

Integration of the Fuel Matrix Degradation Model (embedded)

Virtual Fact-Finding Meeting
Supporting
NWTRB Fall 2021 Meeting
October 13-14, 2021

Paul Mariner
Sandia National Laboratories

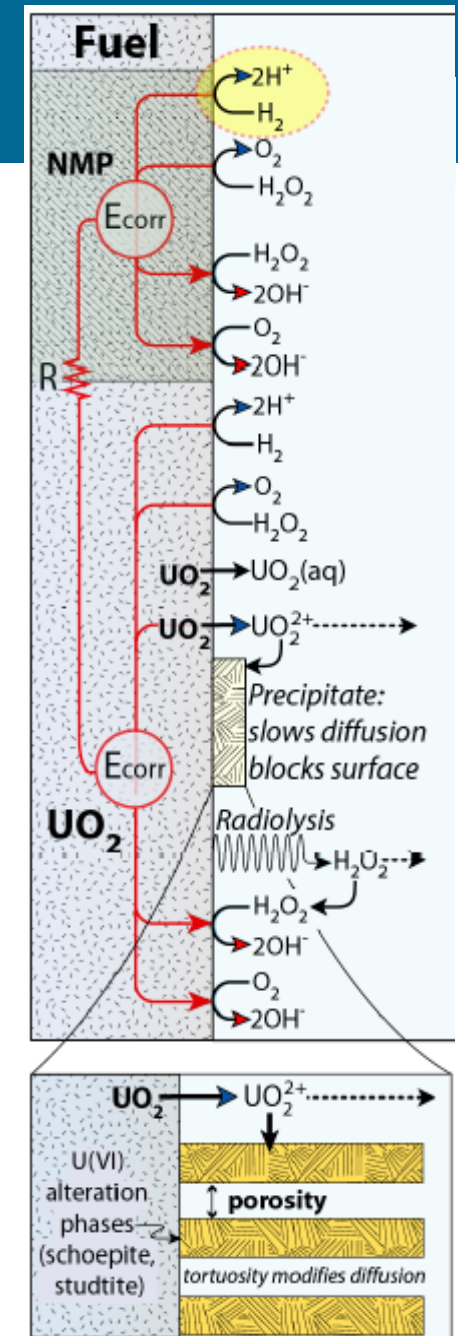


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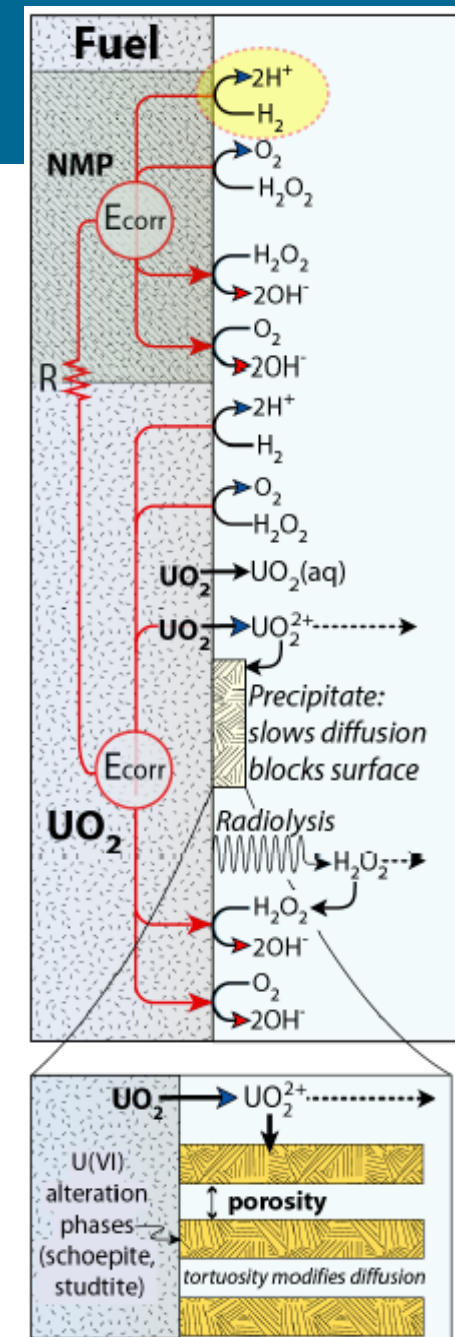
Outline

