DEVELOPMENT OF INTERACTIVE SOFTWARE FOR MODELING AND OPTIMIZING ENERGY PROCESSES USING MACHINE LEARNING AND AI

Pavel.Praks@vsb.cz

IT4Innovations National Supercomputing Center VSB – Technical University of Ostrava Ostrava-Poruba, Czech Republic www.it4i.cz



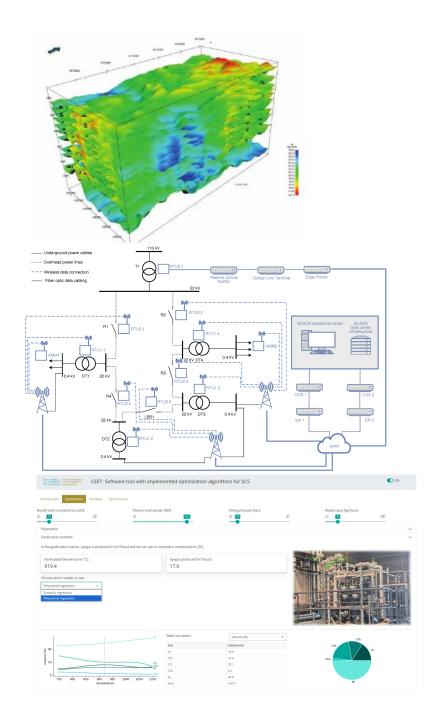


OUTLINE

- Development of interactive software for modeling and optimizing energy processes using machine learning (ML) and AI
- Use cases:
 - Carbon sequestration using OPM Flow (research cooperation IT4Innovations VSB TUO with SINTEF Digital, Oslo, Norway)
 - Interconnected infrastructures (electricity and communication networks – research cooperation with Brno University of Technology, Brno, CZ)
 - Shinyenet software for selected waste-to-energy processes at innovation polygon CEETe VSB TUO

Machine learning and sensitivity analysis on





SELECTED PAPERS FOR USE CASES

Carbon sequestration using OPM Flow

Praks, P., Rasmussen, A., Lye, K. O., Martinovič, J., Praksová, R., Watson, F., & Brkić, D. (2024). Sensitivity analysis of parameters for carbon sequestration: Symbolic regression models based on open porous media reservoir simulators predictions. In Heliyon (Vol. 10, Issue 22, p. e40044). Elsevier BV. https://doi.org/10.1016/j.heliyon.2024.e40044

Interconnected infrastructures (electricity and communication networks)

• Krpelik, D., Vrtal, M., Bris, R., Praks, P., Fujdiak, R., & Toman, P. (2026). Multi-objective optimization of smart grid operations via preventive maintenance scheduling using time-dependent unavailability. Reliability Engineering & System Safety, 265, 111567. https://doi.org/10.1016/j.ress.2025.111567

ZEUS project papers:

- Běloch M., Praks P.; Praksová R.; Fujdiak R.; Vrtal M.; Briš R.; Brkić D. (2025). Evaluating the
 unavailability of interconnected power and communication networks with open-source tools on a petascale
 cluster. Energy Exploration & Exploitation. 2025;0(0). https://doi.org/10.1177/01445987251377791
- Briš, R., Praks, P., Fujdiak, R., Vrtal, M., & Brkić, D. (2025). Maintenance optimization for unavailability enhancement of representative interconnected infrastructure based on minimum cost. Science Progress, 108(3) 1–23. https://doi.org/10.1177/00368504251366357

ML-ACCELERATED GLOBAL SENSITIVITY ANALYSIS

Sobol method:

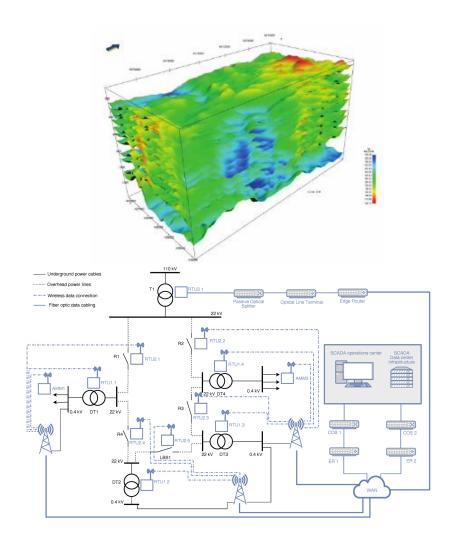
- Probabilistic, global, handles uncertainty
- Works with nonlinear responses & interactions
- Sensitivity indices indicate feature importance

Challenge:

Requires thousands of simulations → costly

Our aim:

- Use machine learning (ML) to predict sensitivity indices
- Reduce simulation cost while preserving accuracy



ACCELERATING SENSITIVITY ANALYSIS OF OPM FLOW WITH ML

- OPM Flow: Open-source reservoir simulator for carbon sequestration
 - Industry-standard formats (ECLIPSE)
 - Easy integration with pre/post-processing tools
- Approach: Model uncertain input parameters to quantify impact on outcomes
- Goal: Speed up global sensitivity analysis (SA) of OPM Flow simulations using machine learning (ML)
- Collaboration: SINTEF Digital (Norway) & IT4Innovations (Czechia) (Praks, P., Rasmussen, A., Lye, K. O., Martinovič, J., Praksová, R., Watson, F., & Brkić, D.)

