

7<sup>th</sup> April 2022

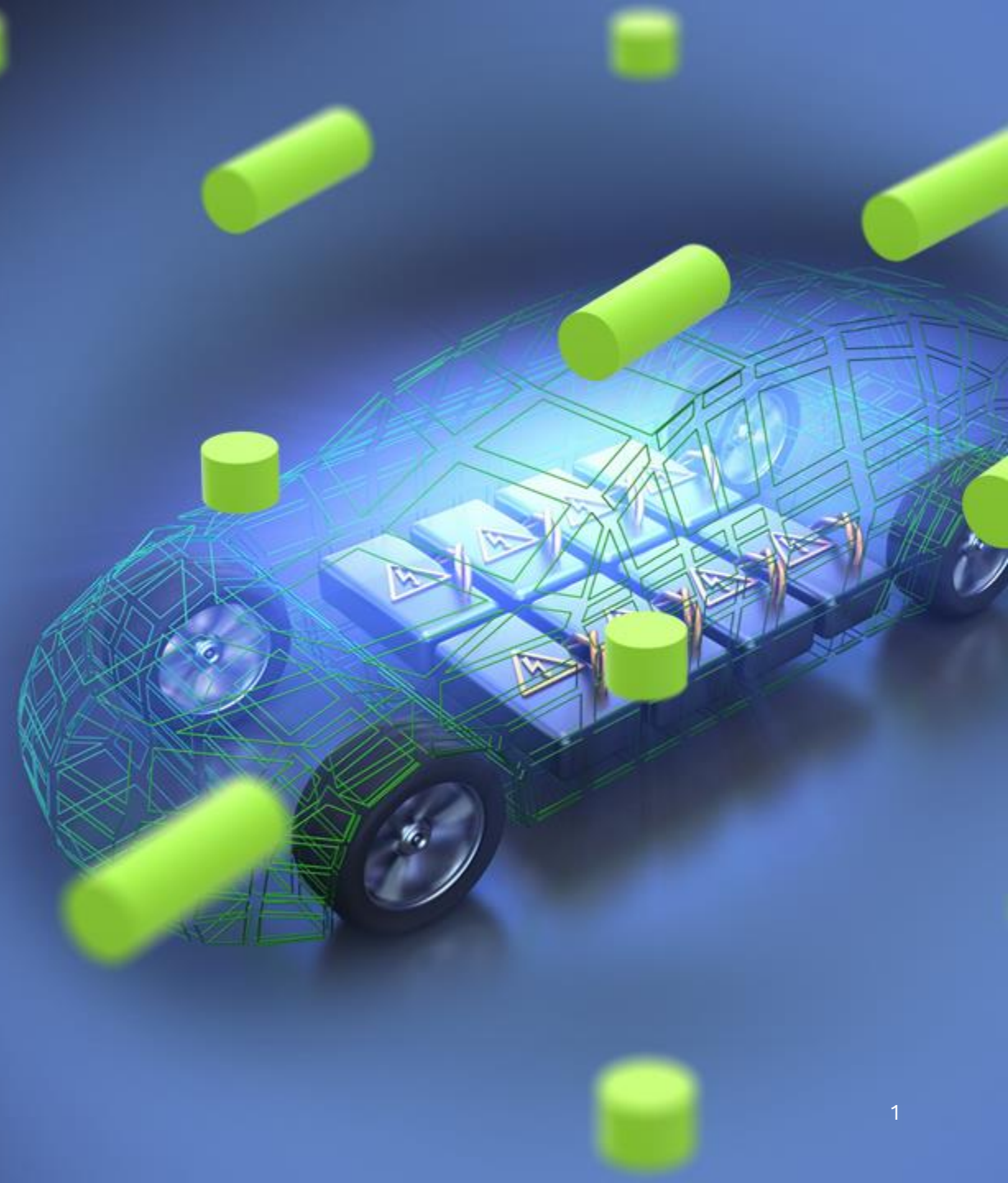
## Battery SOC and SOH Estimation using a Hybrid Machine Learning Approach