- Fuel Matrix Degradation (FMD) process model
- Motivation for surrogate
- Surrogate approach
- Results and future plans



Fuel Matrix Degradation (FMD) Process Model

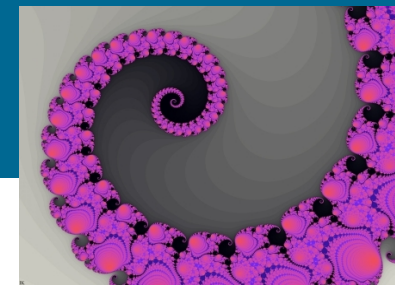
- Complex set of processes
 - Radiolysis
 - Aqueous and surface kinetic reactions
 - Oxidation of H_2 via noble metal particle (NMP) catalyst
 - Growth of an alteration layer
 - 1-D reactive transport through alteration layer
 - Diffusion of reactants and products through the alteration layer
- Expensive to run



(Jerden et al. 2017)

Motivation for FMD Surrogate Model

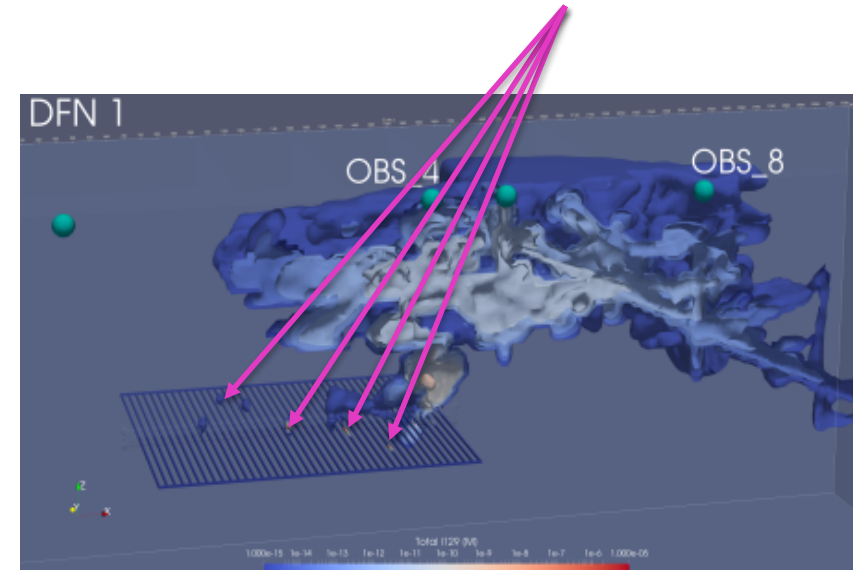
- Fuel degradation rates
 - Change by orders of magnitude
 - Highly sensitive to local conditions (temperature, burnup, dose rate, concentrations of H_2 , O_2 , Fe^{2+} and CO_3^{2-})
- Performance metrics highly sensitive to fuel degradation rates
- Process model too expensive to run on individual waste packages



