

# Table of Contents (LSTM Report)

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## Purpose:

The Aim of the experiments is to determine the relation between the LSTM window & anomaly trends, and also to distinguish between anomaly in sensor data & sensor faults.

## Methodology:

EV dataset is used in different LSTM window sizes & error patterns are injected into the dataset. Later, the LSTM algorithm is tested on the dataset to see its effectiveness in finding the anomaly as well as sensor faults.

Codebase:

<https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Anomaly-Detection>

Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

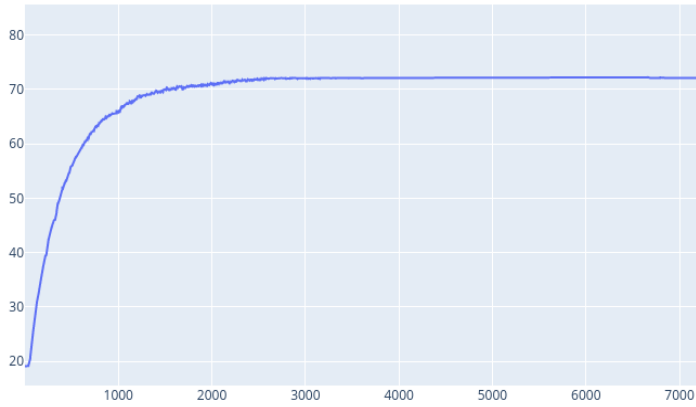
## Part A (LSTM<Error Window)

### 1. Data Vs Anomaly Plots Configuration:

- Dataset: EV stator winding Temp. vs Time (0.5 sec per datapoint)
- LSTM window: 60 data points
- Error window: 200 data points
- Error injected as Sine, Square & Ramp wave
- Error injected as both Positive and Negative values in +Y axis

