

Enhanced Bayesian Network for Reliability Assessment: Application to Salt Domes as Disposal Sites for Radioactive Waste Problem

Andrea Perin

Institute for Risk and Reliability
Leibniz Universität Hannover

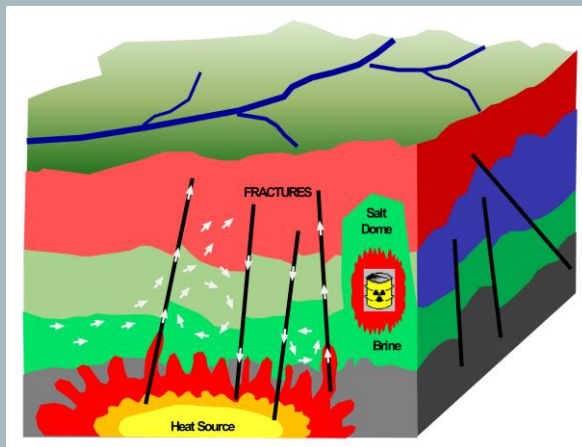
Problem Introduction

Risk Assessment of salt domes as long-term radioactive waste disposal

Project

Project goal is to develop a platform for the probabilistic assessment of unintentional leakage of radioactive materials associated with deep repositories:

- Enhanced Bayesian Network (EBN) for risk assessment
- Thermal-Hydraulic-Components (THC) model for density-driven flow



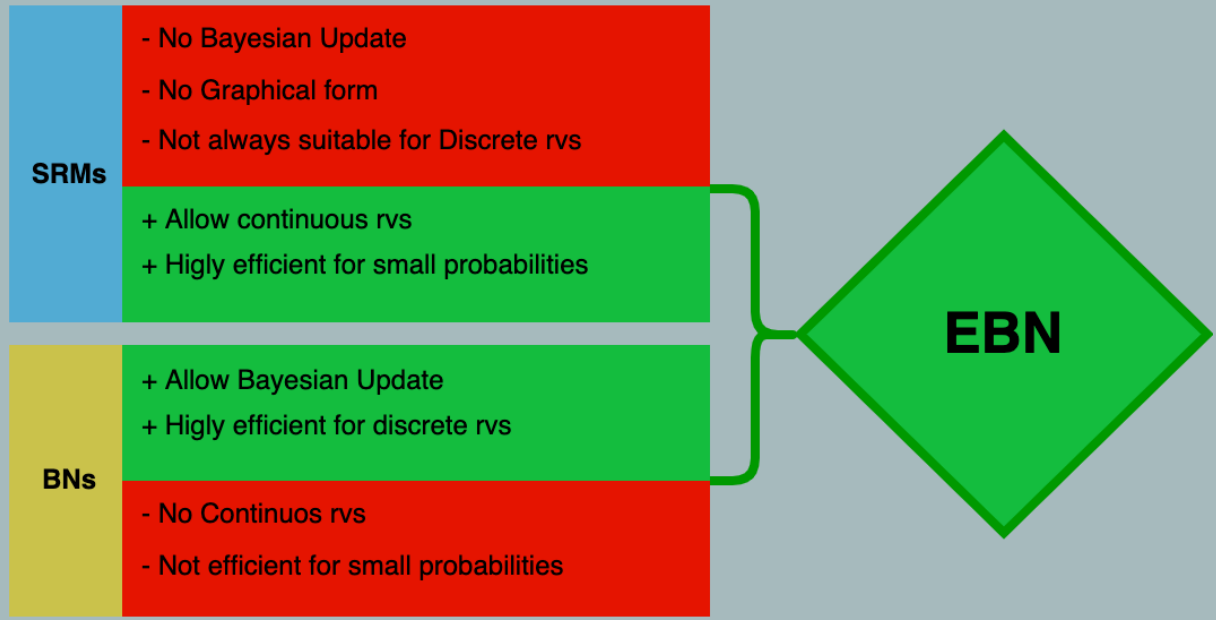
Salt Dome Problem

- Radionuclide transport inside a salt dome is **Thermal-Hydraulic-Components (THC)** transport phenomenon:
 - **Thermal:**
Geothermal gradient + Heat generated by waste decay
 - **Hydraulic:**
Groundwater flow in fractured porous media
 - **Components:**
Salt and radionuclides are the 2 components of interest
 - | | | | |
|---|--------------|---|--|
| { | Salt | → | salt dissolution affects flow velocity and vice-versa (<i>Variable density and viscosity flow</i>) |
| | Radionuclide | → | Quantity of interest for safe/fail state definition |
- Required safety constrain: uncontaminate biosphere over a time span of 1M years

Risk Assessment through Enhanced Bayesian Network (EBN)