Process model



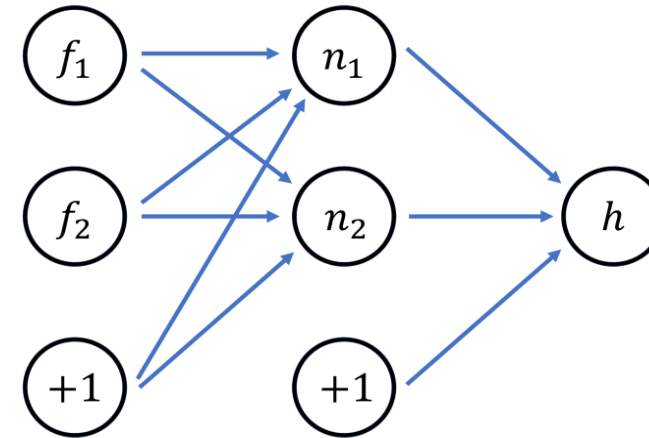
Surrogate model



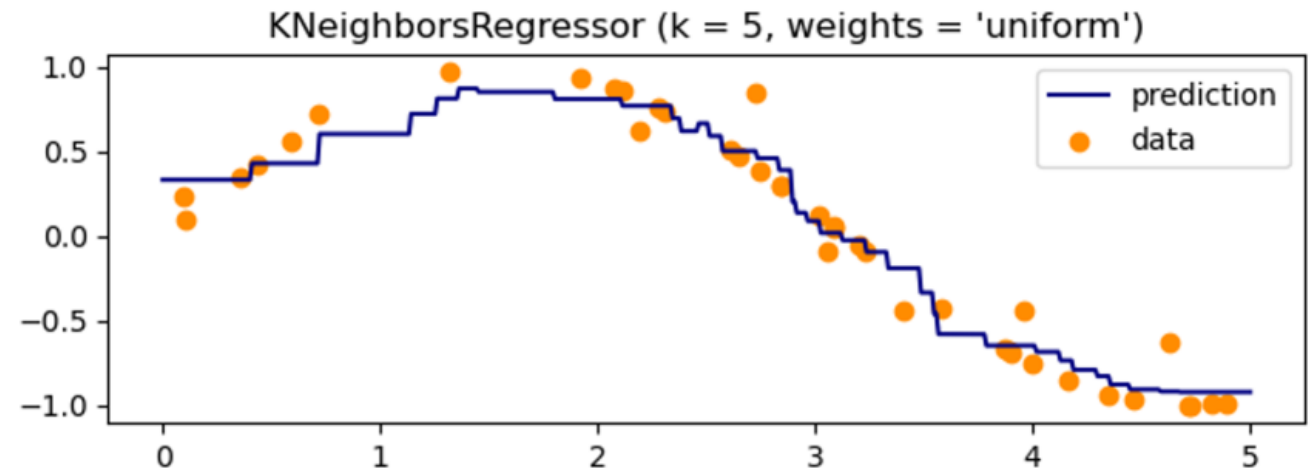
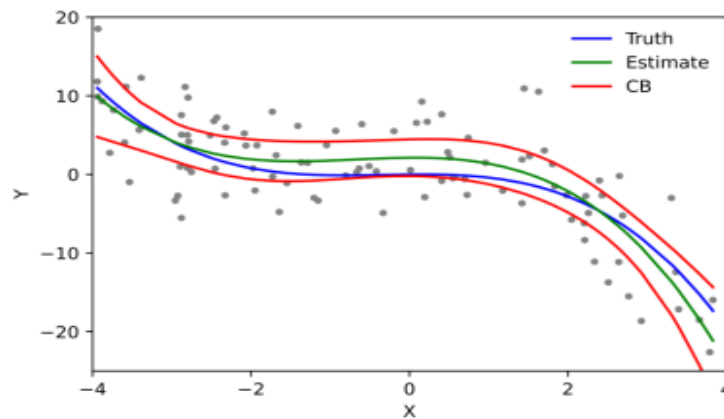
Performance assessment (PA) model

Surrogate Model Approaches

- Three FMD surrogate models tested
 - Artificial Neural Network Regressor (ANN)
 - K-Nearest Neighbors Regressor (kNNr)
 - Polynomial Regression



Single layer feed-forward **neural network** with 2 input features and 2 neurons in the hidden layer



Development and Integration of Surrogate Models

- Surrogate training/testing data
 - Generated from Matlab process model
 - Latin Hypercube Sampling (LHS)
 - 30,000 runs (samples), 100 points per run
 - 3,000,000 training data points
 - 300,000 test data points
- PFLOTTRAN integration
 - Artificial Neural Network (ANN)
 - Trained coefficients read in from file
 - Direct calculation
 - K Nearest Neighbors (kNNr)
 - Scan search tree for k nearest neighbors (e.g., k = 250)
 - Calculate response using inverse distance rule

| Input Features | Training/Testing Range |
|---|-------------------------------------|
| Initial Temp. (K) | 298 – 393 |
| Fuel Burnup (GWd/MTHM) | 20 – 70 |
| Env CO ₃ ²⁻ Conc (mol/m ³)* | 10 ⁻⁴ – 10 ⁻¹ |
| Env O ₂ Conc (mol/m ³)* | 10 ⁻⁷ – 10 ⁻³ |
| Env Fe ²⁺ Conc (mol/m ³)* | 10 ⁻³ – 10 ⁻² |
| Env H ₂ Conc (mol/m ³)* | 10 ⁻⁷ – 10 ⁻¹ |