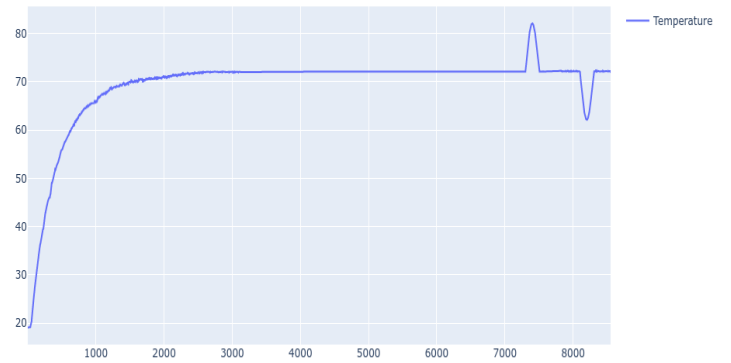
## 1.1 Sine Waveform error

stator\_winding Temp

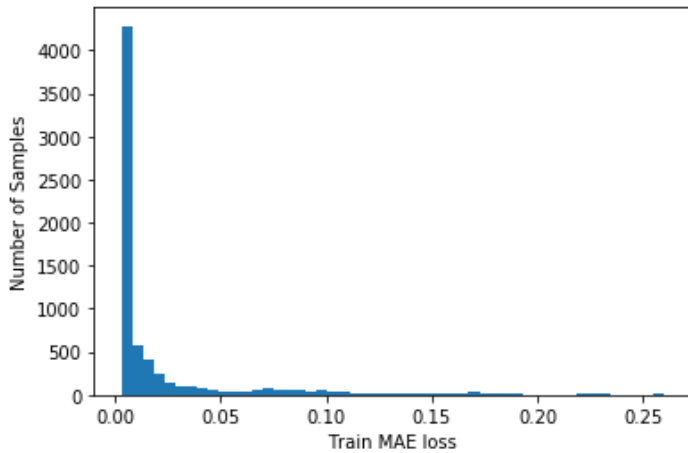


Training Dataset

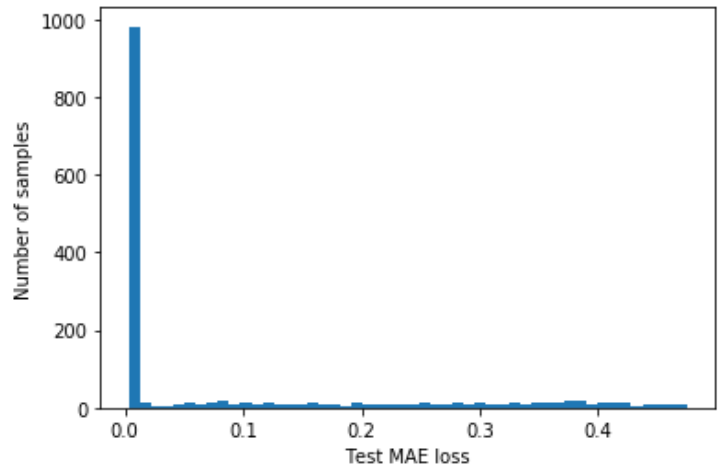
stator\_winding Temp



Sine Dataset

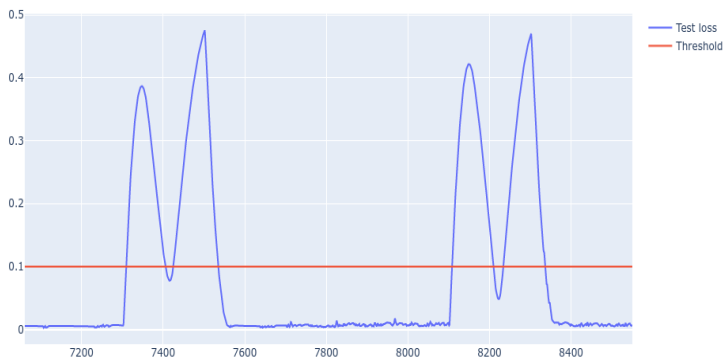


Sine Train MAE loss

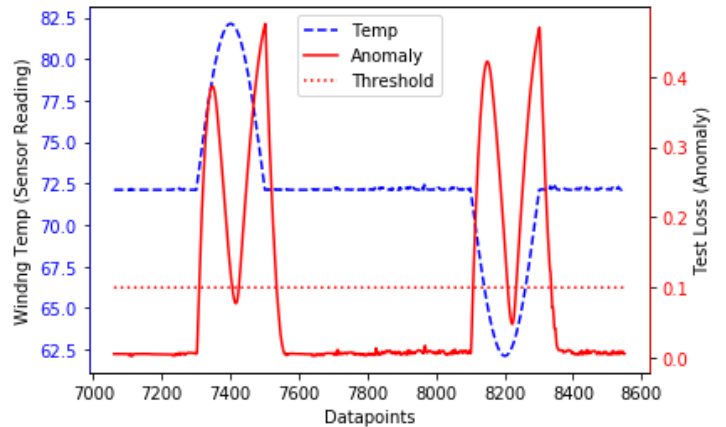


Sine Test MAE loss

Test loss vs. Threshold



Sine Anomaly vs Threshold

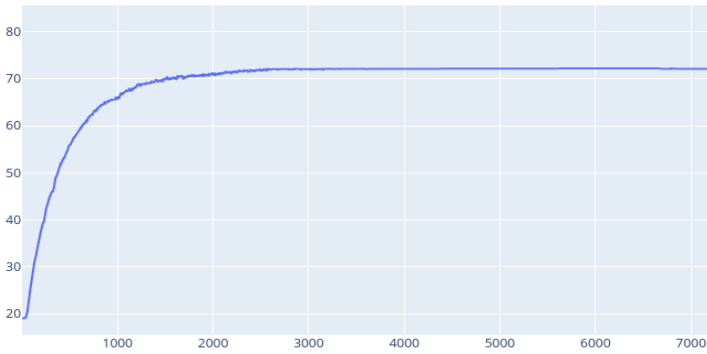


Sine Data vs Anomaly

Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

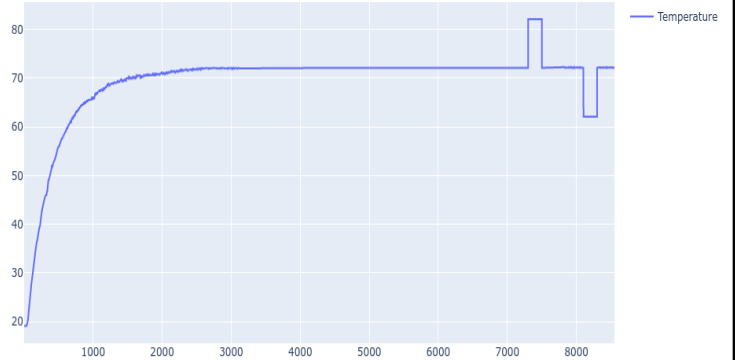
## 1.2 Square Waveform error

stator\_winding Temp

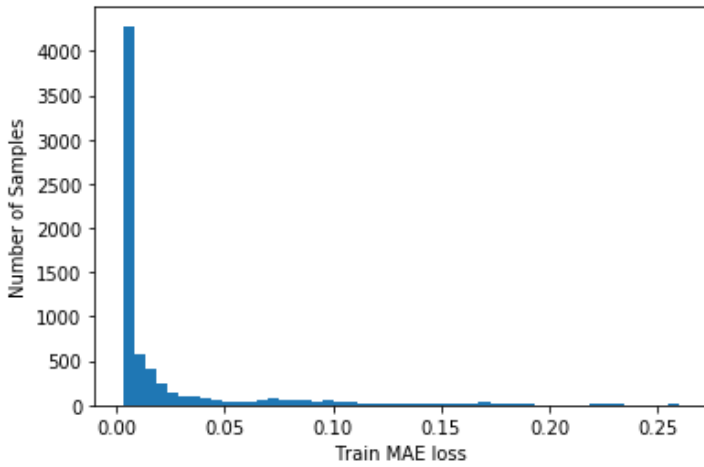


Training Dataset

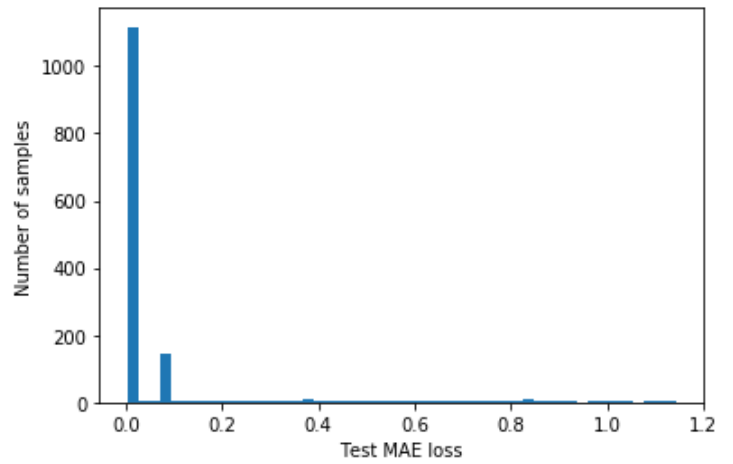
stator\_winding Temp



Square Dataset

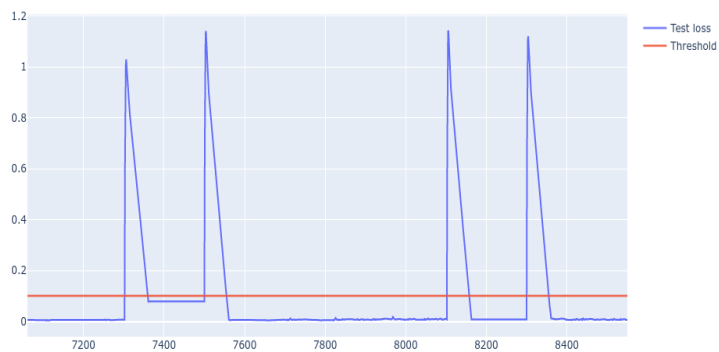


Square Train MAE loss

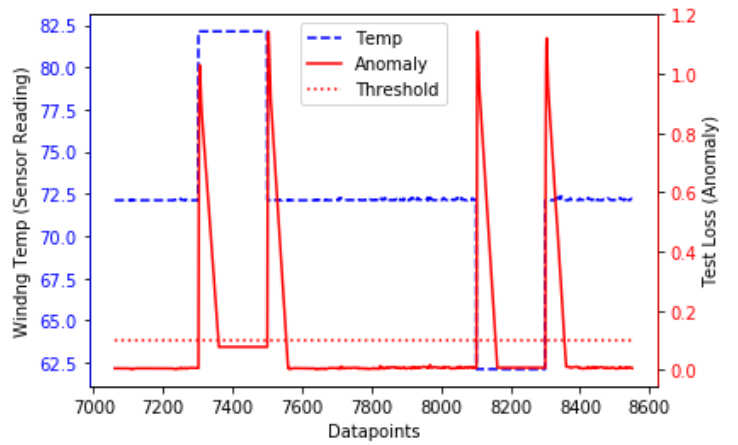


Square Test MAE loss

Test loss vs. Threshold



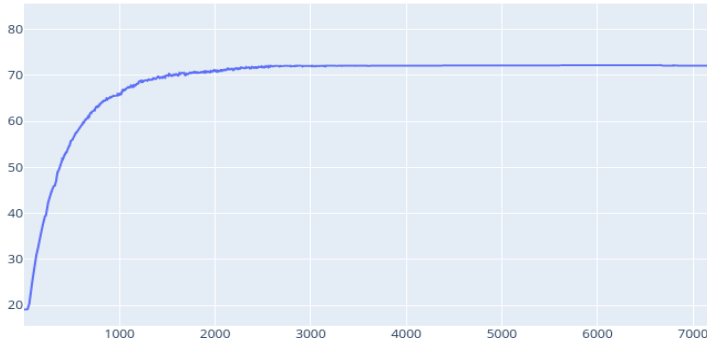