**Mahesh Ghivari**

AVP, Practice Offering  
Leader, Electric &  
Conventional Powertrain

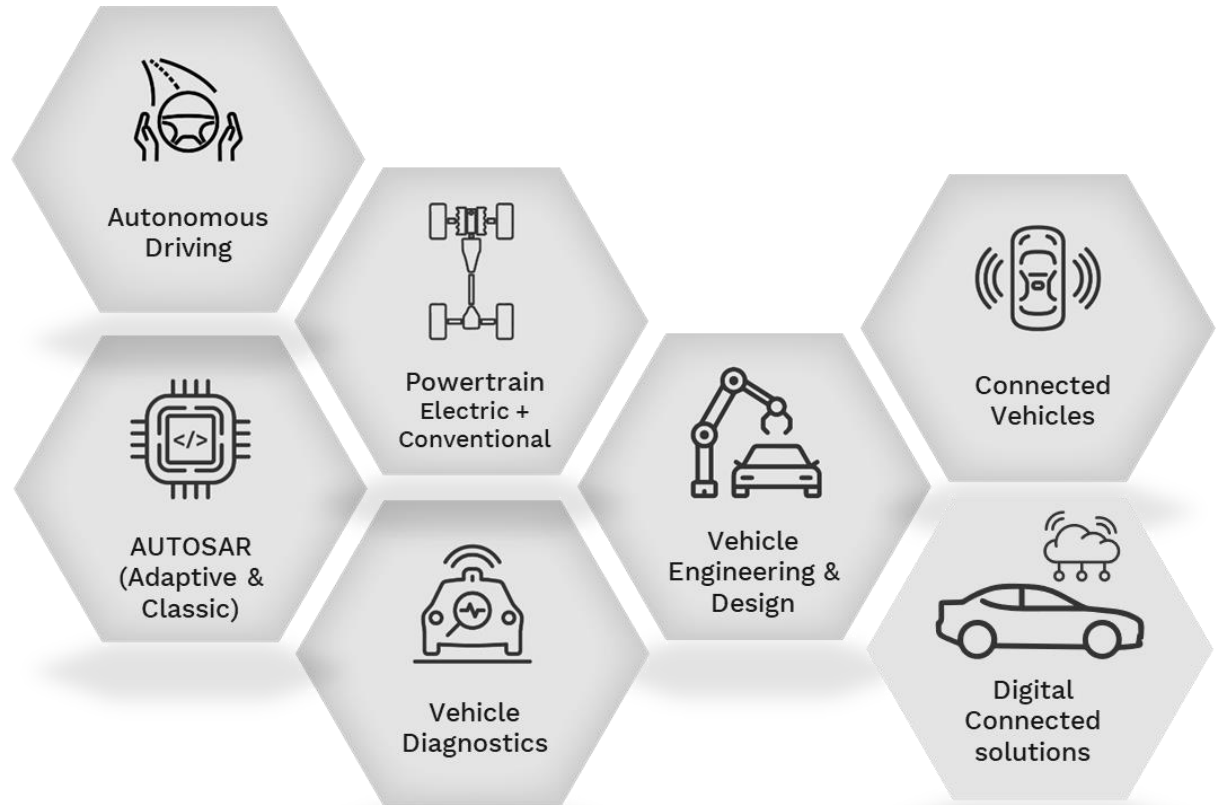
**Debango Chakraborty**

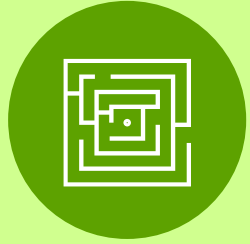
Subject Matter Expert - Battery  
Management System



# Who We Are

**Independent software integrator partner bringing scale and dependability** to build and integrate software features to accelerate the journey from prototype to production





## Challenges in conventional methods for estimating SOC & SOH

- Measurement Noise
- Integration Error
- Initial SOC calculation error
- Peukert's Coefficient
- Incorrect Battery Parameterization of battery models
- Best fit tuning challenge



## Proposed methodology

In the **hybrid approach**, SOC's **generalization** and **nonlinearity approximation** capability are significantly enhanced