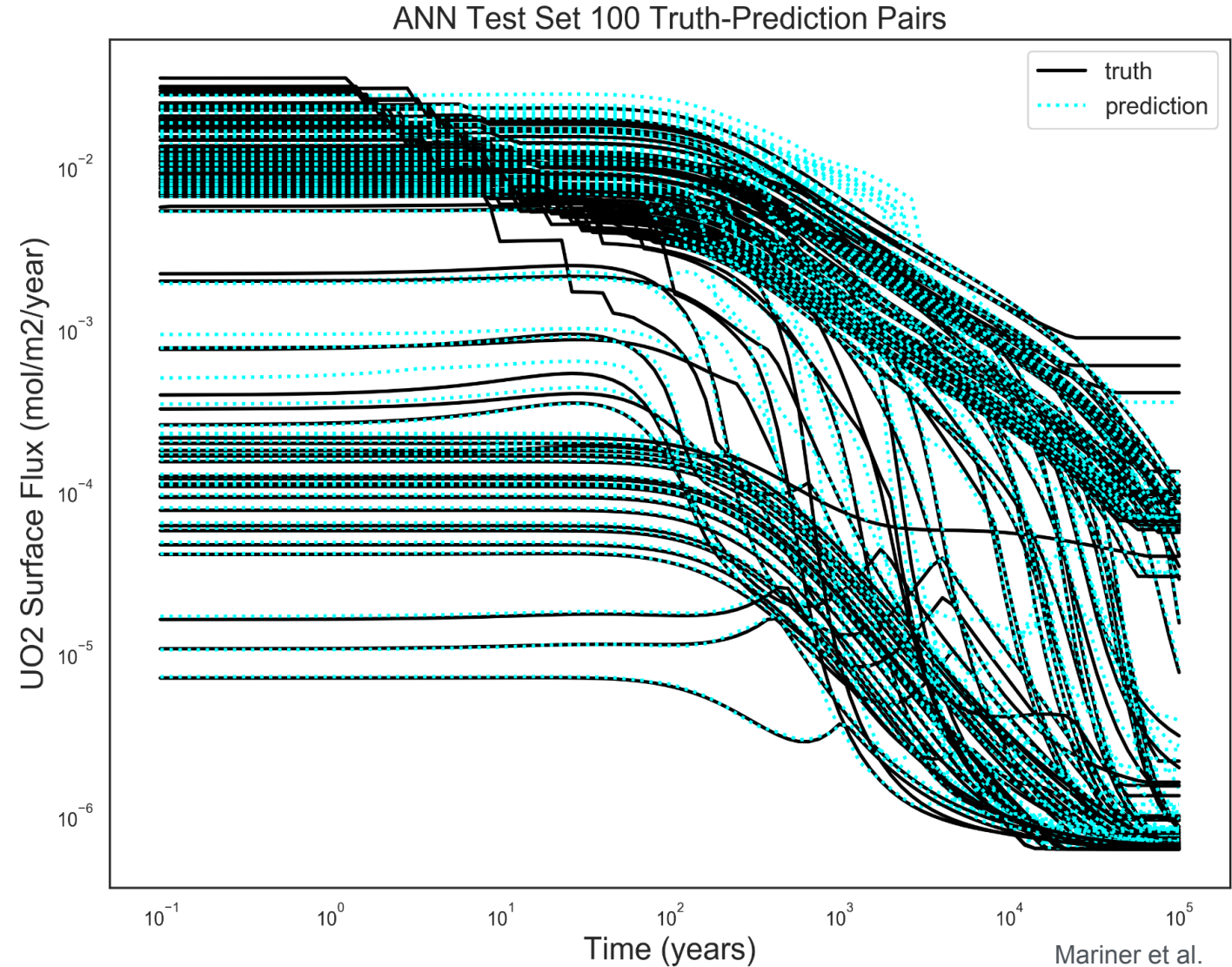
* Log-uniform sampling distribution

Mariner et al.
2020

Surrogate Performance – Accuracy

- Trained surrogates tested against independently generated test data
- ANN and kNNr error much lower than polynomial surrogate error

