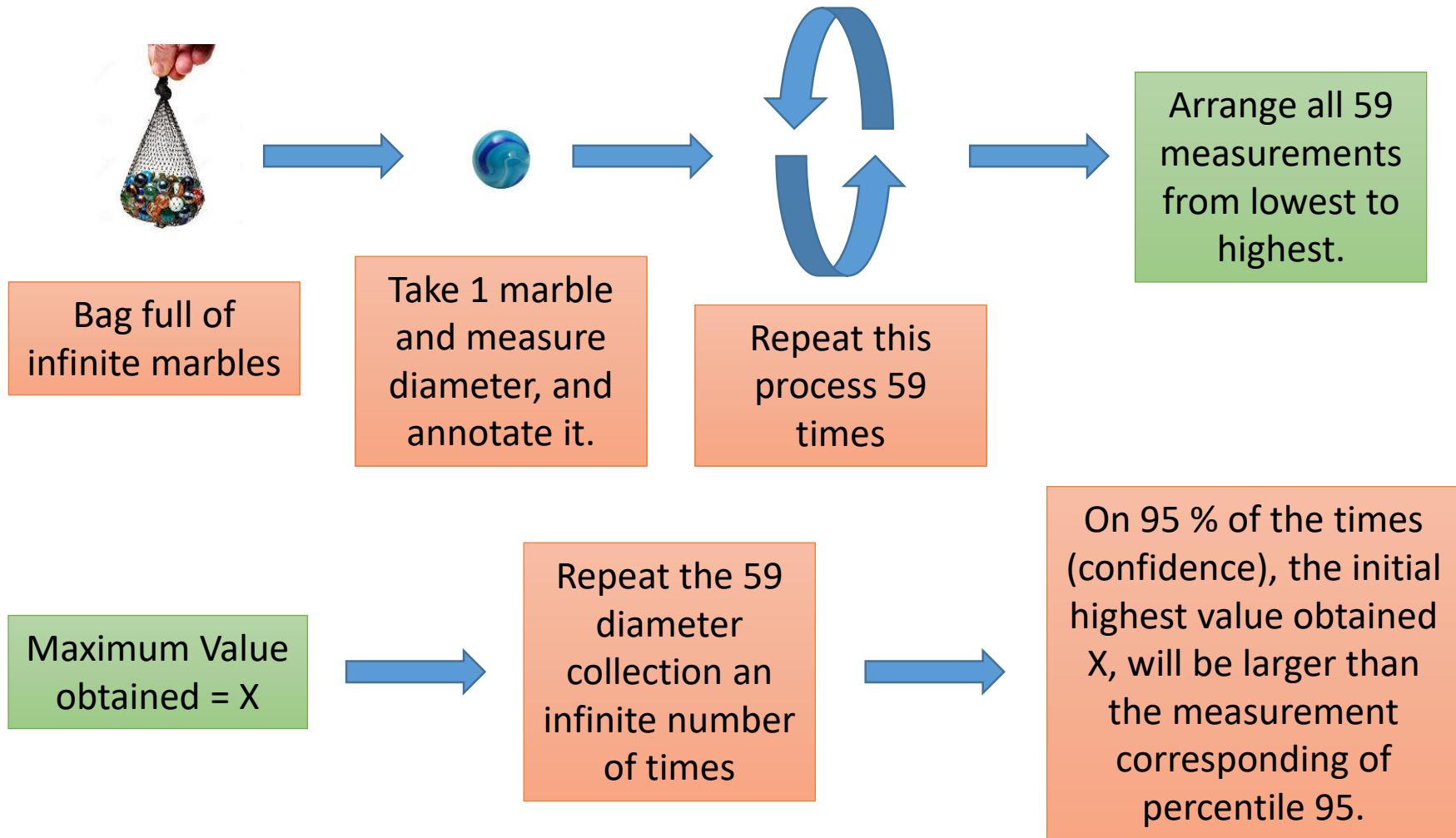


Non-parametric statistics results:

1. **Wald criterium**
2. **Binomial distribution with confidence interval.**

3. BEPU: Methodology

Wilks/Wald statistics: Easy Example



3. BEPU: Methodology

Wilks/Wald statistics:

- Monte-Carlo random sampling of input and internal models parameters (defined by their PDFs) with **DAKOTA** code.
- Number of code calculations **N**:
 - Depends on the probability content γ , confidence level β of the acceptance criteria (PCT, LMO) and order p . Number of bounds (upper and/or lower) = d . $R = n^o$ of FOM
 - Independent of the number of uncertain parameters
- BEPU LBLOCA analysis: $\gamma/\beta = 95/95$, PCT and LMO as output parameters, one sided-tolerance. **93 (p=1) cases are enough to verify the 95/95 acceptance criteria.**