Square Anomaly vs Threshold



Square Data vs Anomaly

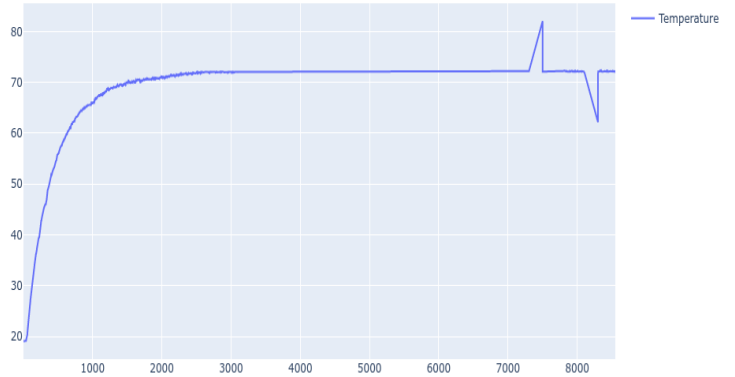
### 1.3 Ramp Waveform error

stator\_winding Temp

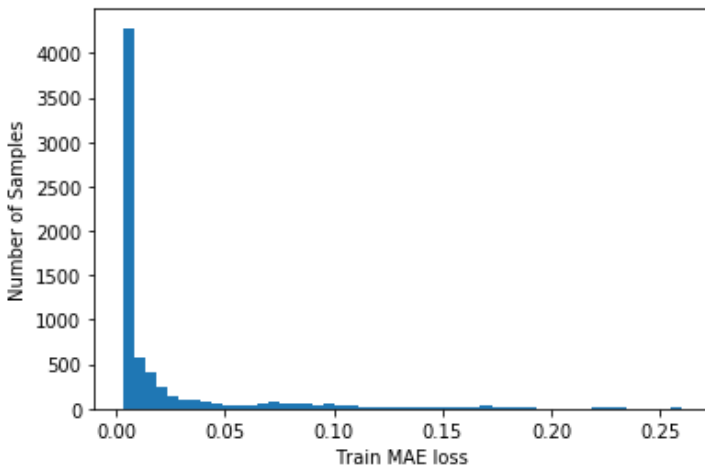


Training Dataset

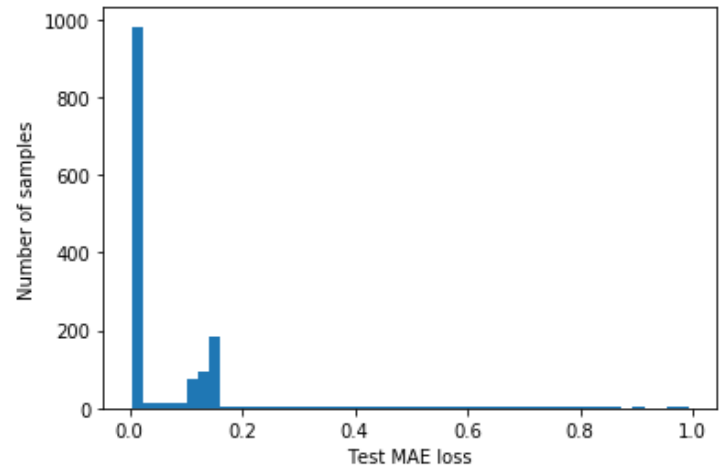
stator\_winding Temp



Ramp Dataset

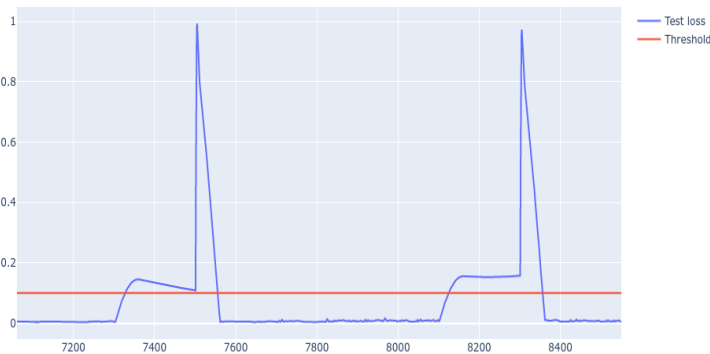


Ramp Train MAE loss

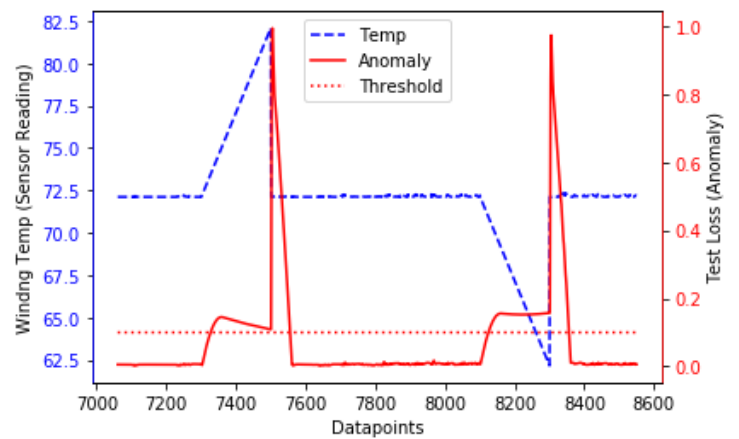


Ramp Test MAE loss

Test loss vs. Threshold



Ramp Anomaly vs Threshold



Ramp Data vs Anomaly

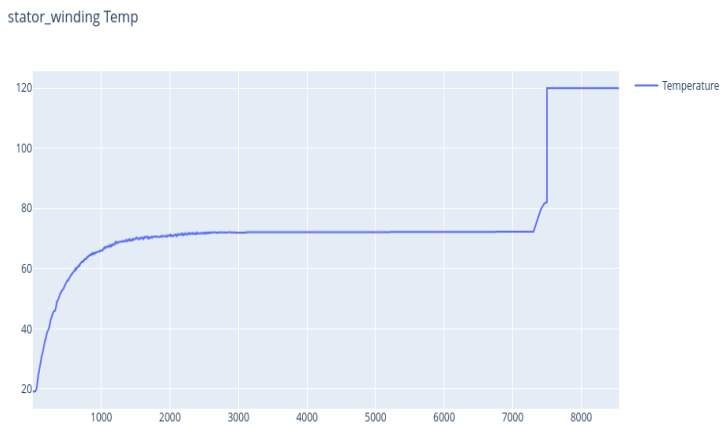
## 2. Data Vs Fault Plots

Configuration:

- Dataset: EV stator winding Temp. vs Time (0.5 sec per datapoint)
- LSTM window: 60 data points
- Error window: 200 data points
- Error injected as Sine, Square & Ramp wave
- Error injected as Positive and Negative values in +Y axis separately
- Operating range of the sensor is assumed from 0°C to 100°C,
- When the sensor is faulty, it will send the temp. that is beyond the operating range, i.e. -20°C and 120°C