| Error Metric | ANN (2020) | kNNr (2020) |
|---|------------|-------------|
| Mean Square Error (mol/m ² /yr) ² | 3.56e-6 | 5.95e-6 |
| Mean Absolute Error (mol/m ² /yr) | 9.30e-4 | 1.25e-3 |
| Mean Absolute Percentage Error | 31.2% | 78.4% |

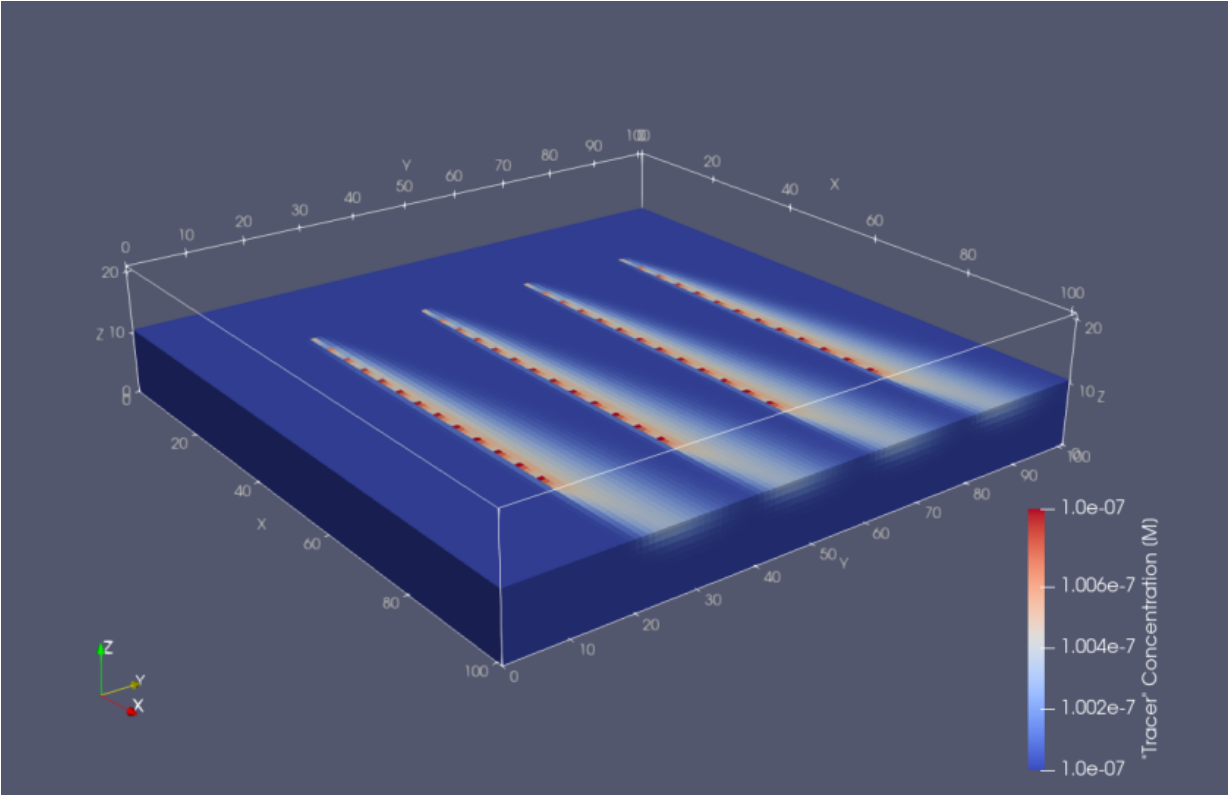


Mariner et al.
2020

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Surrogate Performance – Speed

Test problem: 52 failed waste packages in a flow field (Mariner et al. 2015)