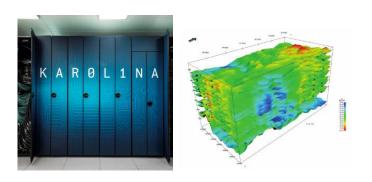
References:

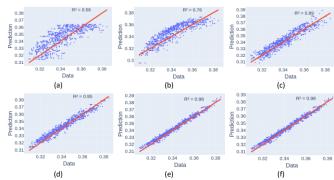
- OPM Flow: Rasmussen et al., Computers & Mathematics with Applications, 2021, https://doi.org/10.1016/j.camwa.2020.05.014
- SA & ML for OPM Flow: Praks et al., Heliyon, 2024. https://doi.org/10.1016/j.heliyon.2024.e40044



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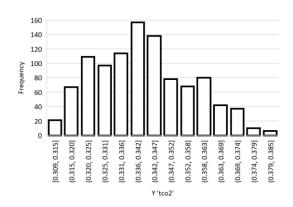
Development of fast interpretable ML approximations of OPM Flow results

(a) 1D model, (b) 2D model, ..., (f) 6D model

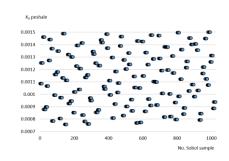
WORKFLOW FOR ML-ACCELERATED SENSITIVITY ANALYSIS IN OPM FLOW CARBON SEQUESTRATION

Uncertainty type	Minimum	Maximum
Parametric/model	30 degrees C	40 degrees C
Parametric/model	1100 mD	5000 mD
Parametric/model	0.00075 mD	0.0015 mD
Parametric/model	1100 mD	5000 mD
Parametric/model	0.27	0.4
Parametric/model	-10 m	+10 m
	Parametric/model Parametric/model Parametric/model Parametric/model Parametric/model	Parametric/model 30 degrees C Parametric/model 1100 mD Parametric/model 0.00075 mD Parametric/model 1100 mD Parametric/model 0.27

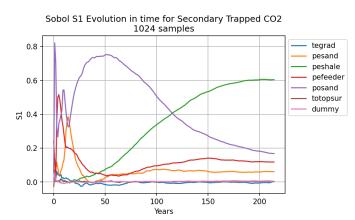
1. Define uncertain inputs



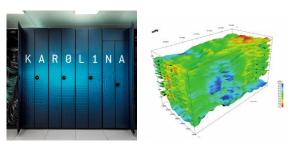
4. Uncertainty analysis of OPM Flow model output (secondary trapped CO₂)



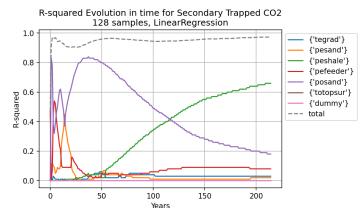
2. Generate quasi-random samples of model inputs



5. Classical approach: Time-dependent global sensitivity analysis Requires many OPM Flow simulations

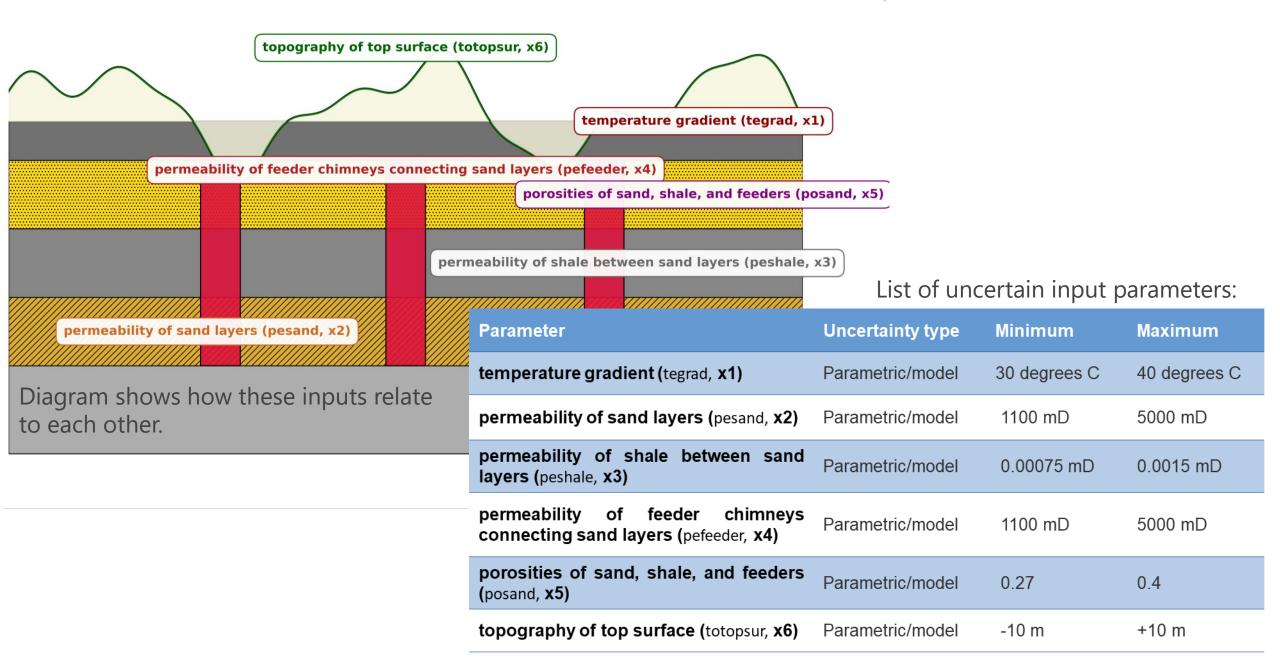


3. Run OPM Flow simulations on HPC cluster and extraction of outputs



 Novel accelerated approach:
 Predict sensitivity indices using ML Reduces computational cost