Risk Assessment (EBN) – General Concepts

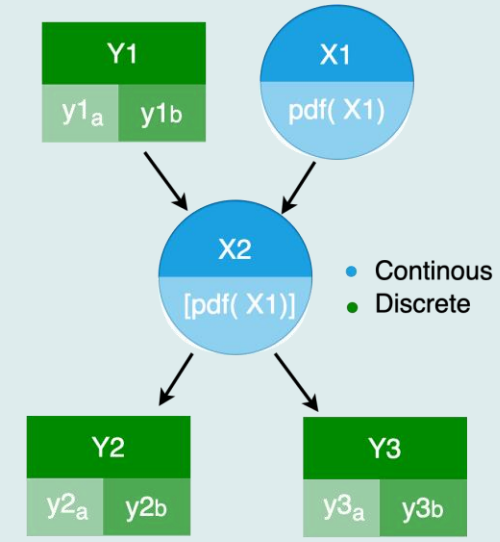
IDEA



eBNs (BNs Enhanced with SRM) are a tool able to:

- Implement Discrete and Continuous rvs
- With arbitrary distributions
- And any dependency

HOW



Formally

- Discrete nodes have a finite sample space
- Continuous nodes are vectors of continuous rvs
- System pdf is expressed by the combined effect of continuous and discrete rvs

System pdf:
$$f(\mathbf{Z}) = \prod_{Y_i \in \mathbf{Y}} f(y_i | pa[Y_i]) + \prod_{X_i \in \mathbf{X}} f(x_i | pa[X_i])$$

The problem of the evaluation of discrete probabilities (or pdf) of each node with at least one continuous parent has the same mathematical form of a Structural Reliability Problem!

Risk Assessment (EBN) – Main Features

Main Structures

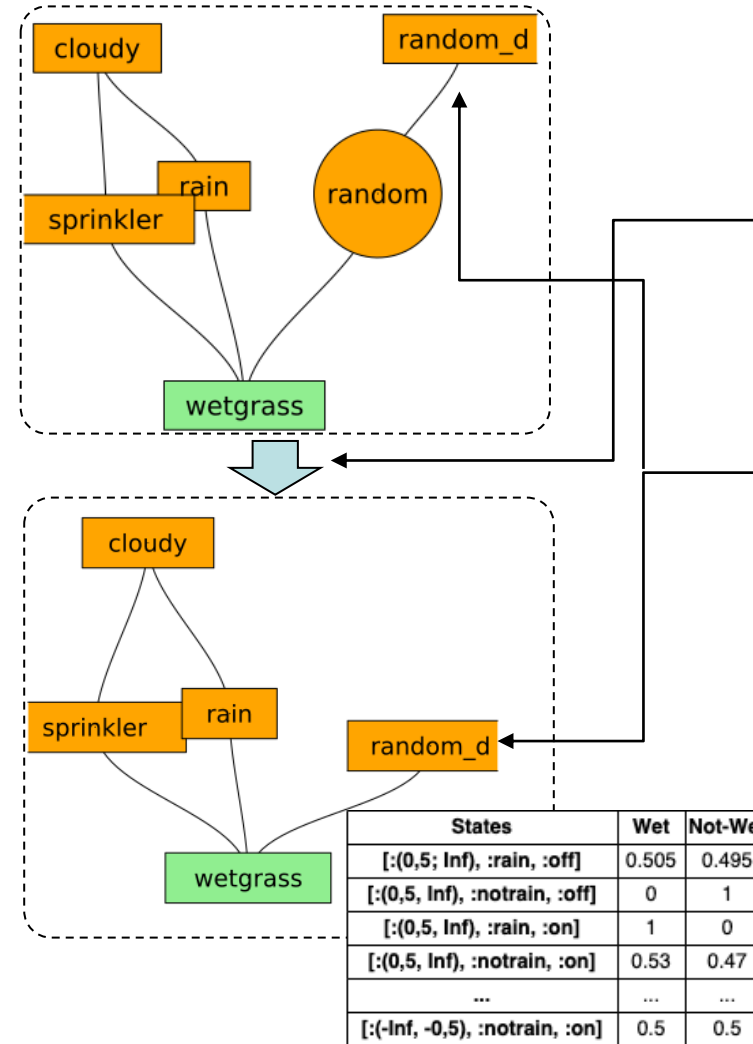
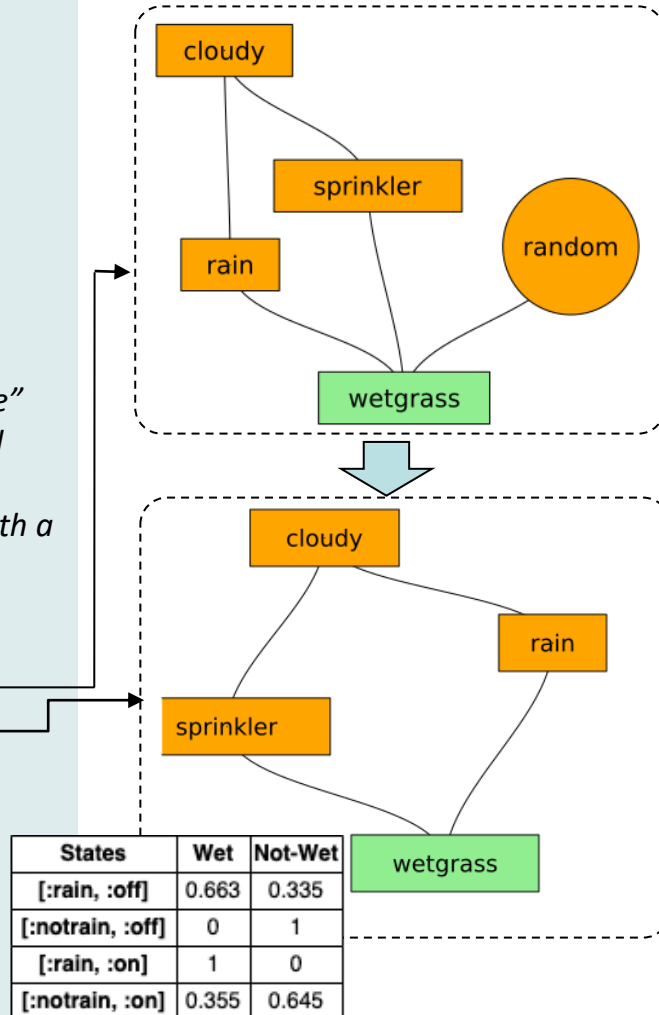
Nodes