$$\beta = \sum_{j=0}^{N-R^*} \binom{N}{j} \gamma^j \cdot (1-\gamma)^{N-j}$$

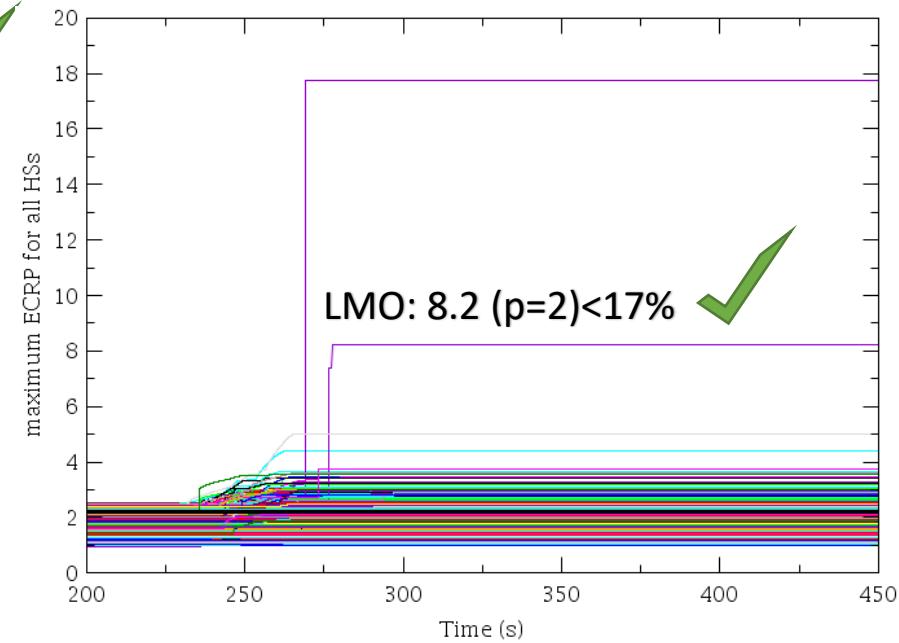
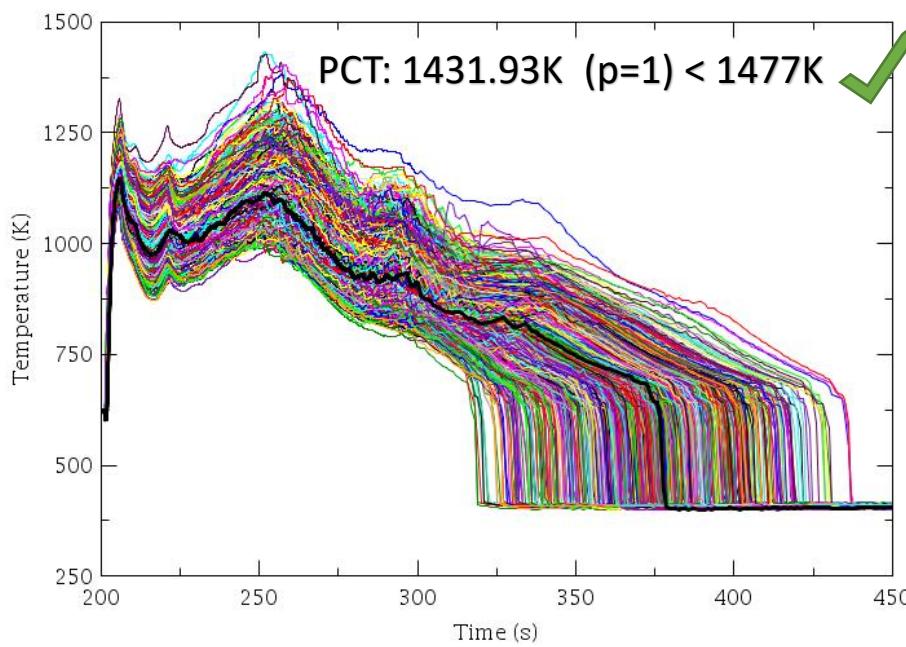
$$R^* = p \cdot \left(\sum_{i=1}^R d_i \right)$$

β	γ values		p
	0.95	0.99	
0.95	93	473	1
	153	773	2
	208	1049	3
0.99	130	662	1
	198	1001	2
	259	1307	3

Wald with two FOM one-sided

3.1 BEPU: Uncertainty Analysis

- Non-parametric statistics results. Wald criterium.



β	γ values	p
	0.95 0.99	
	probability	
93	473	1
153	773	2
208	1049	3
...
0.95	1013	20
0.99	130 198 259	5088

confidence level

95/95 and n=1020 → p=20
 PCT= 1291 K LMO= 3.1 %

For p=2 and n=1020 $\beta = 95\%$, $\gamma = 99,22\%$

Joint Core Damage Probability < 1E-2 with 99% CL

3.1 BEPU: Uncertainty Analysis.

- **Non-parametric statistics** results. **Binomial distribution with confidence interval**
- Binomial distribution. **PCT** or **LMO**
 - > acceptance criterion:
 - Failure/no-failure.
- The 95% confidence interval has been obtained by means of the **Clopper-Pearson**.
- Equivalent to Wilks one-sided.
- Joint Core Damage Probability with 95% confidence interval:

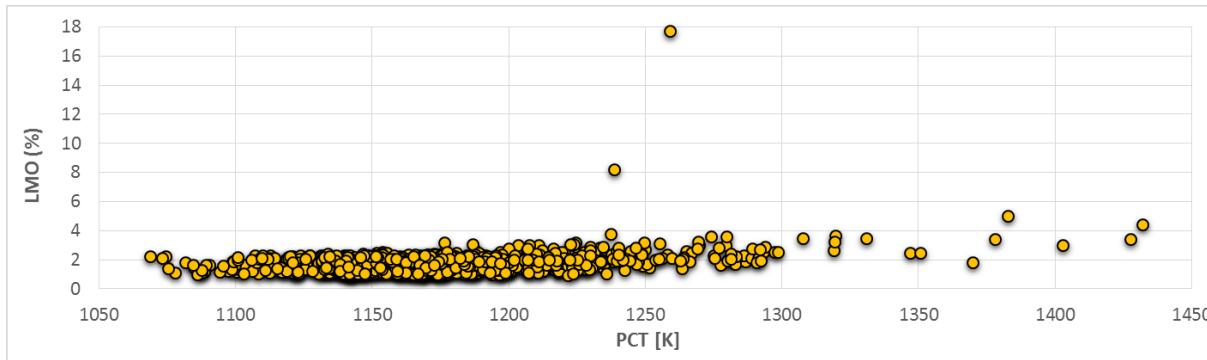
Mean = 0.98E-1%

Upper 95% = 5.5E-1%



3.1 BEPU: Uncertainty Analysis.

- Parametric statistics results. It is quite difficult to obtain a single JPDF for PCT and LMO



- The goodness of fit (**GOF**) tests measure the compatibility of a random sample with a theoretical probability distribution function.
- MC results have been measured against **3 GOF tests** and up to **60 different distributions**:
 - Kolmogorov-Smirnov
 - Anderson-Darling
 - Chi-Squared
- **Anderson-Darling** criterion is used as benchmark due to the test giving more weight to the tails than the Kolmogorov-Smirnov and Chi-Squared tests.
 - More adequate to determine exceedance probabilities.

