

The Quest Towards Analytical Solutions of Linked Fault Tree Models using Binary Decision Diagrams^(*)

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- Motivation and issues with current PSA softwares
- Binary Decision Diagrams (BDD) as an alternative
- Research and development at KKL and ETH Zurich
- Insights and outlook
- Presentation of the *NeuralSpectrum* Software



(*) From the PhD thesis:

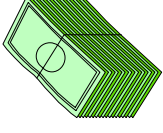
“Analytical Solutions of Linked Fault Tree Probabilistic Risk Assessments using Binary Decision Diagrams with Emphasis on Nuclear Safety Applications”

Motivation and issues with current PSA softwares

- **The (Swiss) NPP have to submit best-estimate, plant-specific PSA models to the regulatory authority**
 - ⚙ Level 1 PSA: Calculation of the Core Damage Frequency (CDF)
 - ⚙ Level 2 PSA: Calculation of the Plant Damage State (PDS) frequencies and associated radiological consequences
 - ⚙ For internal events, area events and external events
 - ⚙ Typically, the PSA modeling techniques are based on the Fault Tree / Event Tree approach (FTA)

Motivation and issues with current PSA softwares

- **Risk Informed Applications**
 - ⚙ Evaluation and support of Plant / TechSpec Modifications (e.g. relaxation)
 - ⚙ Risk Informed In Service Inspection
 - ⚙ Aging Programs
 - ⚙ ILRT frequency relaxation
 - ⚙ Outage optimization



Motivation and issues with current PSA softwares

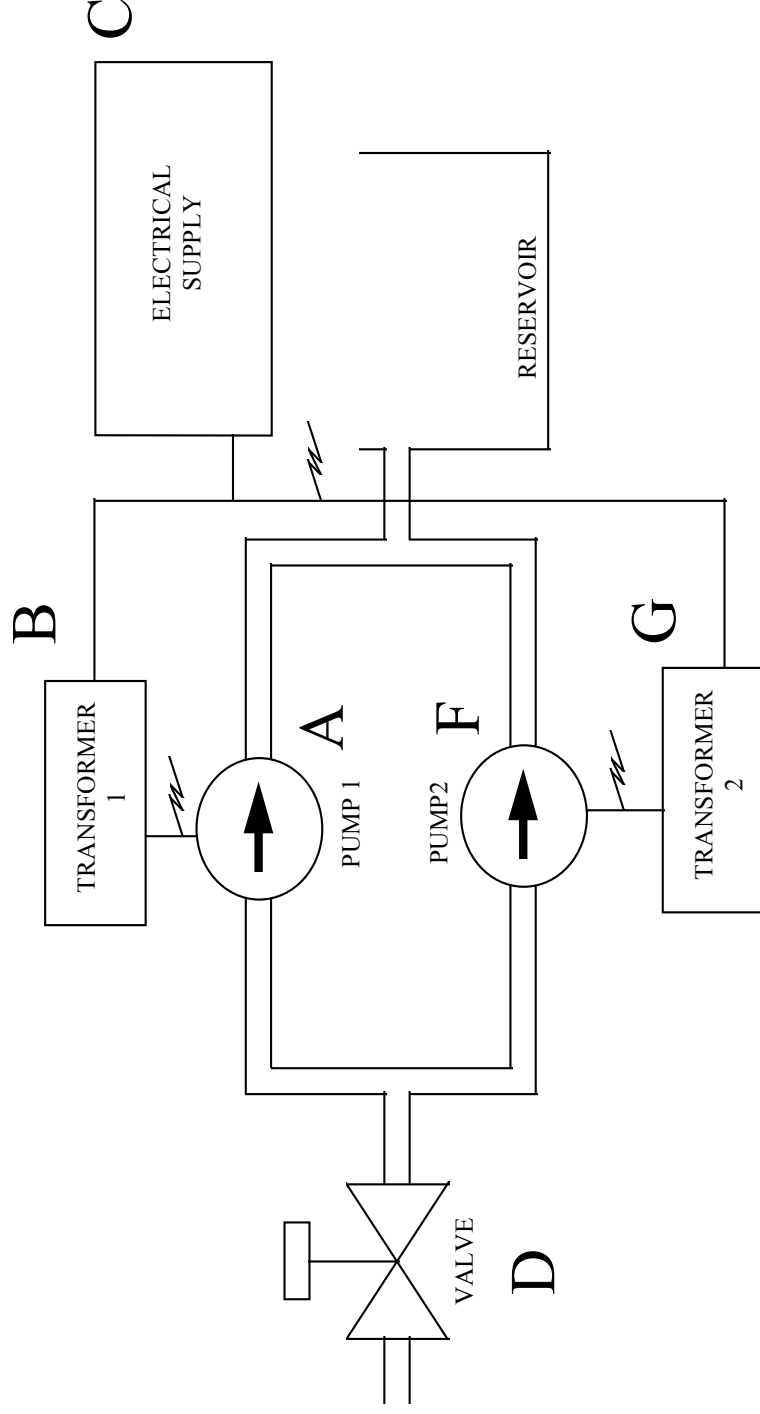
- **Worldwide Fault Tree Analyses (FTA) are performed with codes that produce wrong results**
 - ⚙ Rare event approximation...
 - ⚙ ... but not only!
- **Resulting risk Importance measures are even more wrong (conservative or optimistic, no one can know)**

$$RIF_{x_i} = \frac{CDF(x_i = 1)}{CDF}$$

↖ ↗

Motivation and issues with current PSA softwares

- What is the mean unreliability of the system's function, based on individual Basic Events probabilities ?

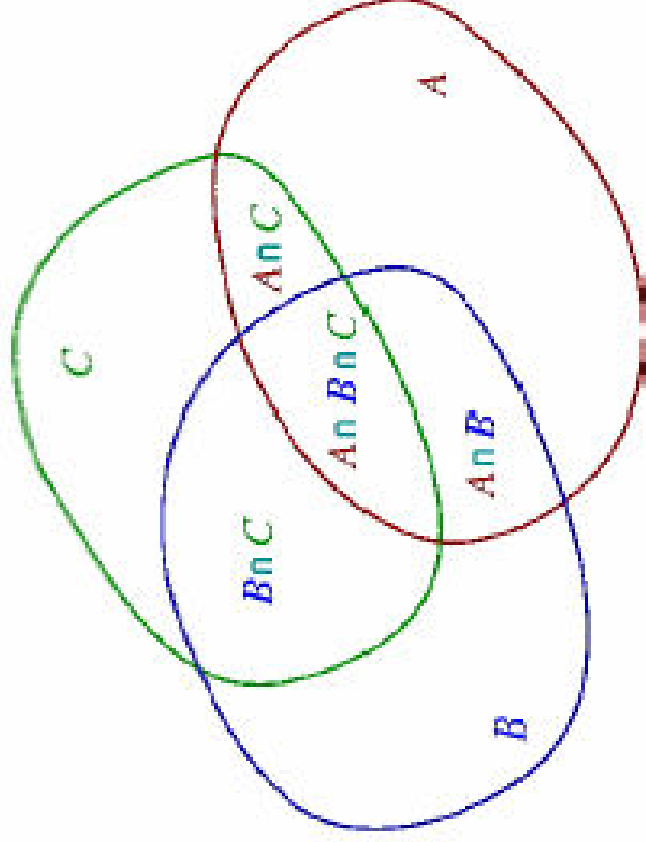


$$P_{top} = D + C + AF + AG + BF + BG$$

Motivation and issues with current PSA softwares

- The rare event approximation (Moivre's equation)