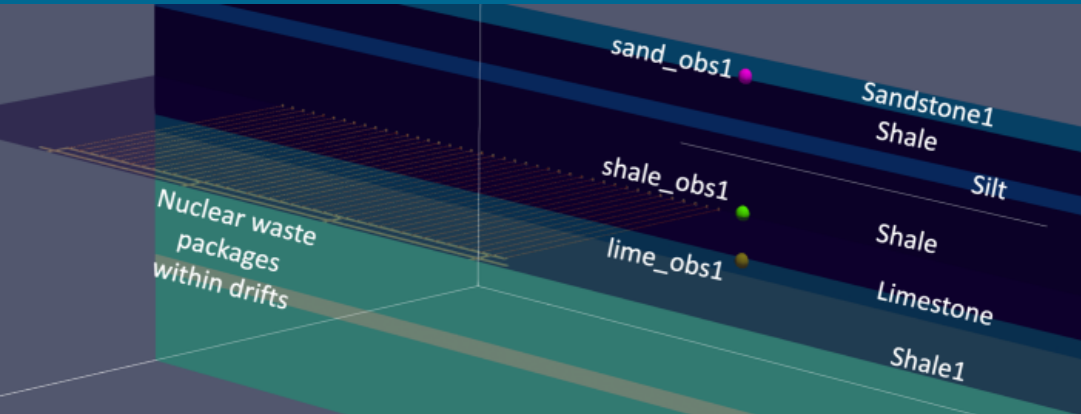
- Speedup of 30,000x (ANN) and 2,800x (kNNr) compared to 2015 coupling of process model

Compute time (sec) and speed-up factor (x)

| Process | Coupled Process Model (2015) | ANN (2020) | kNNr (2020) |
|------------|------------------------------|----------------|---------------|
| Waste Form | 1,522 | 0.05 (30,440x) | 0.54 (2,820x) |
| Flow | 128 | 61.0 (2.1x) | 60.3 (2.1x) |
| Transport | 244 | 147.8 (1.65x) | 117.6 (2.1x) |