# Machine learning based *Hybrid Approach*

## Influencing Factors of SOC



### Current Measurement

Sensor resolutions and inaccuracy



### Voltage Measurement

DC Offset Errors, AC noise



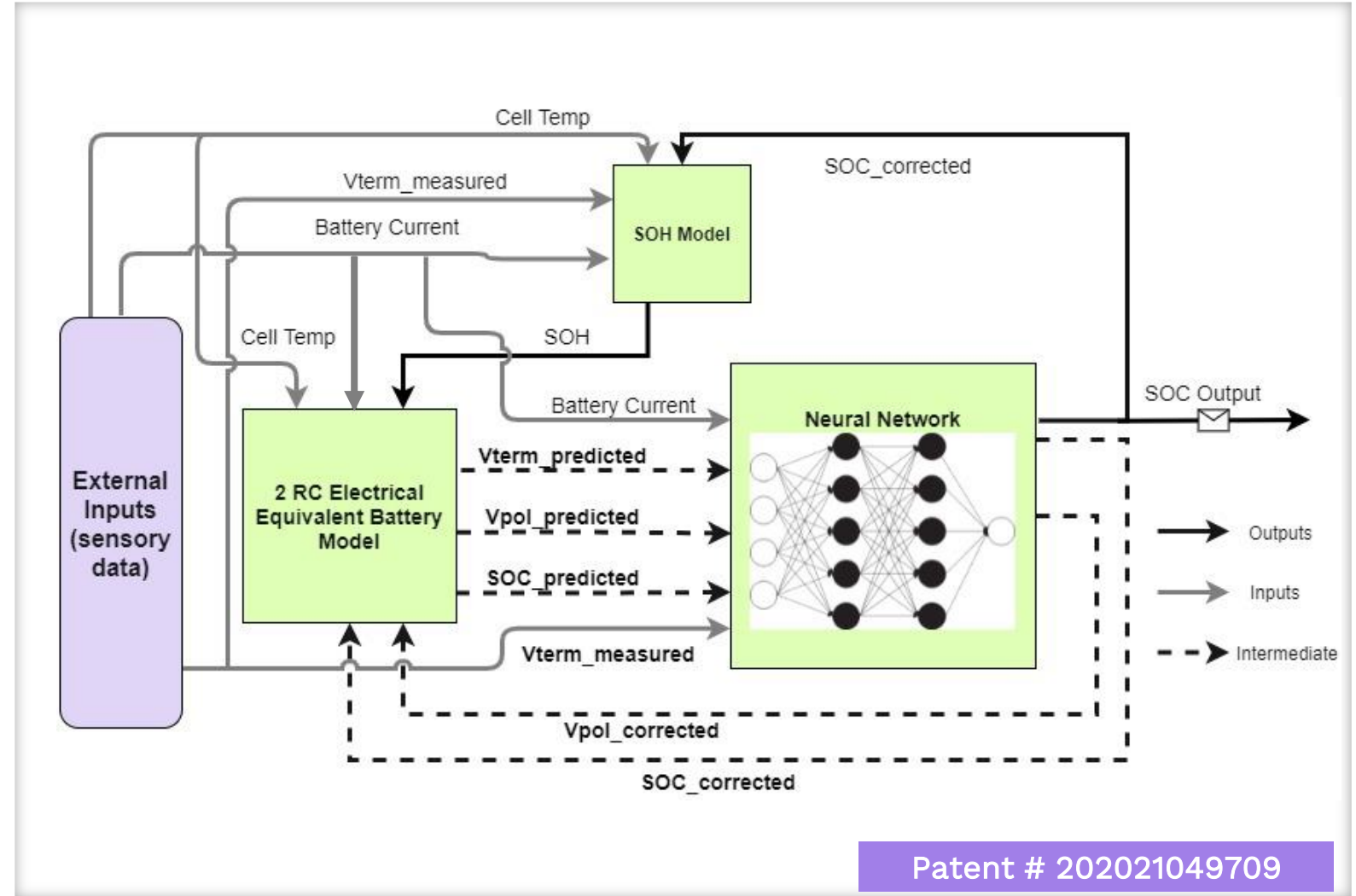
### Temperature Dependency

Incorrect Battery Characterization

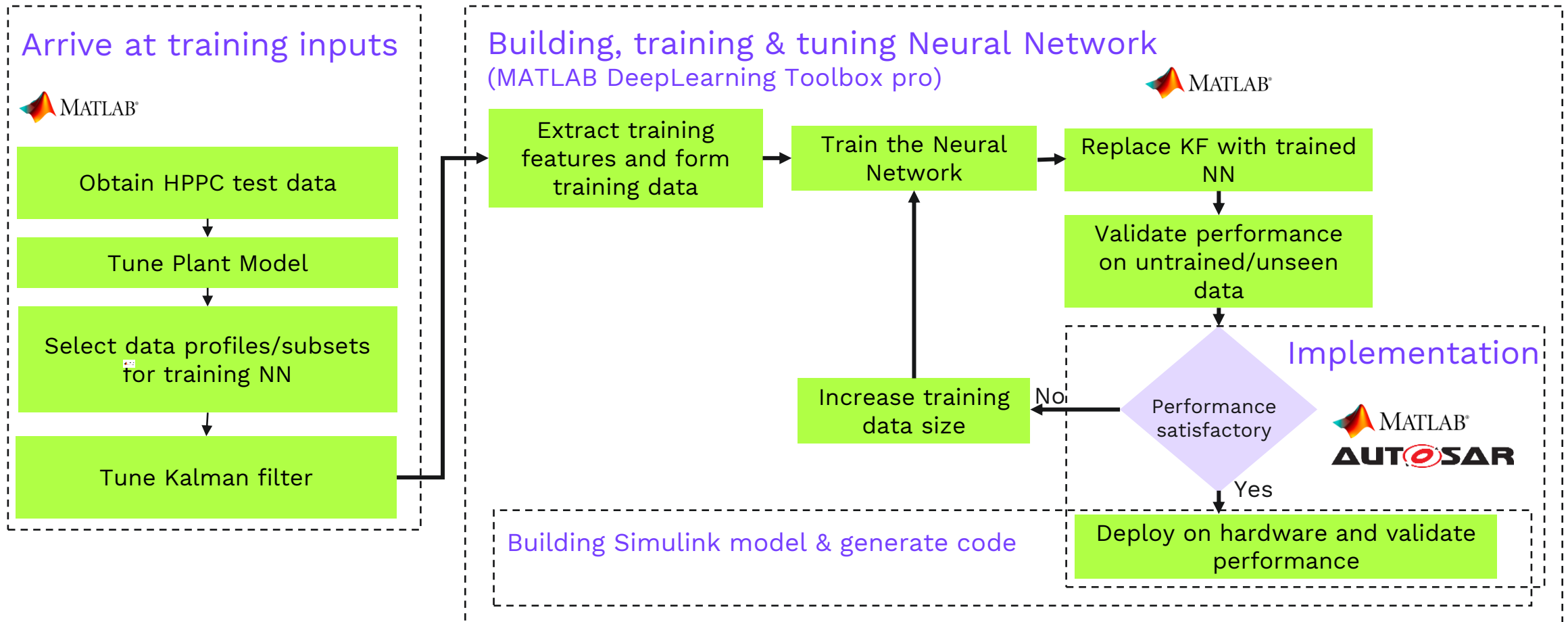


### Internal Resistance

Calculation inaccuracies



# Workflow of proposed hybrid SOC estimation approach



# Training Process for a Neural Network

## Training Stages

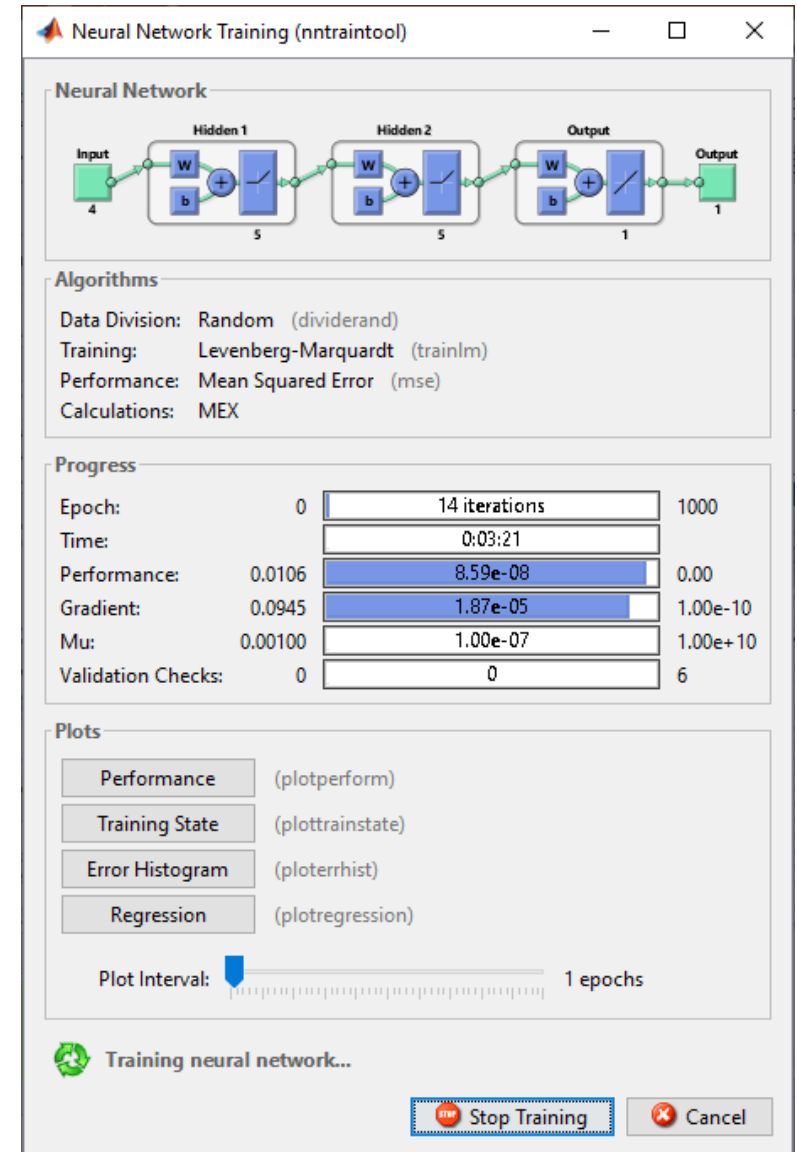
A model with a Kalman filter is used to collect data from two open-source runs

The collected data includes the features and the target variable

MATLAB's Deep Learning Toolbox is used to train a NN to map input features to the target output

To prevent overfitting, a portion of the data is held out for validation checks

Several networks are trained to find the one that generalizes well across all the unseen datasets