INPUTS AND OUTPUTS FOR SA & ML IN OPM FLOW CARBON SEQUESTRATION



INPUTS AND OUTPUTS FOR SA & ML IN OPM FLOW CARBON SEQUESTRATION

✓ Inputs:

Parameter	Uncertainty type	Minimum	Maximum
temperature gradient (tegrad, x1)	Parametric/model	30 degrees C	40 degrees C
permeability of sand layers (pesand, x2)	Parametric/model	1100 mD	5000 mD
permeability of shale between sand layers (peshale, x3)	Parametric/model	0.00075 mD	0.0015 mD
permeability of feeder chimneys connecting sand layers (pefeeder, x4)	Parametric/model	1100 mD	5000 mD
porosities of sand, shale, and feeders (posand, x5)	Parametric/model	0.27	0.4
topography of top surface (totopsur, x6)	Parametric/model	-10 m	+10 m

✓	Οι	ıtp	uts
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 OPM Flow simulations → trapped_co2.csv

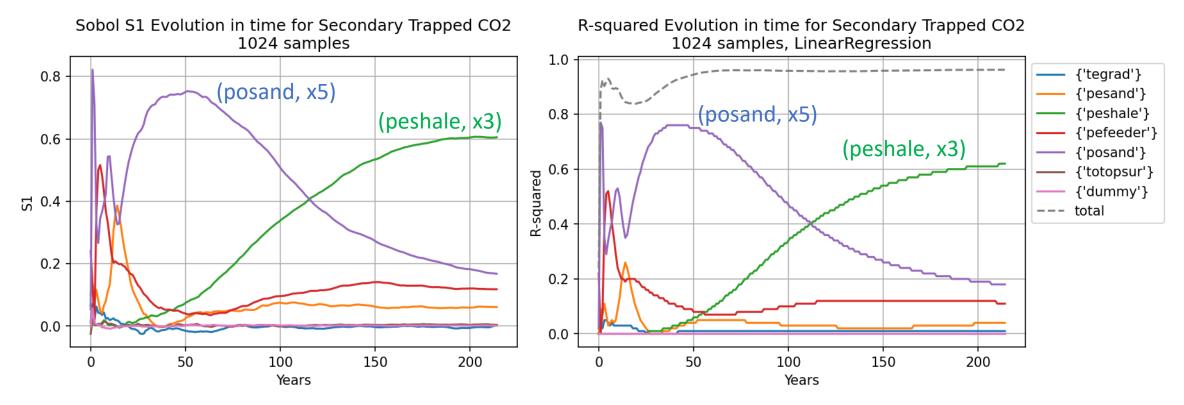
trapped_co2
0.333245
0.336624
0.336523
0.343364
0.330078
0.342297
0.333459

	Quasi Monte-Carlo samples	Tegrad	pesand	peshale	pefeeder	posand	totopsur	dummy
	(input_params.csv)	30.01465	2568.213	0.001086	2998.584	0.342427	6.884766	0.241699
	 Includes dummy 	35.87402	2568.213	0.001086	2998.584	0.342427	6.884766	0.241699
	parameter for validation	30.01465	3817.432	0.001086	2998.584	0.342427	6.884766	0.241699
		30.01465	2568.213	0.001254	2998.584	0.342427	6.884766	0.241699
		30.01465	2568.213	0.001086	4304.932	0.342427	6.884766	0.241699
		30.01465	2568.213	0.001086	2998.584	0.38978	6.884766	0.241699
		30.01465	2568.213	0.001086	2998.584	0.342427	4.130859	0.241699

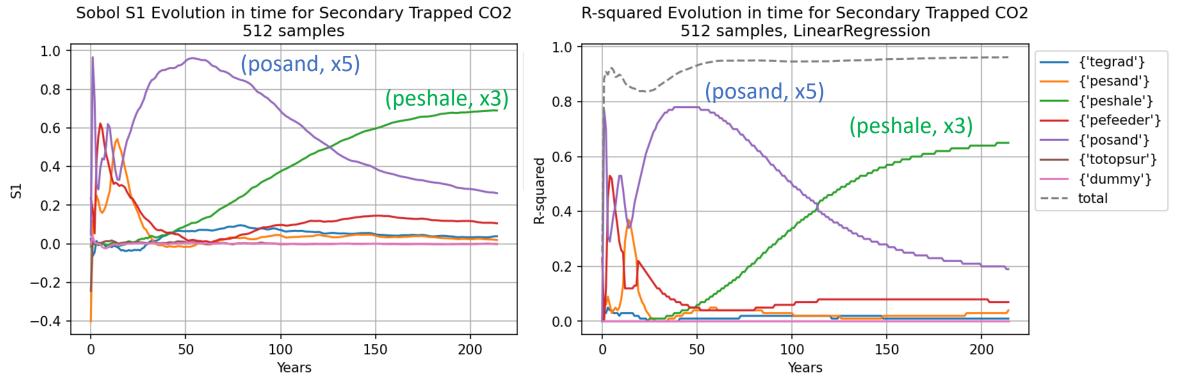
OPM Flow model and simulations:



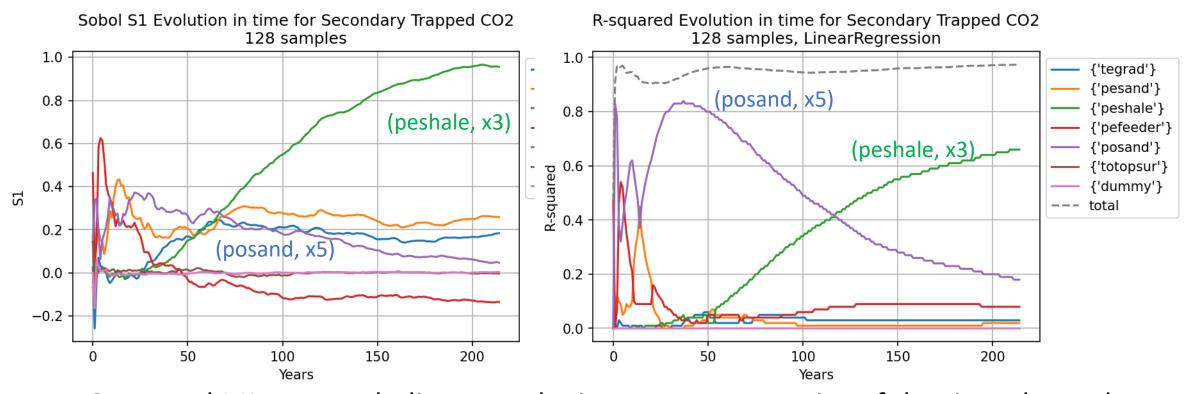
Sensitivity analysis & ML predictions:



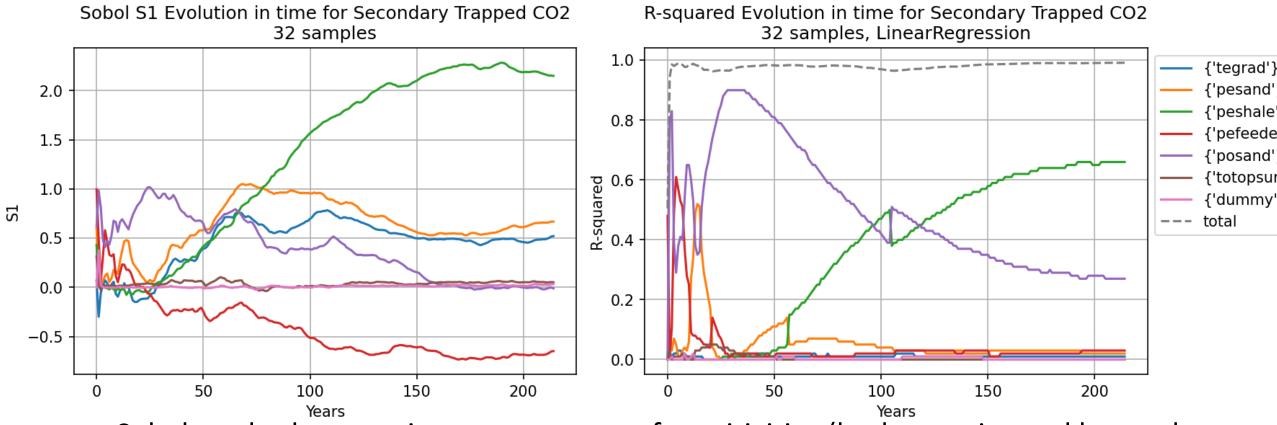
Sobol method and our novel machine learning approach discover the important properties of the time-dependent carbon sequestration process for 1 024 Quasi Monte Carlo (QMC) samples: Sand porosity (posand, x5) is more important than shale permeability (peshale, x3) for about the first 120 years.



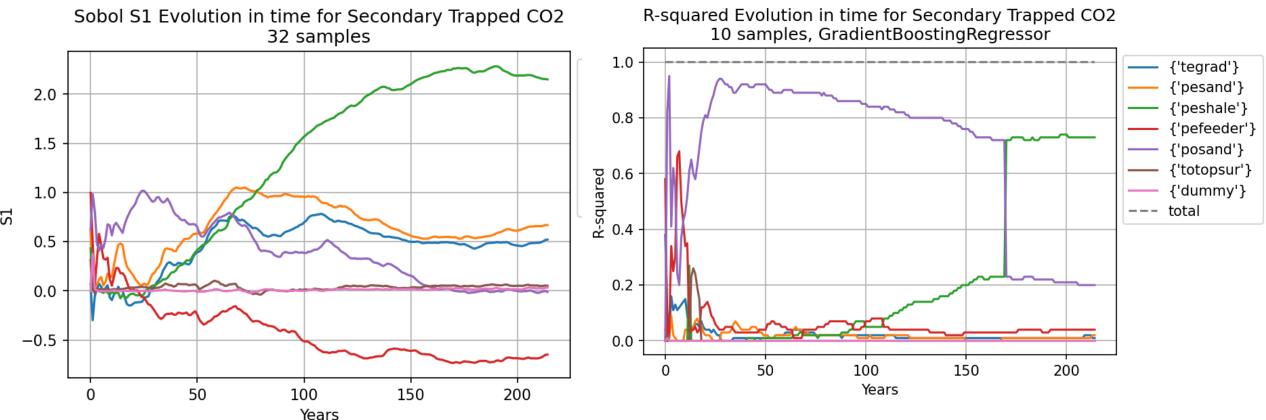
- Sobol method has incorrect negative sensitivity for Year 0 due to an insufficient number (512) of QMC samples.
- Our novel ML approach discovers the important properties of the timedependent carbon sequestration process for 512 QMC samples: Sand porosity (posand, x5) is more important than shale permeability (peshale, x3) for about the first 120 years.