3.1 BEPU: Parametric Statistics - PCT

- Parametric statistics results: Probability Density Functions

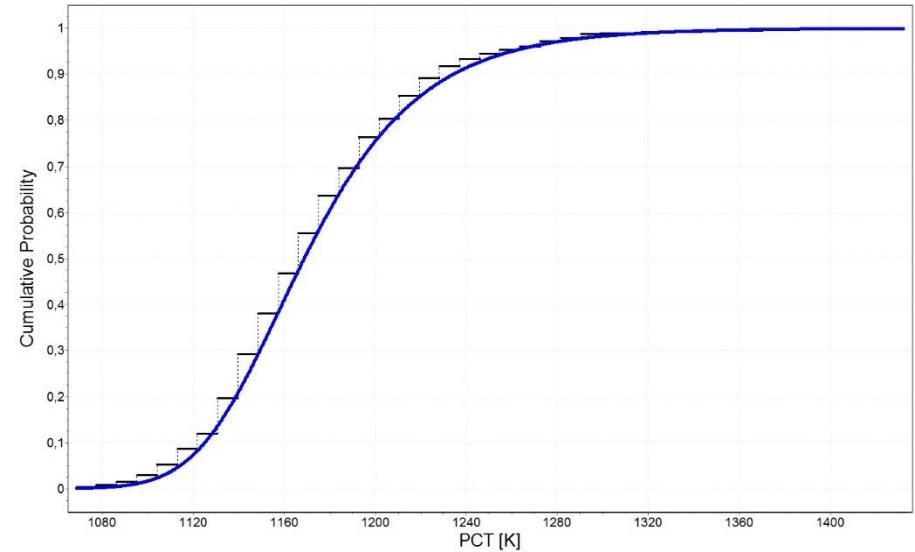
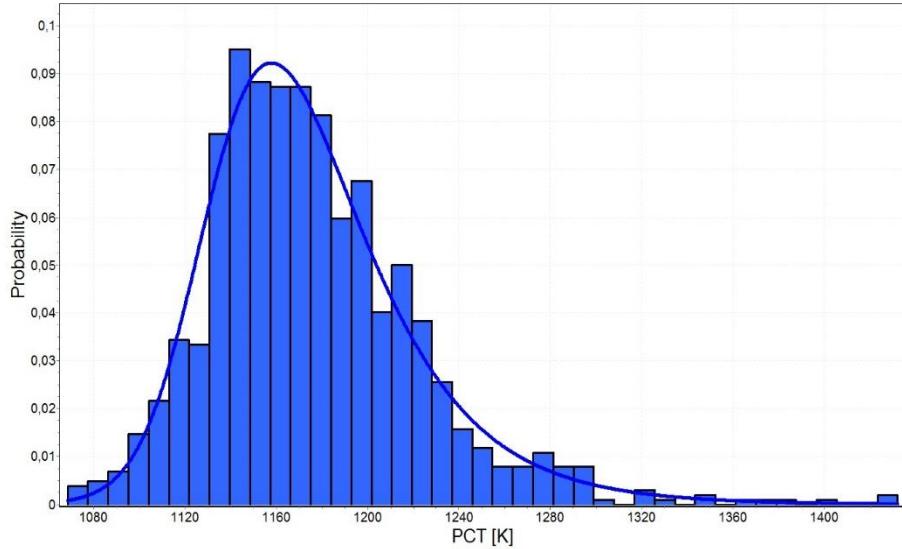


Figure: (Left) PCT histogram and associated PDF. (Right) PCT CDF

- PCT Monte Carlo results fit to the following PDF:
 - PCT (Johnson SU): $P(PCT > 1477\text{K}) = \text{2.7E-2\%} (< 5\%)$.

But this value needs a Confidence interval.

3.1 BEPU: Parametric Statistics - LMO

- Parametric statistics results: Probability Density Functions

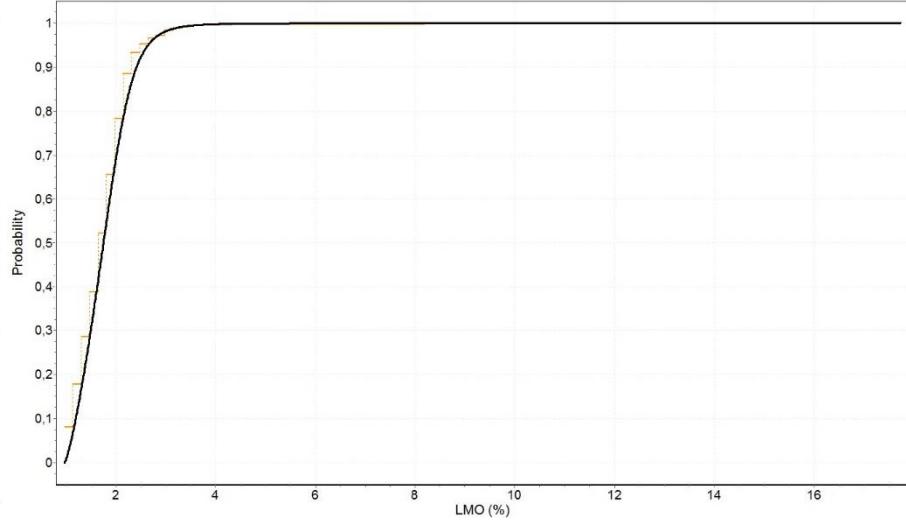
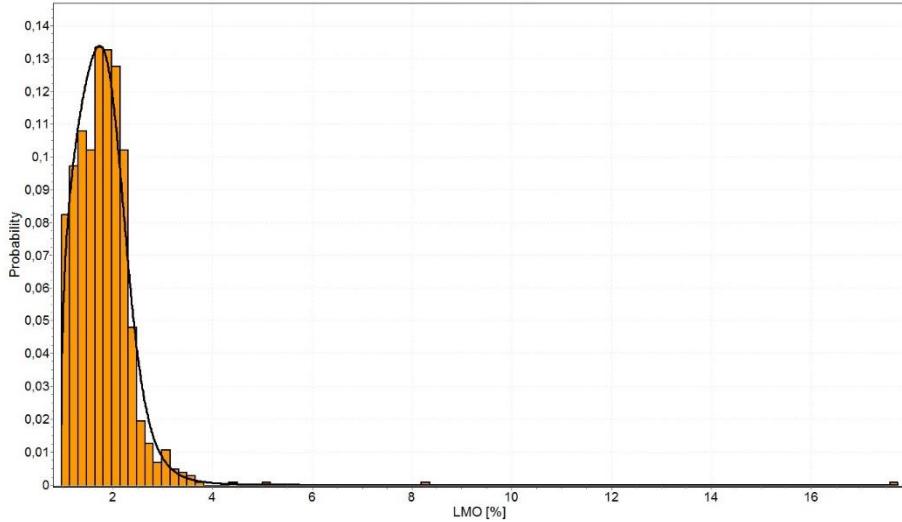


Figure: (Left) LMO histogram and associated PDF. (Right) LMO CDF

- LMO Montecarlo results fit to the following PDF:
 - LMO (**Dagum**): $P(LMO > 17\%) = \textcolor{blue}{1.5E-5\%} (< 5\%)$.

But this value needs a Confidence interval.