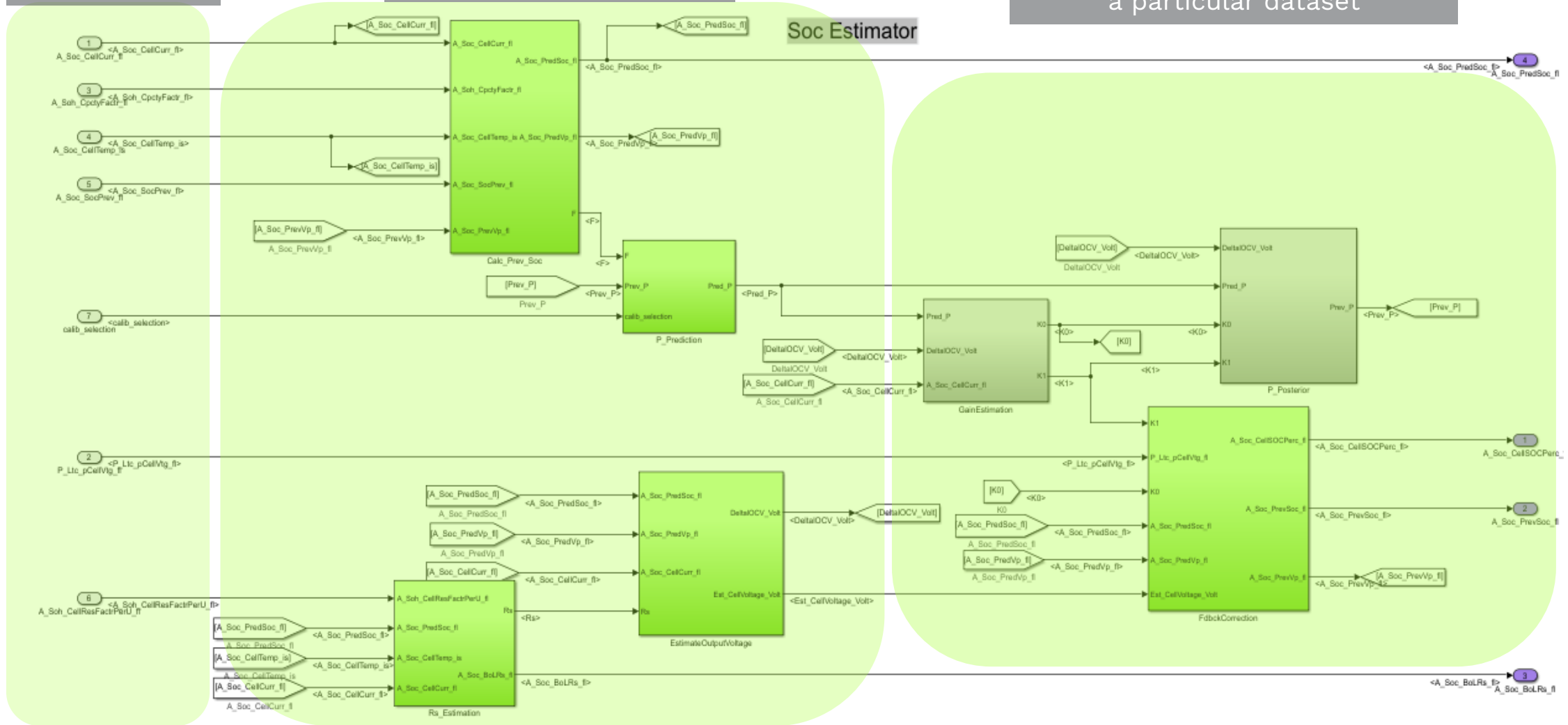


# Arrive at training inputs

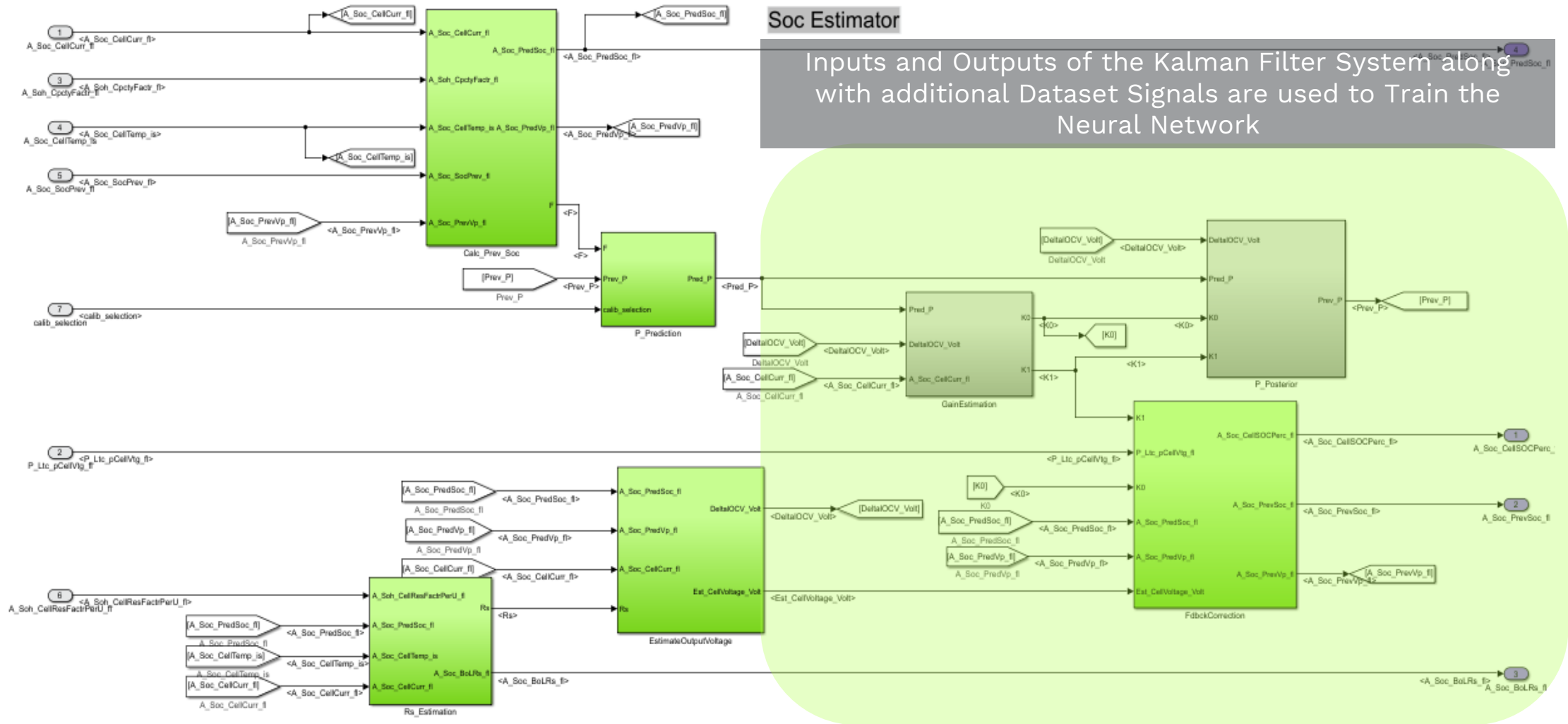
Dataset Inputs

Tuned Battery Plant Model

Kalman Filter Co-efficients tuned for