

**Demonstrating Autonomous Control, Remote
Operation, and Human Factors for
Microreactors:**
*Architecture Development and Training of
ML/DL Algorithms for VSLIM Microreactor*

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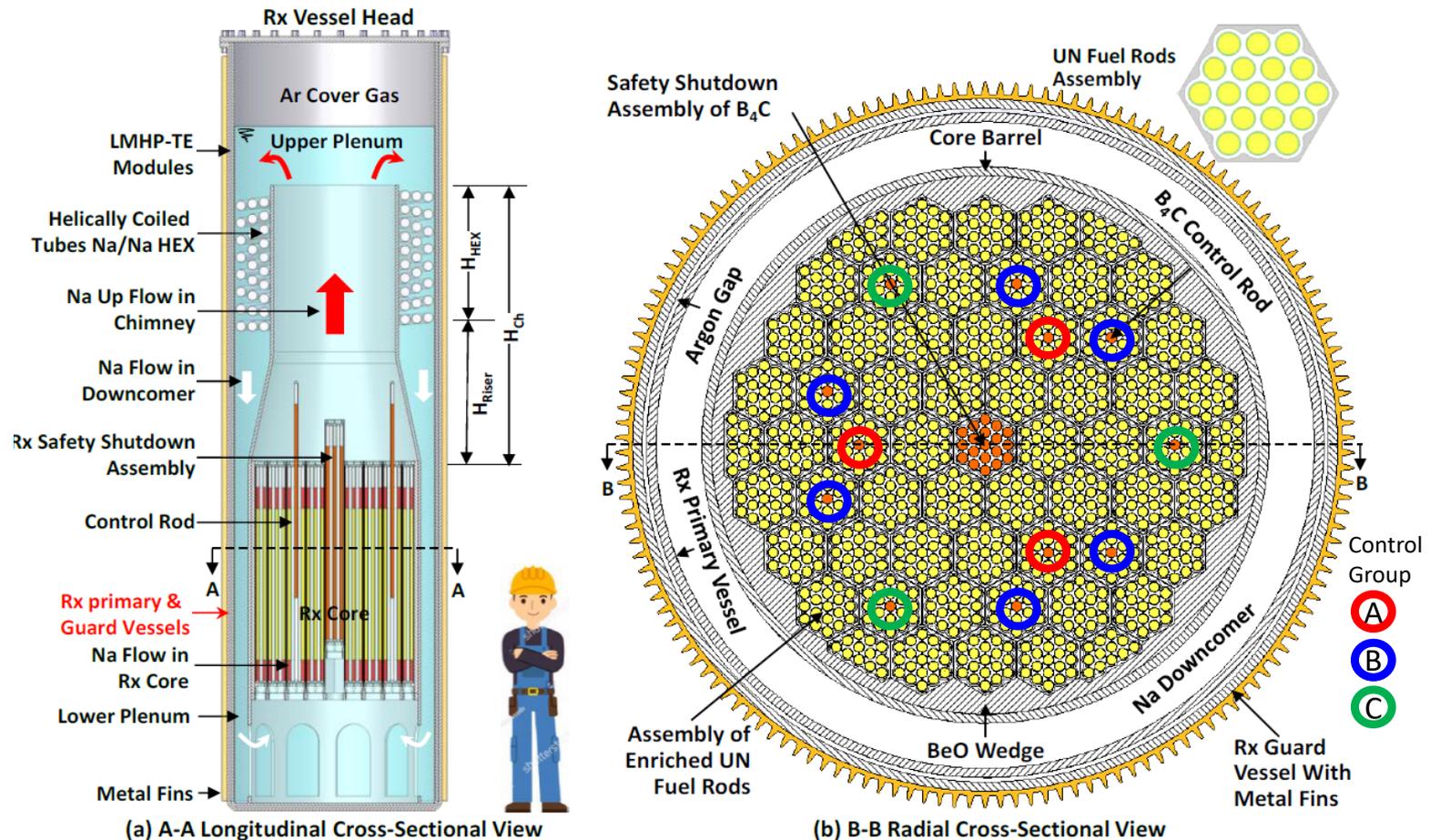
Objectives