3.1 BEPU: Parametric Statistics

- **Parametric statistics** : Different Methods to obtain a confidence Interval.
 - Creating a p-box using confidence interval of the distribution parameters by:
 1. Using **Wald confidence interval** for distribution parameters
 2. Using the **Likelihood Function** to obtain distribution parameters confidence interval
 - Both methods are easily applicable **only for a limited set of distributions** (e.g. normal).
 - Using the computationally demanding Bootstrap Method. (fully used in the present study)

3.1 BEPU: Parametric Statistics

- Parametric statistics results: 500 Bootstrap Samples to obtain the confidence interval

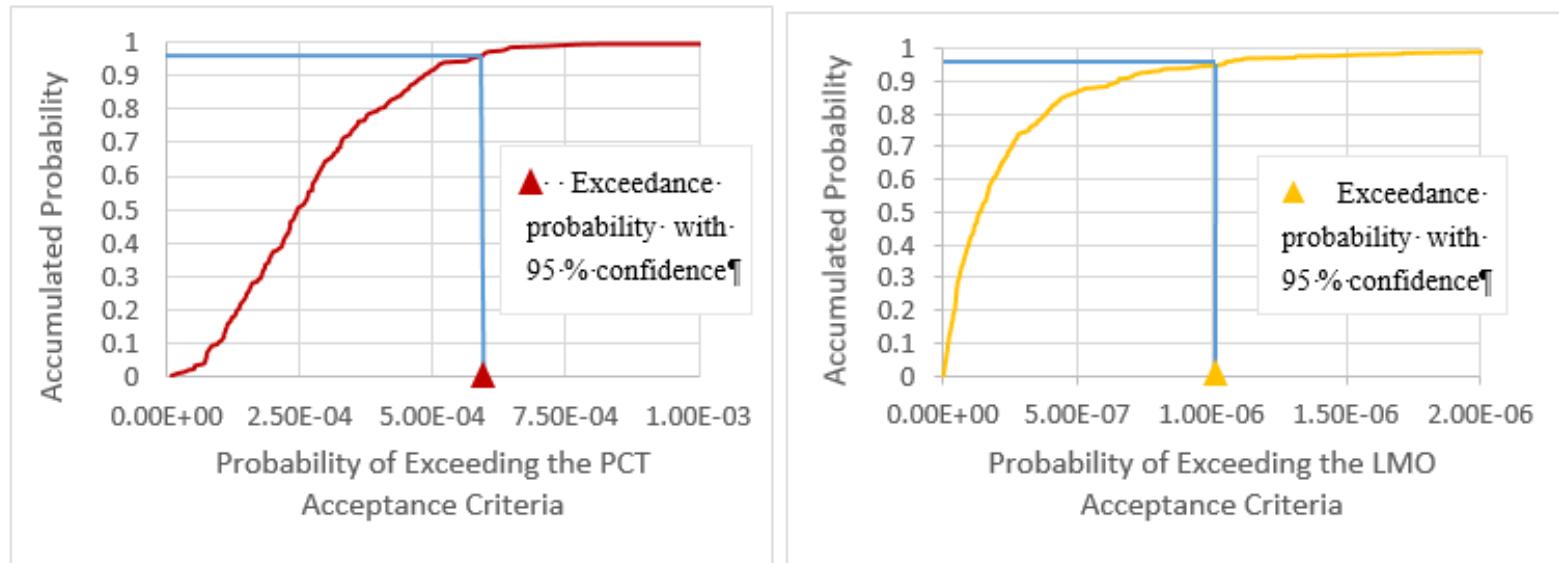


Figure: CDF of the 500 Bootstrap Samples for PCT (left) and LMO (right)

- Upper confidence interval for PCT : **6.00E-02%** for 95 % confidence
- Upper confidence interval for LMO: **1.02E-04%** for 95 % confidence



3.3 BEPU:Comparison Non- parametric .vs. Parametric Statistics

- Non-Parametric statistics results: Wald criterium

Joint Core Damage Probability < 7.8E-01% with 95% CL

- Binomial distribution with 95% confidence interval (Clopper-Pearson)

Joint Core Damage Probability < 5.5E-01% with 95 % CL

- Parametric statistics results: Probability Density Functions with Bootstrap

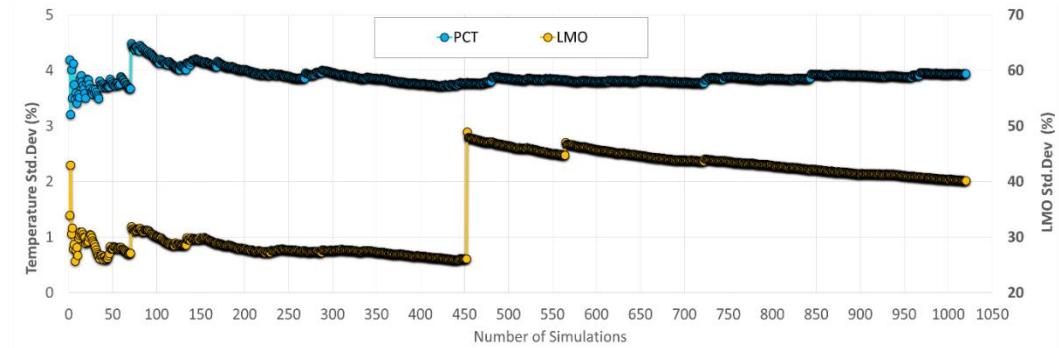
Joint Core Damage Probability < 6.01E-02% with 95% CL

Wald is conservative compared with parametric results

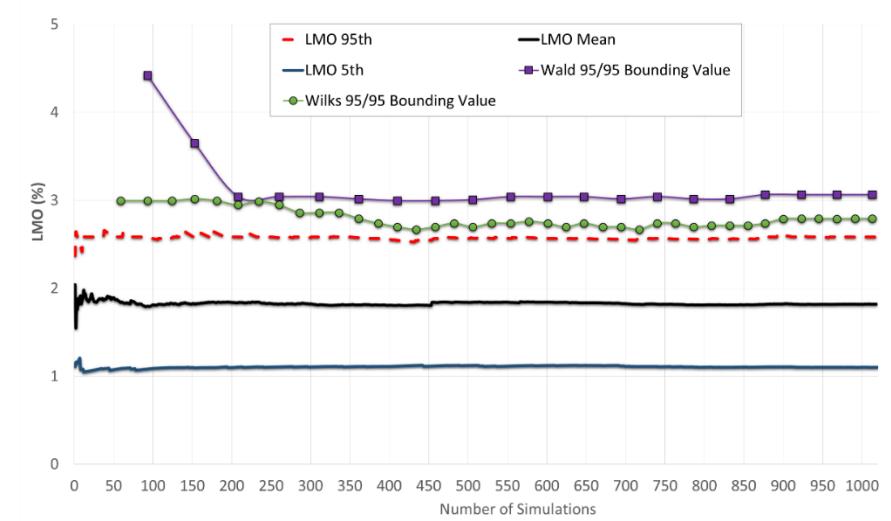
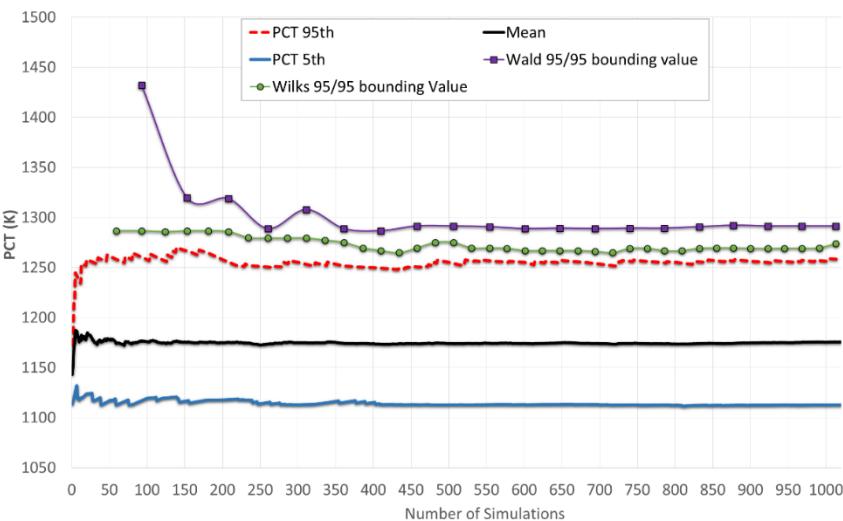
3.1 BEPU: Uncertainty Analysis