- Our novel ML approach discovers the important properties of the time-dependent carbon sequestration process for 128 QMC samples: Sand porosity (posand, x5) is more important than shale permeability (peshale, x3) for about the first 120 years.
- Sobol method does not detect the pattern and has incorrect negative sensitivity for Year 0 due to an insufficient number (128) of QMC samples.



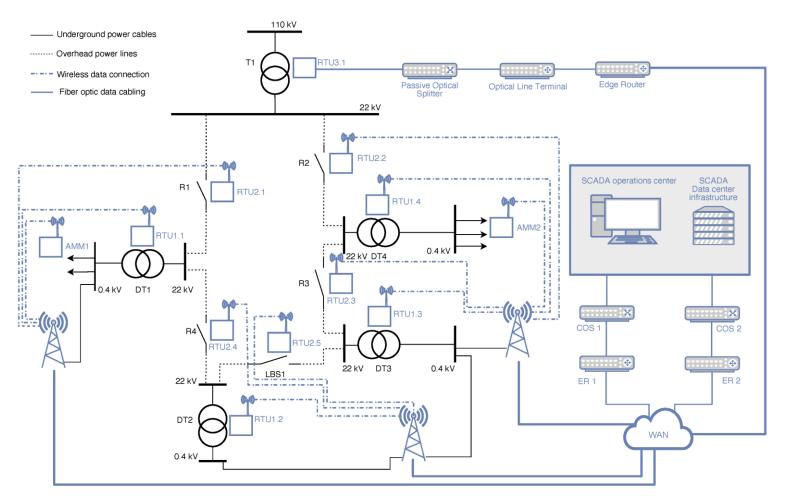
- Sobol method reports incorrect pattern of sensitivities (both negative and larger than one) due to an insufficient number (32) of QMC samples.
- Our novel ML approach discovers the important properties of the time-dependent carbon sequestration process even for 32 QMC samples: Sand porosity (posand, x5) is more important than shale permeability (peshale, x3) for about the first 95-120 years. (1 024 QMC samples estimate it for the first 120 years.)



- Sobol method reports incorrect pattern of sensitivities (both negative and larger than one) due to an insufficient number (32) of QMC samples.
- Our novel ML approach discovers the important properties of the time-dependent carbon sequestration process even for 10 QMC samples: Sand porosity (posand, x5) is more important than shale permeability (peshale, x3) for about the first 170 years. (1 024 QMC samples estimate it for the first 120 years.)

INTERCONNECTED INFRASTRUCTURES

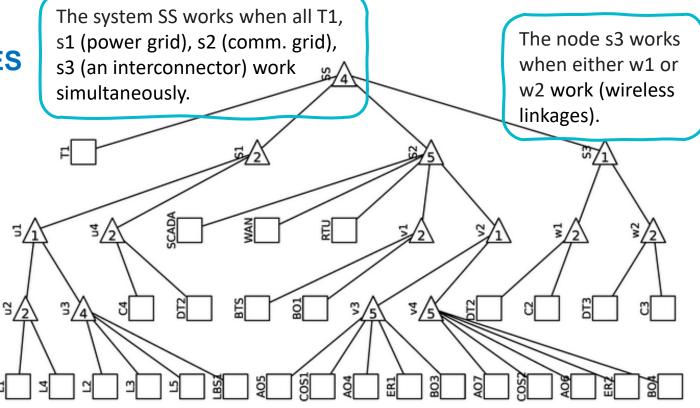
- Two interconnected networks:
 - Power distribution network
 - Communication network
- Power distribution network:
 - Based on Czech distribution grid
 - Ring topology structure



Parameters of the model and the reference solution is given by Běloch M., Praks P.; Praksová R.; Fujdiak R.; Vrtal M.; Briš R.; Brkić D. (2025). Evaluating the unavailability of interconnected power and communication networks with open-source tools on a petascale cluster. Energy Exploration & Exploitation. 2025;0(0). https://doi.org/10.1177/01445987251377791

INTERCONNECTED INFRASTRUCTURES

Components	MTBF (h)	MTTR (h)
Transformer T1	26,310.709	4.403
Overhead line L1	54,750.000	11.417
Overhead line L2	41,714.286	11.417
Overhead line L3	62,571.429	11.417
Overhead line L4	48,666.666	11.417
Overhead line L5	43,800.000	11.417
Load break switch LBS1	224,621.087	5.702
Underground cable C2	57,737.828	85.000
Underground cable C3	38,491.886	85.000
Underground cable C4	153,967.543	85.000
Transformer DT2	43,800.361	0.361
Transformer DT3	43,800.361	0.361
Buried optic fiber BO1	821,875.000	12.000
Buried optic fiber BO3	1,753,333.333	12.000
Buried optic fiber BO4	1,753,333.333	12.000
Edge router ER1	16,246.780	0.780
Edge router ER2	16,246.780	0.780
Aerial optic fiber AO4	500,000.000	6.000
Aerial optic fiber AO5	1,093,750.000	6.000
Aerial optic fiber AO6	500,000.000	6.000
Aerial optic fiber AO7	1,093,750.000	6.000



The interconnected network represented by the success tree.