best case accuracy on a particular dataset



# Arrive at training inputs



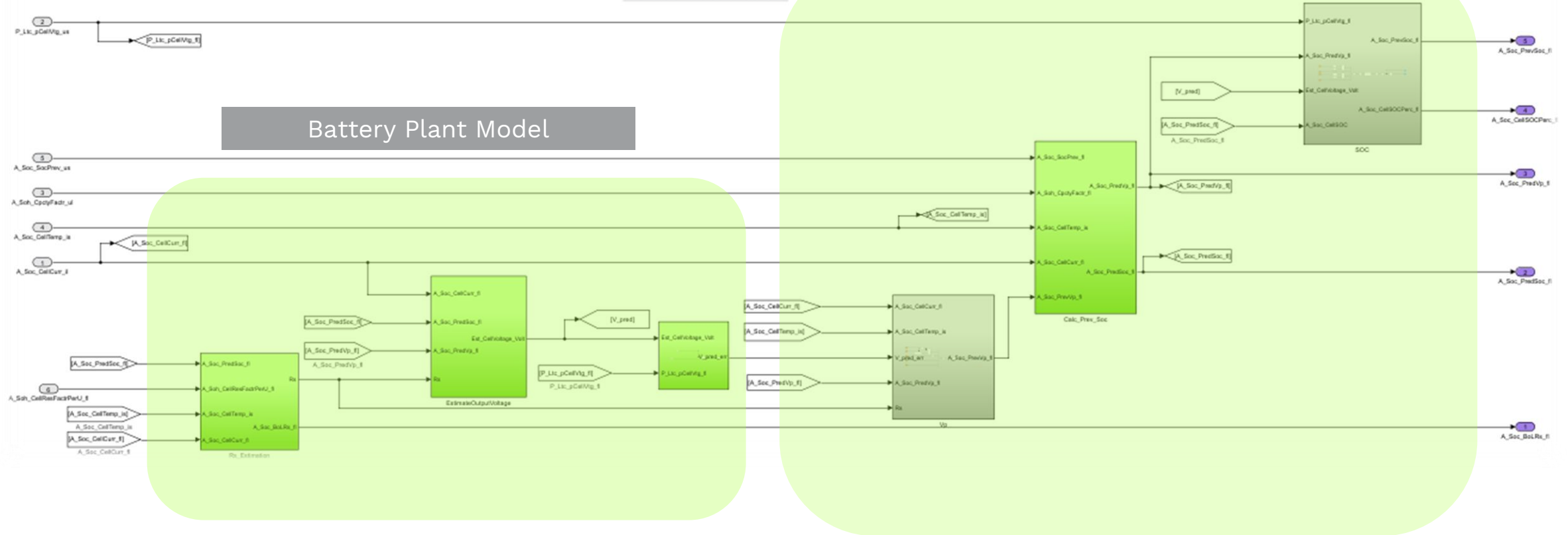


# Final Model design

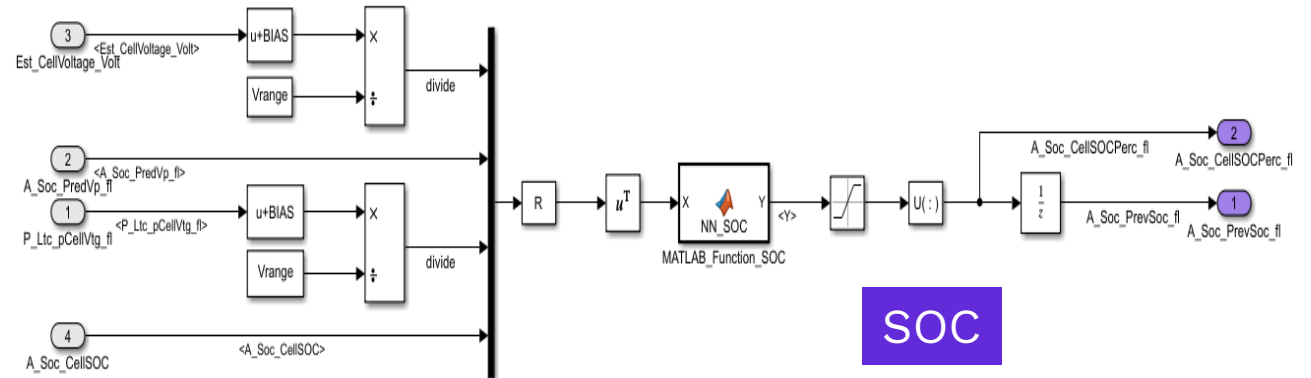