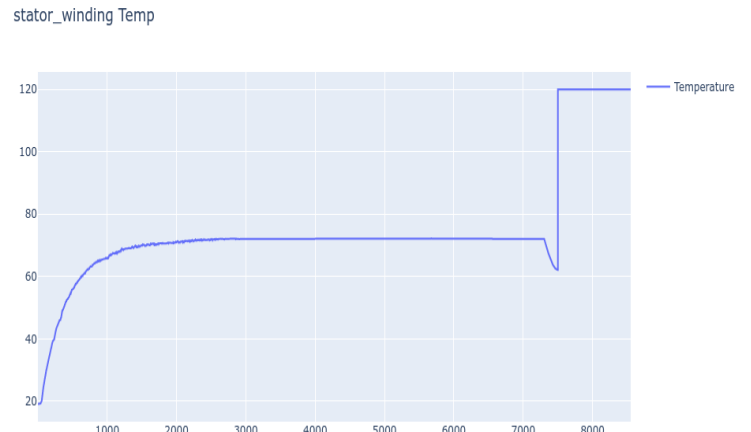
### 2.1.1 Sine, Fault HIGH

Sine Error curve: Rising



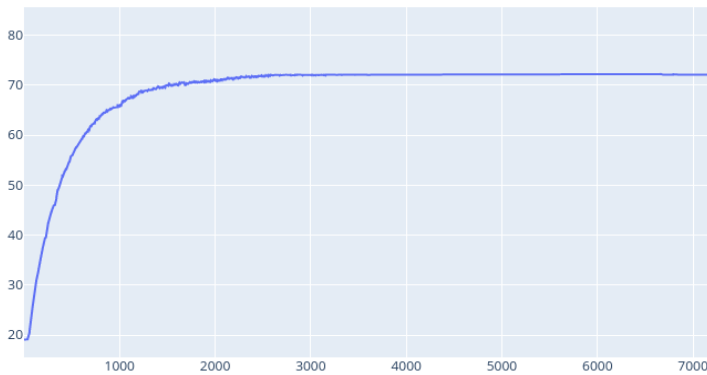
Sine Dataset

Sine Error curve: Falling



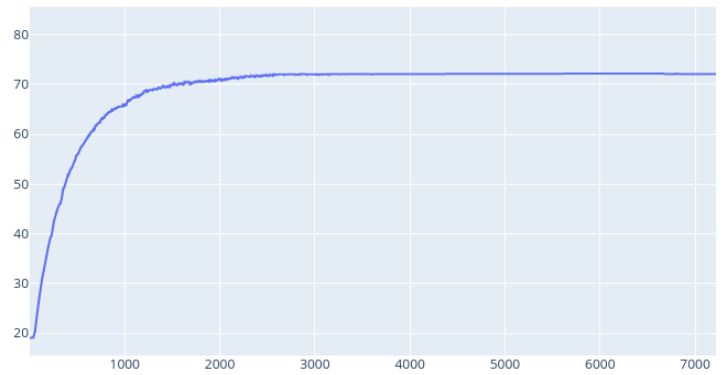
Sine Dataset

stator\_winding Temp

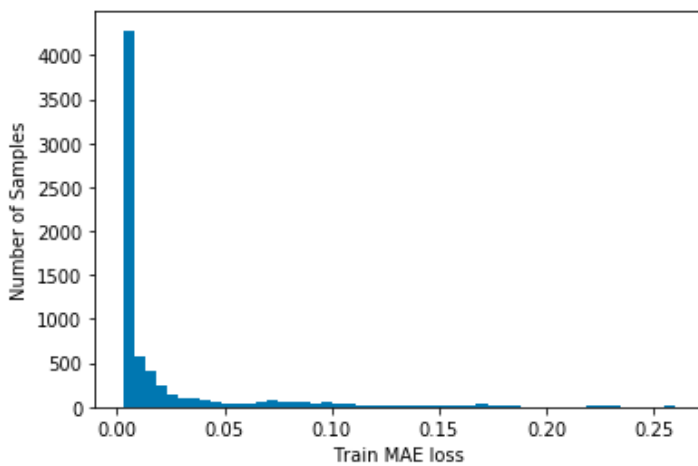


Training Dataset

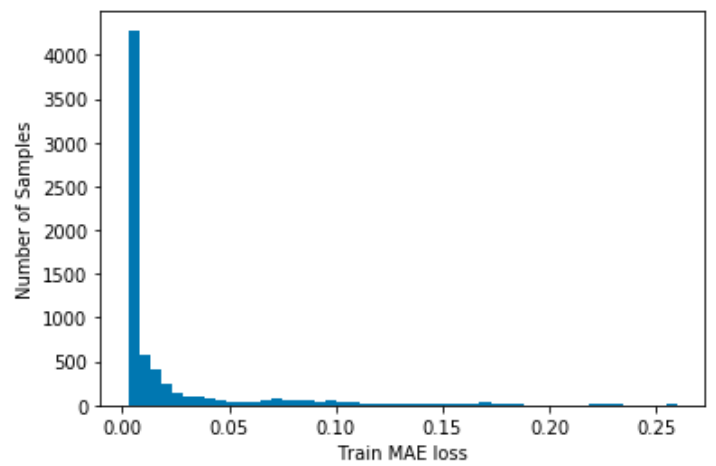
stator\_winding Temp



Training Dataset

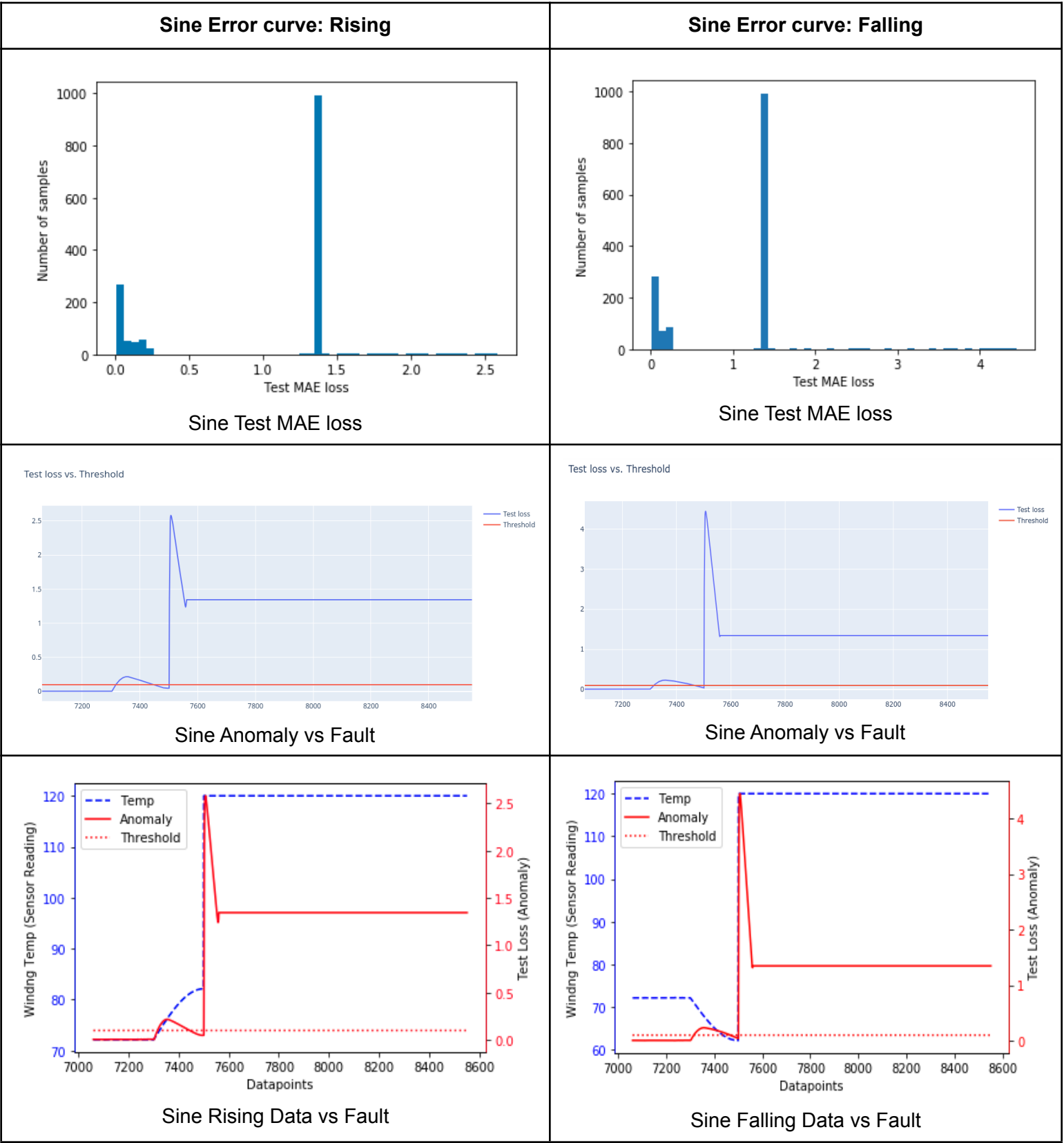


Train MAE loss



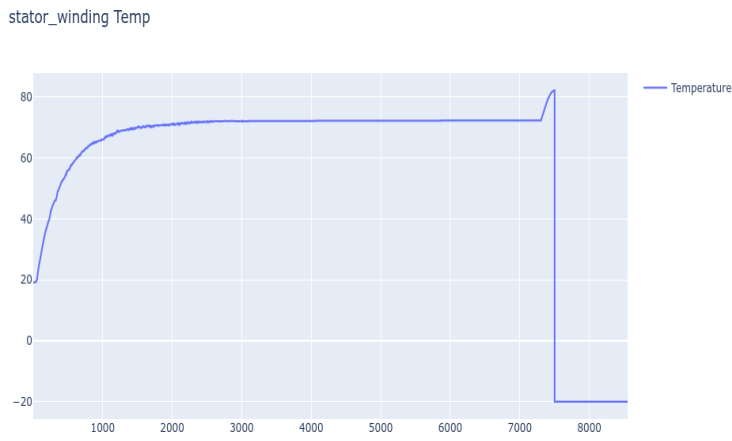
Train MAE loss





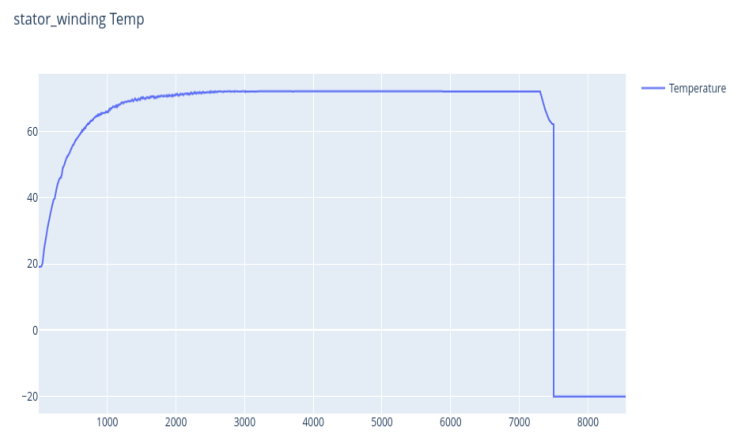
## 2.1.2 Sine, Fault LOW

Sine Error curve: Rising



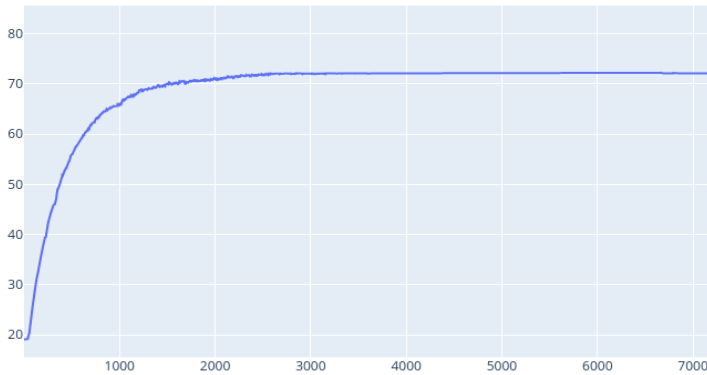
Sine Dataset