- RootNode
- StandardNode
- FunctionalNode

Each node can be "Discrete" with a discrete Conditional Probability Distribution (dCPD) or "Continuous" with a continuous CPD (cCPD)

Networks

- EnhancedBN
- ReducedBN (rBN)
- BN



Main Algorithms

Nodes Elimination Algorithm

- Allow the evaluation of the associated rBN
- Remove continuous nodes
- Transfer info to the rBN's SRPs

Evidence Algorithm

- Discretize continuous node for which future evidence can be available
- Allow Inference over discretized continuous nodes

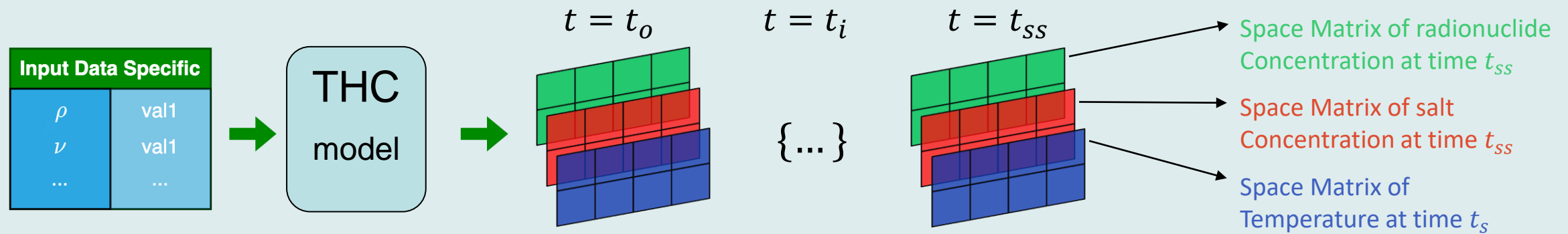
Inference Algorithm

- Allow inference over each node of the resulting BN
- Inference is performed through "variable elimination"
- Computational time is optimized through "minimal increase in complexity algorithm"

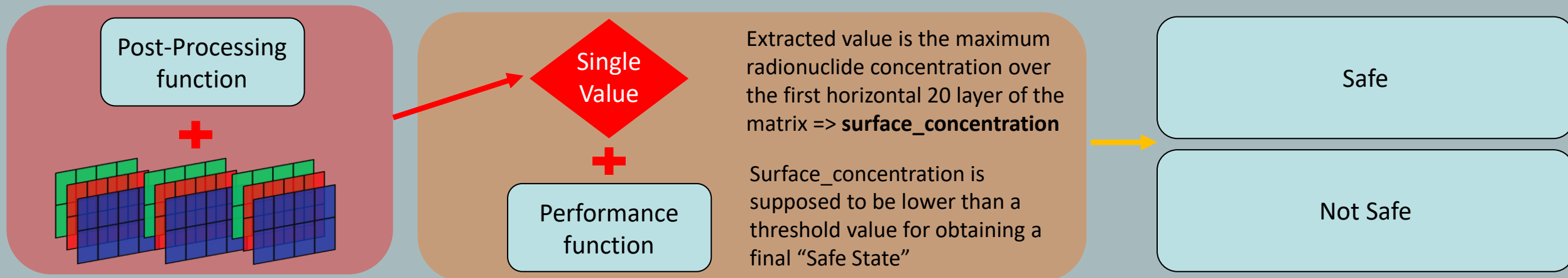
Implementation

Implementation – THC solver

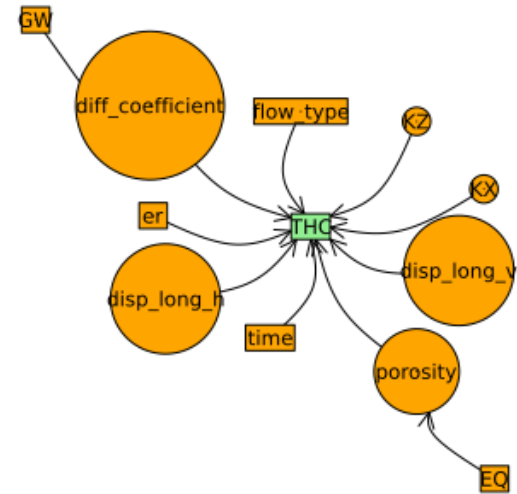
Model's output



Post-Processing + Performance Function



Implementation – EBN



Earthquake (EQ):

