



Accelerating the new hydrogen economy in the Midlands

# SOFC System Longevity: Advanced Modelling, Optimisation, and Machine Learning Predictions

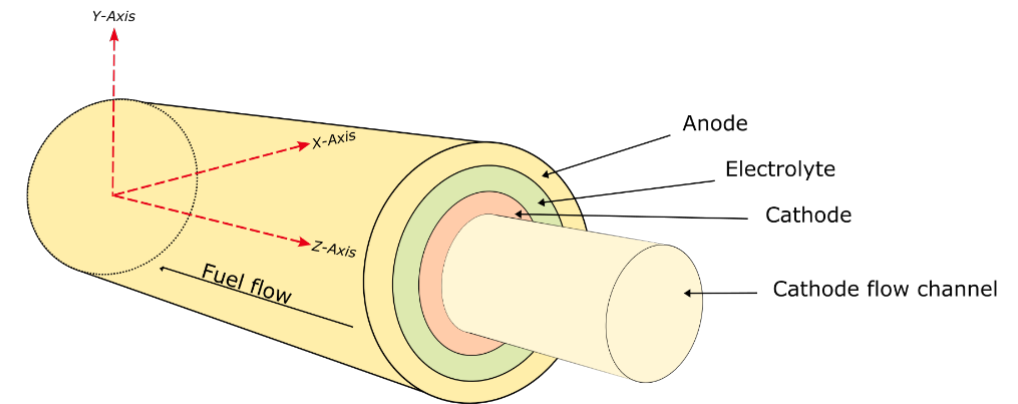
Clyde-Theodore Batista  
28<sup>th</sup> April 2025

Aston University



# Introduction: What are SOFCs?

- Solid Oxide Fuel Cells (SOFCs) convert chemical energy directly into electrical energy through electrochemical reactions.
- Offer high electrical efficiencies (typically range from 40% to 60%), reaching up to 80% in CHP applications.
- Primarily utilise hydrogen as fuel but can also operate on various hydrogen carriers with appropriate reforming and impurity removal strategies.
- Commonly applied in off-grid power generation, Combined Heat and Power (CHP) systems, and emergency back-up power.



# RESEARCH MOTIVATION

## Challenges:

- Durability issues: thermal mismatch, multifaceted degradation, local instabilities.
- Lack of large, high-quality experimental datasets.
- Difficulty replicating real-world degradation in system models.

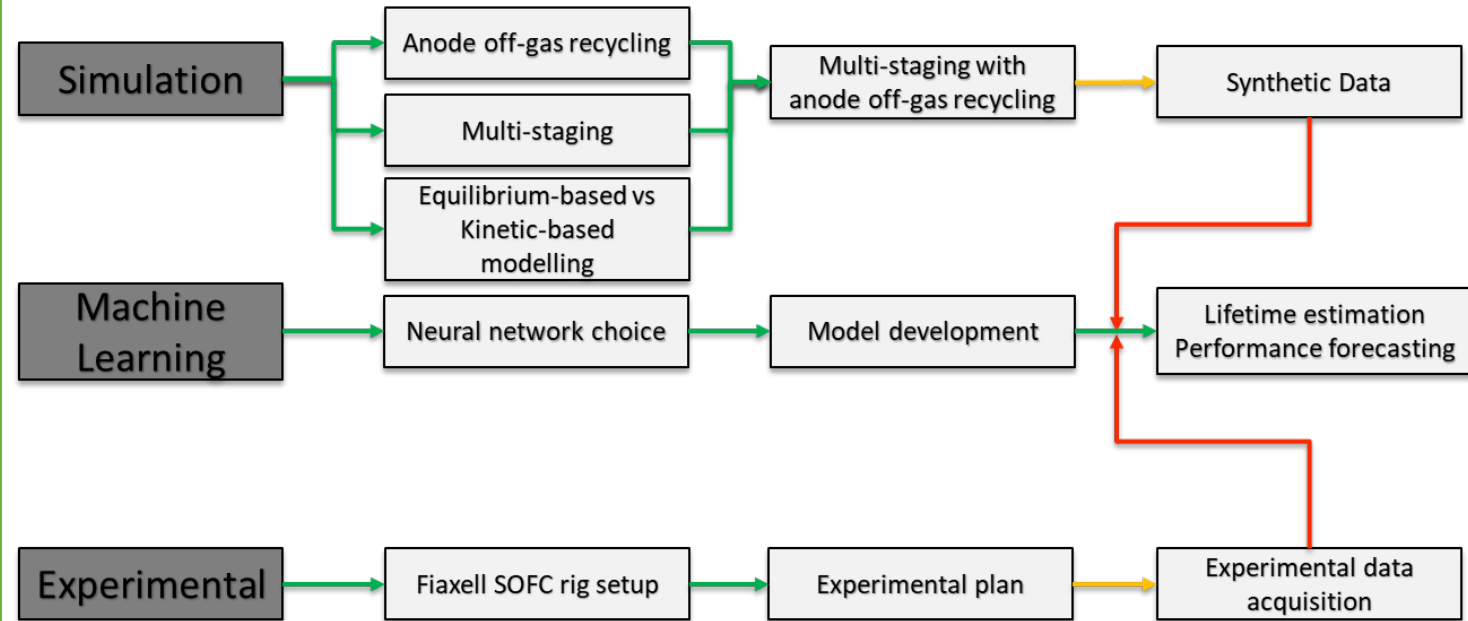
## Motivation:

- Commercialisation hindered by durability and lifetime uncertainties.
- End-of-life and second-life pathways poorly defined.
- Insufficient data limits accurate lifetime prediction and maintenance planning.

# RESEARCH AIMS

- **Optimise system configurations** (multi-staging, AOGR) to enhance fuel utilisation and **extend system lifetime**.
- Build and validate a machine learning framework (**BiLSTM**) for predicting voltage degradation and Remaining Useful Life (RUL).
- Commission experimental equipment and establish baseline performance characteristics of SOFCs.
- Evaluate initial SOFC durability to inform and define future experimental testing plan.
- Propose an **integrated methodology** combining experimental, simulation, and machine learning approaches for future SOFC lifetime prediction.