$$|A_1 \cup \dots \cup A_p| = \sum_{1 \leq i \leq p} |A_i| - \sum_{1 \leq i_1 < i_2 \leq p} |A_{i_1} \cap A_{i_2}| + \dots + (-1)^{p-1} |A_1 \cap \dots \cap A_p|$$

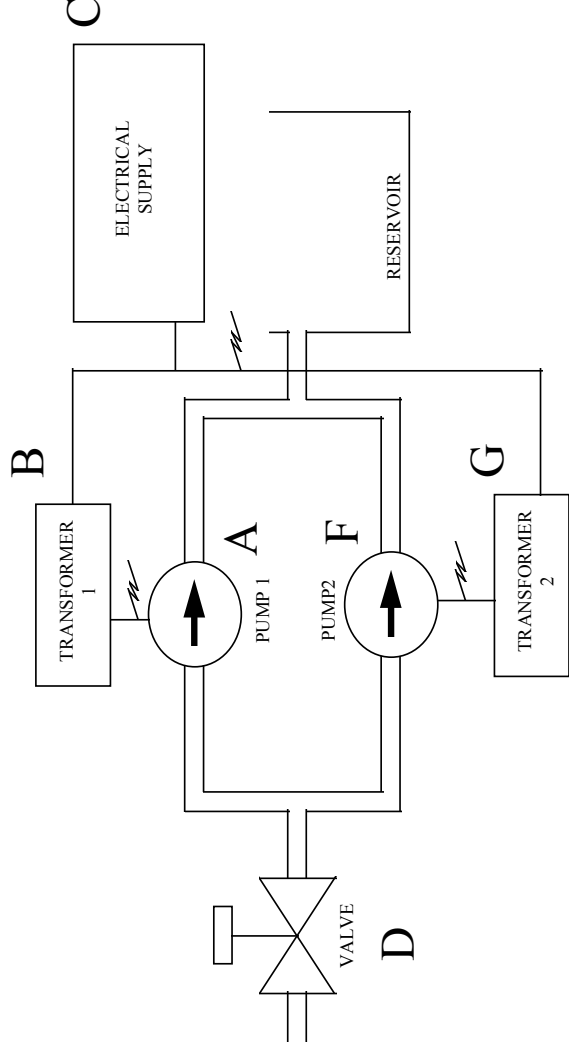


Motivation and issues with current PSA softwares

- Analytically correct result yields:

$$P_{\text{top}} = (d + f + g + c - d f - d g - d c - f g - f c - g c + d f g + d f c + d g c + f g c - d f g c) (a + b + c + d - a b - a c - a d - b c - b d - c d + a b c + a b d + a c d + b c d - a b c d)$$

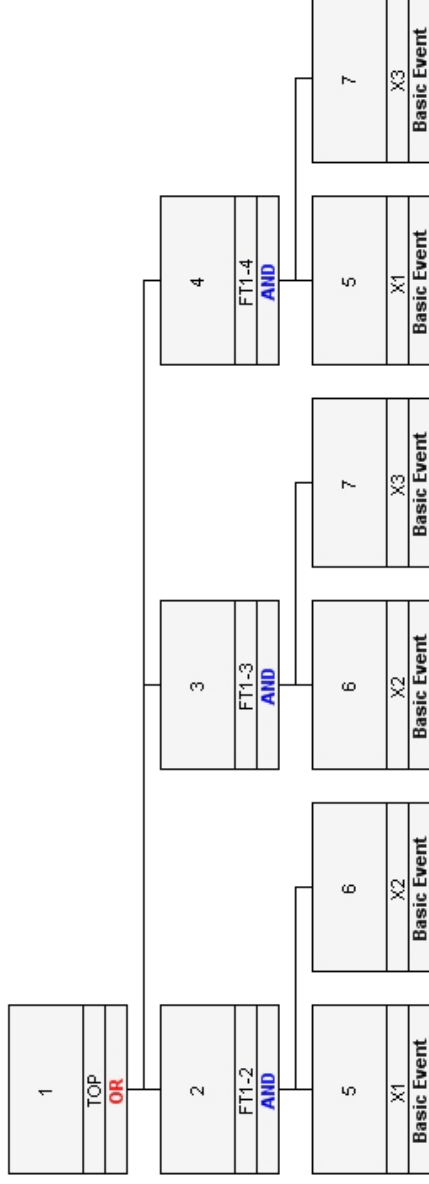
$$= [c - a (-1 + b) (-1 + c) (-1 + d) + b (-1 + c) (-1 + d) + d - c d] \cdot [f - c (-1 + d) (-1 + f) (-1 + g) + d (-1 + f) (-1 + g) + g - f g]$$



Motivation and issues with current PSA softwares

- Consider a « 2 out of 3 system » given by the Boolean equation:

$$T = (x_1 \wedge x_2) \vee (x_1 \wedge x_3) \vee (x_2 \wedge x_3) \quad p(x_1) = p(x_2) = p(x_3) := q$$



$$p(T) \neq p(x_1)p(x_2) + p(x_1)p(x_3) + p(x_2)p(x_3) = 3 \cdot q^2$$

$$p(T) = p(x_1)p(x_2) + p(x_1)\underbrace{(1 - p(x_2))}_{\text{Success}}p(x_3) + \underbrace{(1 - p(x_1))}_{\text{Success}}p(x_2)p(x_3) = 3q^2 - 2q^3$$

Motivation and issues with current PSA softwares

- **Other issues include:**
 - ⚙ Wrong treatment of negative logic (e.g. forbidden maintenance unavailabilities according to TechSpec)
 - ⚙ Quantification cutoff (typically 1E-12 to 1E-14)
 - ⚙ Wrong interpretation of risk importance measures of components, systems and safety divisions (RIF, FV, etc.)
 - ⚙ Treatment of exchange events
- **Advanced PSA models include HRA, CCF, seismic and phenomenological events, where failure probabilities approach 1**
- **It is accepted that current quantification tools have reached their own limits [Rauzy, 2001]**