- Develop a dynamic model of the Very Small, Long-Life Modular (**VSLIM**) microreactor developed at UNM-ISNPS
 - A walk away safe microreactor design for generating 1.0-10 MW(t)
 - Cooled by natural circulation of in-core liquid sodium
 - Offers passive and redundant decay heat removal, redundant reactor operation and control
 - Factory assembled and sealed and requires no onsite storage of fresh or spent nuclear fuel
 - Offers passive auxiliary electric power generation after reactor shutdown, independent of on-site and off-site power sources
- Use simulation results of VSLIM dynamic model to train reactor controllers using ML algorithms
 - Generate data sets of reactor startup scenarios at different initial and final power for training neural networks of the ML algorithms
 - Implement trained neural network into a real-time reactor controller coupled to VSLIM dynamic Simulink model
 - Test and validate accuracy of neural network for determining control rods displacements during simulated transients using the VSLIM microreactor controller

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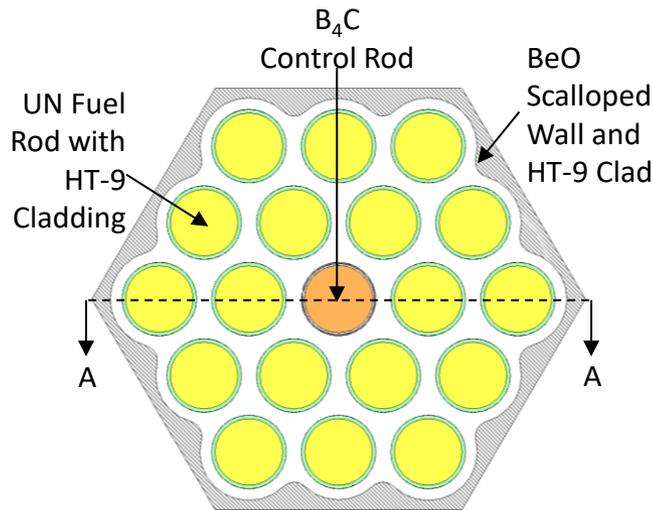
Design Highlights: VLLIM Microreactor

- 5.8 Full Power Year (FPY) lifetime at 10 MW_{th} and > 92 FPY at 1.0 MW_{th} .
- Cooled by natural circulation of in-vessel liquid sodium during nominal operation and after shutdown, aided by in-vessel chimney and compact Na-Na heat exchanger (HEX)

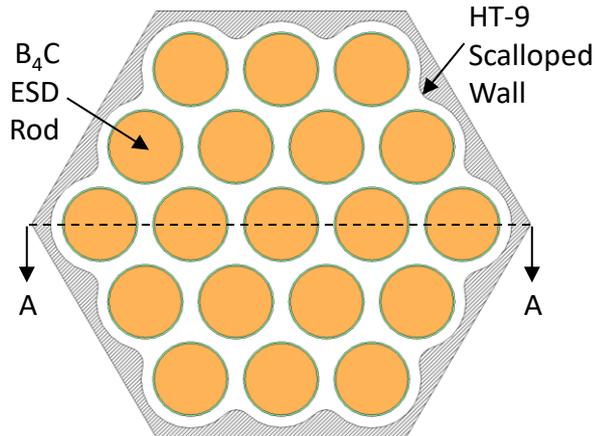


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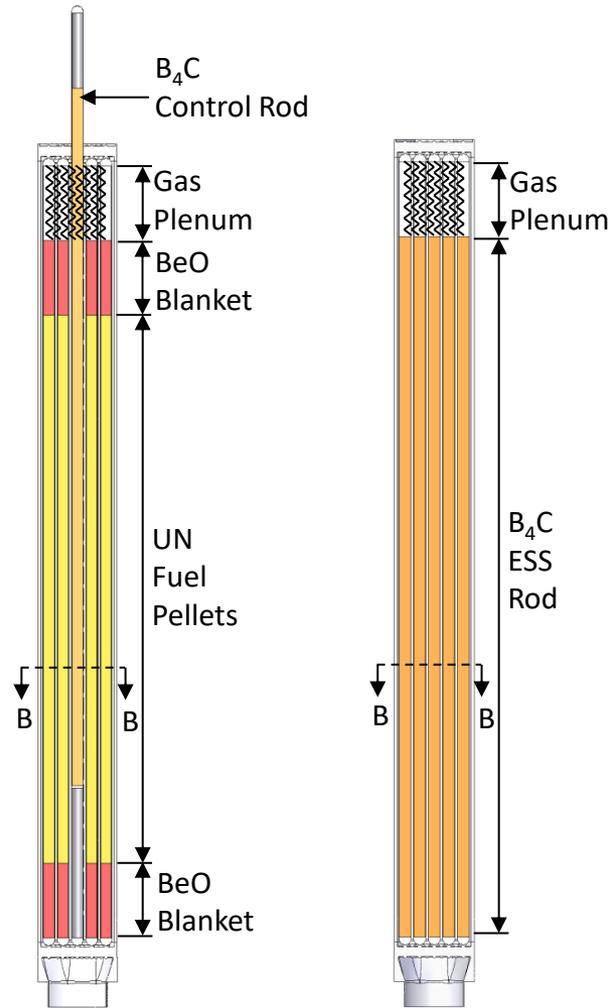
VSLIM: Control & Emergency Shutdown