- DiscreteRootNode
 - :happen = $10e^{-5}$
 - :not_happen = $1 - 10e^{-5}$

Porosity (porosity)

- Child of Earthquake
- ContinuousStandardNode
 - :happen => $\text{trunc } N(3; 0.5)$
 - :not_happen => $\text{trunc } N(1; 0.5)$

Global-warming (GW)

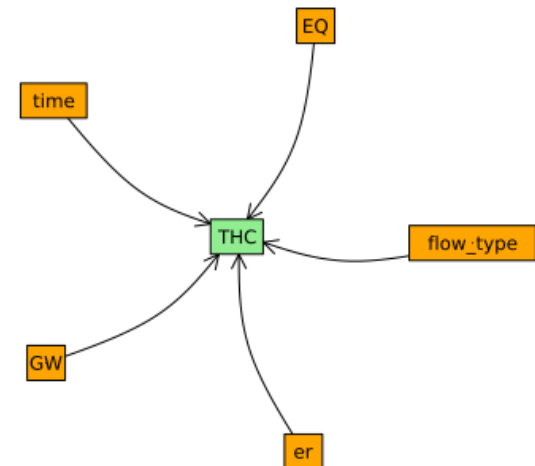
- DiscreteRootNode
 - :warming => 0.7
 - :astoday => 0.2
 - :cooling => 0.1

Diffusion (diff_coefficient)

- Child of Global-Warming
- ContinuousStandardNode
 - :warming => $\text{trunc } N(2e^{-6}; e^{-6})$
 - :astoday => $\text{trunc } N(2e^{-8}; e^{-7})$
 - :cooling => $\text{trunc } N(2e^{-9}; e^{-6})$

Extreme Rain (er)

- DiscreteRootNode
 - :extremerain => 0.4
 - :no_extremerain => 0.6
- Influenced parameters
 - :extremerain => :head = 1.2
 - :no_extremerain => :head = 0.8



Hydraulic Conductivity x-direction (KX)

- ContinuousRootNode($\text{truncated } N(9.81e^{-6}; e^{-4})$)

Hydraulic Conductivity z-direction (KZ)

- ContinuousRootNode($\text{truncated } N(9.81e^{-6}; e^{-4})$)

Time Scenario

- Short (10^5 days) or Long (10^7 days)

Longitudinal Dispersivity vertical (disp_long_v)

- ContinuousRootNode($\text{Uniform}(10; 60)$)

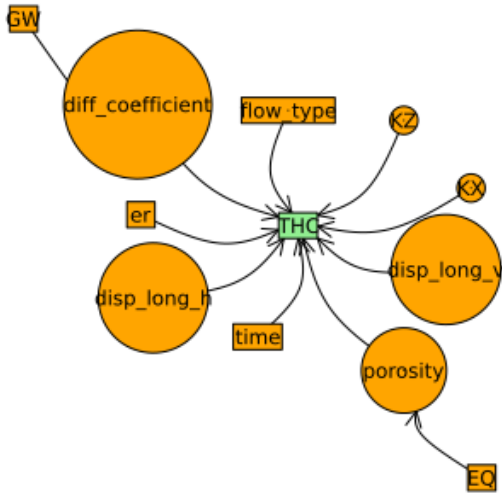
Longitudinal Dispersivity horizontal (disp_long_h)

- ContinuousRootNode($\text{Uniform}(1; 6)$)

Flow Type

- Steady-State or Transient Solution

Implementation – EBN



THC Model's Node

Model:

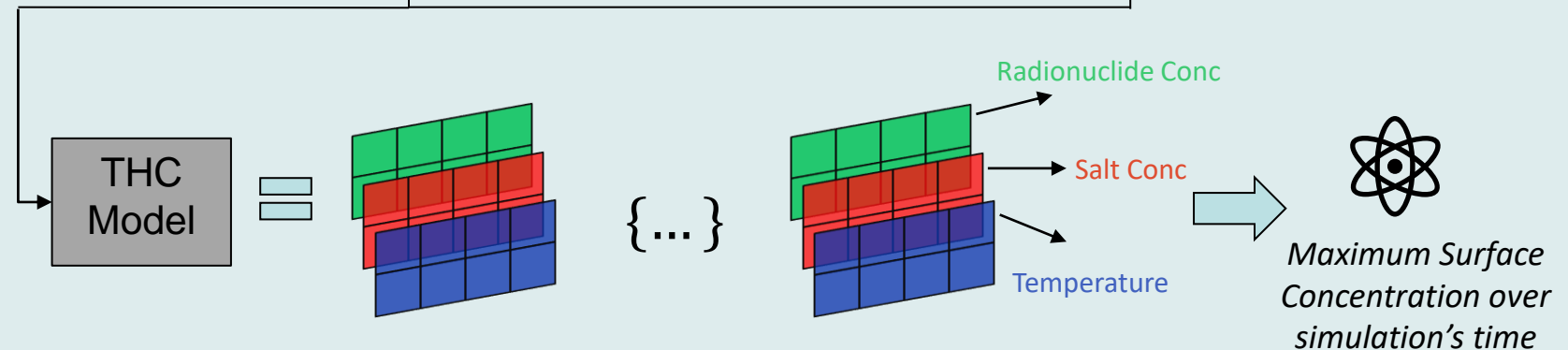
- FE model (60x30)
- No heat source
- Salt and Radionuclide Transport
- Density driven flow simulation

Continuous Parents

- Diffusion Coefficient
- Hydraulic Conductivity X
- Hydraulic Conductivity Z
- Dispersivity Longitudinal Vertical
- Dispersivity Longitudinal Horizontal
- Porosity

Discrete Parents

- Extreme Rain
- Time Scenario
- Flow type



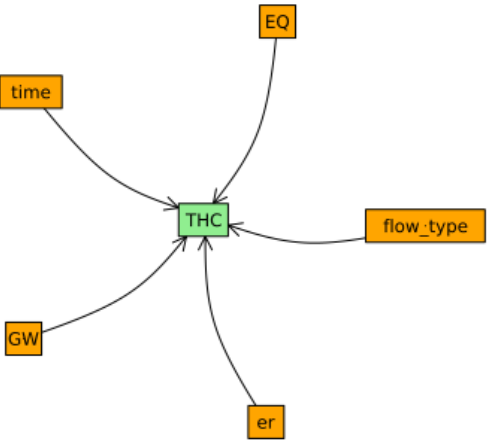
Implementation – Results

rBN – THC model's Node

- Extreme Rain => 2 states
- Time Scenario => 2 states
- Flow type => 2 states
- Earthquake => 2 states
- Global Warming => 3 states

$$2^4 * 3 = 48 \text{ SRPs (200 simulations each)}$$

Performance => Surface Concentration > 0



state	fail	safe
comb 1	0.995	0.005
comb 2	1	0
comb 3	0.87	0.13
comb 4	0.885	0.114
comb 5	0.92	0.079
comb 6	0.875	0.125
comb 7	1	0
comb 8	0.83	0.17
comb 9	0.985	0.015
comb 10	0.99	0.01
comb 11	0.885	0.114
comb 12	0.92	0.079

{ ... }

state	fail	safe
comb 22	0.985	0.015
comb 23	0.94	0.06
comb 24	0.98	0.02
comb 25	1	0
comb 26	1	0
comb 27	0.985	0.015
comb 28	0.82	0.18
comb 29	0.93	0.069
comb 30	0.835	0.165
comb 31	0.905	0.094
comb 32	0.895	0.104
comb 33	0.905	0.094

Implementation – Results

rBN – Inference

Time Scenario $\left\{ \begin{array}{l} p(\text{time_long}|\text{THC_fail}) = 0.57 \\ p(\text{time_short}|\text{THC_fail}) = 0.43 \end{array} \right.$

Earthquake $\left\{ \begin{array}{l} p(\text{earthquake}|\text{THC_fail}) = 10^{-5} \\ p(\text{no_earthquake}|\text{THC_fail}) = 0.99999 \end{array} \right.$

Global Warming $\left\{ \begin{array}{l} p(\text{cooling}|\text{THC_fail}) = 0.11 \\ p(\text{astoday}|\text{THC_fail}) = 0.19 \\ p(\text{warming}|\text{THC_fail}) = 0.70 \end{array} \right.$

Extreme Rain $\left\{ \begin{array}{l} p(\text{er}|\text{THC_fail}) = 0.54 \\ p(\text{no_er}|\text{THC_fail}) = 0.46 \end{array} \right.$

- Deal with multidisciplinary and continuous node
- Allow "What-if" analysis through Inference algorithm

- Allow "model Update" through Bayesian update.
- Computational cost depends on model only

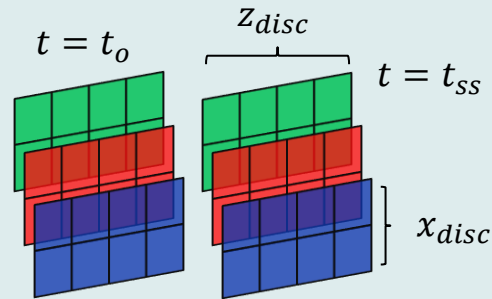
Analysis speed-up through Surrogate Model

FE Models are too computational expensive in a framework where are required to be run several times in different scenarios, especially when low probability of failure have to be established