Motivation and issues with current PSA softwares

- **Develop a new PSA quantification methodology that:**
 - ⚙ Overcomes the deficiencies of the rare approximation, i.e. credit success branches, calculate the rare event up to **infinite order**
 - ⚙ Yields a correct evaluation of Risk Importance Factors (RIFs)
 - ⚙ Support the treatment of negative logic
 - ⚙ Do not apply cutoff when generating the sequences
 - ⚙ Improve calculation speed and result consistency

Binary Decision Diagrams (BDD) as an alternative

- Shannon expansion

$$x \rightarrow y_0, y_1 := (x \wedge y_0) \vee (\bar{x} \wedge y_1) := \text{ite}(x, y_0, y_1)$$

- Shannon expansion of t with respect to x

$$t = x \rightarrow t[1/x], t[0/x] \Rightarrow t = (x \wedge t[1/x]) \vee (\bar{x} \wedge t[0/x])$$

- ⊛ $t[0/x]$ and $t[1/x]$ both contain one less variable than the expression t
- ⊛ We can recursively find ITEs up to the basic elements 0 (false) and 1 (true)

Binary Decision Diagrams (BDD) as an alternative

- Example for the „2 out of 3“ system $t = AB + BC + AC$

$$t = A \rightarrow (t_0, t_1)$$

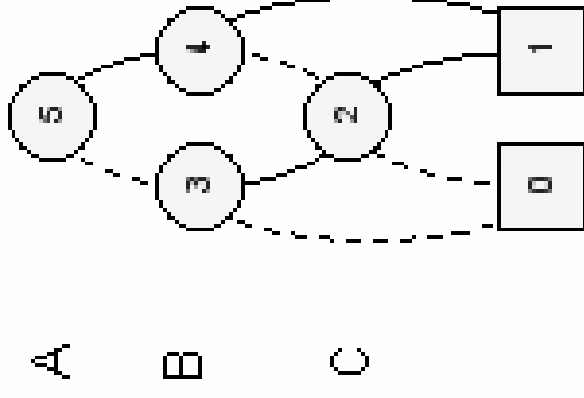
$$\odot t_0 = B \rightarrow (0, t_{01})$$

$$\odot t_1 = B \rightarrow (1, t_{10})$$

$$\odot t_{01} = C \rightarrow (1, 0)$$

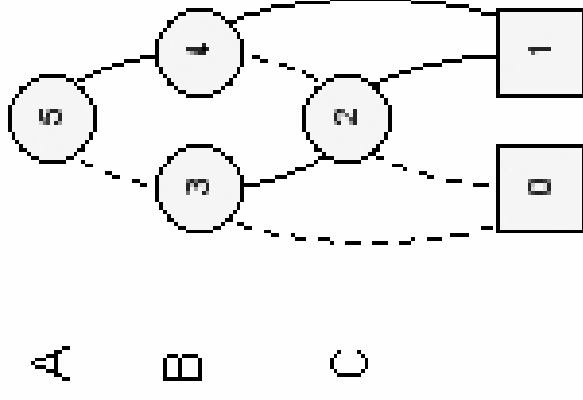
$$\odot t_{10} = C \rightarrow (1, 0)$$

$$\Rightarrow t = A \rightarrow (B \rightarrow (0, C \rightarrow (1, 0)), B \rightarrow (1, C \rightarrow (1, 0)))$$



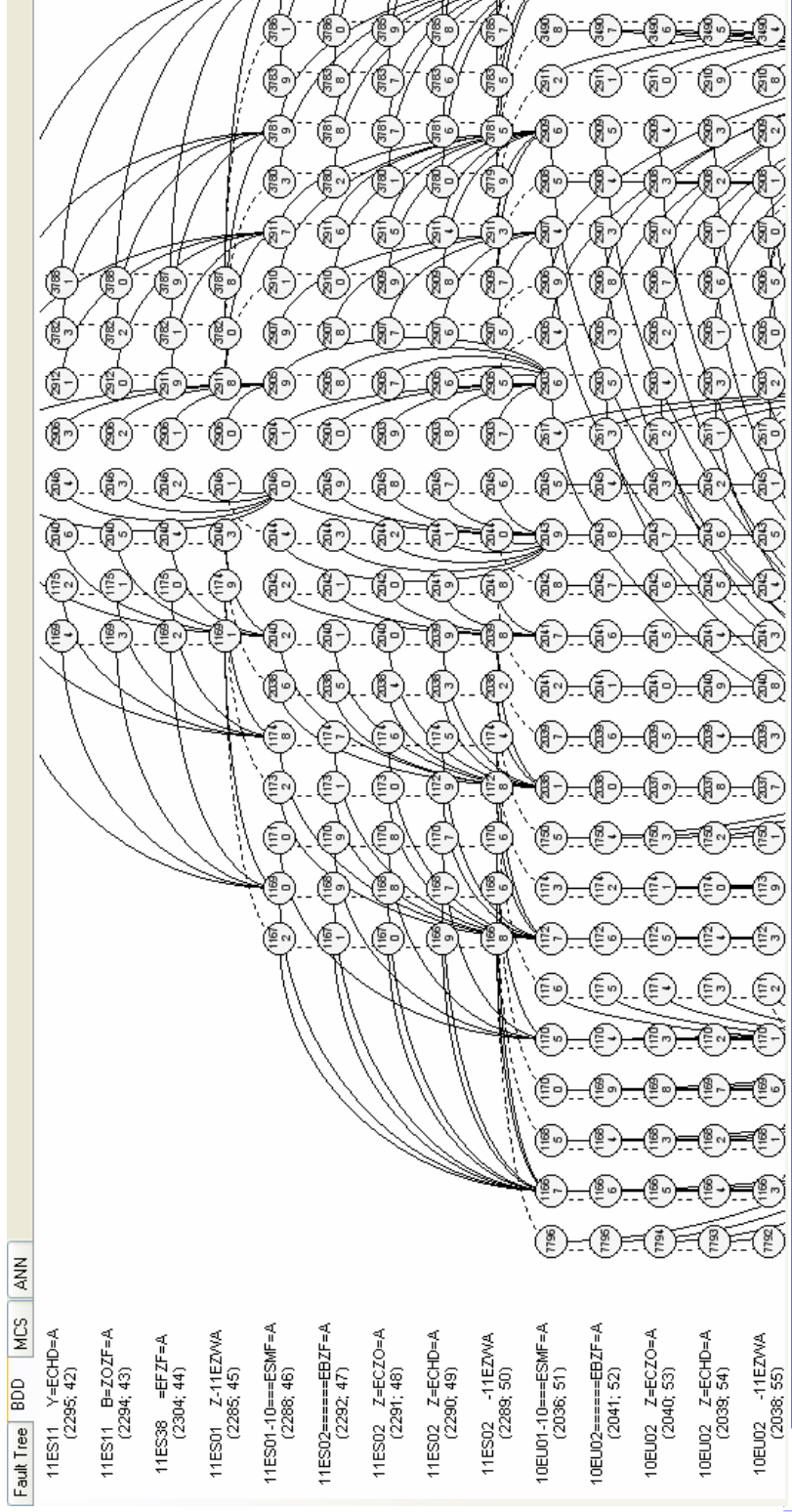
Binary Decision Diagrams (BDD) as an alternative