Sine Error curve: Falling



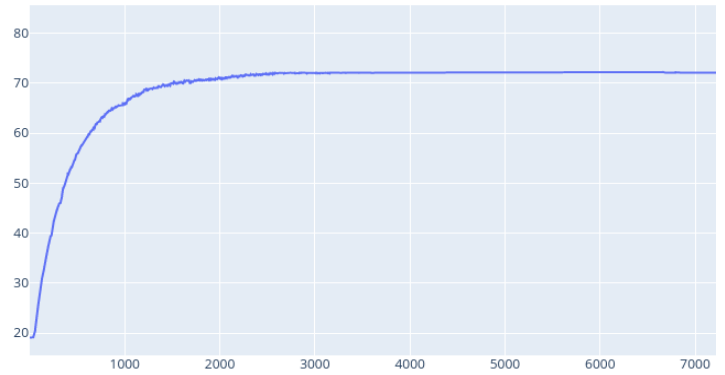
Sine Dataset

stator\_winding Temp

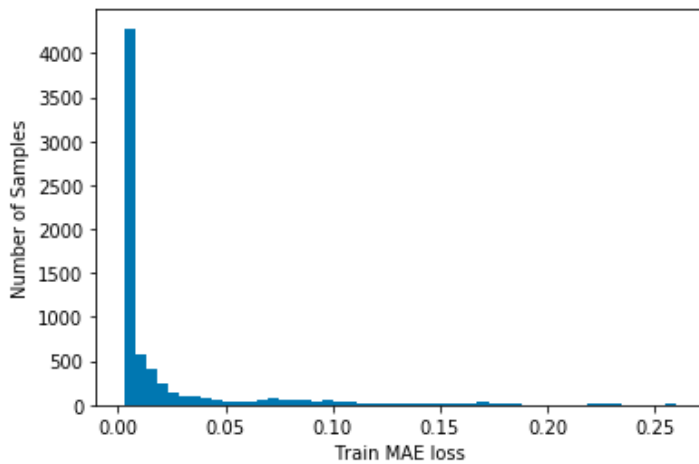


Training Dataset

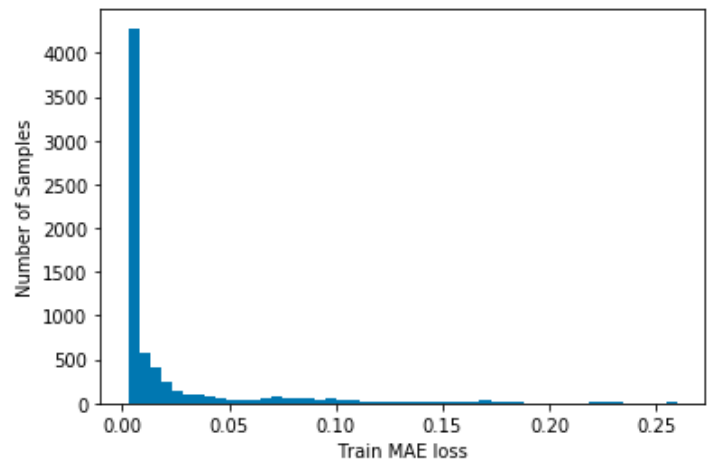
stator\_winding Temp



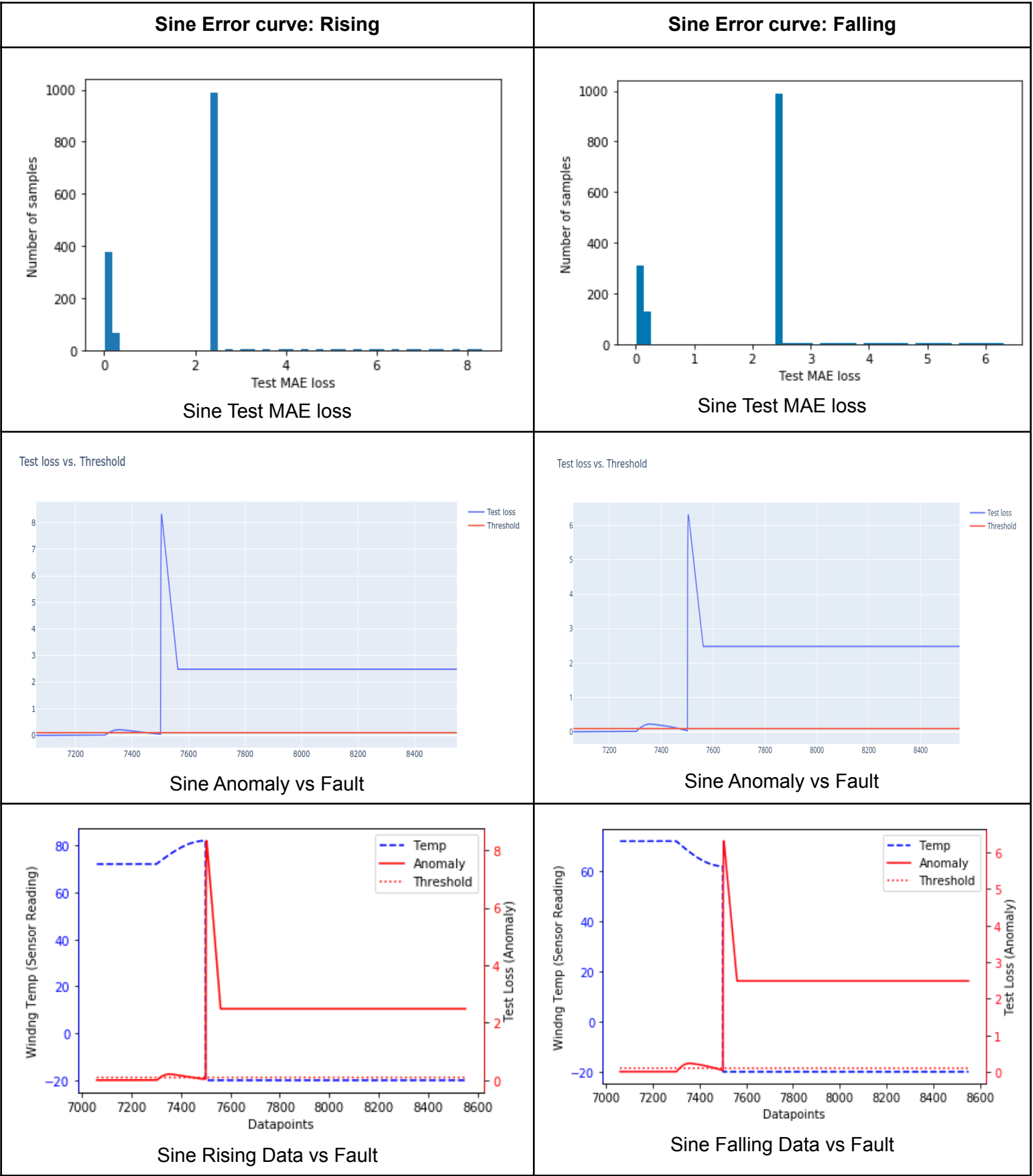
Training Dataset



Train MAE loss

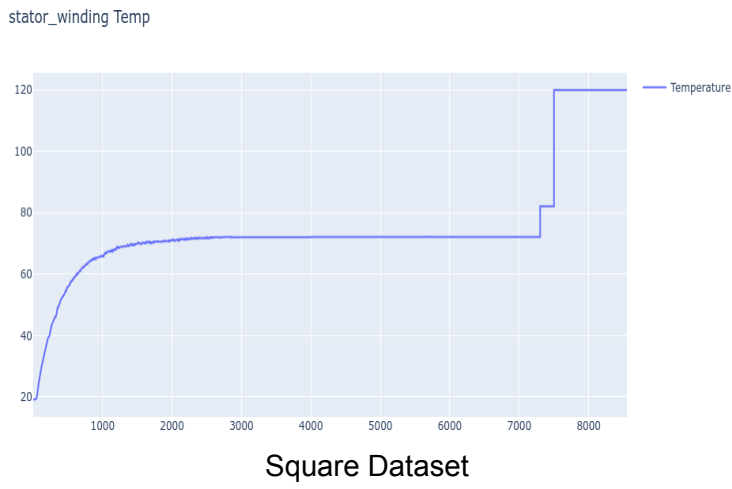


Train MAE loss

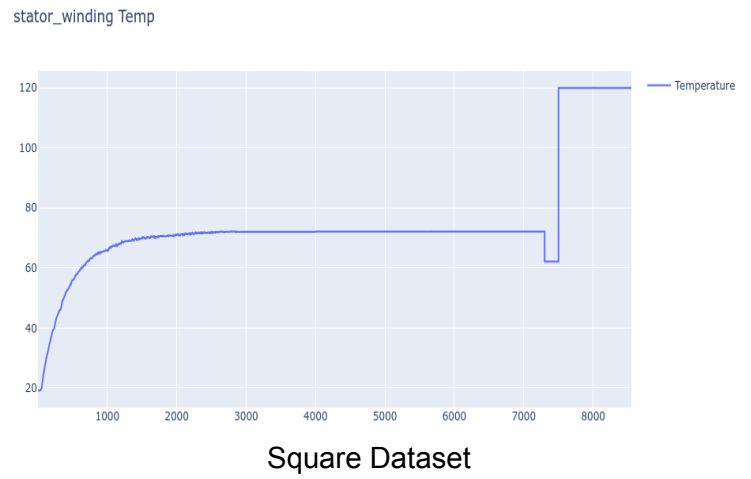


### 2.2.1 Square, Fault HIGH

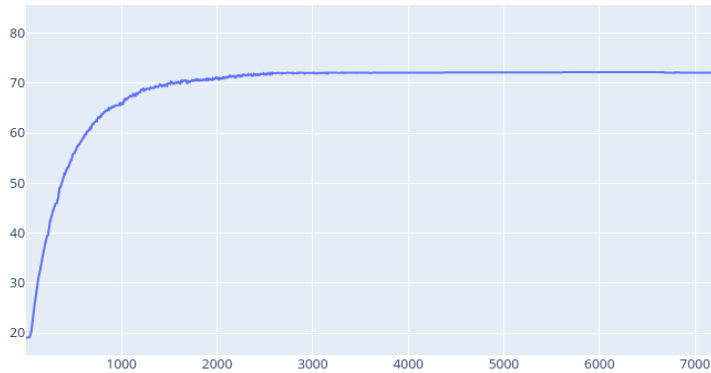
**Square Error curve: Rising**



**Square Error curve: Falling**

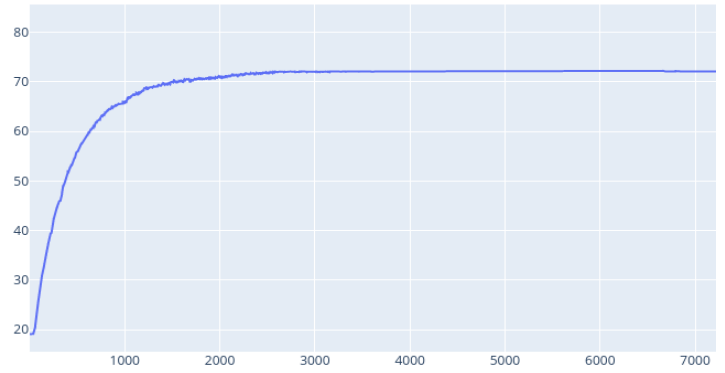


stator\_winding Temp

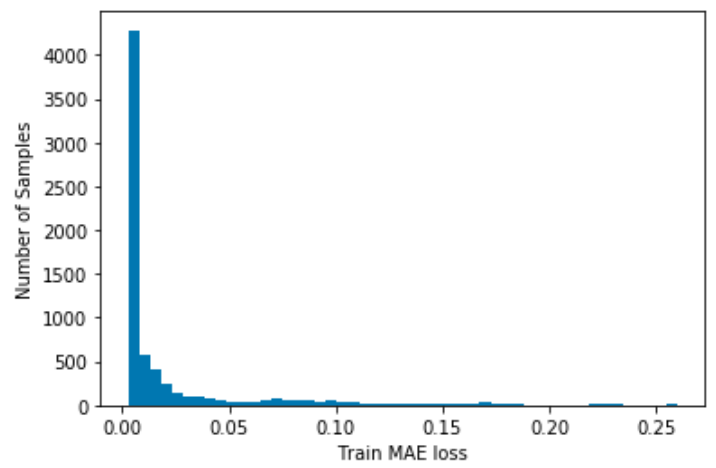
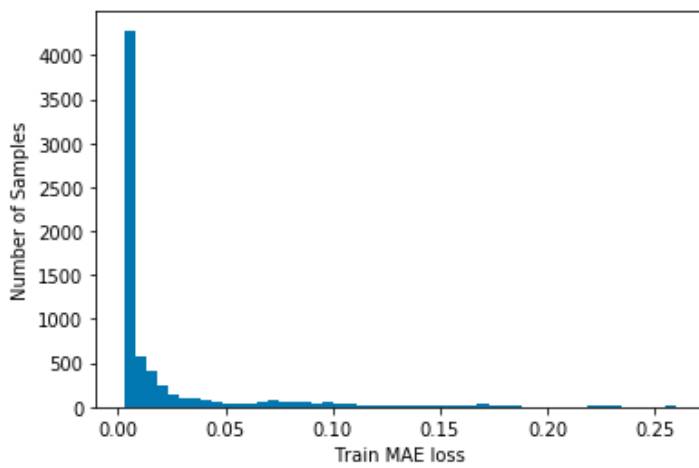