(a) RC Cross-Section View B-B



(b) ESD Cross-Section View B-B



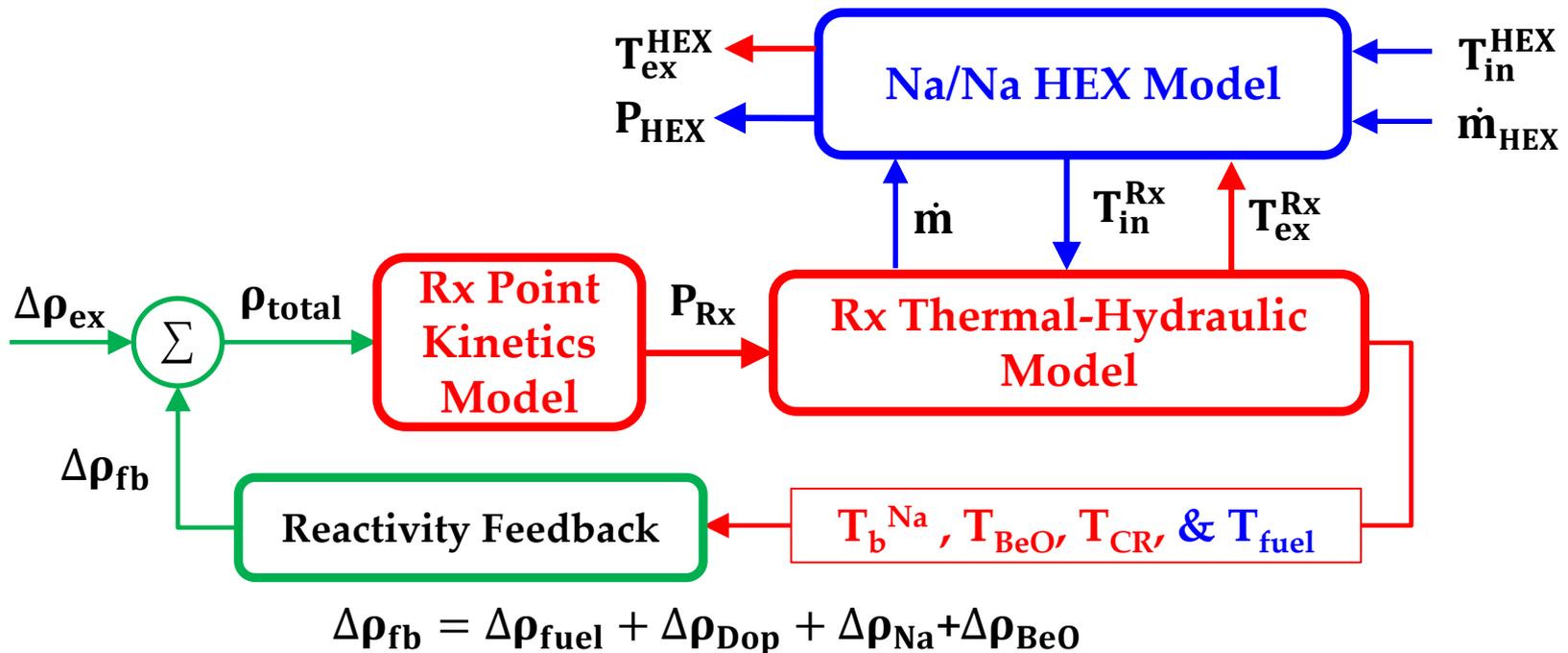
(c) RC Cross-Section View A-A

(d) ESD Cross-Section View A-A

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VSLIM: Simulink Transient Model