TH Model
not simplified version

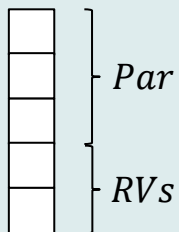


~ 24h ∇ simulation



$Dim = Par + RVs$

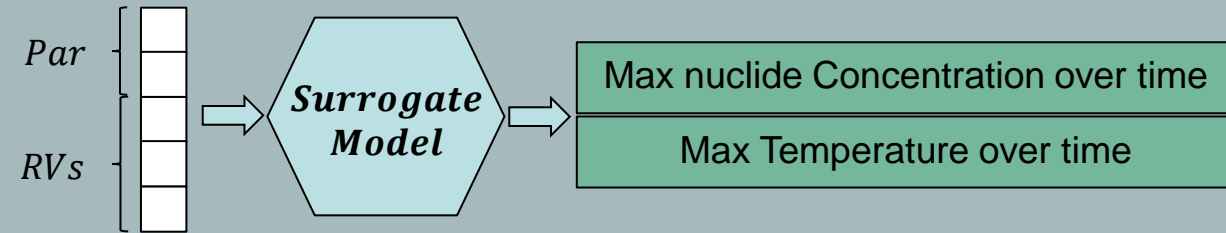
$Dim = t_{disc} * x_{disc} * z_{disc} * outputs$



Output	
t_{disc}	1 or 2
x_{disc}	10^2
z_{disc}	10^2
outputs	3 [c_s;T;c_r]

Surrogate Output

Reduced Output Dimensionality



Instead of predicting matrices, the Surrogate Model will predict a specific value for Radionuclide Concentration and a specific value for the Temperature

With a 24h simulation we obtain 1 output sample of 10^4 dimension!

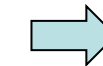
Analysis speed-up through Surrogate Model

Interval Predictor Model (IPM): No feasible solution for the optimization problem

Polynomial Chaos Expansion (PCE) with Iso-probabilistic transformation

Sensitivity Analysis

Variables	FirstOrder	FirstOrderStdError	TotalEffect	TotalEffectStdError
head_factor	0.0045	0.0504	0.0101	0.0653
KZ	0.1224	0.0532	0.548	0.0826
KX	0.2319	0.0539	0.6133	0.099
disp_long_h	0.1791	0.0503	0.279	0.0645
disp_long_v	0	0.0488	0	0.0636
porosity	0	0.0476	0	0.0665
diff_coefficient	-0.0016	0.0479	0.0008	0.0625



PCE Random Variables

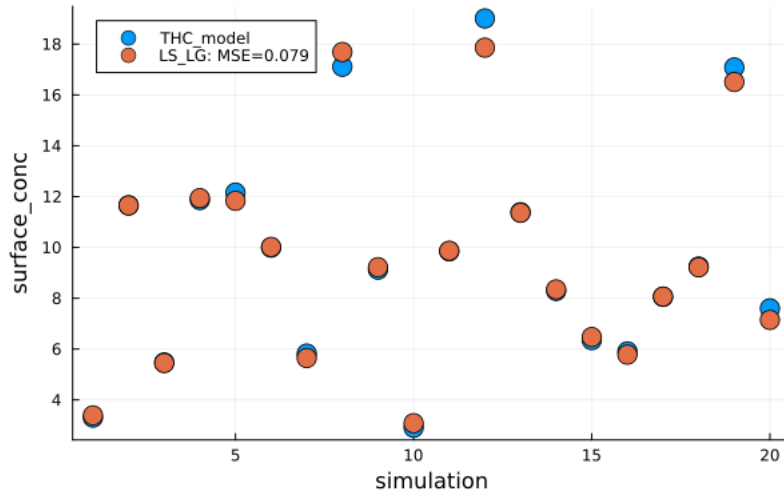
Variable	Distribution
KX	truncated Normal
KZ	truncated Normal
disp_long_v	Uniform

PCEs have been tested with 4 different combination of basis/point determination strategies:

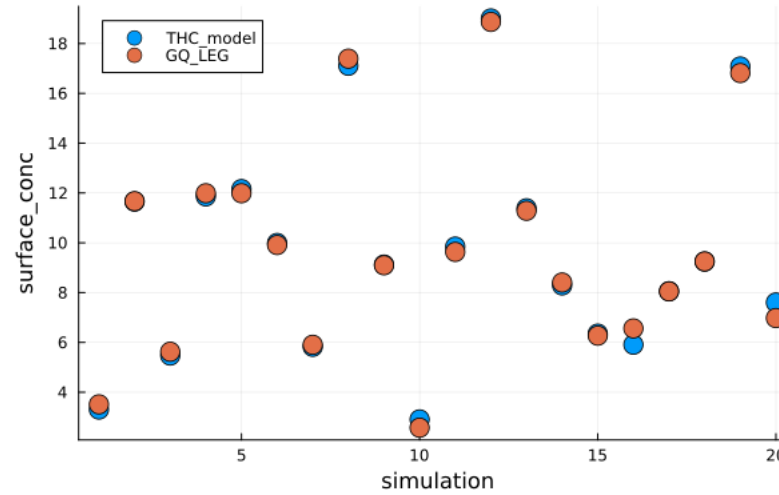
- Least Square – Legendre Basis Only
- Least Square – Mixed Legendre/Hermite Basis
- Gaussian Quadrature – Legendre Basis Only
- Gaussian Quadrature – Mixed Legendre/Hermite Basis