- Canonical formulation of Boolean equations !
- Example: $P_{t=true} = AB + A(1-B)C + (1-A)BC$



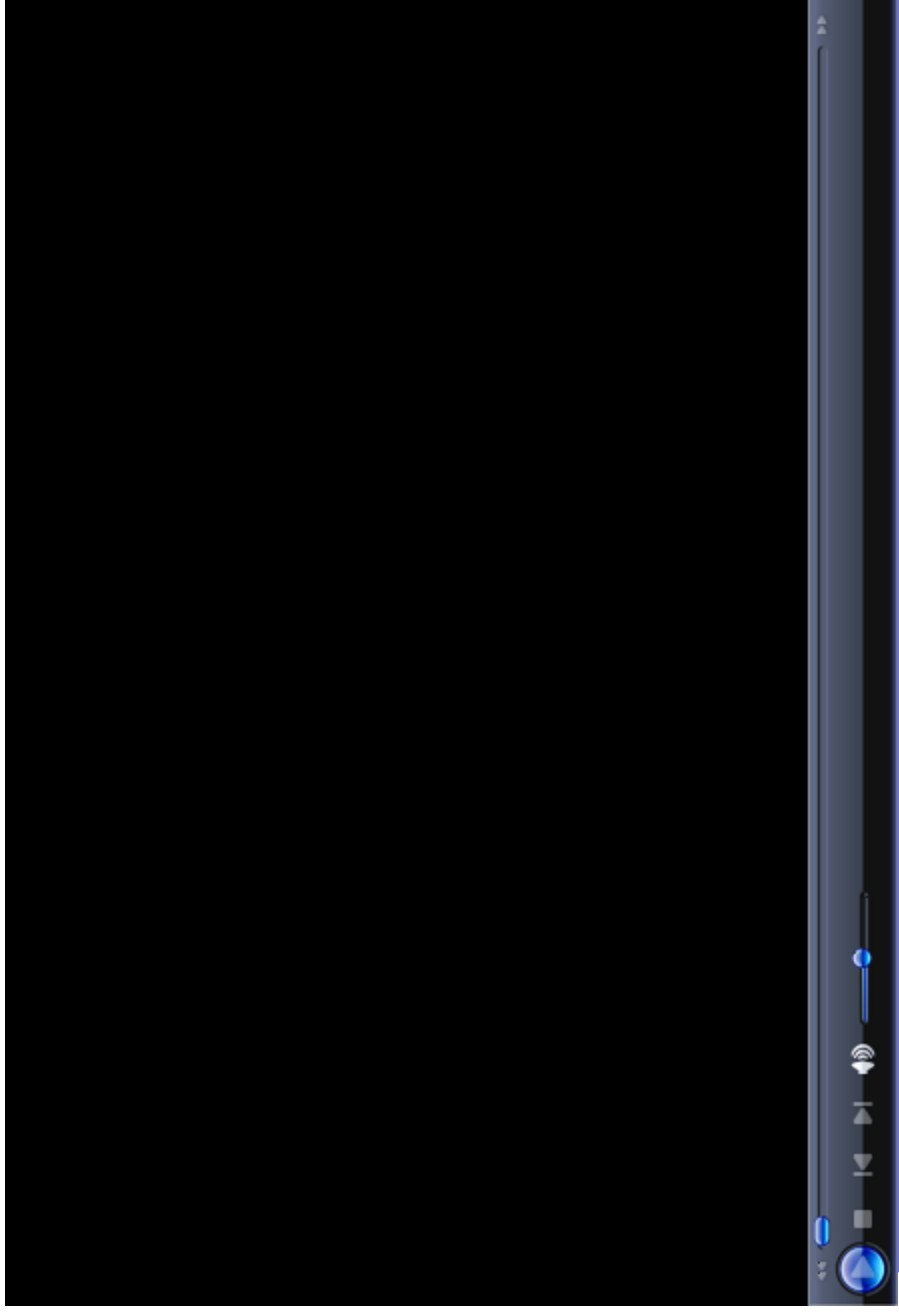
Binary Decision Diagrams (BDD) as an alternative

- BDD complexity is not related to the number of prime implicants of the encoded formula
- This small BDD (37620 nodes) encodes a total of 10^9 cutsets



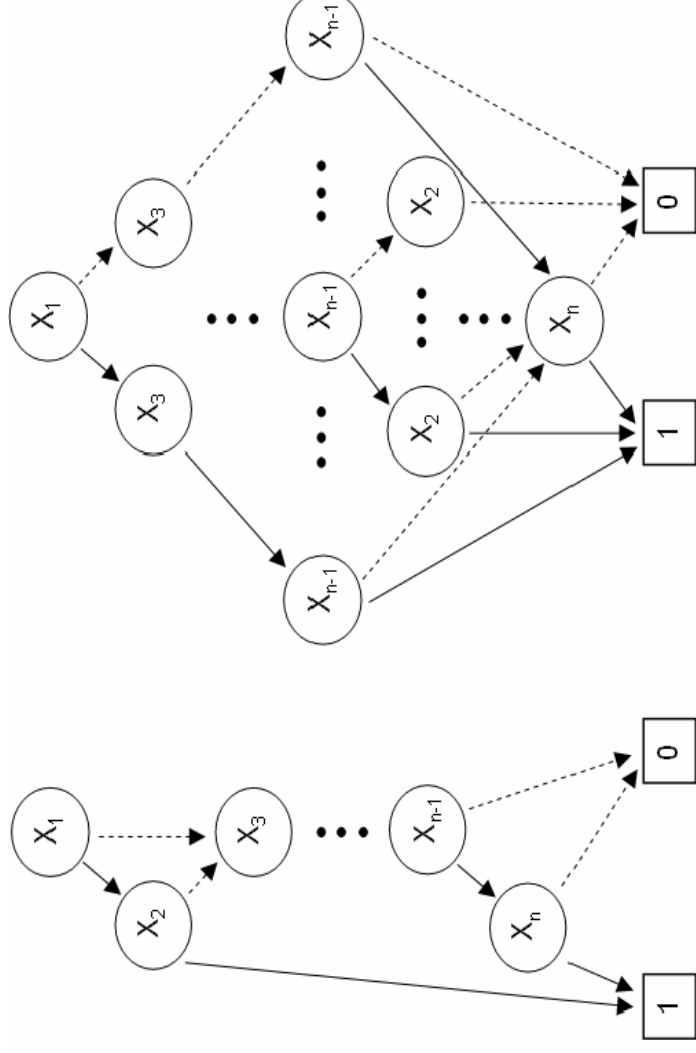
Binary Decision Diagrams (BDD) as an alternative