# RESEARCH OVERVIEW



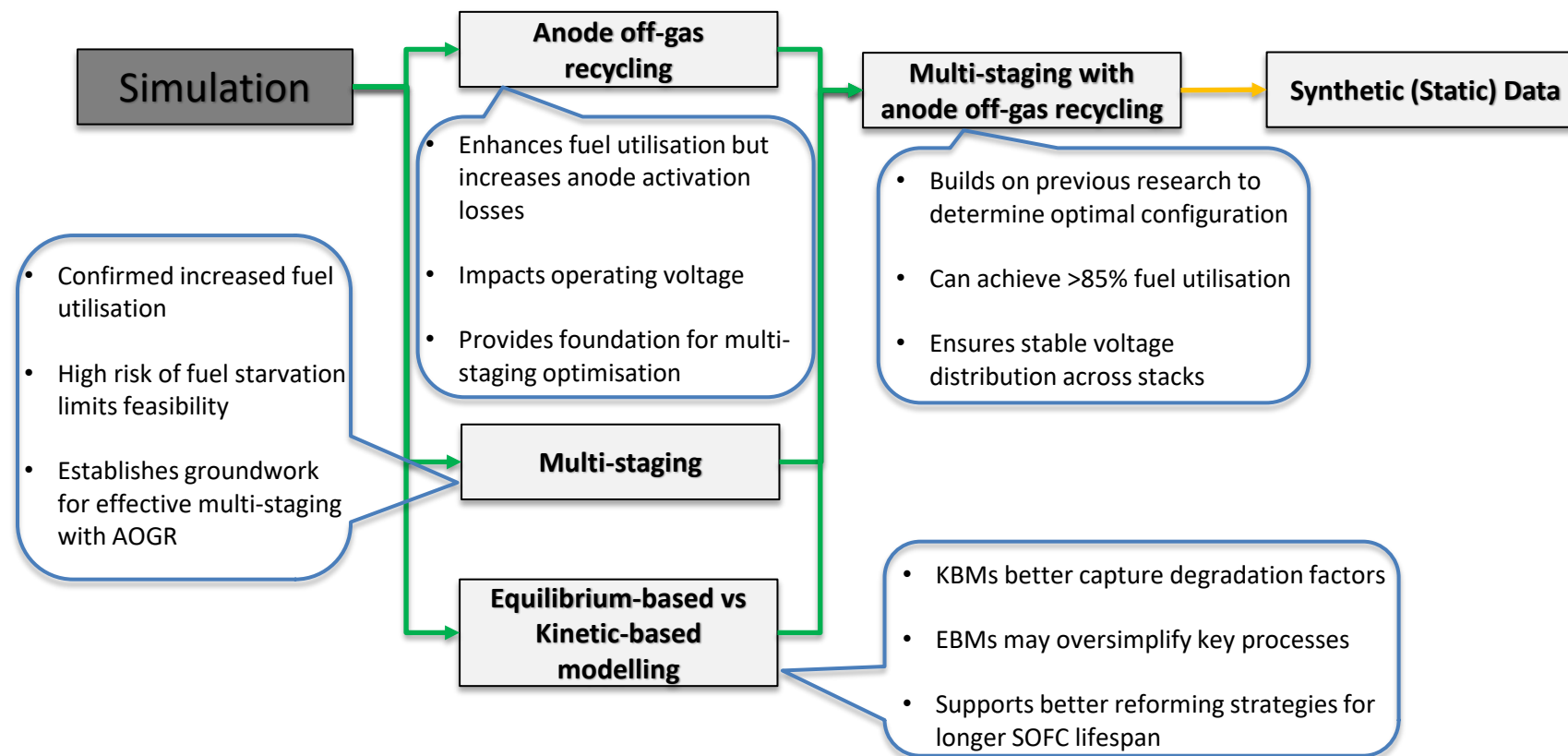
# Aspen Simulation: Base Model

- Electrochemical parameters were calculated using **semi-empirical** relations from Zhang et al.
- The base design is based on the Siemens-Westinghouse SOFC platform.
- All subsequent system configurations were developed by adapting this model.

| Variable/Parameter                     | W. Zhang et al., 2005 | W. Doherty et al., 2010 | T. R. Tanim, 2012 | This work    | % Deviation  |
|--|-----------------------|-------------------------|-------------------|--------------|--------------|
| Voltage, (V)                           | 0.70                  | 0.68                    | 0.69              | <b>0.69</b>  | <b>0</b>     |
| Current density, (mA/cm <sup>2</sup> ) | 178.0                 | 182.9                   | 182.0             | <b>180.7</b> | <b>-0.15</b> |
| Pre-reforming temperature, (°C)        | 536.0                 | 535.1                   | 535.0             | 535.1        | -0.05        |
| Cathode inlet temperature, (°C)        | 821.3                 | 823.7                   | 826.0             | 826.0        | 0.28         |
| Stack exhaust temperature, (°C)        | 834.0                 | 833.7                   | 836.0             | 793.9        | -4.87        |
| Anode outlet composition               |                       |                         |                   |              |              |
| H <sub>2</sub> O                       | 50.9%                 | 50.9%                   | 50.9%             | 50.9%        | 0.00         |
| CO <sub>2</sub>                        | 24.9%                 | 24.9%                   | 24.9%             | 24.9%        | 0.00         |
| H <sub>2</sub>                         | 11.6%                 | 11.6%                   | 11.6%             | 11.6%        | 0.00         |
| CO                                     | 7.4%                  | 7.4%                    | 7.4%              | 7.4%         | 0.00         |
| N <sub>2</sub>                         | 5.1%                  | 5.1%                    | 5.1%              | 5.1%         | 0.00         |
| Stack exhaust composition              |                       |                         |                   |              |              |
| N <sub>2</sub>                         | 77.3%                 | 77.3%                   | 77.2%             | 77.3%        | 0.04         |
| O <sub>2</sub>                         | 15.9%                 | 15.9%                   | 15.7%             | 15.9%        | 0.42         |
| H <sub>2</sub> O                       | 4.5%                  | 4.5%                    | 4.7%              | 4.5%         | -1.46        |
| CO <sub>2</sub>                        | 2.3%                  | 2.3%                    | 2.4%              | 2.3%         | -1.43        |
| Gross electrical efficiency (LHV), (%) | 52.0%                 | 51.3%                   | 51.7%             | <b>52.0%</b> | <b>0.65</b>  |

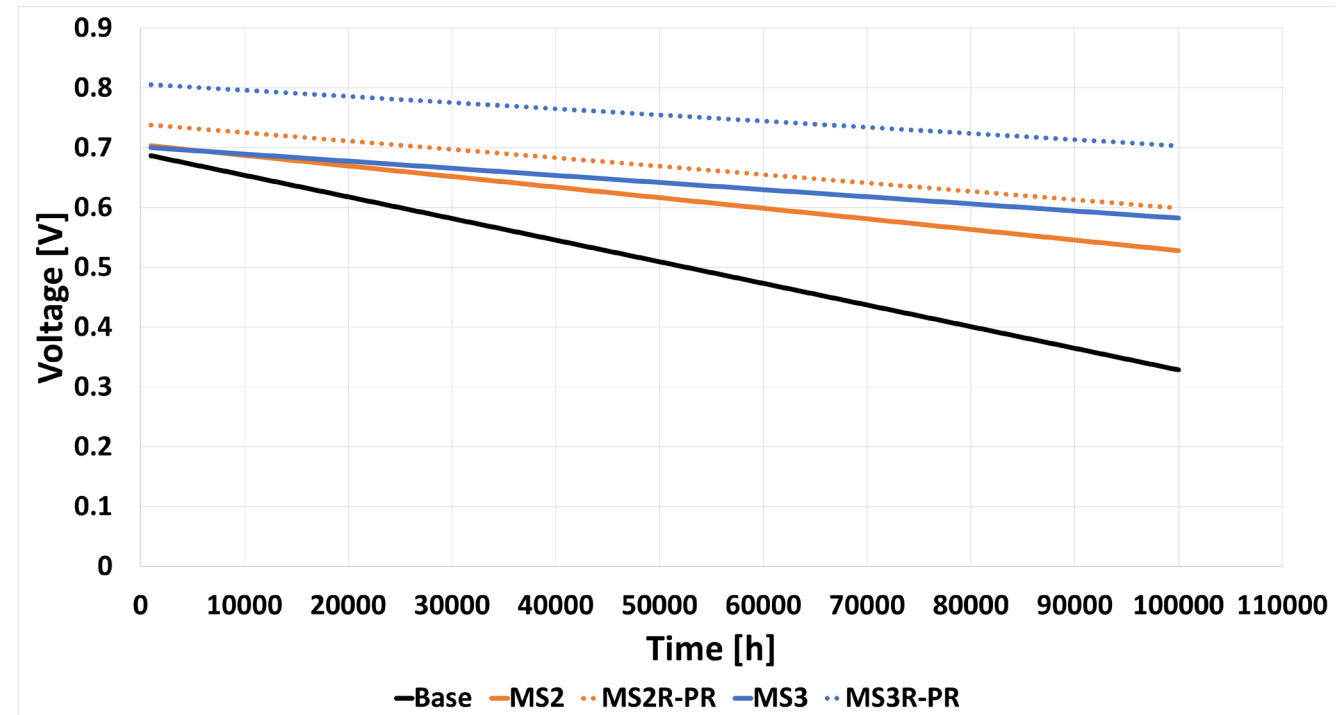
Zhang, W., Chan, S.H., and Li, J., 2005. Simulation of a tubular solid oxide fuel cell stack using AspenPlus™ unit operation models. *Energy Conversion and Management*, 46(2), pp.181–196.  
Doherty, W., Reynolds, A., and Kennedy, D., 2010. Computer simulation of a biomass gasification-solid oxide fuel cell power system using Aspen Plus. *Energy*, 35(12), pp.4545–4555.  
Tanim, T.R., 2012. Modelling of a 5kW Solid Oxide Fuel Cell Based Auxiliary Power Unit Operating on JP-8 Fuel.