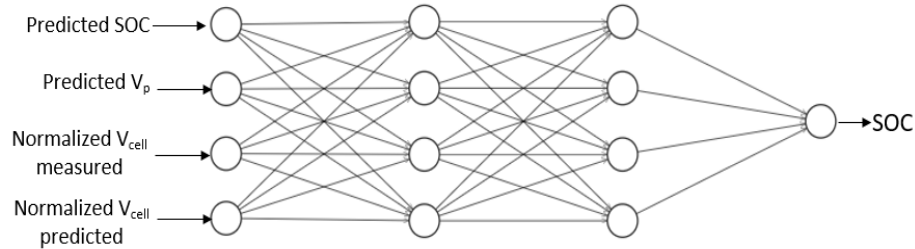
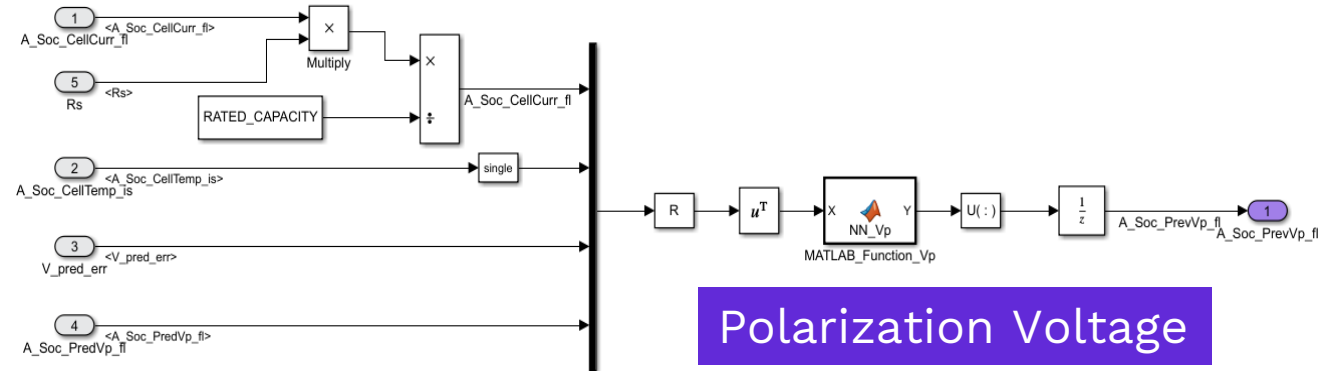
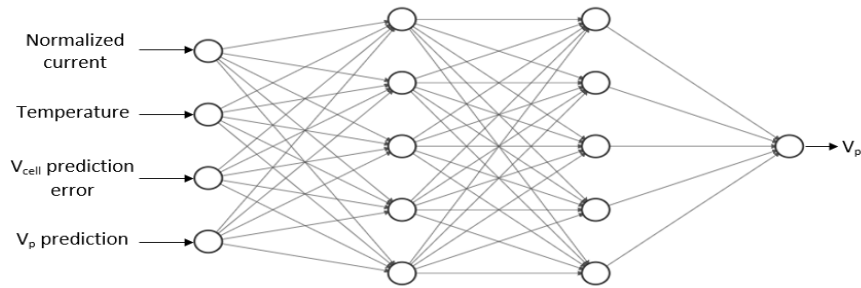
Neural Network Simulink Systems replace the Kalman Filter of previous design