- **Let's have a closer look at it !**
- ⚙ HPCS System of the Leibstadt NPP



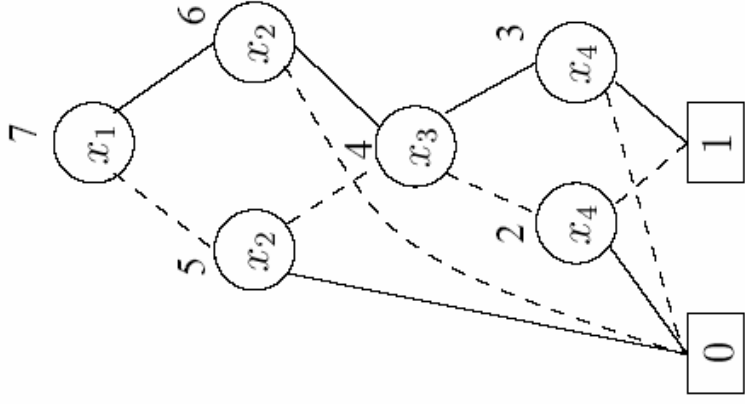
Binary Decision Diagrams (BDD) as an alternative

- **Impact of variable order on BDD size**
 - ⚙ From linear ☺ to exponential ☹
 - ⚙ Finding the best order is of **NP-Complete complexity** (Bollig / Wegener, 1996)



Binary Decision Diagrams (BDD) as an alternative

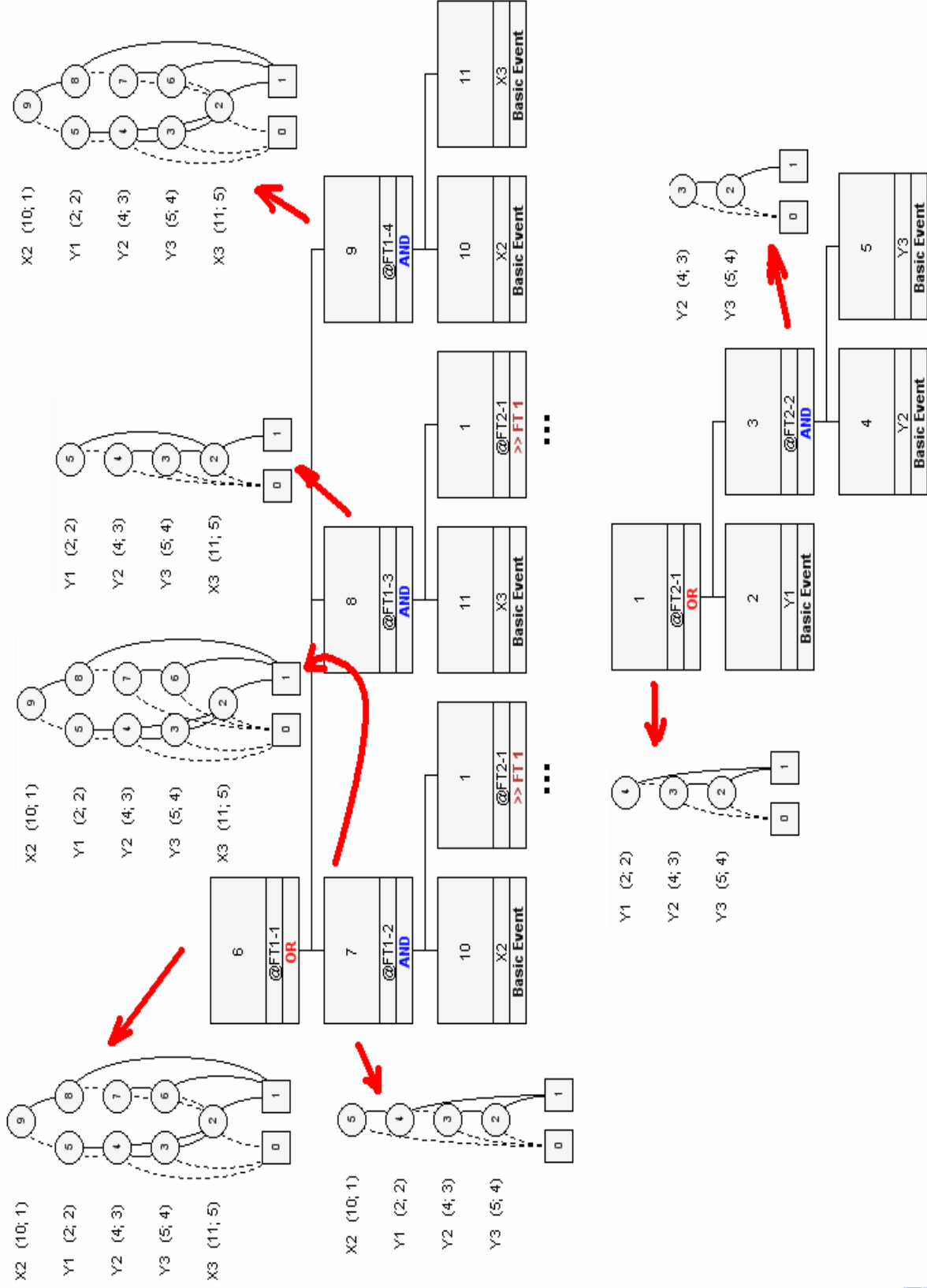
- Computer encoding of BDDs
 - ⚙ Open-ended node table (dynamic)
 - ⚙ Open-ended Hashtable (e.g. unique table)