# System Configuration: Optimisation



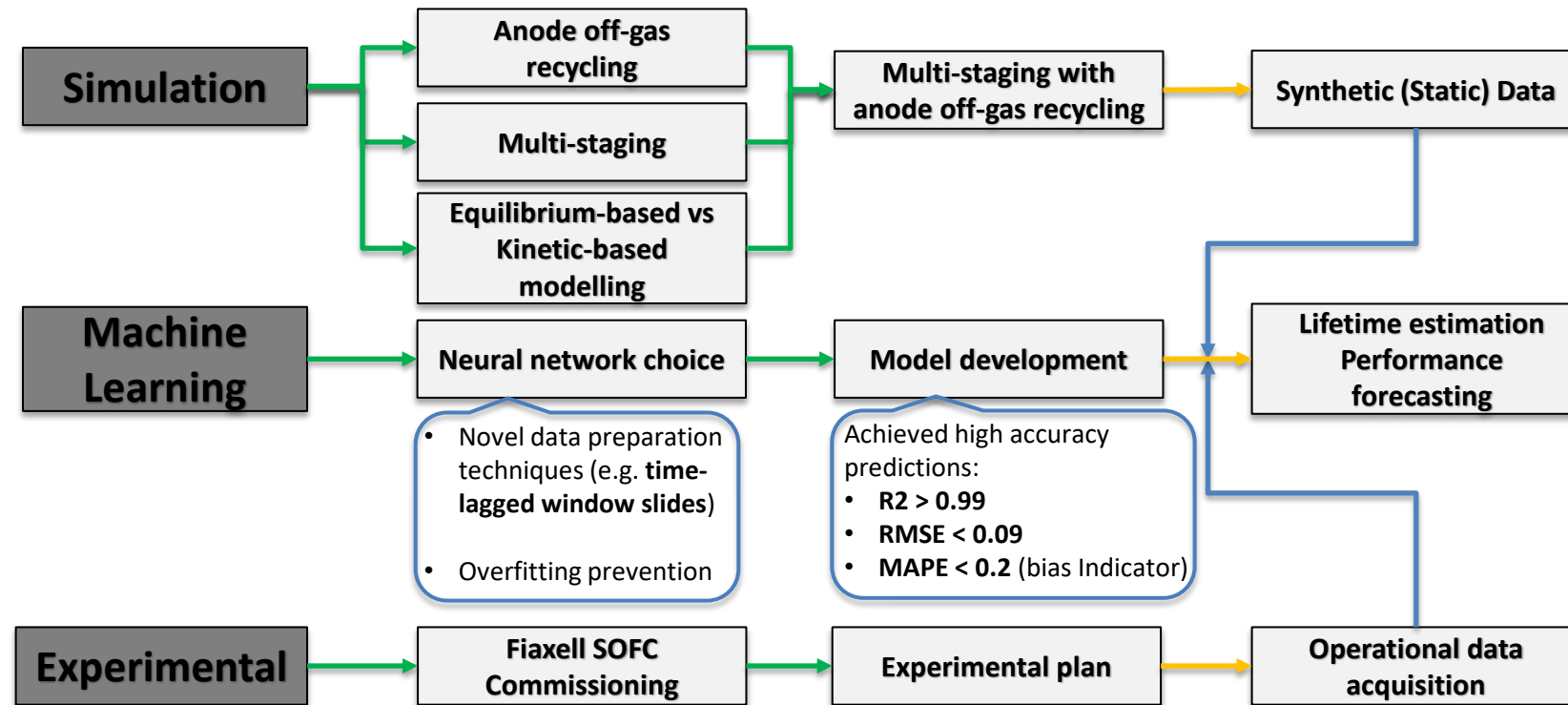
# System Configuration: ASR RUL

|         |         | Voltage | Current Density | Efficiency | Fuel Utilisation |
|---------|---------|---------|-----------------|------------|------------------|
| Base    | System  | 0.69    | 180.7           | 52%        | 0.85             |
| MS2     | Stack 1 | 0.77    | 81.3            | 24%        | 0.32             |
|         | Stack 2 | 0.71    | 88.5            | 36%        | 0.49             |
|         | System  | 1.47    | 84.9            | 52%        | 0.65             |
| MS2R-PR | Stack 1 | 0.77    | 94.9            | 36%        | 0.40             |
|         | Stack 2 | 0.74    | 70.4            | 43%        | 0.57             |
|         | System  | 1.51    | 82.6            | 62%        | 0.85             |
| MS3     | Stack 1 | 0.79    | 52.8            | 18%        | 0.32             |
|         | Stack 2 | 0.77    | 54.1            | 26%        | 0.32             |
|         | Stack 3 | 0.70    | 59.4            | 39%        | 0.46             |
|         | System  | 2.26    | 55.4            | 54%        | 0.75             |
| MS3R-PR | Stack 1 | 0.82    | 50.8            | 22%        | 0.32             |
|         | Stack 2 | 0.84    | 49.6            | 33%        | 0.32             |
|         | Stack 3 | 0.81    | 51.6            | 48%        | 0.51             |
|         | System  | 2.47    | 50.7            | 67%        | 0.85             |

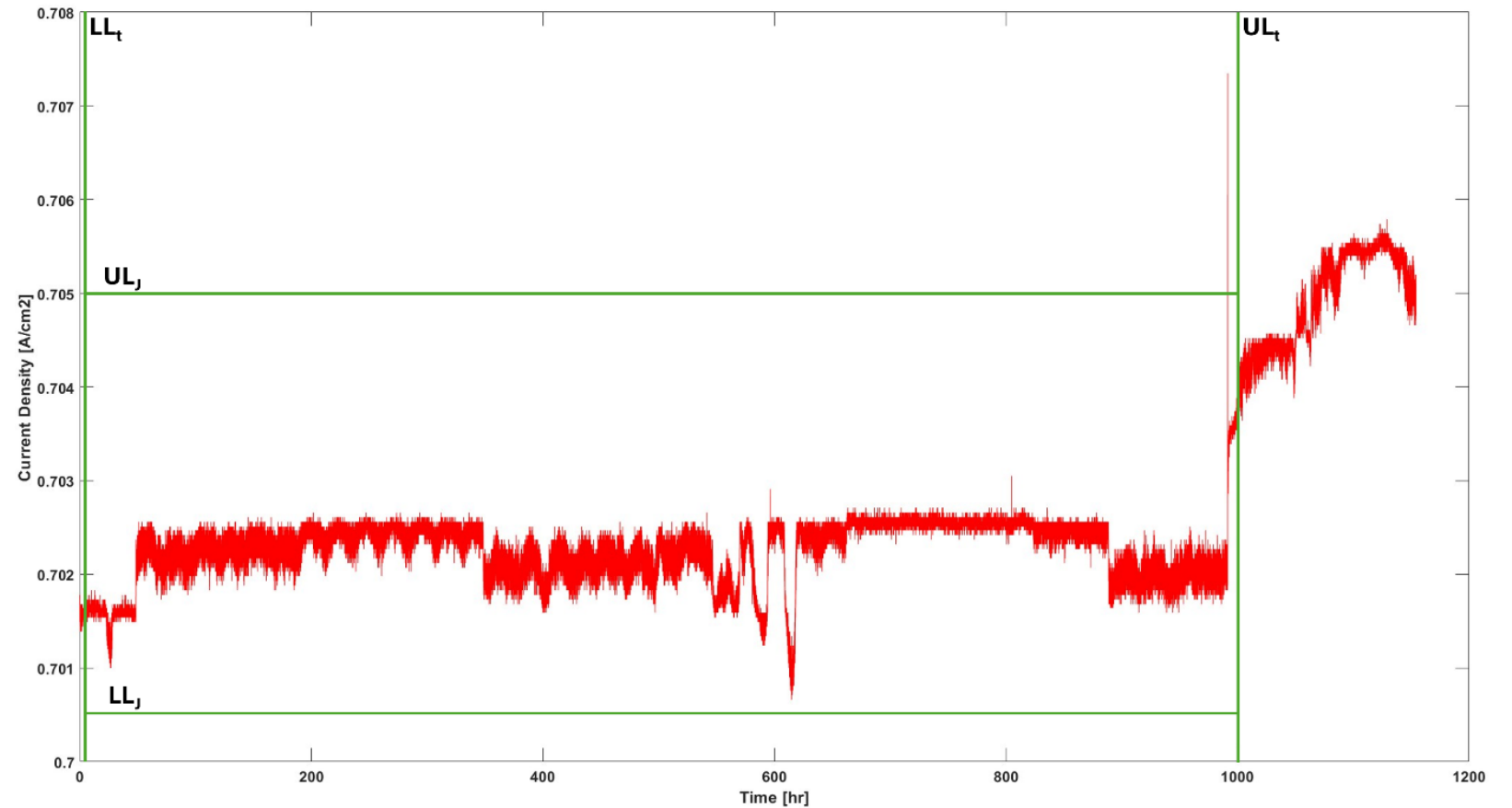
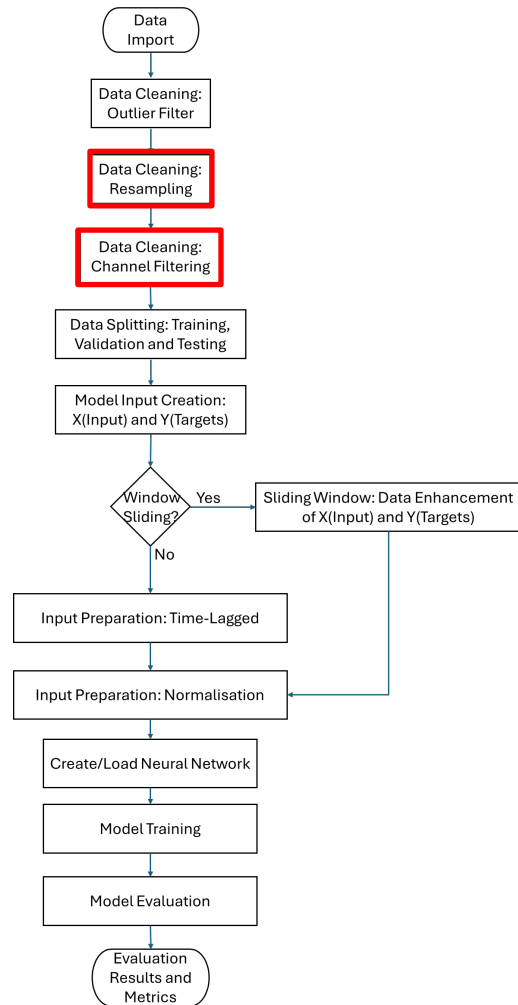