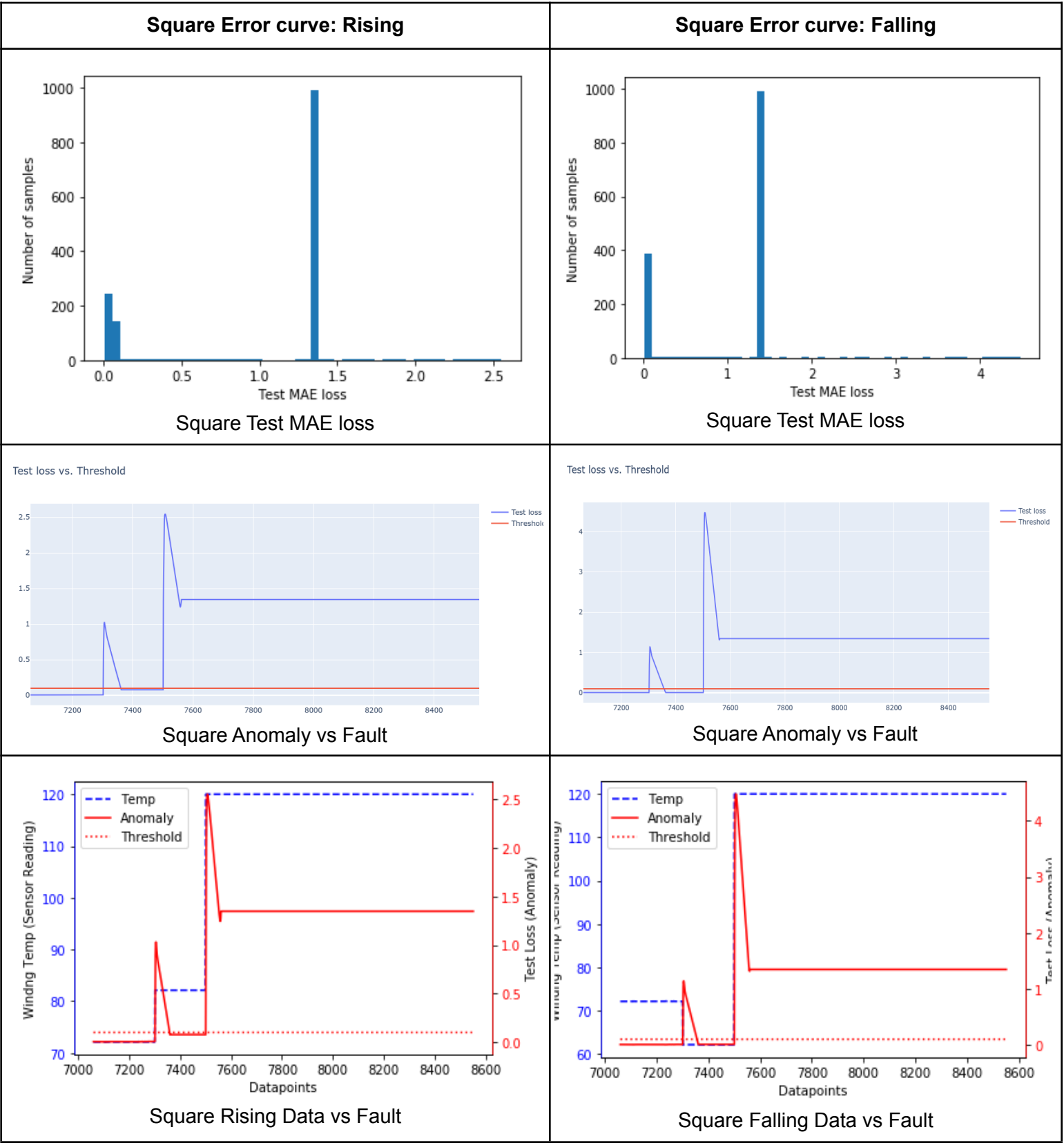
Training Dataset

stator\_winding Temp



Training Dataset

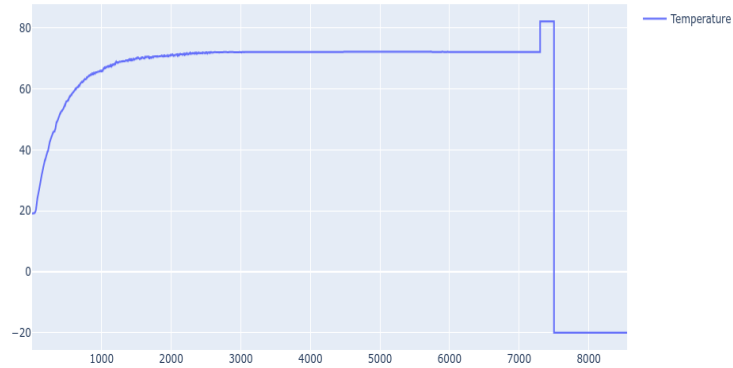




## 2.2.2 Square, Fault LOW

**Square Error curve: Rising**

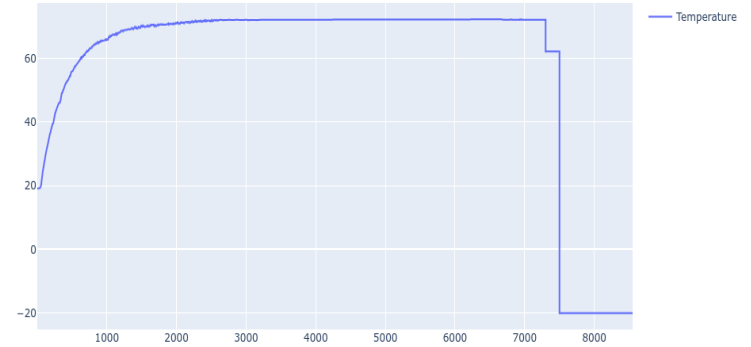
stator\_winding Temp



Square Dataset

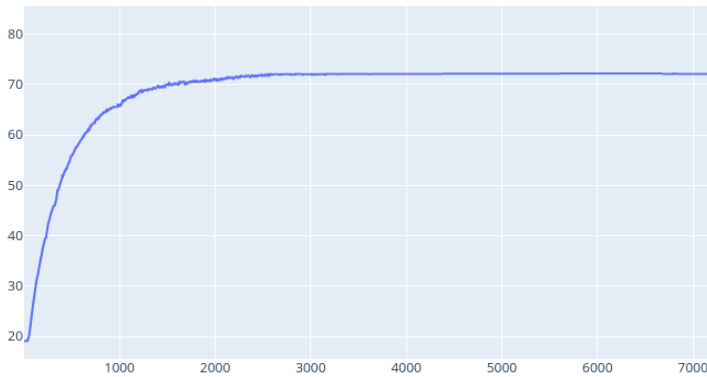
**Square Error curve: Falling**

stator\_winding Temp



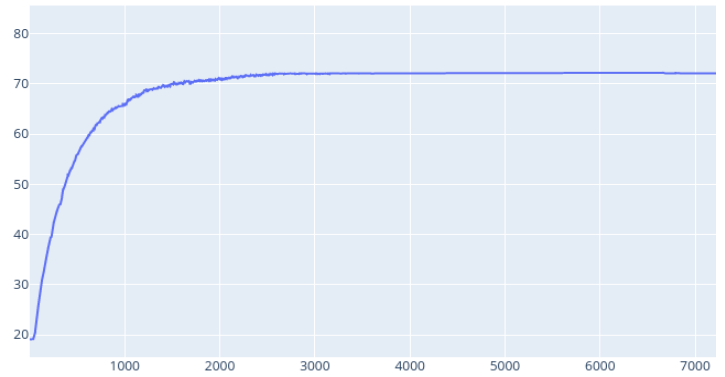
Square Dataset