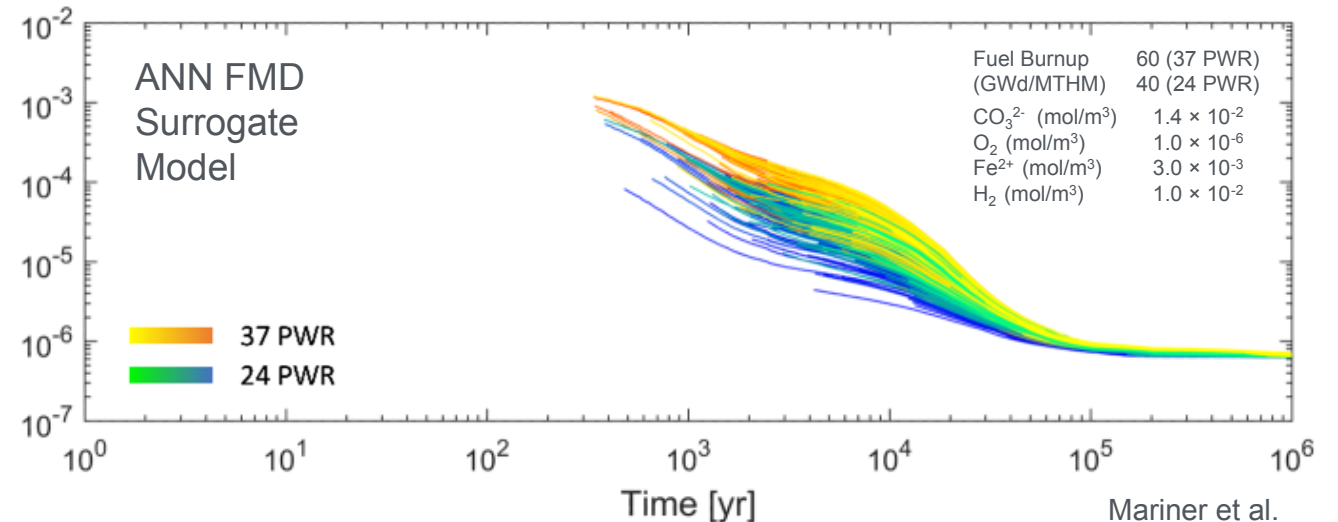
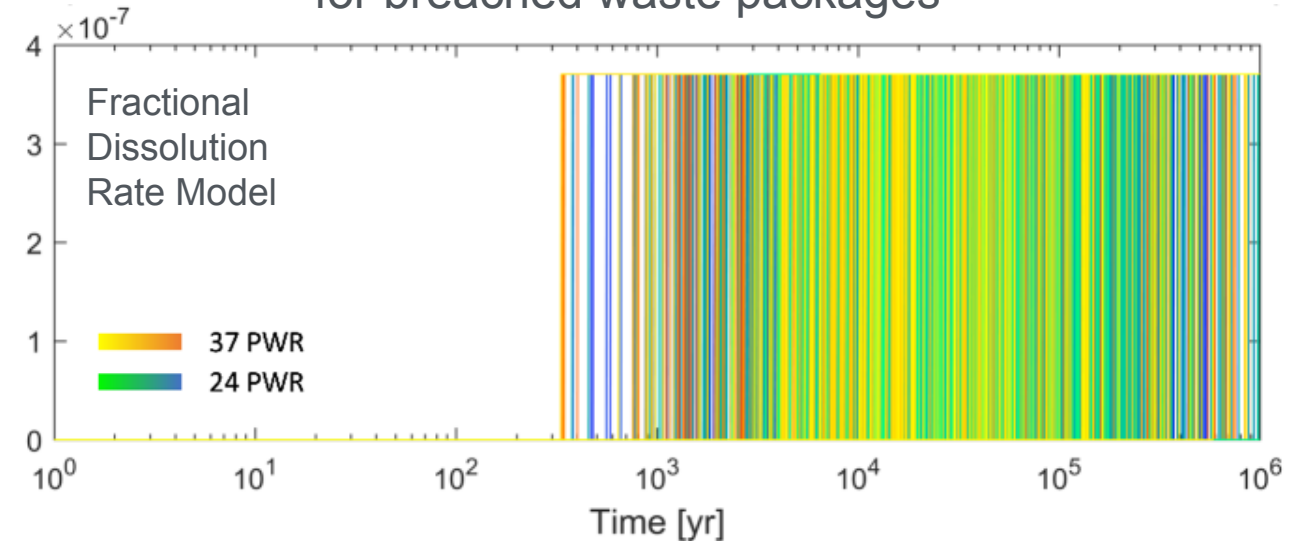
Mariner et al. 2020

Shale Reference Case Demonstration



- Traditional Fractional Dissolution Rate (FDR) model ignores local conditions and fuel age
- FMD accounts for fuel age and local conditions

Degradation rates ($\text{mol m}^{-2} \text{yr}^{-1}$) for breached waste packages



Summary and Future Work

■ Summary

- Surrogate modeling has allowed us to emulate fuel matrix degradation processes at each waste package in full-scale reference case simulations
 - Allows accounting for local environmental conditions (temperature and chemistry) and individual burnups and dose rates
- Artificial neural network (ANN) and k nearest neighbors regression (kNNr) are quite accurate and extremely fast

■ Future work

- Awaits the next major upgrade of the process model

References

- Jerden, J., V. K. Gattu and W. Ebert (2017). *Progress Report on Development of the Spent Fuel Degradation and Waste Package Degradation Models and Model Integration*. SFWD-SFWST-2017-000091, SFWD-SFWST-2017-000095. Lemont, Illinois, Argonne National Laboratory.
- Mariner, P.E., Berg, T.M., Chang, K.W., Debusschere, B.J., Leone, R.C., and Seidl, D.T. (2020), *Surrogate Model Development of Spent Fuel Degradation for Repository Performance Assessment*. SAND2020-10797 R. Sandia National Laboratories.
- Mariner, P.E., Gardner, W.P., Hammond, G.E., Sevougian, S.D., and Stein, E.R. (2015), *Application of Generic Disposal System Models*. SAND2015-10037R. Sandia National Laboratories.

Questions