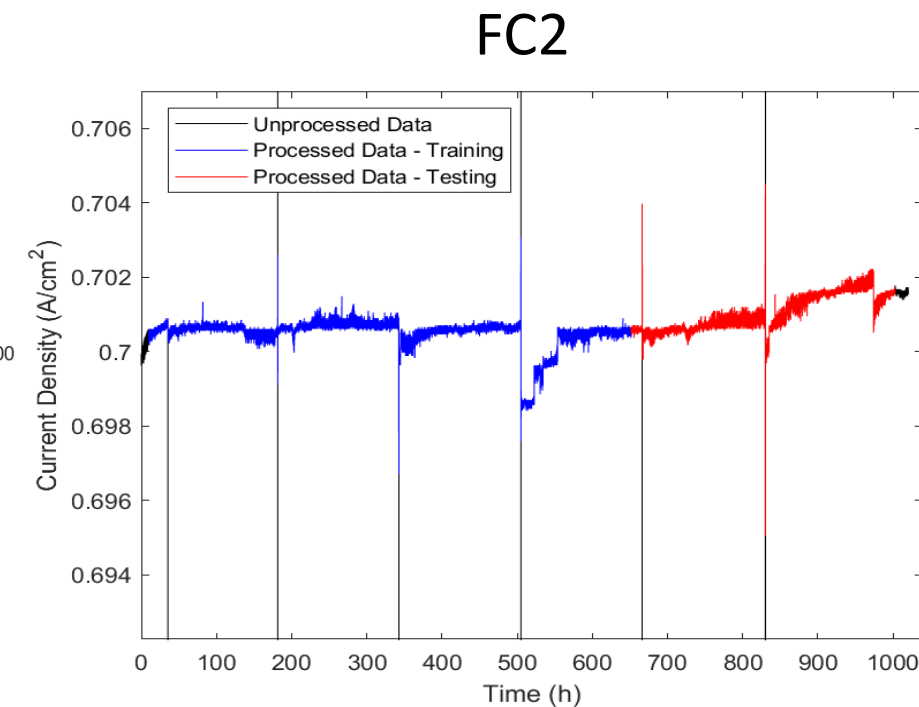
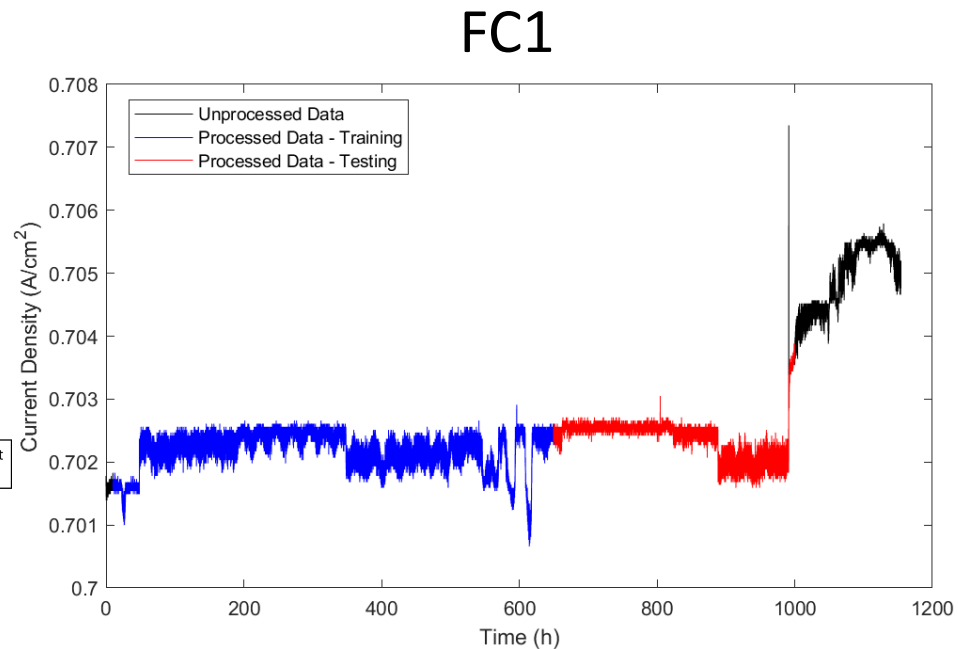
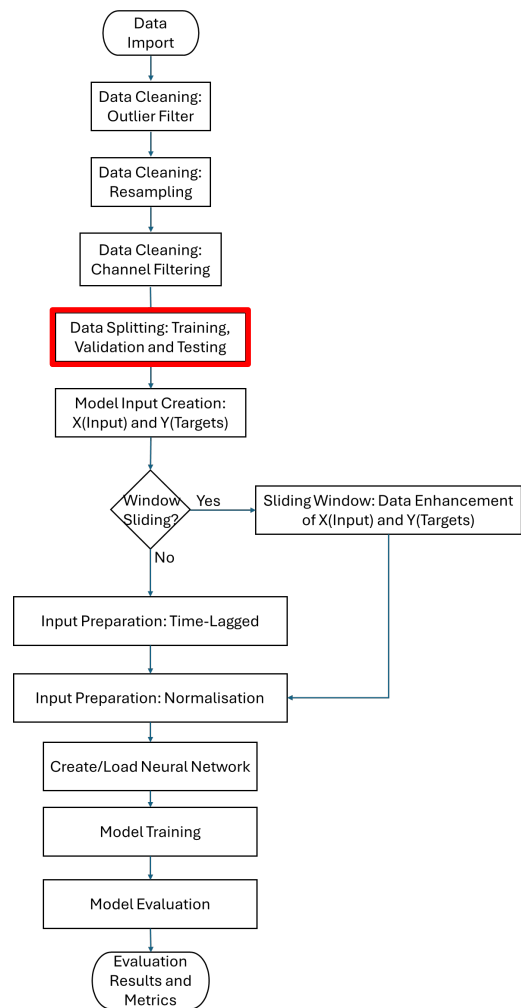
# AI-Based Machine Learning