Components	MTBF (h)	MTTR (h)	
Core optical switch COS1	5,000,014.000	14.000	
Core optical switch COS2	5,000,014.000	14.000	
SCADA operation and data center	175,200.000	184.600	
Wide area network WAN	100,000.000	4.000	
Remote terminal unit RTU	100,048.000	48.000	
Base transceiver station BTS	100,000.000	4.000	

Characteristic values of the power grid (and the communication network. In all cases, the shape parameter of Weibull β =2 is used. See Briš, R., Praks, P., Fujdiak, R., Vrtal, M., & Brkić, D. (2025). Maintenance optimization for unavailability enhancement of representative interconnected infrastructure based on minimum cost. Science Progress, 108(3) 1–23. https://doi.org/10.1177/00368504251366357

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Aerial optic fiber AO7	1,093,750.000	6.000

Reliability and maintainability metrics of components:

MTTR (Mean Time to Repair)

Average time to repair or replace a component.

- → Low MTTR = quick recovery after failure → high maintainability
- MTTF (Mean Time to Failure)

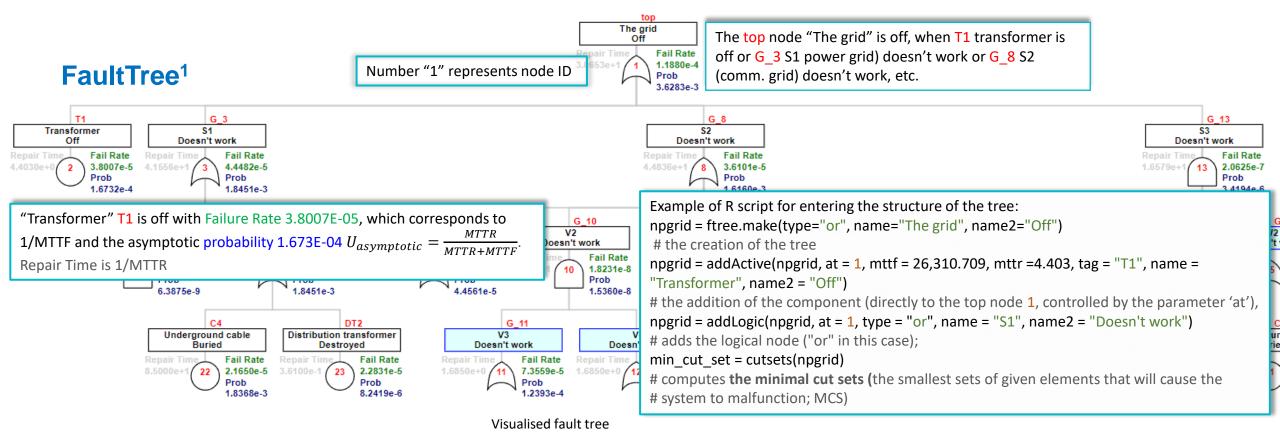
Average operational time before a component fails.

- → **High MTTF** = long-lasting operation → **high reliability**
- MTTR measures repair speed, while MTTF measures expected lifespan.

Simulation setup:

- We simulate component downtimes across a wide range:
 0.1×MTTR to 10×MTTR
- How do these variations affect system unavailability at t = 5 years?

Characteristic values of the power grid and the communication network. In all cases, the shape parameter of Weibull β =2 is used. See Briš, R., Praks, P., Fujdiak, R., Vrtal, M., & Brkić, D. (2025). Maintenance optimization for unavailability enhancement of representative interconnected infrastructure based on minimum cost. Science Progress, 108(3) 1–23. https://doi.org/10.1177/00368504251366357



- Provides an interactive and easy-to-understand visualization of the fault tree structure.
- The tree structure is defined by addActive and addLogic commands.
- addActive individual elements (leaves) are described by Mean Time to Failure (MTTF) and Mean Time To Repair (MTTR)
- addLogic adds the logical node ("and", "or").
- For all logical nodes, FaultTree computes failure rates and asymptotic probabilities that the given node is not operating.
- The FaultTree uses the Binary Decision Diagrams (BDD) to efficiently compute the minimal cut sets of a fault tree.
- Given the asymptotic unavailability probabilities of the leaves (terminal events), BDD also computes the probability of the root event "The grid Off" by the Shannon's decomposition².

Used logic symbols (gates)

¹ Silkworth, D. (2023). FaultTree: Fault Trees for Risk and Reliability Analysis. R package version 1.0.1. URL: https://CRAN.R-project.org/package=FaultTree

² Rauzy, A. (1993). New algorithms for fault trees analysis. Reliability Engineering & System Safety, 40(3), 203-211. https://doi.org/10.1016/0951-8320(93)90060-c

Ftaproxim³

Component	MTTF (h)	MTTR (h)	
Transformer T1	26,310.709	4.403	

```
T1 = list(
states=c("OK","F"),
G=rbind(c(NA,1), c(1,NA)), # G is a transition matrix describing the probabilities of going from the state "OK" to "F" and vice versa.
dist=c("weibull", "unif"),
# definition of probability density function (pdf) of time to failure and time to repair
# weibull represents pdf of the time to failure and uniform represents the pdf of time to repair.
param=list(c(beta, T1_mttf/gamma(1+1/beta)), c(0, 2*T1_mttr)))
# parameters of pdfs: (shape β, scale θ) for Weibull
# and start and end of the interval for uniform pdf.
```

Definition of the component Transformer T1, which has 2 states.