- Sample discussion: Parameters Convergence.

Convergence of Values across the simulation



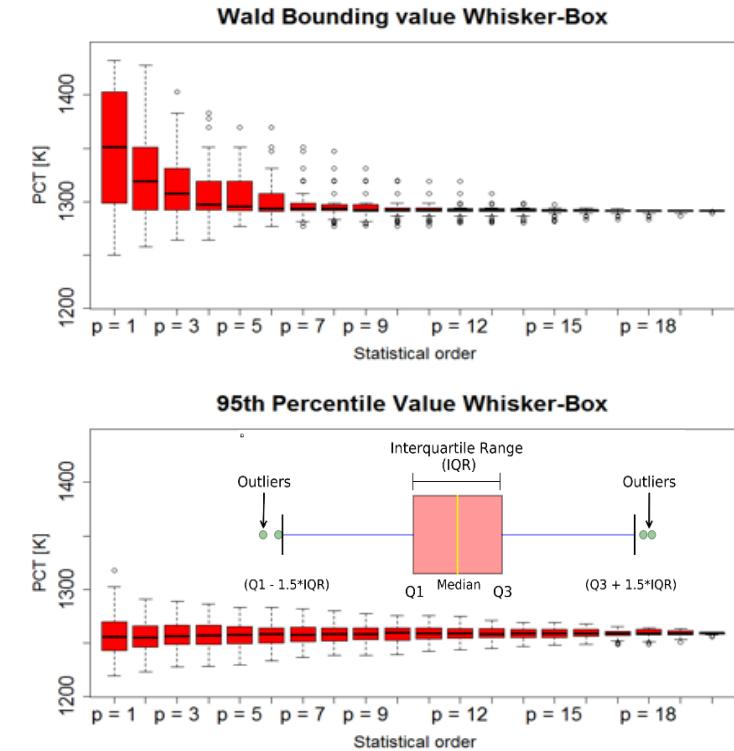
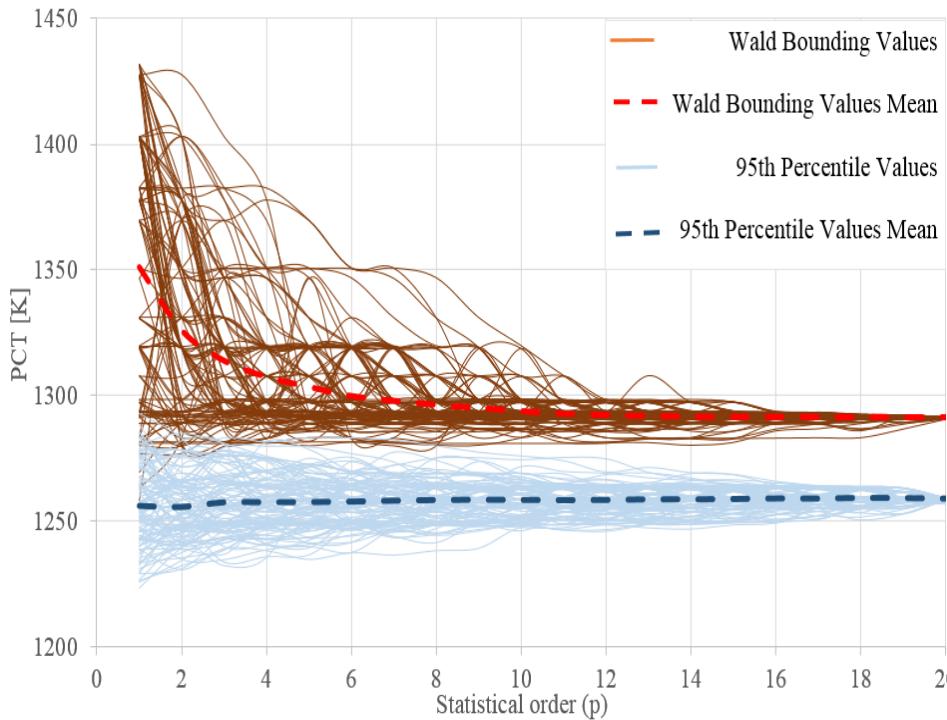
Low order shows higher variance in all parameters



3.1 BEPU: Uncertainty Analysis

- **Sample discussion:** Parameters Convergence.