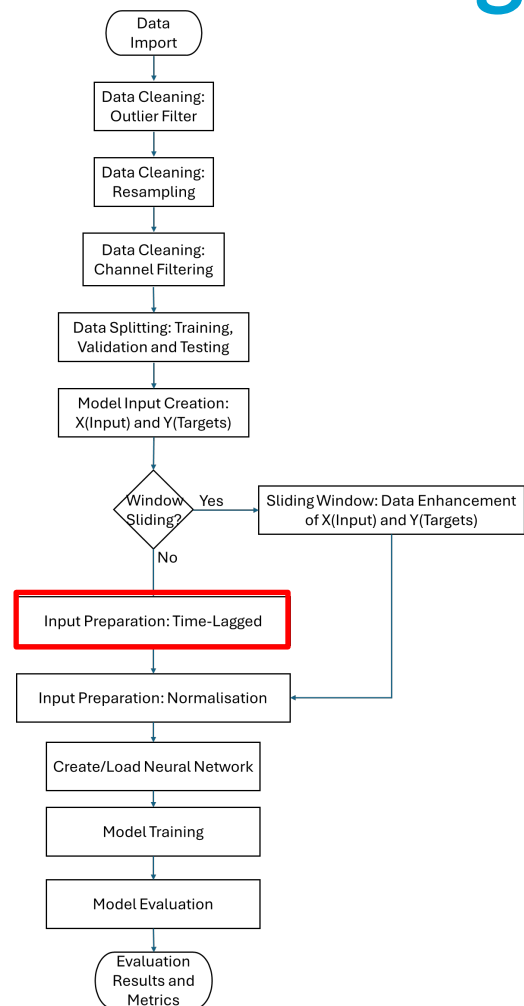
# ML: Data Pre-Processing



# ML: Data Pre-Processing



# ML: Time-Lagged Inputs



| Time (h) | Current (A) | H <sub>2</sub> , T <sub>in</sub> (°C) | Voltage (V) |
|----------|-------------|---------------------------------------|-------------|
| 0        | 70.164      | 25.93                                 | 3.317       |
| 0.000156 | 70.164      | 25.945                                | 3.317       |
| 0.000321 | 70.164      | 25.943                                | 3.316       |
| 0.000475 | 70.164      | 25.945                                | 3.316       |
| 0.000624 | 70.164      | 25.939                                | 3.316       |
| 0.000794 | 70.164      | 25.956                                | 3.317       |
| 0.000951 | 70.164      | 25.951                                | 3.317       |
| 0.001103 | 70.164      | 25.939                                | 3.317       |
| 0.001258 | 70.164      | 25.956                                | 3.317       |
| 0.001409 | 70.164      | 25.951                                | 3.316       |

Predictors (X)

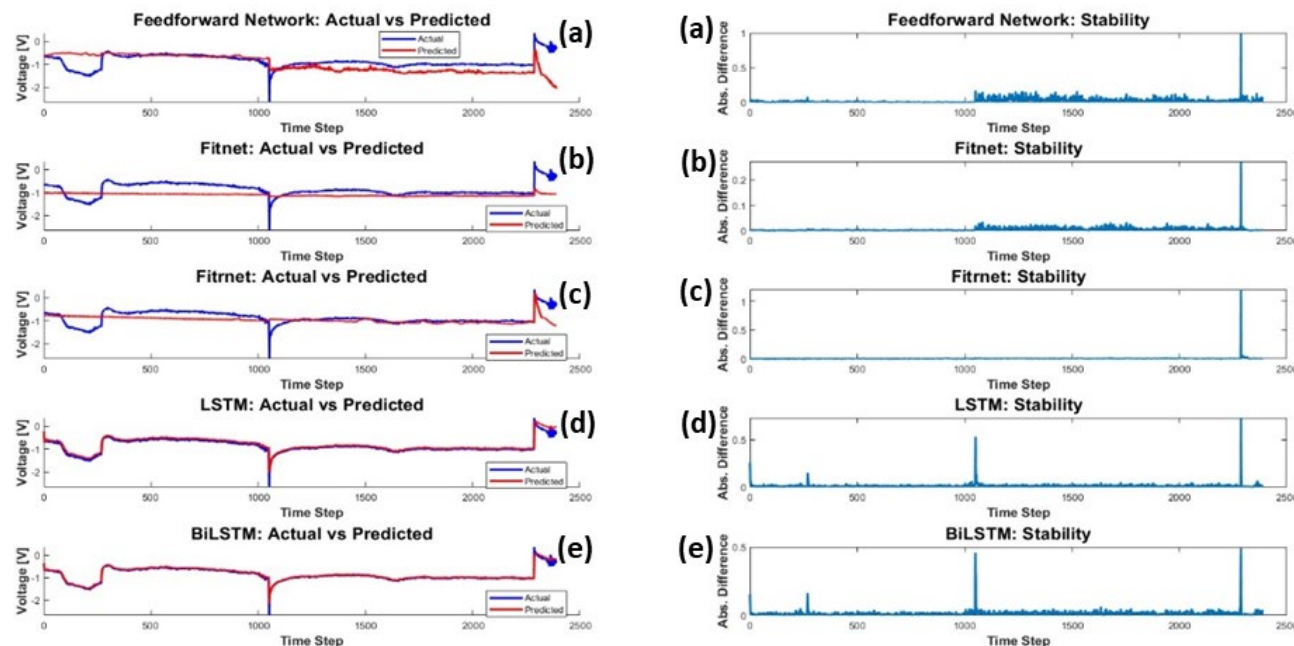
Target (Y)

| Time (h) | Current (A) | H <sub>2</sub> , T <sub>in</sub> (°C) | Voltage (V) |
|----------|-------------|---------------------------------------|-------------|
| 0        | 70.164      | 25.93                                 | 3.317       |
| 0.000156 | 70.164      | 25.945                                | 3.317       |
| 0.000321 | 70.164      | 25.943                                | 3.316       |
| 0.000475 | 70.164      | 25.945                                | 3.316       |
| 0.000624 | 70.164      | 25.939                                | 3.316       |
| 0.000794 | 70.164      | 25.956                                | 3.317       |
| 0.000951 | 70.164      | 25.951                                | 3.317       |
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| 0.000951 | 70.164      | 25.951                                | 3.317       |
| 0.001103 | 70.164      | 25.939                                | 3.317       |
| 0.001258 | 70.164      | 25.956                                | 3.317       |
| 0.001409 | 70.164      | 25.951                                | 3.316       |

# ML: Neural Network Selection

Prediction Stability - FC1



FC1

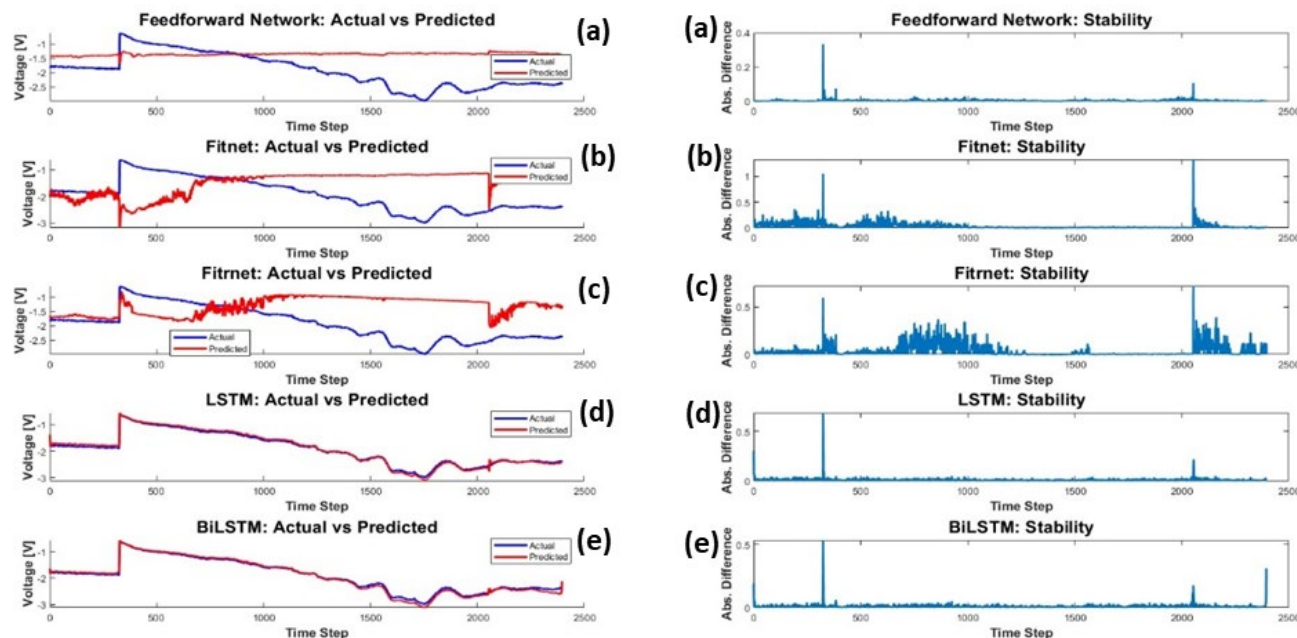
| METRIC      | Feedfor<br>ward | FitNet  | FitrNet | LSTM  | BiLSTM |
|-------------|-----------------|---------|---------|-------|--------|
| <b>R2</b>   | -1.293          | -0.564  | 0.022   | 0.977 | 0.991  |
| <b>RMSE</b> | 0.419           | 0.346   | 0.274   | 0.090 | 0.053  |
| <b>MSE</b>  | 0.176           | 0.120   | 0.075   | 0.009 | 0.004  |
| <b>MAE</b>  | 0.291           | 0.287   | 0.197   | 0.064 | 0.033  |
| <b>MAPE</b> | 106.155         | 110.939 | 58.237  | 0.125 | 0.065  |