- A deterministic iterative method for unavailability estimation for multistate systems
- Uses the minimal cut sets (MCSs) of the fault tree; individual elements (leaves) are described by density functions of time to failure and time to repair.
- MTTF of a given Weibull component is computed as $MTTF = \theta \cdot \Gamma(1 + 1/\beta)$, where θ is a scale, β is a known shape and Γ is a gamma function. As the θ parameter of the distribution is unknown, we have to factor it out.
- Inputs: list of the components and their failures and repair characteristics, MCS, mission time, time step, tolerance
- The output: the values of unavailabilities of each element in every time step (computed by the command FTUna)
- Very versatile tool, can work with all commonly used density functions (Weibull, exponential, uniform, normal)
- The time and space complexity of the algorithm is exponential in the number of discrete time steps⁴

³-Niloofar, P., Haghbin, H., & Lazarova-Molnar, S. (2022). Ftaproxim - R Package for Proxel-Based Simulation of Fault Trees: A Case Study. In 2022 6th International Conference on System Reliability and Safety (ICSRS). 2022 6th International Conference on System Reliability and Safety (ICSRS). IEEE. https://doi.org/10.1109/icsrs56243.2022.10067562

⁴G. Horton (2002), A new paradigm for the numerical simulation of stochastic petri nets with general firing times. Proceedings of the European Simulation Symposium (2002), pp. 129-136

GRID SEARCH OF INPUT PARAMETERS IN FTAPROXIM

Objective: Analyze how **time step** and **tolerance** parameters influence the **computation time** and **result accuracy** of unavailability estimation in Ftaproxim's unavailability estimation.

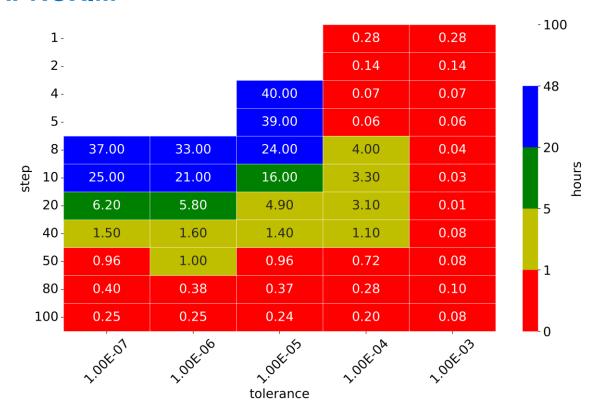
Methodology:

- Grid Search on a distribution network model using Barbora cluster.
- 55 combinations:
 - Step sizes: {1, 2, 4, 5, 8, 10, 20, 40, 50, 80, 100}
 - **Tolerances:** {1e-03, 1e-04, 1e-05, 1e-06, 1e-07}
- Executed as 55 parallel processes.
- Max runtime per test: 48 hours

Code availability: Python and R open-source

implementations provided, see (Běloch, Praks et al., 2025):

- Appendix A: Python code
- Appendix B: R code



Heatmap shows impact of parameter choices on **computational time.** White cells: combinations exceeding 48h limit.

Parameters of the model and the reference solution is given by Běloch M., Praks P.; Praksová R.; Fujdiak R.; Vrtal M.; Briš R.; Brkić D. (2025). Evaluating the unavailability of interconnected power and communication networks with open-source tools on a petascale cluster. Energy Exploration & Exploitation. 2025;0(0). https://doi.org/10.1177/01445987251377791

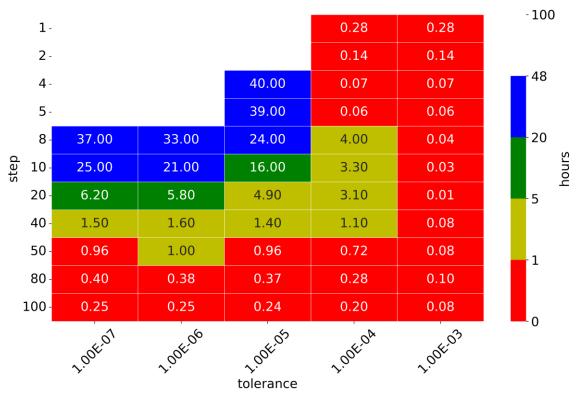
RESULTS OF GRID SEARCH ON BARBORA CLUSTER

Observations:

- Higher tolerance → faster computation.
- Larger time steps → reduced computation time.
- Trade-off: Fastest runs often yield invalid results

Key findings:

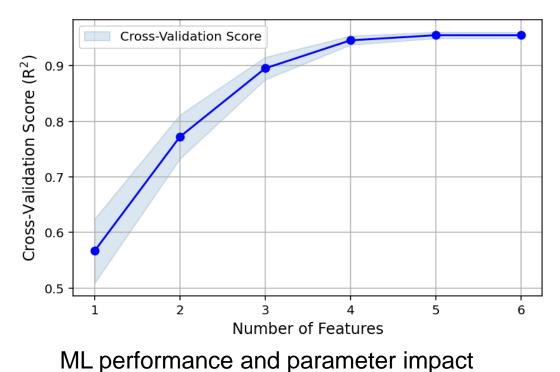
- Tolerance = 1×10⁻³: Fastest, but invalid results (no transitions, unavailability = 0).
- Optimal combinations:
 - Tolerance = 1×10^{-4} , Step = $50 \to 0.72 \text{ h}$
 - Tolerance = 1×10^{-6} , Step = $8 \rightarrow 33 \text{ h}$
 - These combinations provide plausible results, balancing accuracy and performance (Běloch, Praks et al., 2025).