stator\_winding Temp

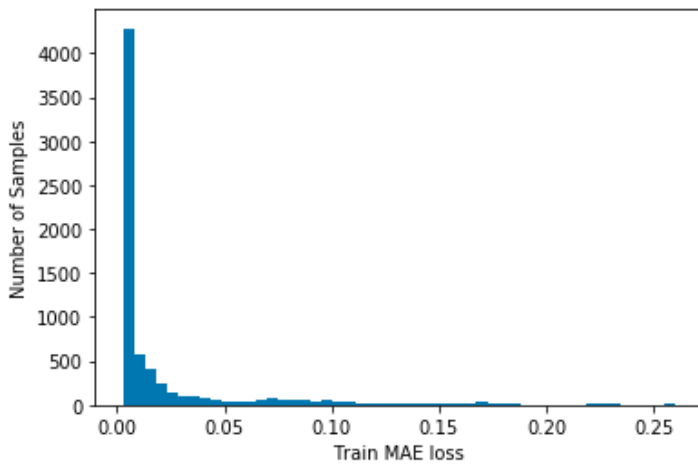


Training Dataset

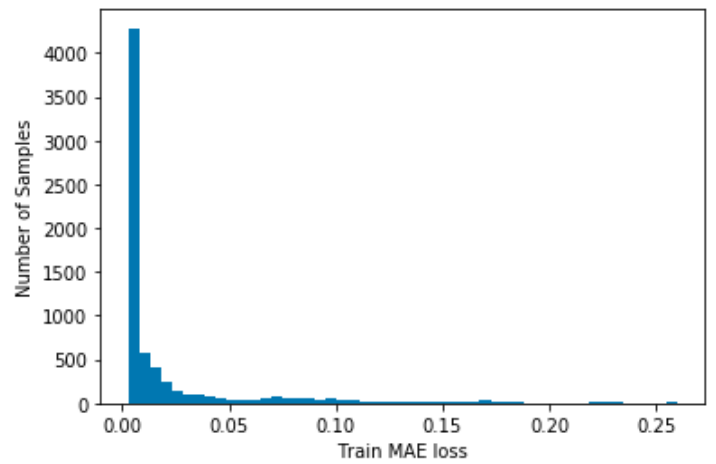
stator\_winding Temp



Training Dataset

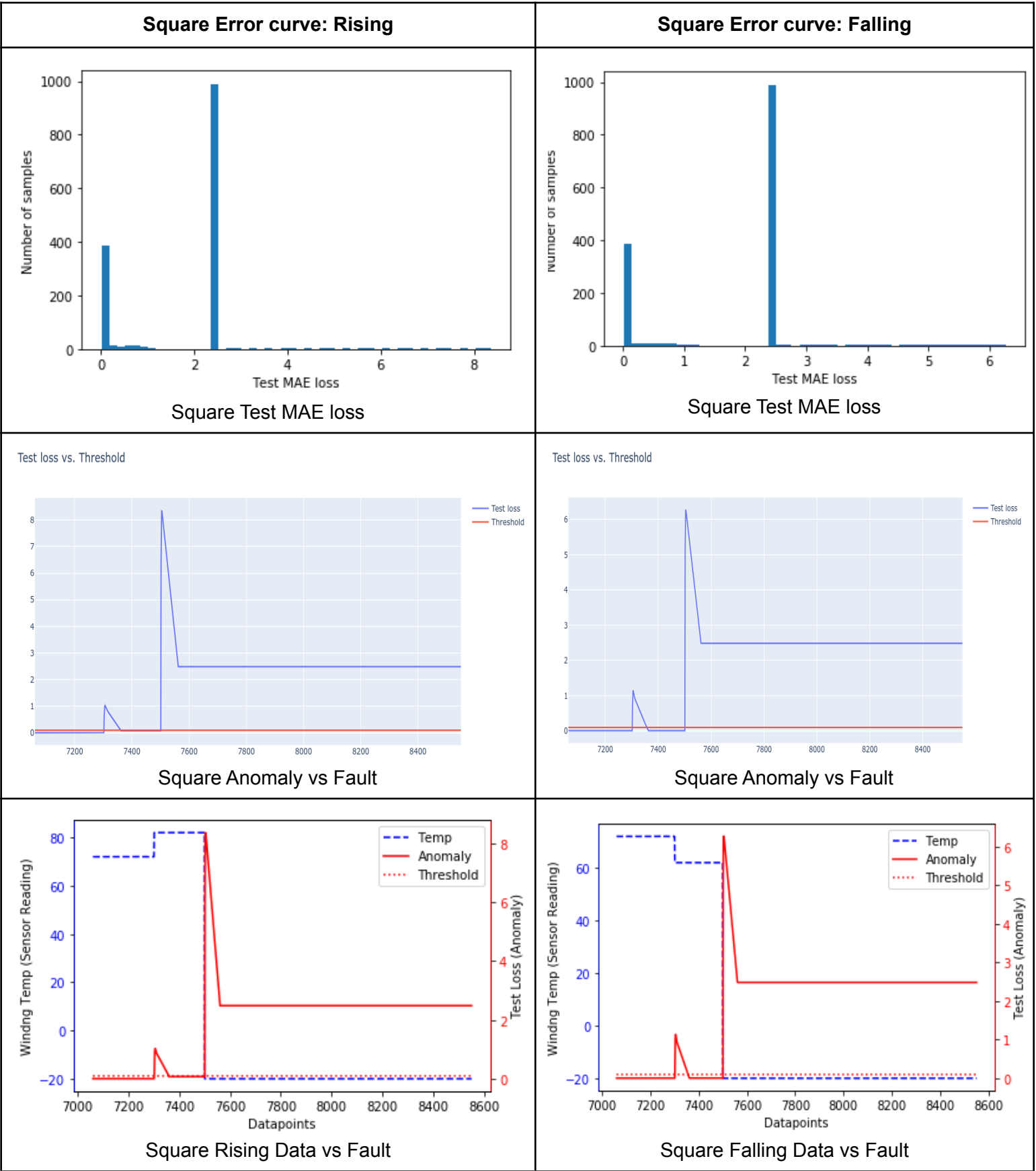


Train MAE loss



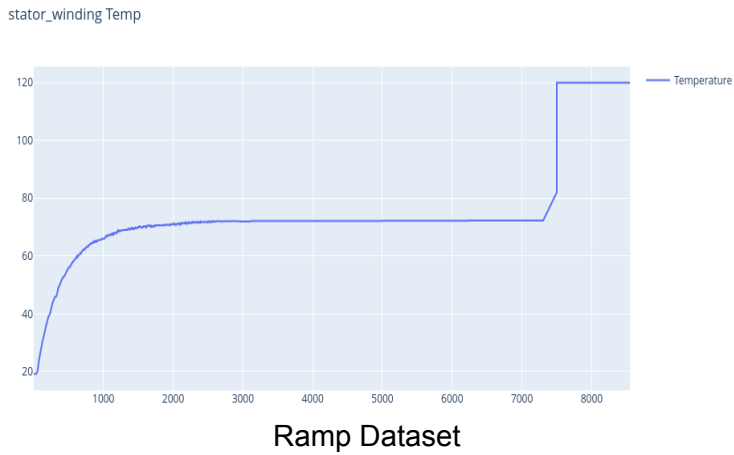
Train MAE loss

Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

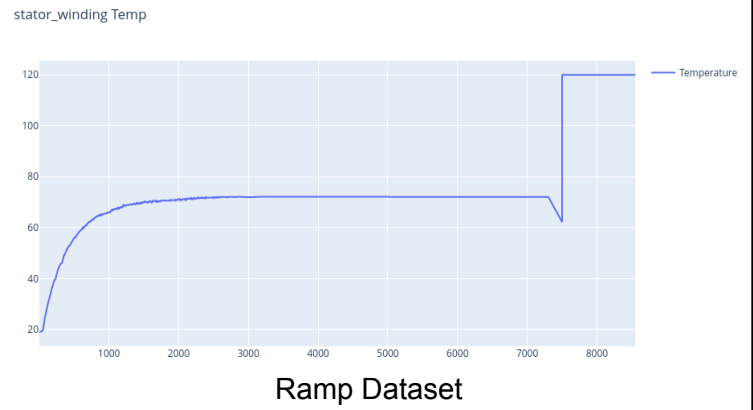


### 2.3.1 Ramp, Fault HIGH

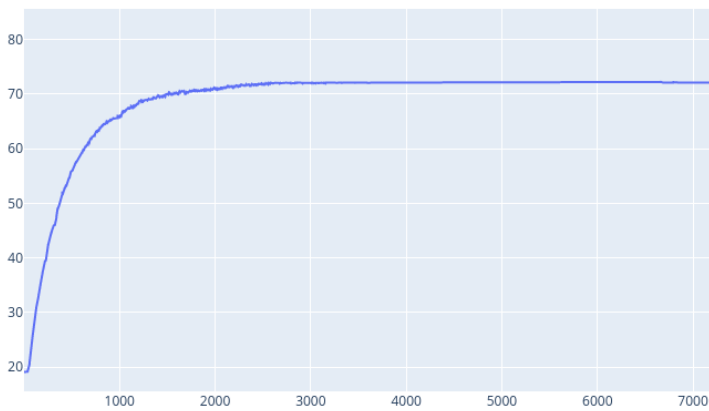