FC2

|             |        |        |        |       |       |
|-------------|--------|--------|--------|-------|-------|
| <b>R2</b>   | -0.990 | -2.016 | -1.766 | 0.991 | 0.988 |
| <b>RMSE</b> | 0.828  | 1.019  | 0.976  | 0.062 | 0.066 |
| <b>MSE</b>  | 0.685  | 1.038  | 0.952  | 0.014 | 0.018 |
| <b>MAE</b>  | 0.696  | 0.859  | 0.826  | 0.049 | 0.052 |
| <b>MAPE</b> | 34.147 | 49.517 | 41.990 | 0.413 | 0.359 |

# ML: Neural Network Selection

Prediction Stability - FC2



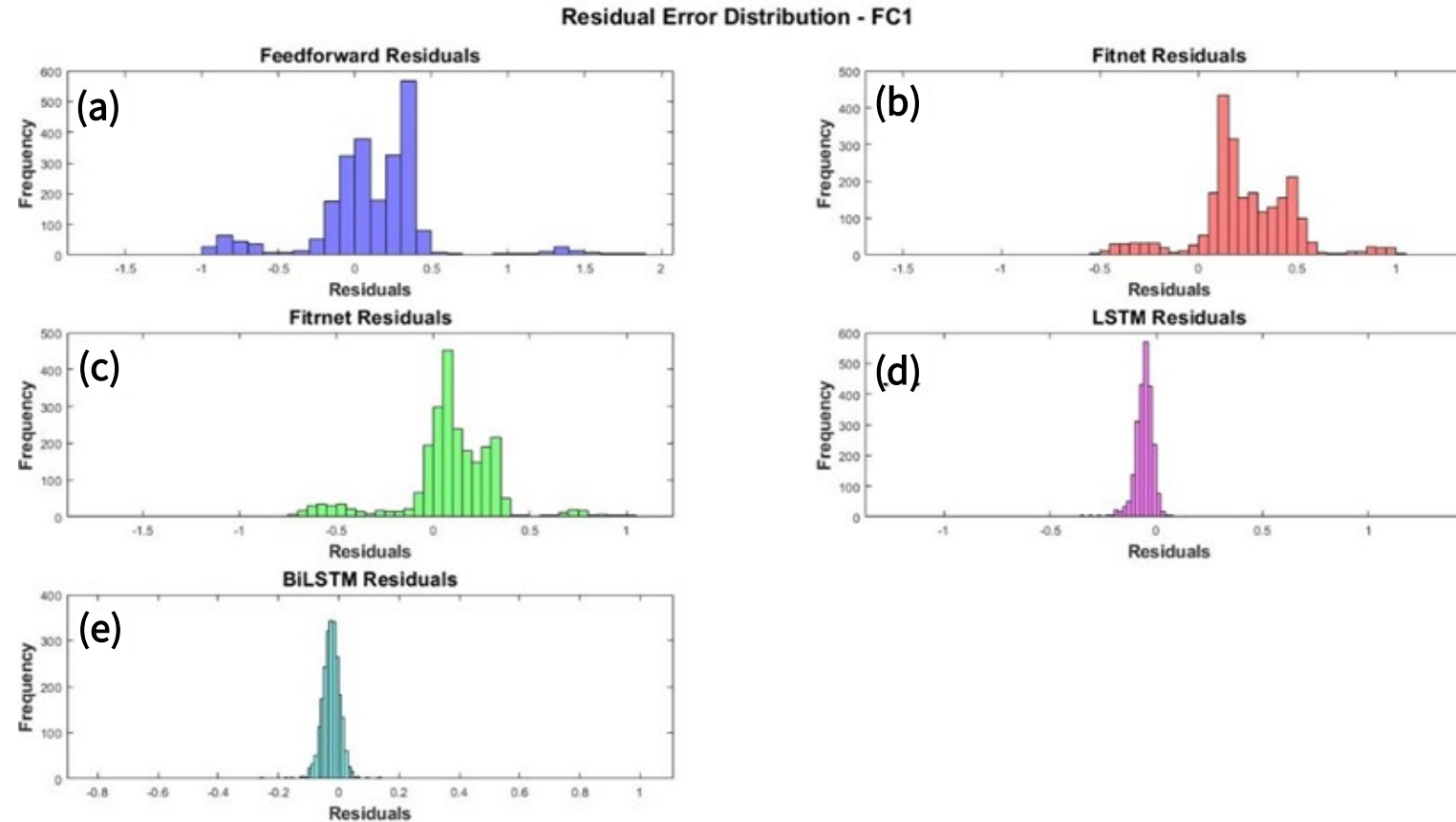
FC1

| METRIC | Feedfor<br>ward | FitNet  | FitrNet | LSTM  | BiLSTM |
|--------|-----------------|---------|---------|-------|--------|
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FC2

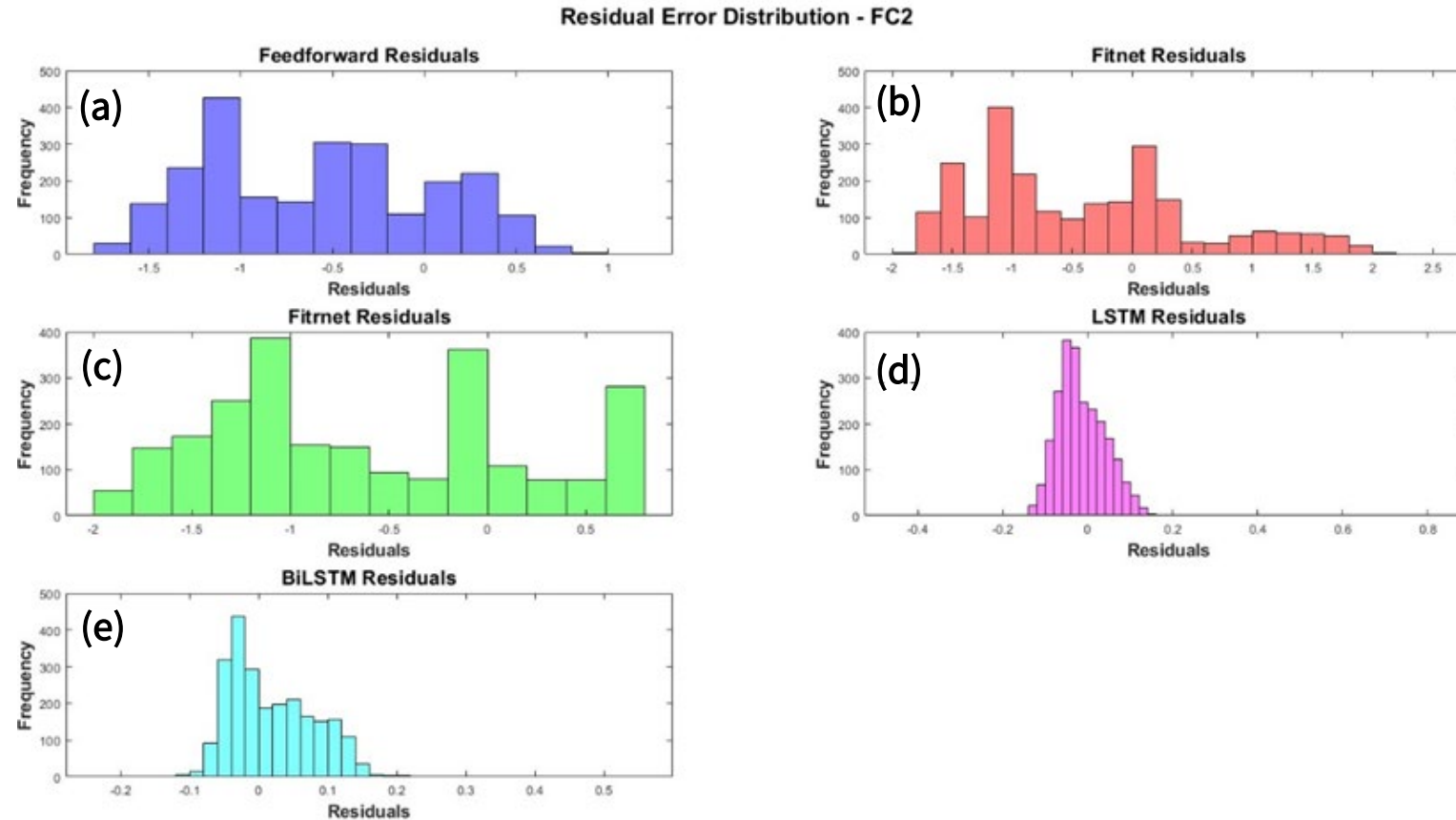
|      |        |        |        |       |       |
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# ML: Model Results and Validation