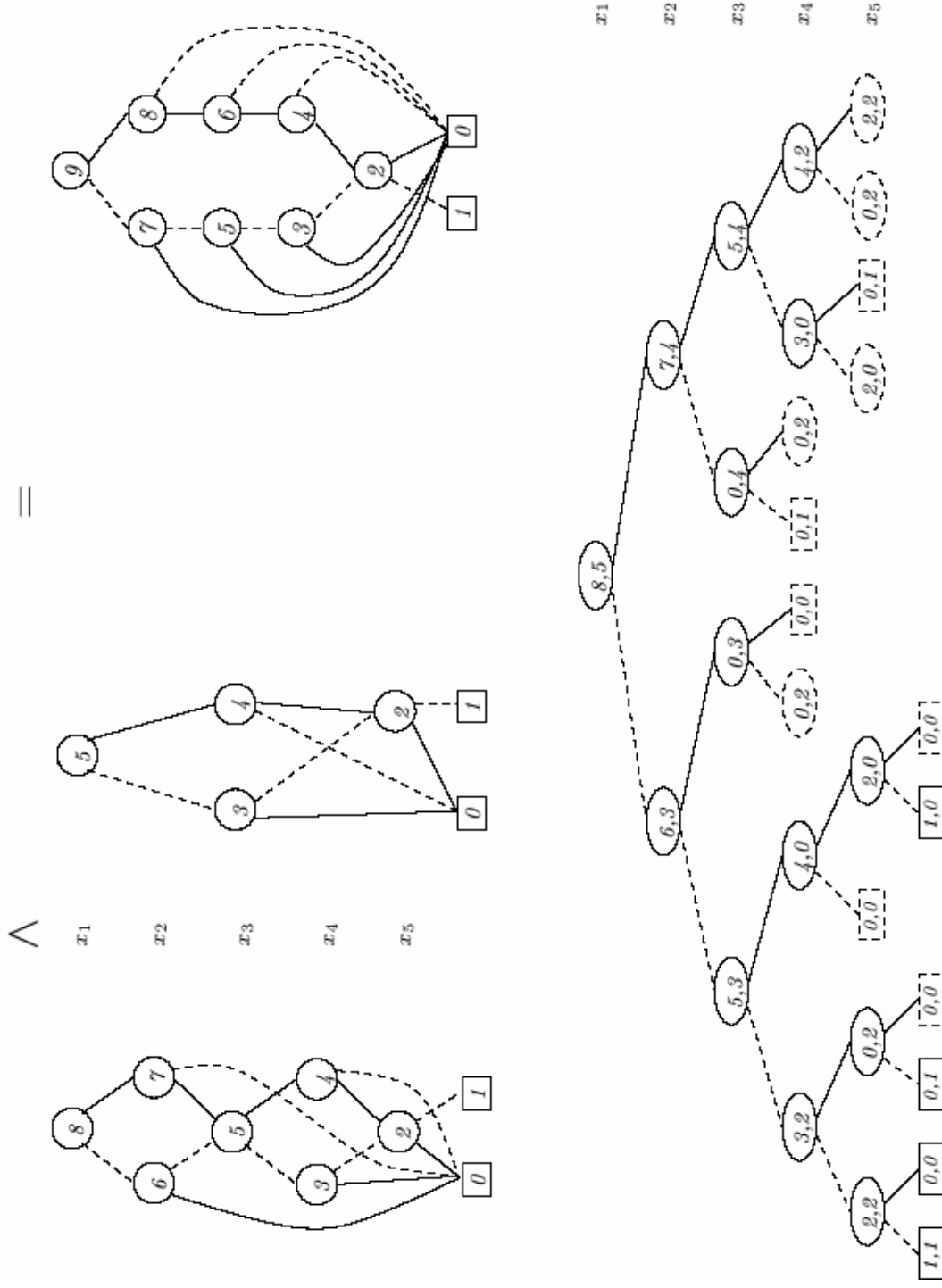
$$T : u \mapsto (i, l, h)$$

u	var	low	$high$
0	5		
1	5		
2	4	1	0
3	4	0	1
4	3	2	3
5	2	4	0
6	2	0	4
7	1	5	6

Binary Decision Diagrams (BDD) as an alternative



Binary Decision Diagrams (BDD) as an alternative



Binary Decision Diagrams (BDD) as an alternative

Algorithm *Apply*[*T*, *H*](*op*, *u*₁, *u*₂)

Require: *u*₁ and *u*₂ the top nodes of the BDD to assemble.

Ensure: The resulting BDD.

- 1: **if** $G(u_1, u_2) \neq \emptyset$ **then**
 - 2: **return** $G(u_1, u_2)$
 - 3: **else if** $u_1 \in \{0, 1\}$ and $u_2 \in \{0, 1\}$ **then**
 - 4: $u = op(u_1, u_2)$
 - 5: **else if** $var(u_1) = var(u_2)$ **then**
 - 6: $u = newnode(var(u_1), apply(low(u_1), low(u_2)), apply(high(u_1), high(u_2)))$
 - 7: **else if** $var(u_1) < var(u_2)$ **then**
 - 8: $u = newnode(var(u_1), apply(low(u_1), u_2), apply(high(u_1), u_2))$
 - 9: **else**
 - 10: $u = newnode(var(u_2), apply(u_1, low(u_2)), apply(u_1, high(u_2)))$
 - 11: **end if**
 - 12: $G(u_1, u_2) \leftarrow u$ {Add to computation table}
 - 13: **return** u {Returns node index}
-

Binary Decision Diagrams (BDD) as an alternative

- **Previous Work**
 - ⚙ BDD have been implemented in the early 90's for Integrated Circuits (IC) checking and IC optimization (16 and 32 bits)
 - ⚙ Some attempts to convert small to medium size models (typically with a few hundreds Basic Events) have succeeded
 - ⚙ All attempts with more Basic Events (>1000) have failed due to the exponential growth in complexity (BDD blow up)

Research and development at KKL and ETH Zurich

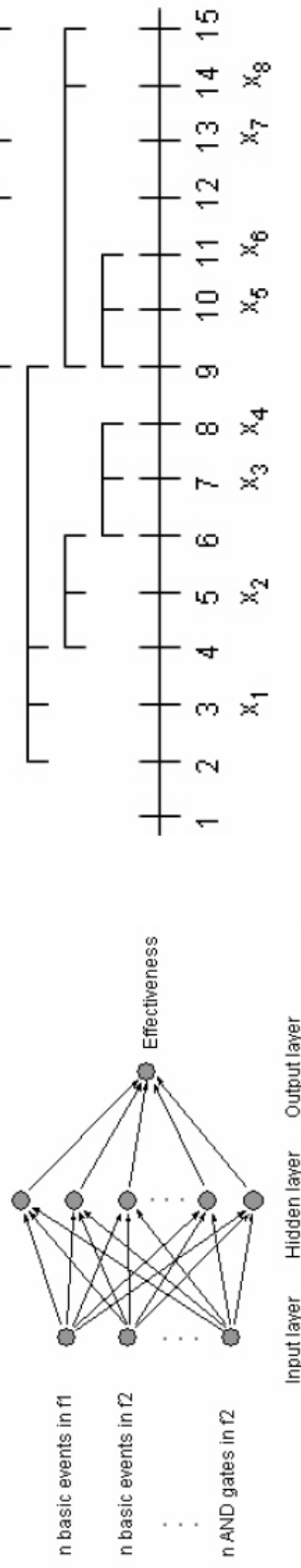
- Development of Fault Tree to BDD conversion engine
- Development of statistical measures
- Analysis of Fault Tree model pre-processing (rewriting) techniques

⚙ Basic Event occurrence based ordering

⚙ Weights fan-out pre-processing $\rightarrow W(v) = \begin{cases} 1 & \text{for basic events} \\ \sum_i W(v_i) & \text{for gates} \end{cases}$