Heatmap shows impact of parameter choices on **computational time.** White cells: combinations exceeding 48h limit.

Parameters of the model and the reference solution is given by Běloch M., Praks P.; Praksová R.; Fujdiak R.; Vrtal M.; Briš R.; Brkić D. (2025). Evaluating the unavailability of interconnected power and communication networks with open-source tools on a petascale cluster. Energy Exploration & Exploitation. 2025;0(0). https://doi.org/10.1177/01445987251377791

MACHINE LEARNING FOR IDENTIFYING KEY PARAMETERS IN RELIABILITY MODELING



Objective:

Use machine learning to identify parameters that most influence system availability in interconnected **power and communication networks**.

Method:

An optimization algorithm selects input parameters that maximize the R² (coefficient of determination).

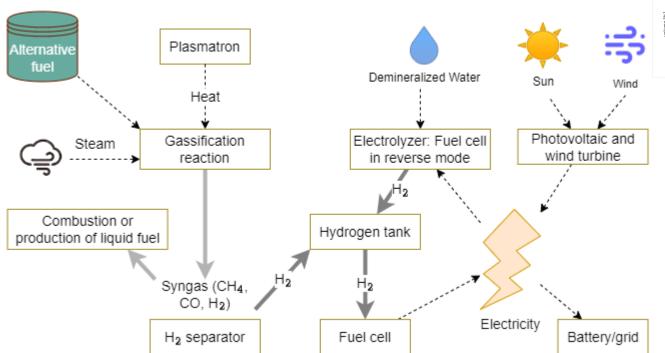
Key result:

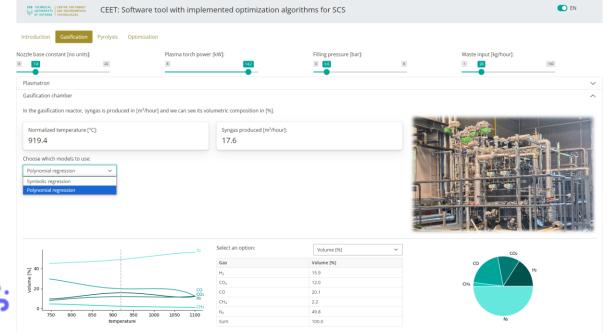
Out of **27 parameters**, using only **four** – SCADA, RTU, C4, T1 – achieves an $\mathbb{R}^2 = 0.9938$, explaining **99.38%** of the variability in system availability over a time horizon $\mathbf{t} = \mathbf{5}$ years.

Number of Features	Selected Features	Selected Features Names	R-squared	Std. Error
1	(24)	('SCADA')	0.4394	0.0149
2	(23, 24)	('SCADA', 'RTU')	0.7971	0.0082
3	(11, 23, 24)	('SCADA', 'RTU', 'C4')	0.9057	0.0041
4	(0, 11, 23, 24)	('SCADA', 'RTU', 'C4', 'T1')	0.9938	0.0003
5	(0, 11, 23, 24, 25)	('BTS', 'SCADA', 'RTU', 'C4', 'T1')	0.9968	0.0002
6	(0, 11, 23, 24, 25, 26)	('WAN', 'BTS', 'SCADA', 'RTU', 'C4', 'T1')	0.9981	0.0001

Developed Tool: Shinyenet

- Developed Tool: Shinyenet an online software tool implementing optimization algorithms for selected waste-to-energy processes.
- Application Context: Designed for the innovation polygon CEETe at VSB – Technical University of Ostrava.
- Software is running on the IT4Innovations infrastructure.







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The software will be **open-source** and is available for testing:

https://shinyenet.vsb.cz/

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OF OSTRAVA

IT4INNOVATIONS
NATIONAL SUPERCOMPUTING
CENTER



Thank You for Your attention!

Pavel.Praks@vsb.cz

This work was co-funded by the European Union in the Increasing the resilience of power grids in the context of decarbonisation, decentralisation, and sustainable socioeconomic development project (CZ.02.01.01/00/23_021/0008759) under the OP JAC.





Increasing the resilience of power grids in the context of decarbonisation, decentralisation, and sustainable socioeconomic development

• Acronym: ZEUS

• 2025 – 2028

Project ID: CZ.02.01.01/00/23_021/0008759

Provider: Programme Johannes Amos Comenius (MEYS)

• The project focuses on research and development of software tools for analysing installation efficiency and resilience of new energy sources and storage capacities for use within energy communities and research in the field of hydrogen production, transport and storage. This includes social science research to promote acceptance of these specialised technologies by the general public, development of cooperation with the application sector, project preparation, infrastructure upgrade, and compliance with RIS3.

https://www.it4i.cz/en/research/research-projects