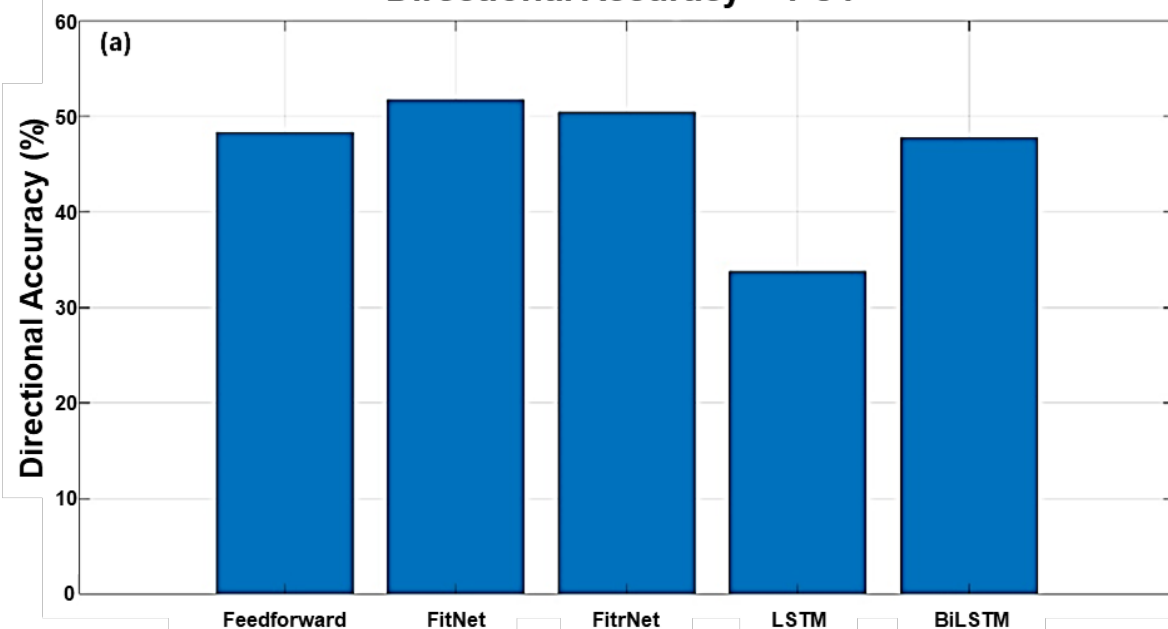
# ML: Model Results and Validation



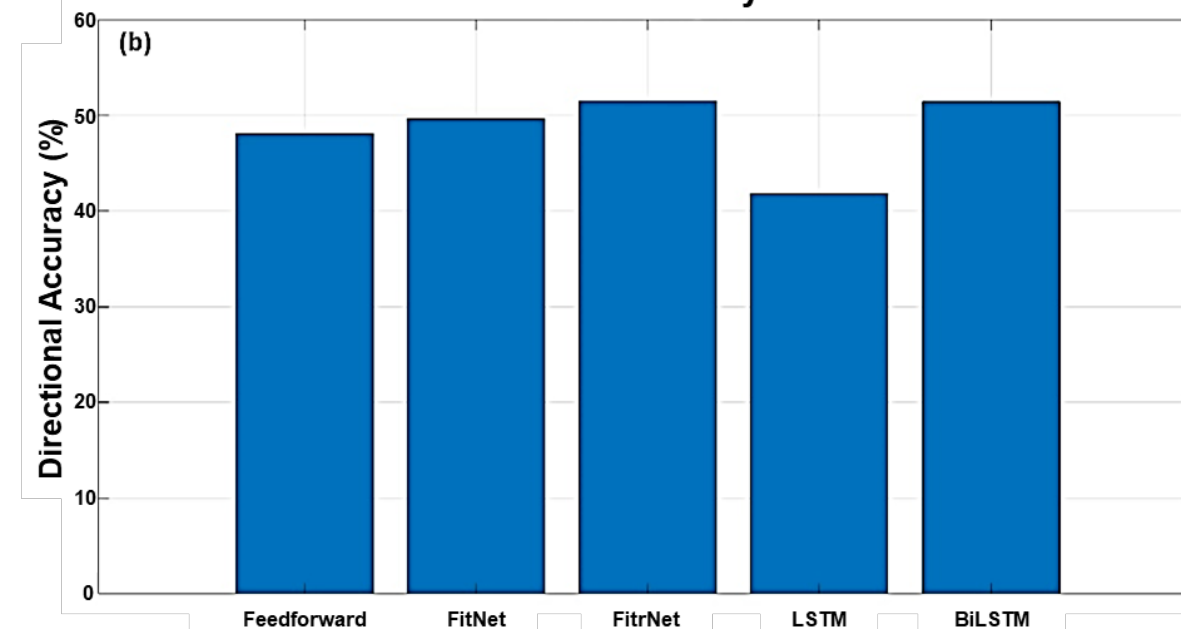


# ML: Model Results and Validation

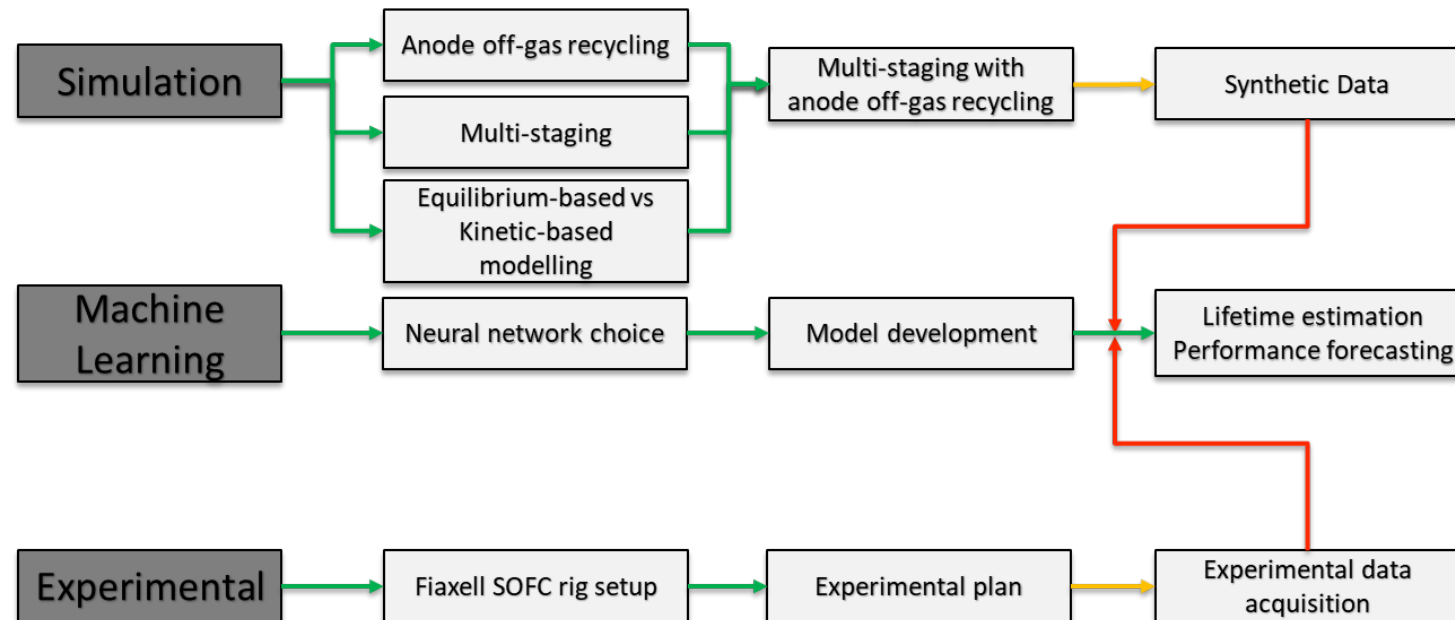
Directional Accuracy – FC1



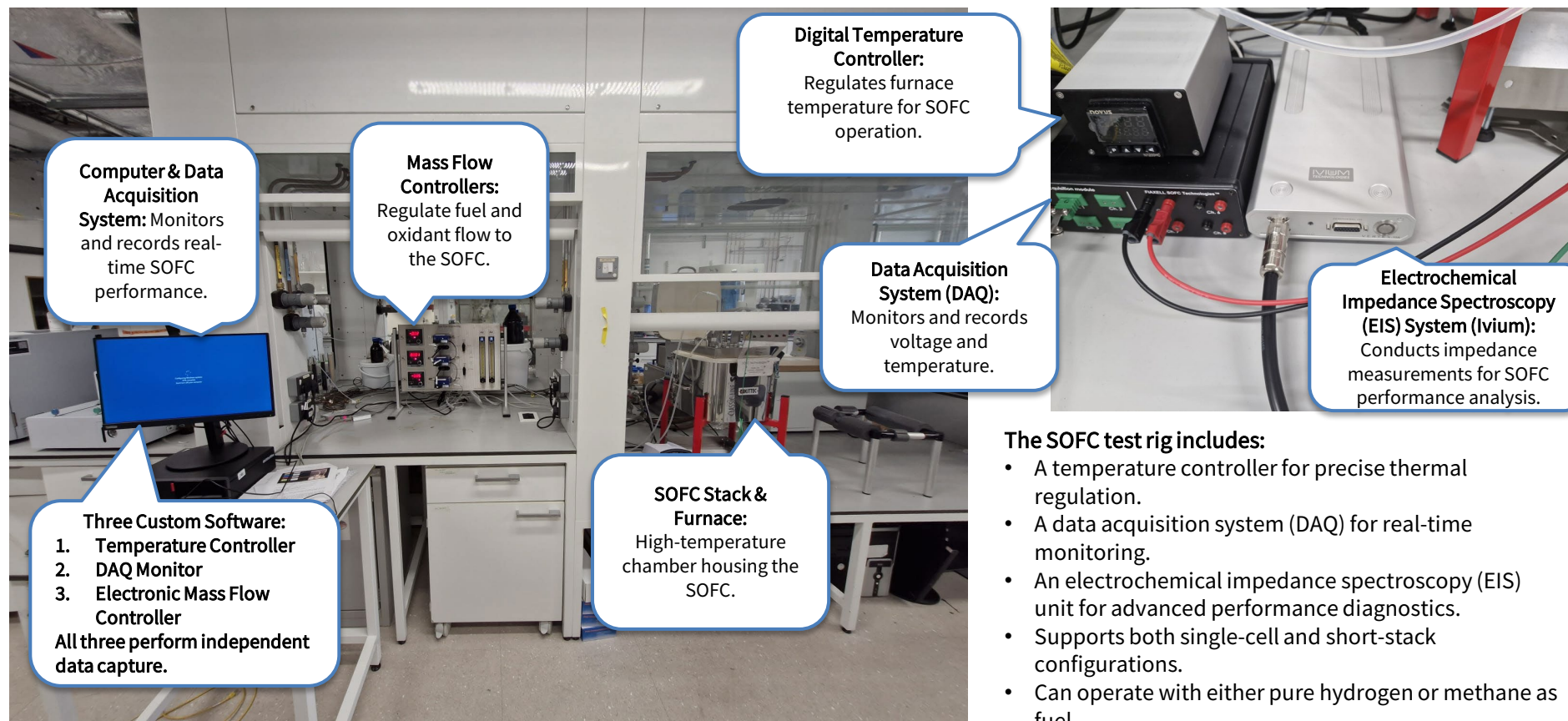
Directional Accuracy – FC2



# Experimental Work



# Experimental Work

**Computer & Data Acquisition System:** Monitors and records real-time SOFC performance.

**Mass Flow Controllers:** Regulate fuel and oxidant flow to the SOFC.

**Digital Temperature Controller:** Regulates furnace temperature for SOFC operation.

**Data Acquisition System (DAQ):** Monitors and records voltage and temperature.

**Electrochemical Impedance Spectroscopy (EIS) System (Ivium):** Conducts impedance measurements for SOFC performance analysis.

**Three Custom Software:**

1. Temperature Controller
2. DAQ Monitor
3. Electronic Mass Flow Controller

All three perform independent data capture.