500 samples taken in random order (without substitution)

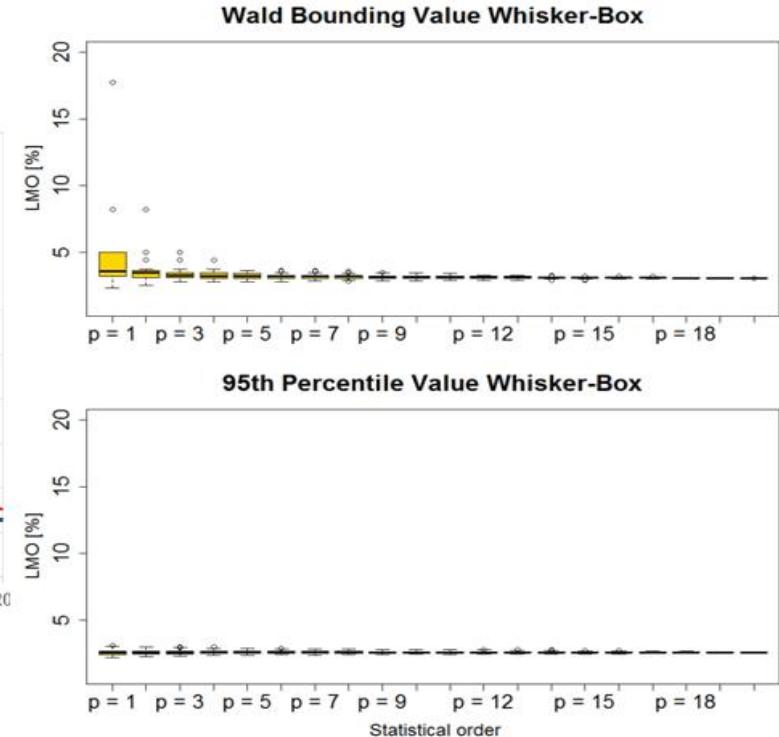
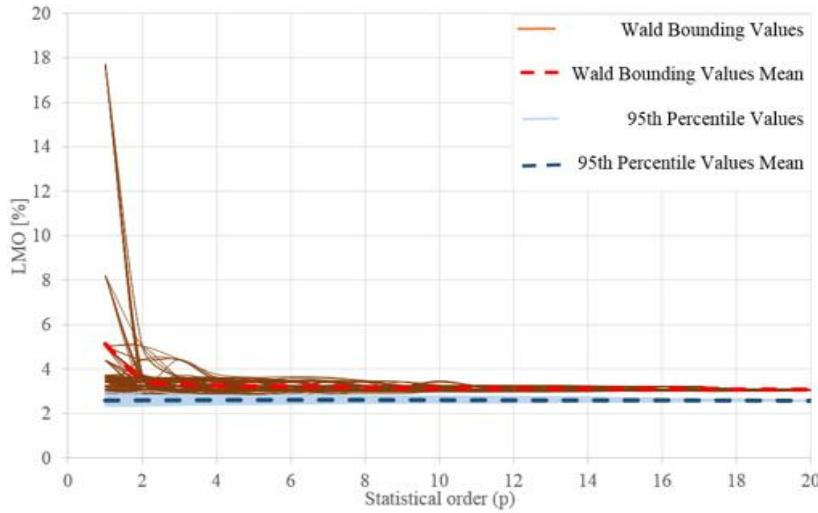


With a low order, the Wald bounding value tends to be more conservative

3.1 BEPU: Uncertainty Analysis

- **Sample discussion:** Parameters Convergence.

500 samples taken in random order (without substitution)

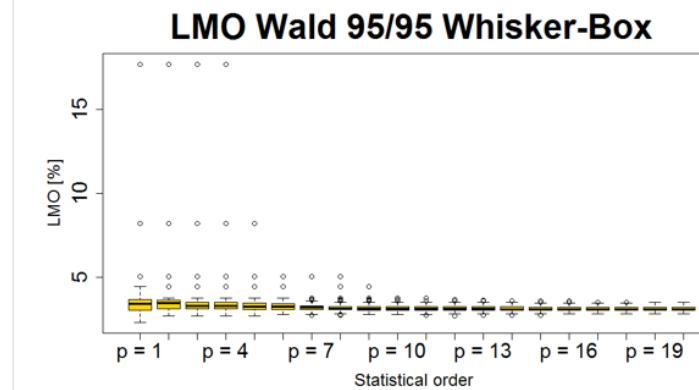
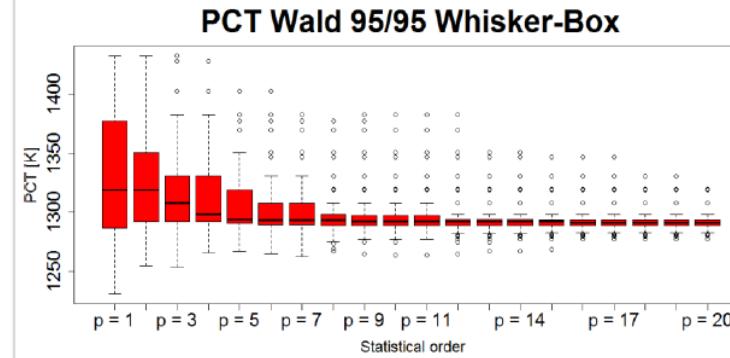
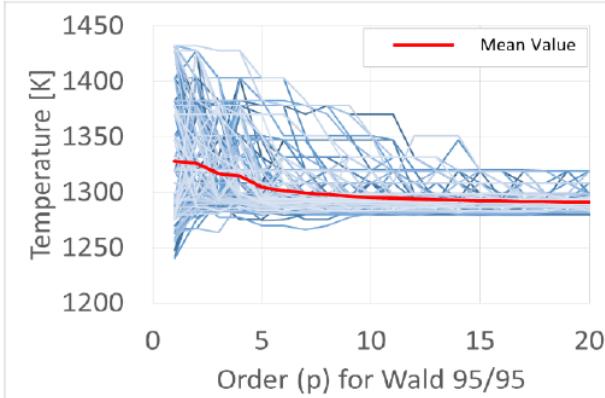