- **Couples 6-group point kinetics and Rx thermal-hydraulics models**
 - Solves steep point kinetics equations using robust exponential matrix technique using the 7th order Padé(3,3) function
 - Solution efficient, accurate and stable independent of timestep size
- **Point kinetics reactivity feedback**
 - Doppler broadening and thermal expansion of UN fuel, Na coolant, BeO shrouds, and B₄C control rods

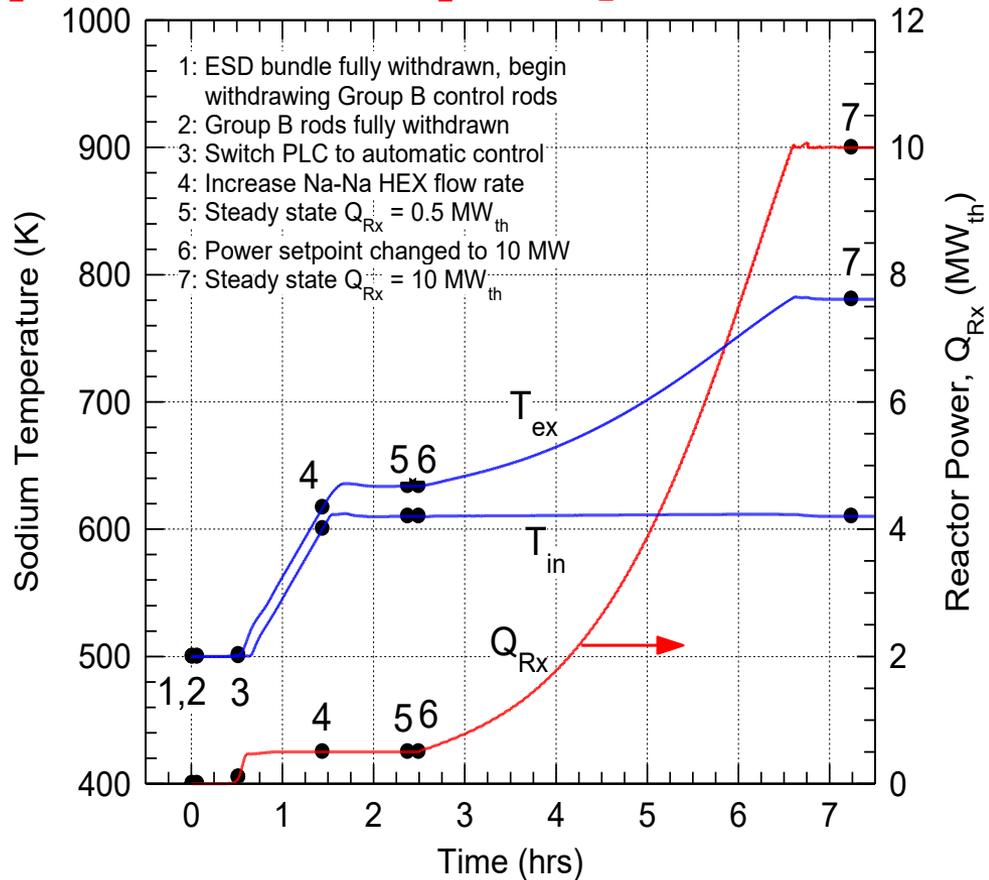
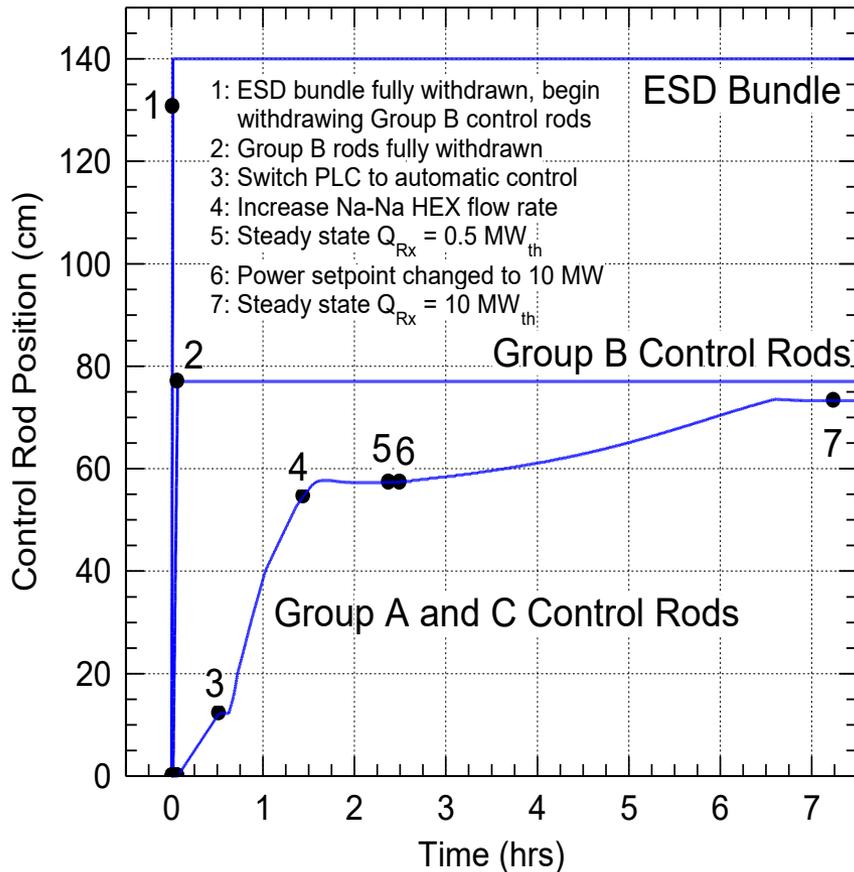