**SOFC Stack & Furnace:** High-temperature chamber housing the SOFC.

**The SOFC test rig includes:**

- A temperature controller for precise thermal regulation.
- A data acquisition system (DAQ) for real-time monitoring.
- An electrochemical impedance spectroscopy (EIS) unit for advanced performance diagnostics.
- Supports both single-cell and short-stack configurations.
- Can operate with either pure hydrogen or methane as fuel.

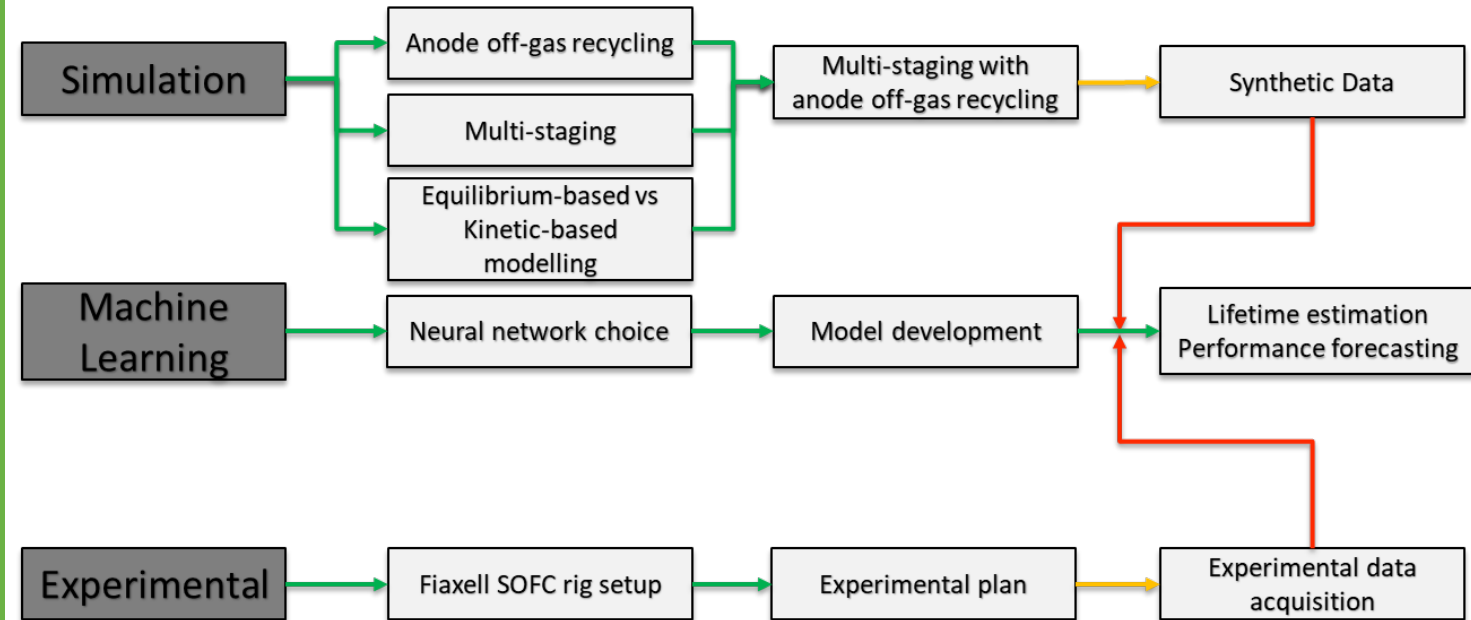
# Experimental Work: Future Plans



- Clearly demonstrate system performance degradation over time.
- Cover multiple operating modes.
- Include multiple scenarios (e.g., thermal cycling, dynamic loading).
- Provide data sufficient to extract reaction kinetics and degradation rates.

*Industrial standards typically require 40,000–90,000 hours of operational data for accurate kinetic modelling. However, for laboratory-based accelerated degradation studies, approximately 4,000 hours should be sufficient to develop conceptual models.*

# RESEARCH OVERVIEW





# Thank you!

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