Analysis speed-up through Surrogate Model

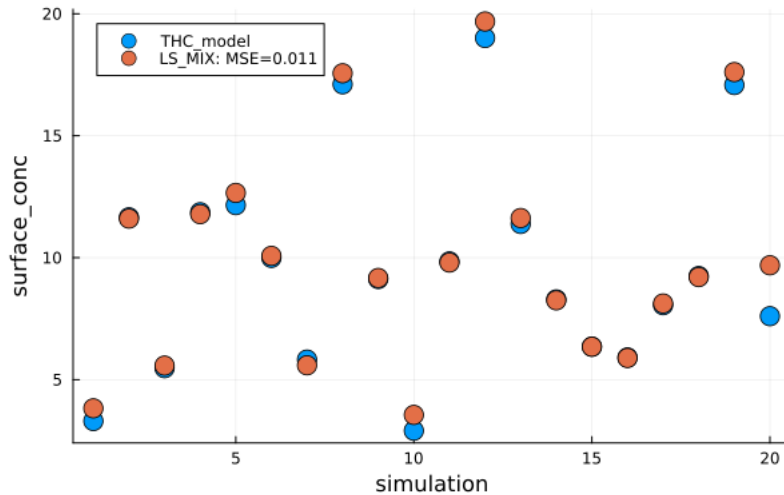
PCE Least Squared - Legendre Basis



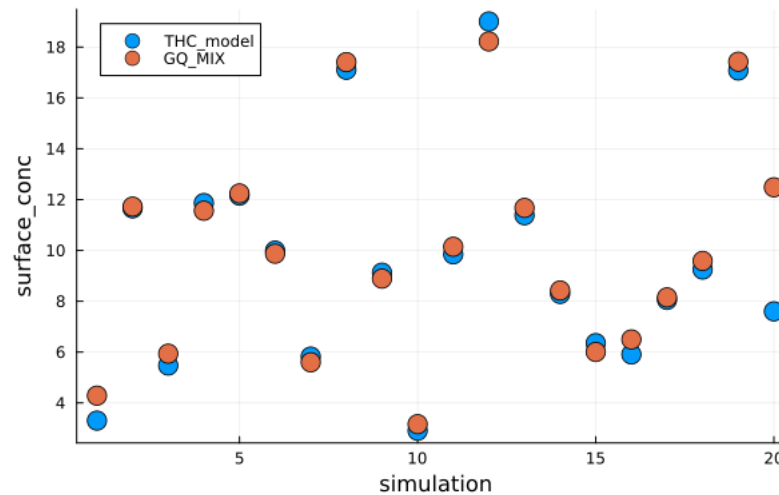
PCE Gaussian Quadrature - Legendre Basis



PCE Least Squared - Mix Basis



PCE Gaussian Quadrature - Mix Basis



Test MSE with 150 random samples

LS_leg	LS_mix	GQ_leg	GQ_mix
0.339	0.268	0.269	0.352

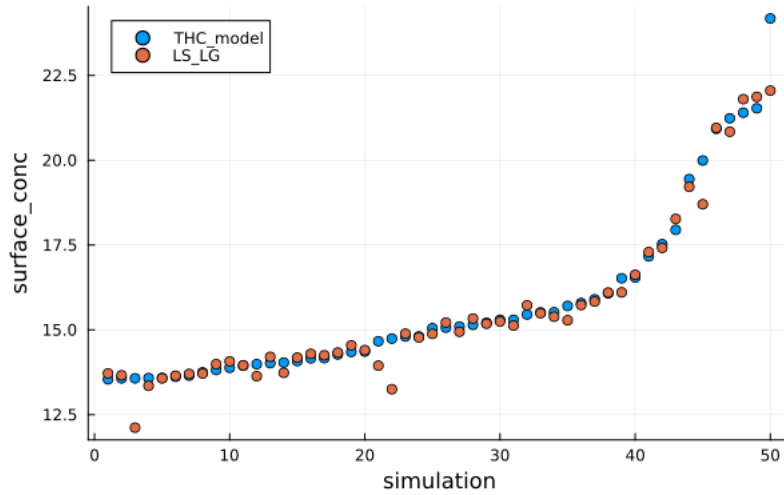
All the the combinations of basis and points determination algorithm are obtained with:

- degree 6
- 216 samples (equal to ones required by Gaussian Quadrature)