With a low order, the Wald bounding value tends to be more conservative

3.1 BEPU: Uncertainty Analysis

- Sample discussion: Parameters Convergence

500 Bootstrap (with substitution) samples for LMO and PCT



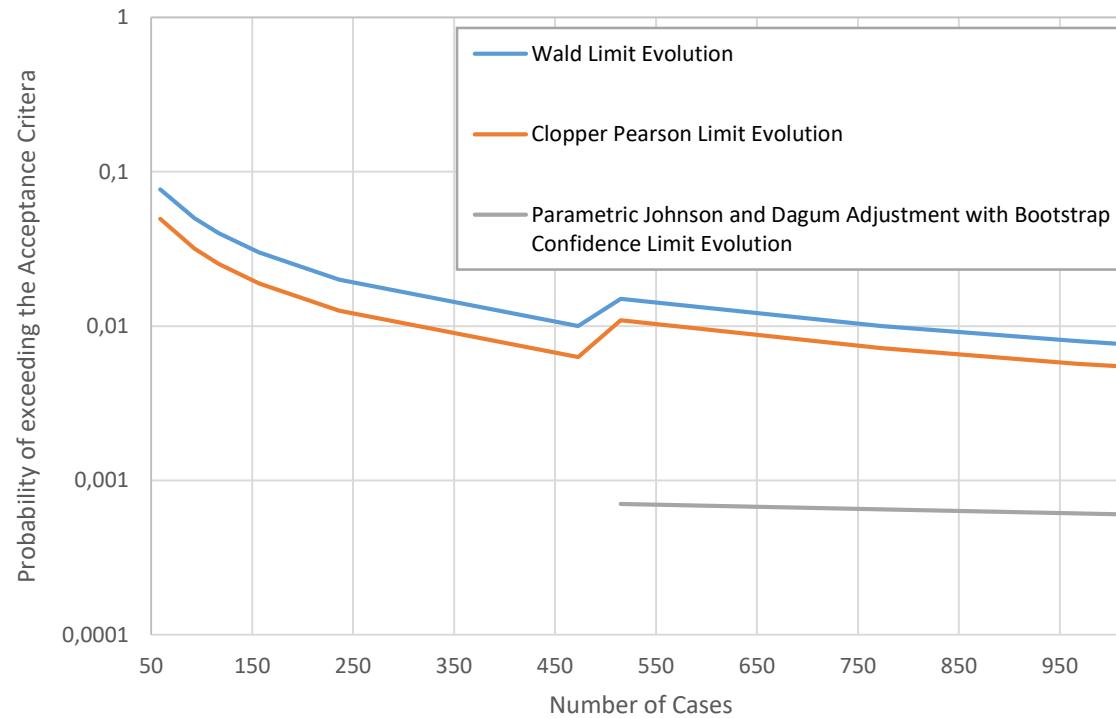
Confidence interval for Wald bounding values can be obtained with Bootstrap approach.

With a low order, the Wald bounding value tends to be more conservative

3.1 BEPU: Uncertainty Analysis

- **Sample Discussion:**

Evolution of the exceedance probability 95% confidence level with increasing number of simulations



With a low order, the Wald bounding value tends to be more conservative



4. EBEPU - Introduction

Can methods from PSA
be integrated in DSA to
obtain deeper insights of
a NPP sequence?

TABLE 1. OPTIONS FOR PERFORMING DETERMINISTIC SAFETY ANALYSIS

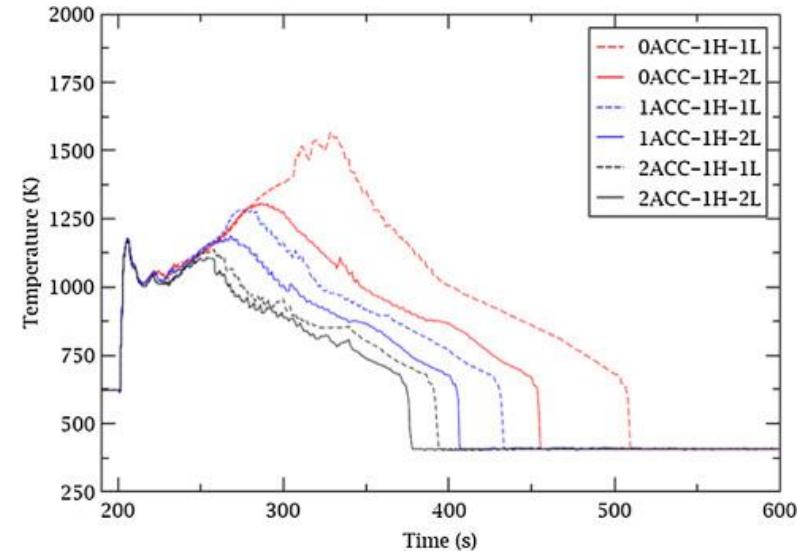
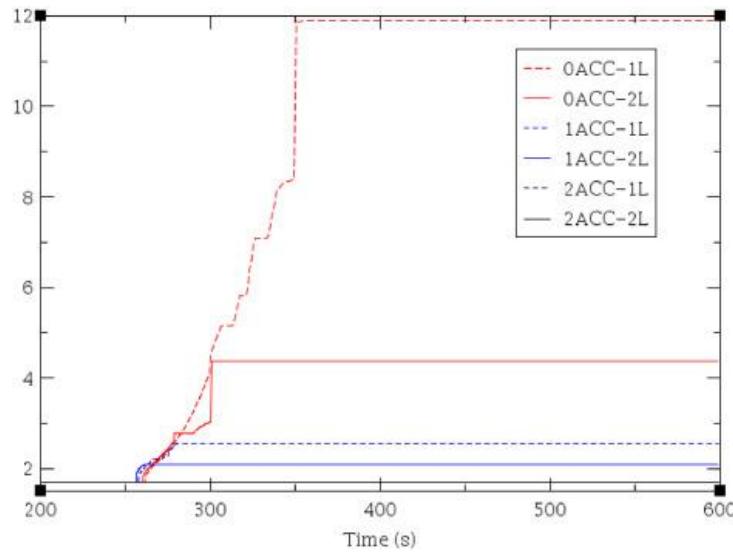
Option	Computer code type	Assumptions about systems availability	Type of initial and boundary conditions
1. Conservative	Conservative	Conservative	Conservative
2. Combined	Best estimate	Conservative	Conservative
3. Best estimate plus uncertainty	Best estimate	Conservative	Best estimate Partly most unfavourable conditions
4. Realistic*	Best estimate	Best estimate	Best estimate

Deterministic safety analysis for nuclear power plants. Specific Safety Guide No. SSG-2. IAEA 2019