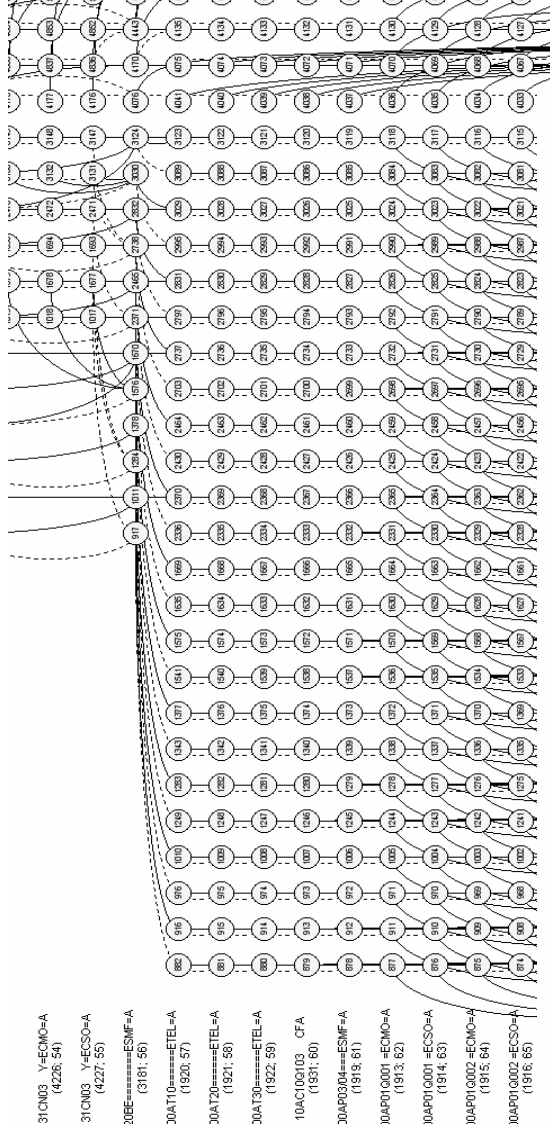
⚙ Hypergraph optimization techniques

⚙ Artificial Neural Network



Research and development at KKL and ETH Zurich

- Development of dynamic optimization techniques (e.g. Sifting, p-cut variable arrangement)
- Development of Group-Sifting for FTA



	DFLM	Regular sifting	Group-sifting
HPCS	6545	3204	497
LPCS	206'503	40'656	7763
RHR/A	306'339	99'945	11'948

Research and development at KKL and ETH Zurich

- **Development and analysis of modularization techniques**

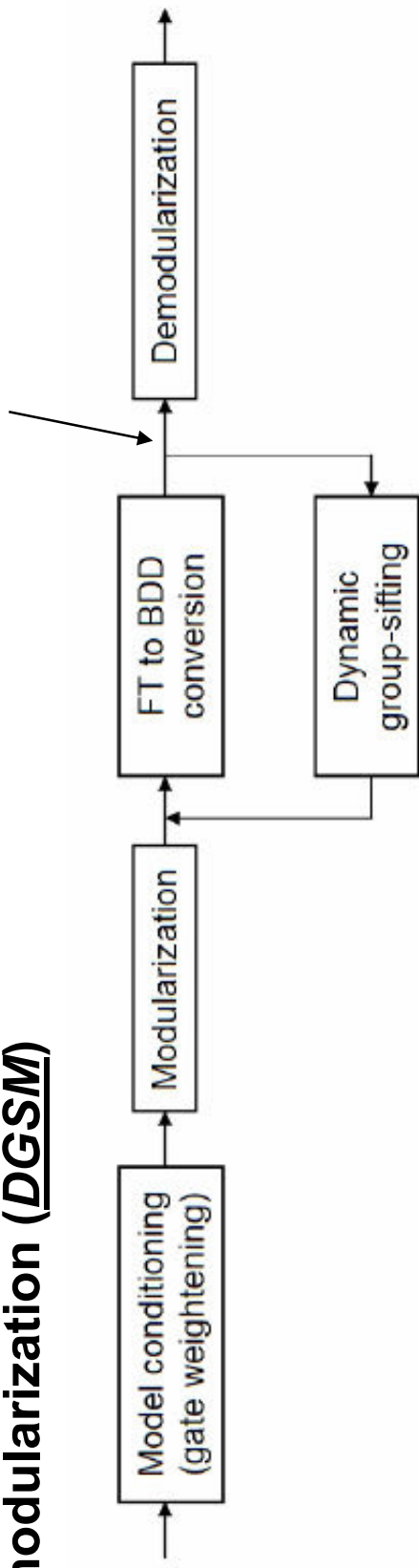
- ⚙ Occurrence vectors and detection criteria



Research and development at KKL and ETH Zurich

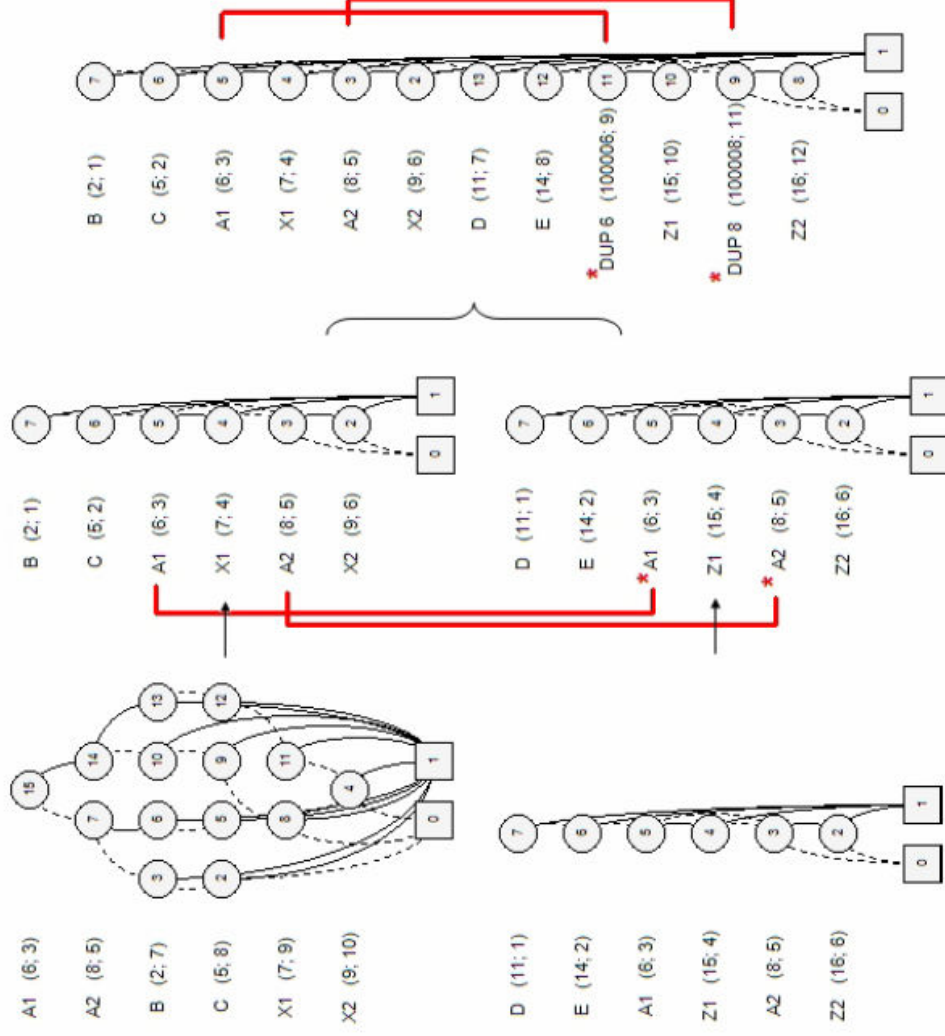
- **Dynamic Group-Sifting using modularization (DGSM)**