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Simulation Results: Startup Transient

VSLIM dynamic model simulates startup from subcritical state to nominal steady state operation at different thermal power levels (1.0 -10 MW(t))

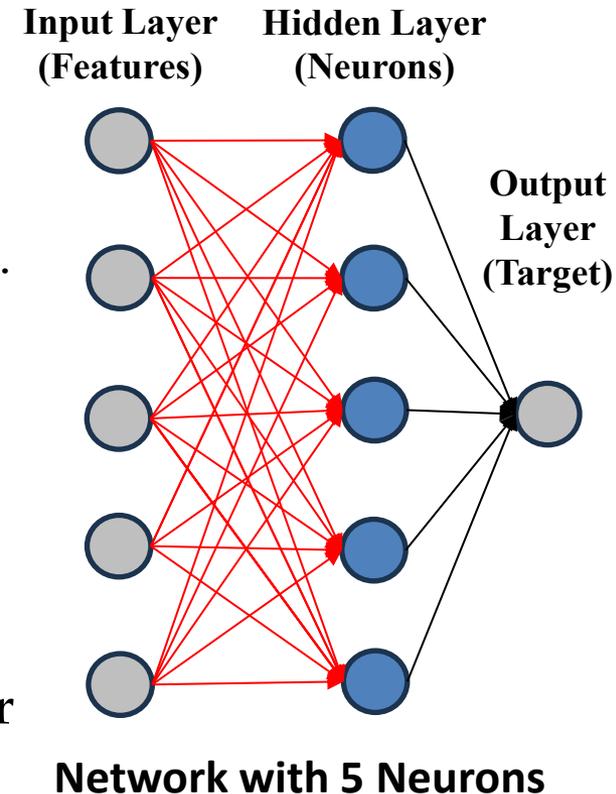
- First, Controller brings reactor steady state power to an Initial Setpoint P_1
- It subsequently increases reactor power to a final Setpoint P_2