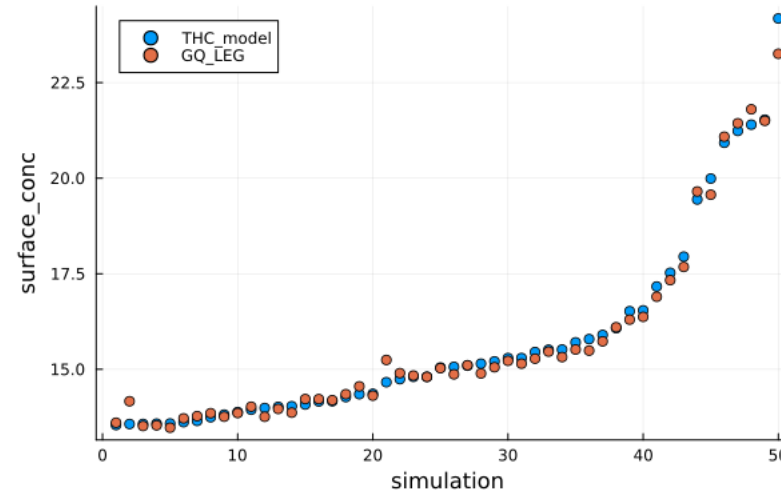
All the the combinations of basis and points determination algorithm seems to perform good with a slightly better MSE for Least Squared and Mix Basis (Legendre for Uniform distributions and Hermite for Gaussian distributions)

Analysis speed-up through Surrogate Model

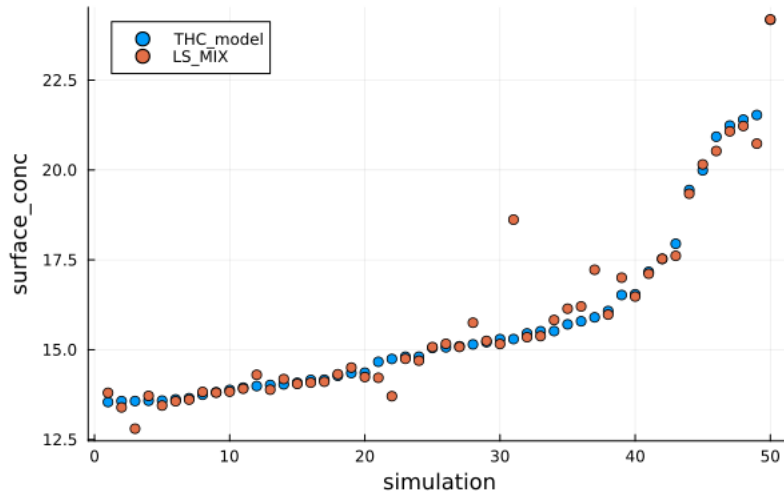
PCE Least Squared - Legendre Basis



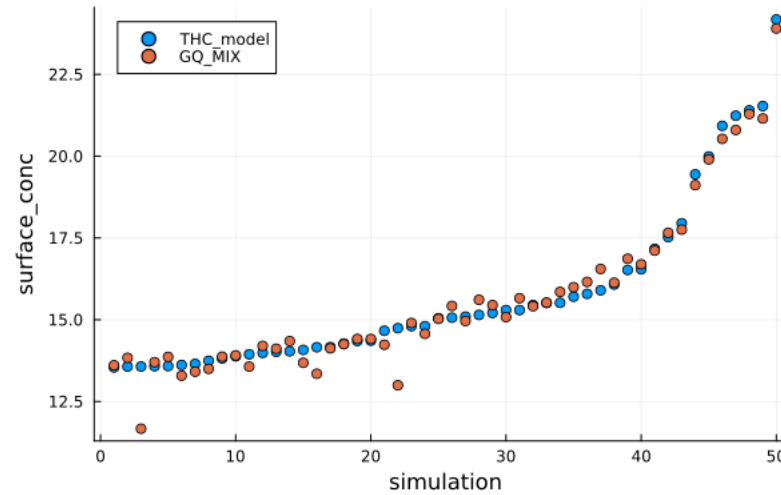
PCE Gaussian Quadrature - Legendre Basis



PCE Least Squared - Mix Basis



PCE Gaussian Quadrature - Mix Basis



Test MSE 100 samples in failure region

LS_leg	LS_mix	GQ_leg	GQ_mix
0.255	0.342	0.057	0.213

All the the combinations of basis and points determination algorithm are obtained with:

- degree 6
- 216 samples (equal to ones required by Gaussian Quadrature)

PCE with Gaussian Quadrature and degree 6 (216 samples) is the one that perform better in prediction of system failure state.

Conclusions & Outlooks

Conclusions & Outlooks

RADON Project

EBN

- In-depth analysis of the events (eBN nodes) and their influences on THC model's inputs:
 - *NEA report - Updating the NEA International FEP List An Integration Group for the Safety Case (IGSC) Technical Note*
 - *Projekt ANSICHT FEP Katalog für das Endlagerstandortmodell SÜD*

- *Interval Probability for Discrete Nodes
=> CredalNetwork*
- *P-Boxes for Continuous Nodes*
- *Gaussian Process to model the error of PCE
=> Interval model*

ACKNOWLEDGMENTS

Funded by *Bundesgesellschaft für Endlagerung (BGE)*
the federal company for radioactive waste disposal