$$size_{BDD_R} > \left(\left(\frac{size_{BDD_R}}{N} \right)^\alpha + 1 \right) \times (size_{BDD_1} + size_{BDD_2})$$



Systems	DFLM	Dynamic group-sifting	Dynamic group-sifting using modularization
HPCS	6545	761	497
LPCS	206'503	3050	3053
RHR/A	306'339	3117	3117
RHR/B	impossible	52'447	21'177
ECCS	impossible	impossible	1'781'100 (CPU= 11 hours)

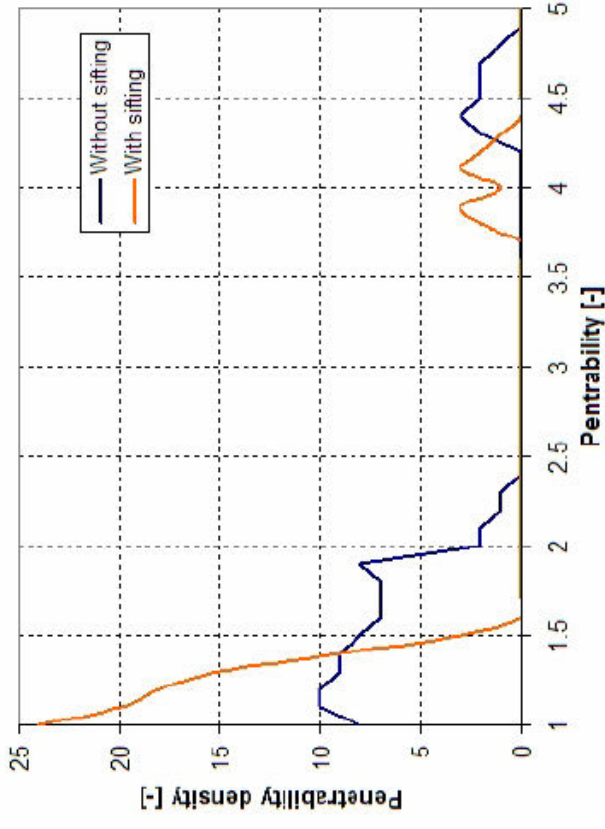
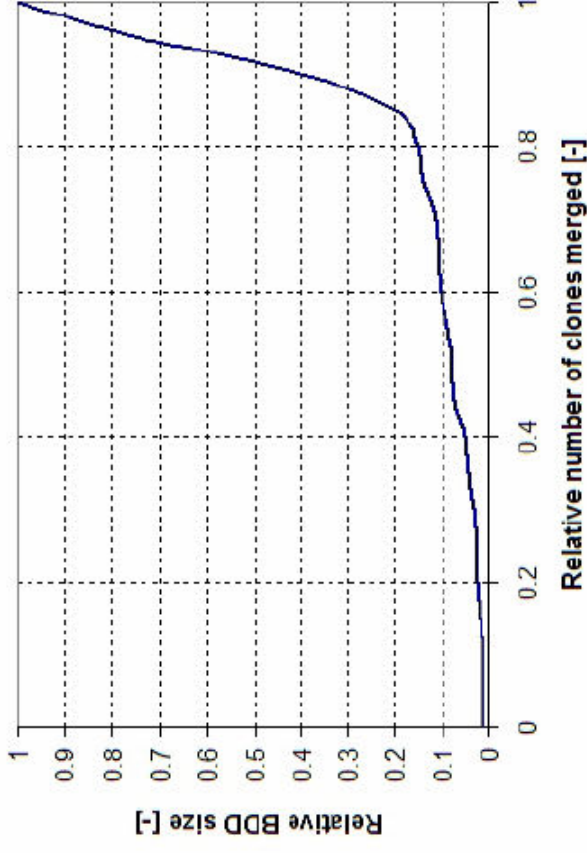
- Algorithm FUSION (BDD as objects)



Research and development at KKL and ETH Zurich

- **Performance of FUSION**
 - ⚙ About 90% of the variables can be merged without major impact on BDD size
 - ⚙ Penetrability p : effective identification of mathematical “hot spots”

$$p(x) = \frac{\text{Size}(\text{Opt}(BDD_{x \text{ merged}}))}{\text{Size}(BDD)}$$



Research and development at KKL and ETH Zurich

- **Improved Dynamic Group-Sifting Using Modularization (IDGSM)**
 - ⚙ Further limitation of global perturbation when optimizing locally
 - ⚙ Online identification and treatment of “hot spots” using penetrability spectrum
 - ⚙ Improvement in the Group-Sifting algorithm
 - ⚙ Use of genetic optimization algorithms
 - ⚙ Generation and treatment of “clones” (Algorithm FUSION)

Insights and outlook

- **Results and insights:**
 - ⚙ FTA quantification using BDD requires complex algorithms and programming techniques
 - ⚙ The combination of global, static, dynamic and BDD objects techniques proved to be effective when dealing with large models
 - ⚙ We succeeded in converting the Leibstadt PSA model to a BDD form of more than 1'500'000 with 30 clones, for a total of about 3500 basic events
 - ⚙ The BDD covers a complete event tree sequence that includes reactor shutdown and reactor cooling with the eight Emergency Core Cooling Systems of the Leibstadt Nuclear Power Plant (including support systems)

Insights and outlook

- **The nuclear industry is facing a major issue (and is not yet fully aware of it):**
 - ⚙ Worldwide probabilistic analyses are performed with codes that produce wrong results (conservative or optimistic, no one can know)
 - ⚙ New IAEA requirements are impossible to address with existing FTA quantifiers (e.g. seismic assessments)
 - ⚙ Utilities and authorities *still* trust the results of existing FTA quantifiers, applying approximations where they should not be applied (→ blind-trust in numbers)
 - ⚙ The techniques developed in this study raises the BDD approach to a mature technology for PSA model solving