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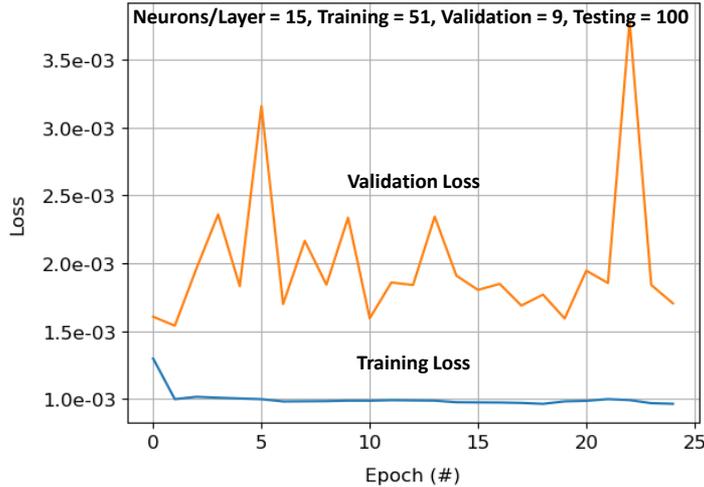
Neural Networks and ML Algorithm

- **Long Short-Term Memory (LSTM) algorithm**
 - Implements PyTorch library LSTM functions in Python code
- **Hyperparameters**
 - **5 Features:** Rx Power setpoints, **transient** Rx power, and liquid sodium flow rate and inlet and exit temperatures.
 - **One Target:** Group A & C control rods position.
 - **Neural Network:** single layer of 5, 10, and 15 neurons
 - **Learning Rate:** 0.001
 - **Optimizer:** AdamW with 0.1 weight decay
 - **Lookback window:** 20 (4 s)
- **Supplied ML Training data sets:**
 - **797** sets of simulated startup transient, with more than **956 million data points** for different Rx power set points, P_1 and P_2 :
 - P_1 : 0.5 - 9.75 MW
 - P_2 : 1.0 - 10.0 MW



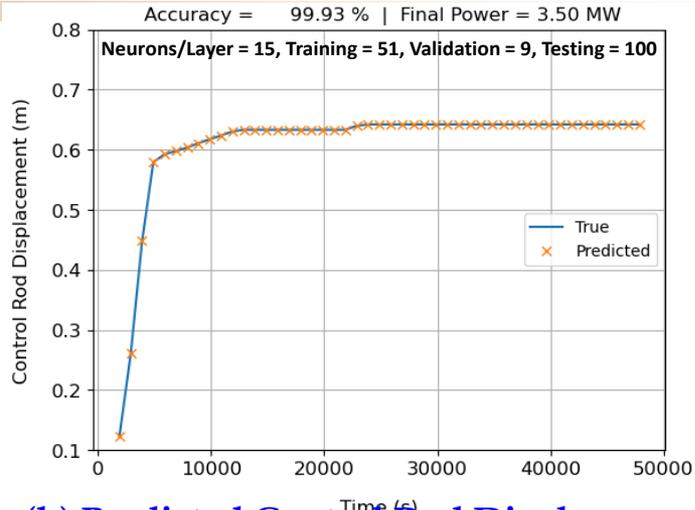
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Results: LSTM ML Training