SocEstimator

Battery Plant Model



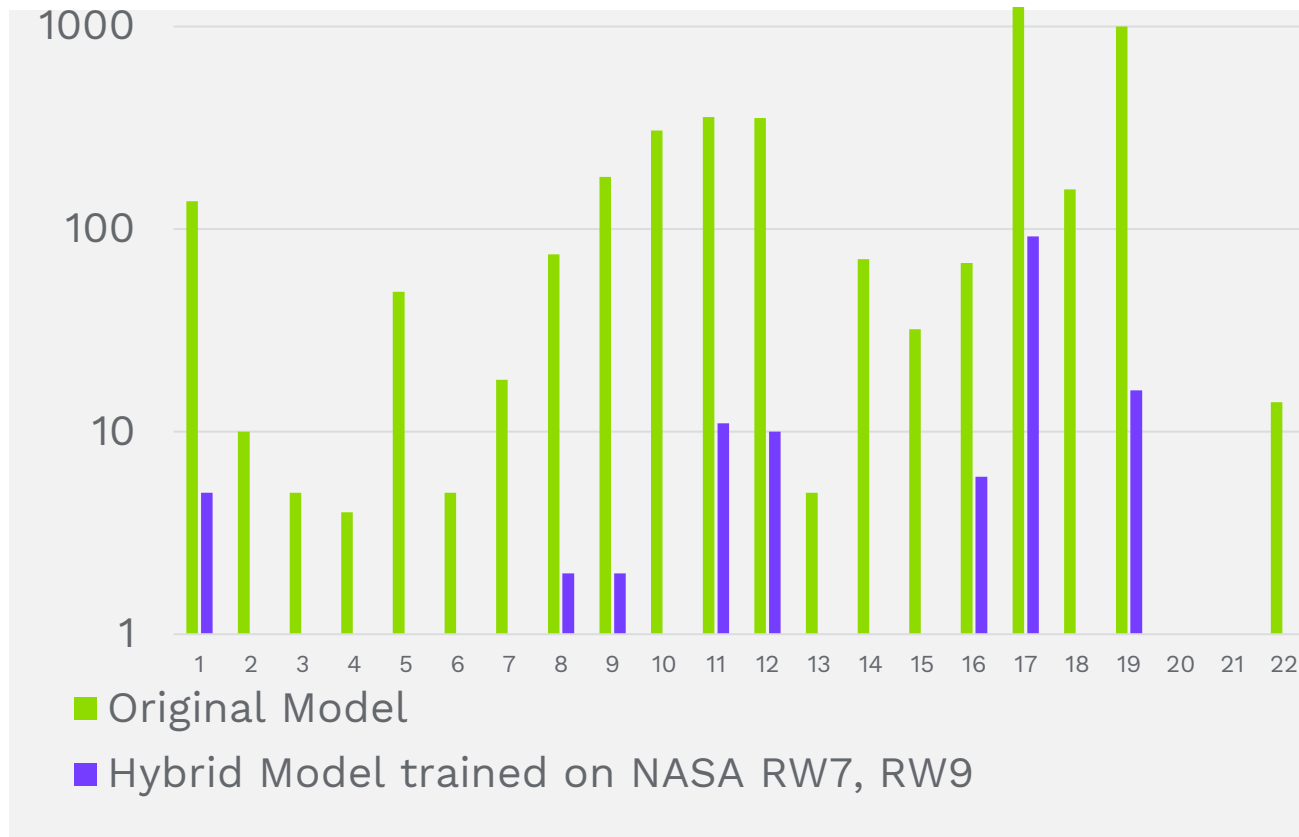
# Neural Network design



Simulink NN Model

# Model performance

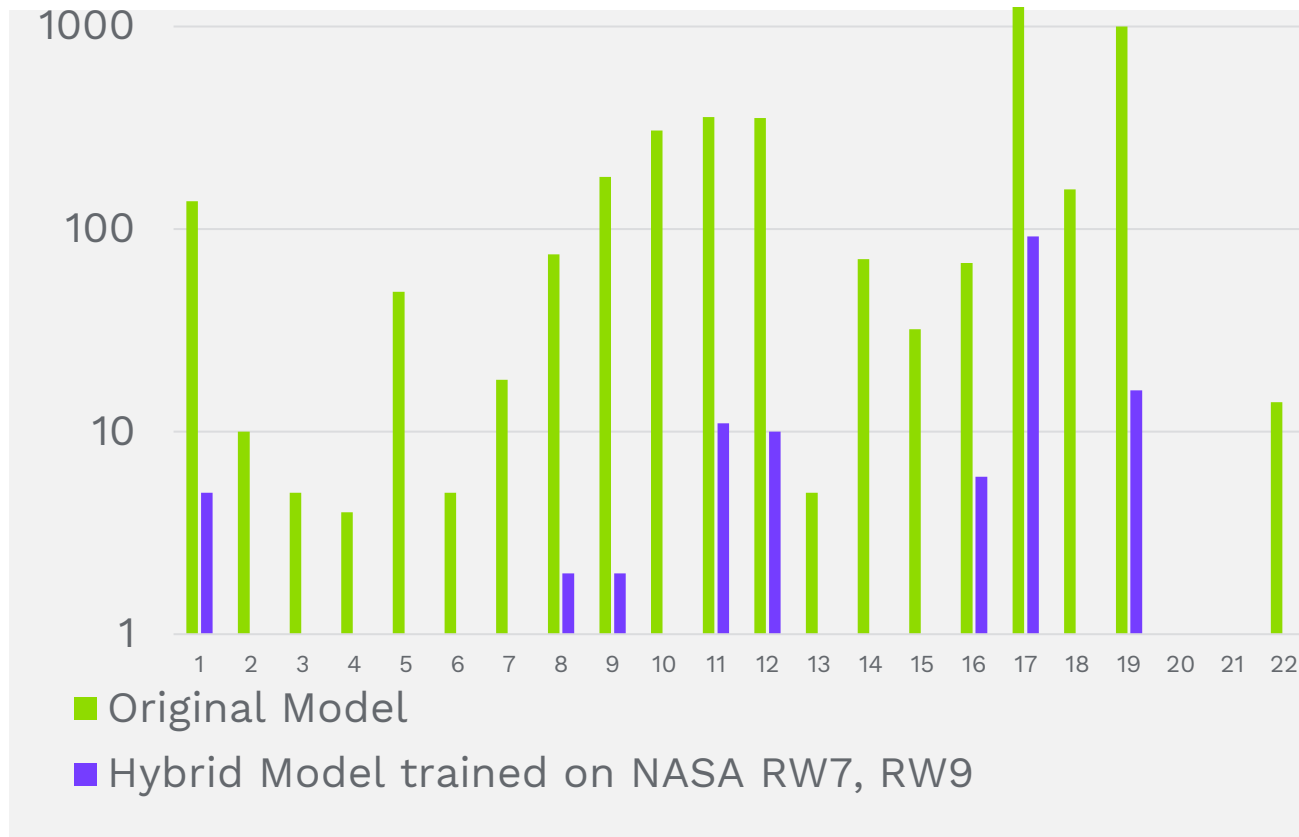
Number of error occurrences (log scale)



Error =  $\text{abs}(V_{\text{pred}} - V_{\text{meas}})$   
Error\_Occurrence =  $\text{focv}(\text{error}) > 5\%$

# Model performance

Number of error occurrences (log scale)



## ACG Performance using Simulink Embedded Coder

Memory Segment	Memory Sub-Segments	NN-SW Design Hosted on MPC5746R
RAM		515 Bytes
ROM	Code	8730 Bytes
	Constants	388 Bytes
Stack		623 Bytes

# Summary

- Application of AI/ML along with domain has consistently yielded the desired estimate of cell SOC at acceptable accuracy levels.
- Conventional methods were less complex and accurate, the increase in computational power and usage of powerful toolchain from MathWorks has encouraged to explore the complex techniques to enhance the algorithms
- The MathWorks environment provided the required computational and design toolboxes to seamlessly enable this workflow of designing and training the neural network, simulating in a closed-loop environment, and generating production-grade embedded code for deployment on hardware.
- Further, this study was mainly for mild hybrid applications; for fully battery electrical, the computational aspect becomes more challenging. Further, these algorithms of SOC and SOH can be extended to integrate with a vehicle control unit and thereby improve the electrical range and achieve better fuel economy. And all this development is possible in a shorter time due to powerful simulation toolchains from Mathworks.

# Thank You

Your partner for software development and integration for electrification



**12**

OEMs & Tier1s rely on KPIT for HEV/EV technologies including

**30+**

Programs executed on Hybrid and Electric Vehicles

**Domains**

BMS, Charger, Inverter, DC-DC Converter, VCU – BEV, HEV

Software integrator partner, managing vehicle programs with multiple clients



Production ready software components for EVCC, CCS, GB/T, CHAdeMO and BMS



Global delivery model (Germany, US, India, Thailand)



End-to-end validation partner for xEVs - SIL, MIL and HIL testing | Automation strategy