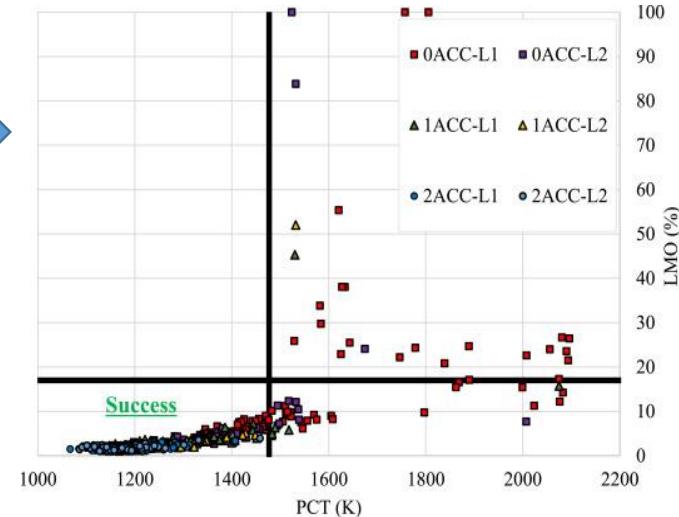
4. EBEPU - Example



4. EBEPU



100 Cases with Monte-carlo Sampling for assessing uncertainties

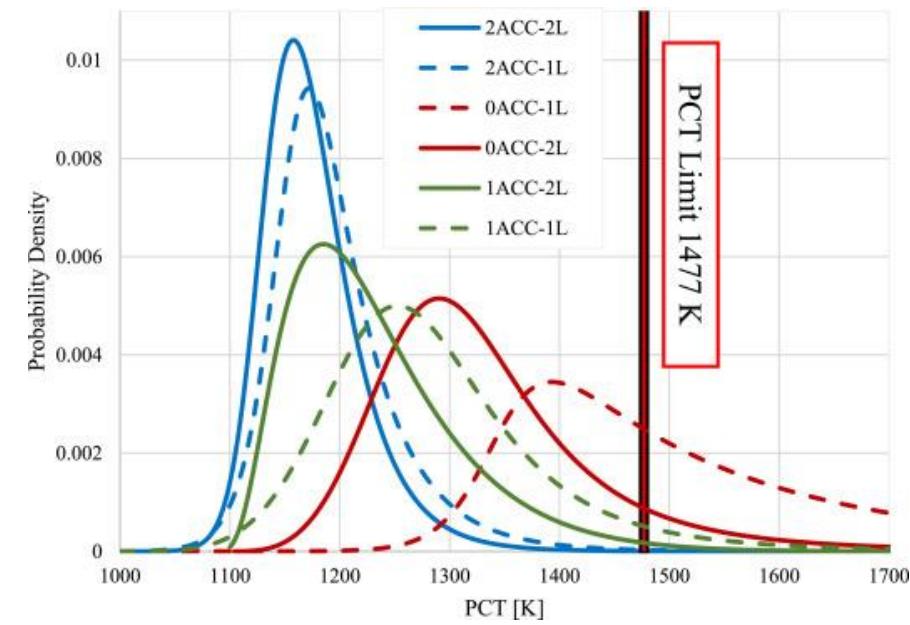
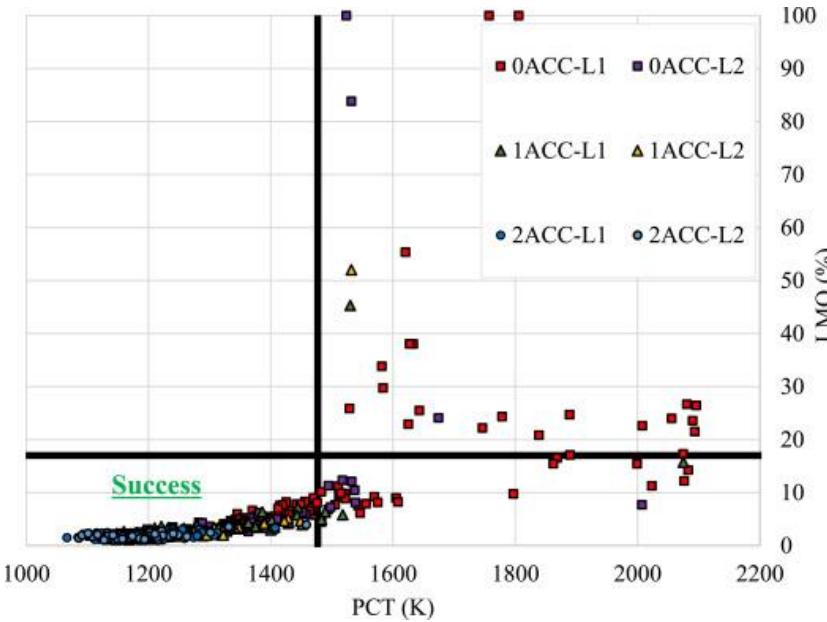


100 Cases with Monte-carlo Sampling for assessing uncertainties





4. EBEPU



Non-Parametric
Statistics



Parametric
Statistics



4. EBEPU

FROM PSA

FROM DSA BEPU

Sequence	Pseq	IE · Pseq [year $^{-1}$]	100% Power Damage probability	CDF [year $^{-1}$]
2ACC-2L	9.69E-01	1.11E-03	2.95E-02	3.27E-05
2ACC-1L	3.00E-02	3.45E-05	2.95E-02	1.02E-06
2ACC-0L	9.00E-04	1.03E-06	1.0	1.03E-06
1ACC-2L	2.58E-04	2.97E-07	4.66E-02	1.38E-08
1ACC-1L	8.00E-06	9.20E-09	1.02E-01	9.38E-10
1ACC-0L	2.40E-07	2.76E-10	1.0	2.76E-10
0ACC-2L	2.91E-05	3.34E-08	1.80E-01	6.01E-09
0ACC-1L	9.00E-07	1.04E-09	5.40E-01	5.62E-10
0ACC-0L	2.70E-08	3.11E-11	1.0	3.11E-11
No-recirc	6.41E-05	7.31E-08	1.0	7.31E-08
Total	1.00E + 00	1.15E-03	3.04E-02	3.49E-05



5. Conclusions – LBLOCA BEPU

- The analyses performed have allowed to evaluate the new thermomechanical models included in **TRACE**.
- Results show the compliance with the **10 CFR 50.46** requirements for the PCT and LMO:
 - Non-parametric analysis proves the compliance with a 95/95 probability and confidence.
 - PCT and LMO PDFs obtained by means of parametric analyses show marginal probabilities of acceptance criteria exceedance.
- Employing Extended-BEPU technique to assess a transient can provide interesting outcomes not seen with just PSA or DSA.
- All these discussions **can be seen in:**



Reliability Engineering & System Safety
Volume 205, January 2021, 107246



Application of Expanded Event Trees combined with uncertainty analysis methodologies

Annals of Nuclear Energy
Volume 144, 1 September 2020, 107505



Reliability Engineering & System Safety
Volume 193, January 2020, 106607



Statistical characterization of NPP transients:
Application to PWR LBLOCA

Uncertainty and sensitivity analysis of a PWR LOCA sequence using parametric and non-parametric methods



Statistical Analysis of NPP transients: LBLOCA in a PWR; Parametric, non-parametric methods and EBEPU

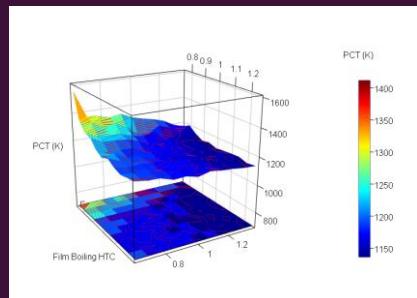
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Seminarium Zakładu Energetyki Jądrowej i Analiz Środowiska, Jan 2022



Extra Slides