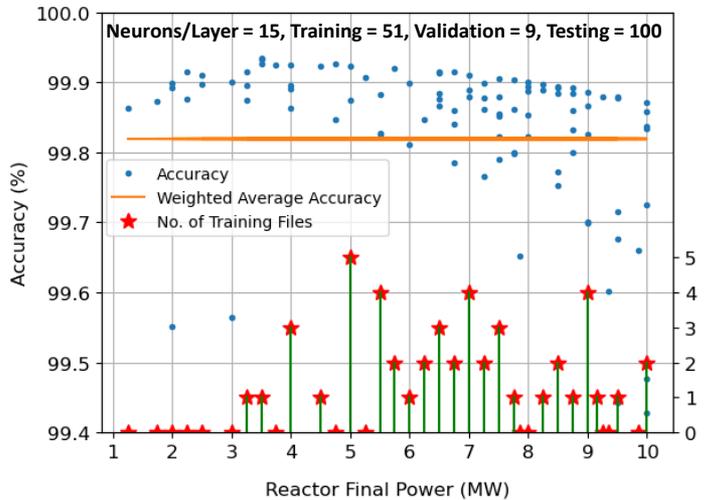
(a) Training Loss Convergence

Max = 99.93 % | Min = 99.43 % | Weighted Average Accuracy = 99.82 %

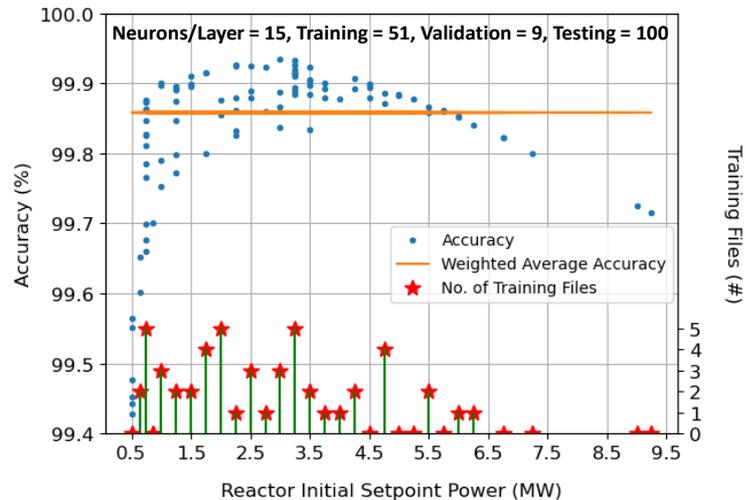


(b) Predicted Control Rod Displacement

Max = 99.93 % | Min = 99.43 % | Weighted Average Accuracy = 99.86 %



(c) Accuracy with Final Power P_2

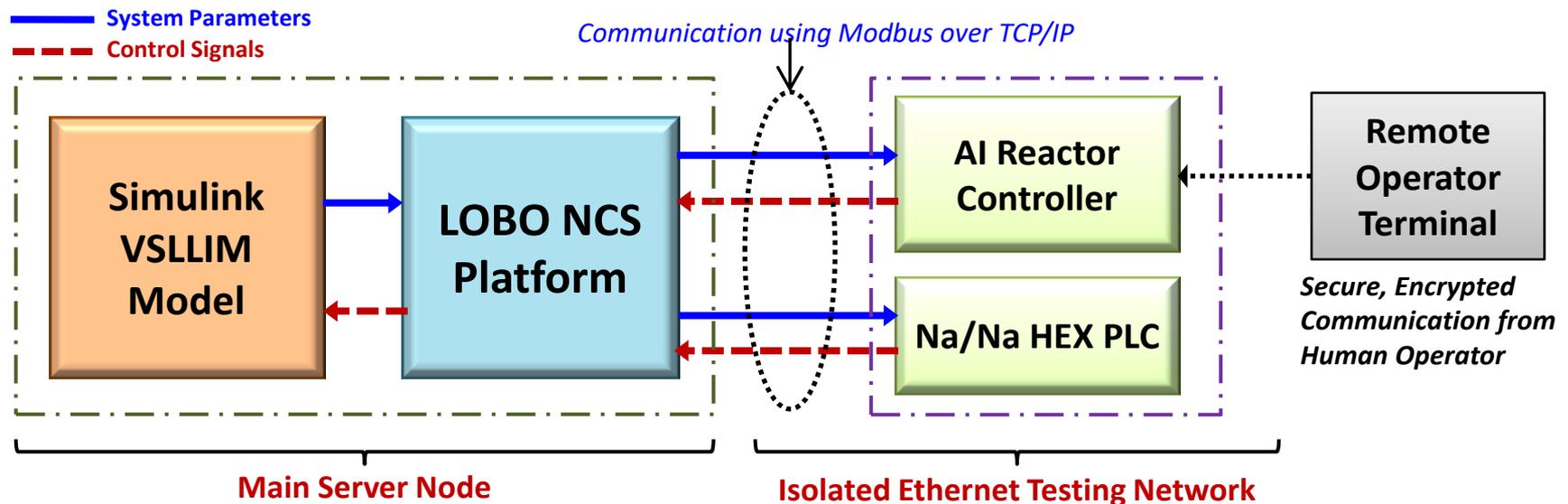


(d) Accuracy with Initial Power Setpoint P_1

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Ongoing & Future Work: Integrate AI Controller

- Use LOBO Nuclear Cybersecurity (NCS) Platform, Developed by UNM-ISNPS in collaboration with Sandia National Laboratories (SNL).
 - Versatile for testing advanced digital I&C systems and cybersecurity analysis.
 - Couples physical hardware and emulated controllers to real-time Simulink model.
- Integrate trained neural network into reactor control PLC
- Begin testing the AI controller coupled to VSLLIM Simulink model for startup simulation sequences



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The End

Happy to answer
your questions



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