Ramp Error curve: Rising



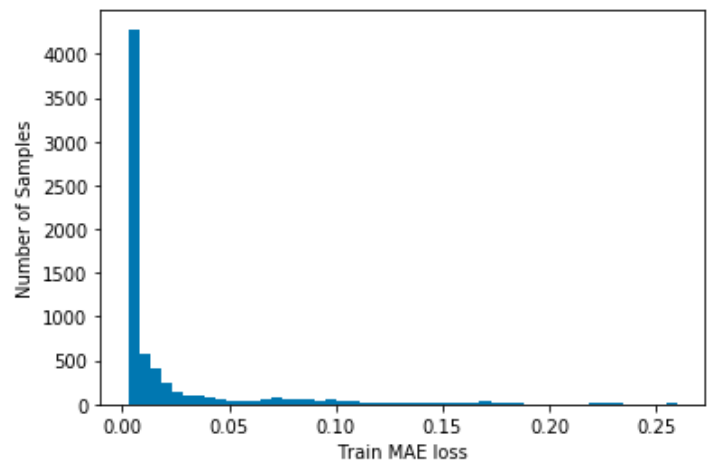
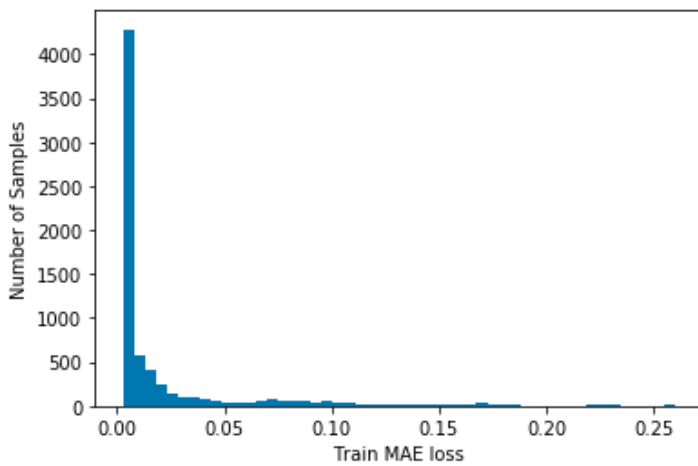
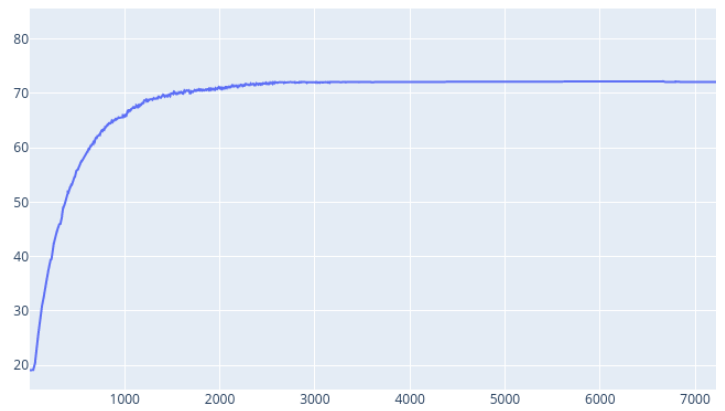
Ramp Error curve: Falling



stator\_winding Temp

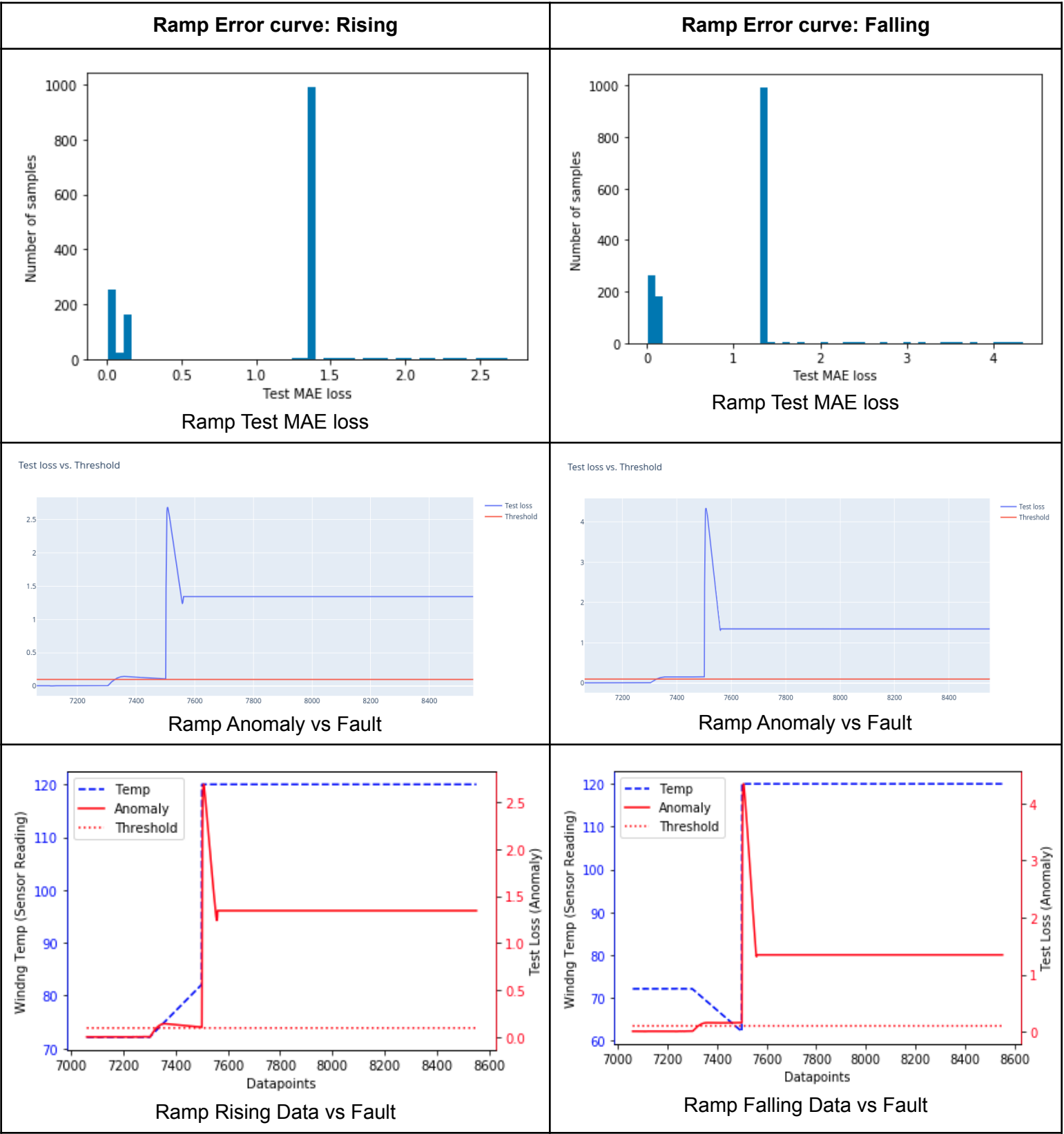


stator\_winding Temp



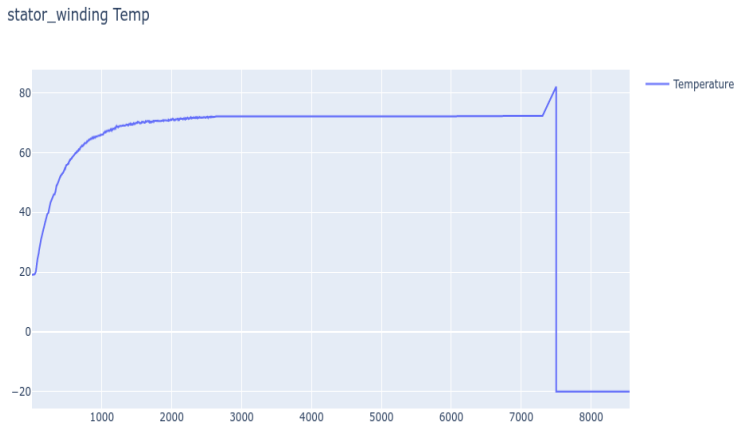
Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>



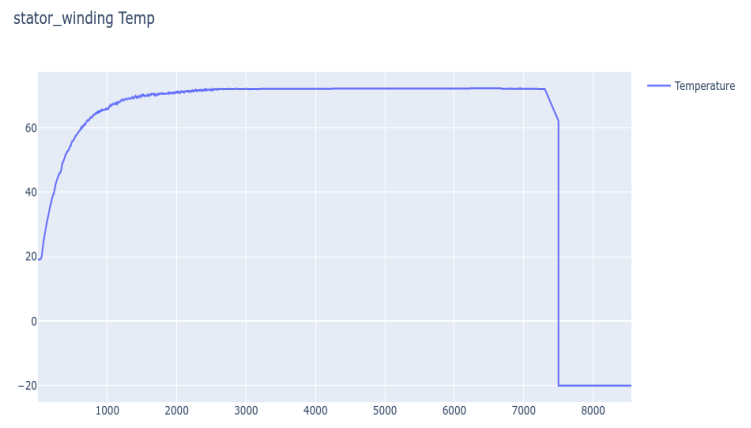


### 2.3.2 Ramp, Fault LOW

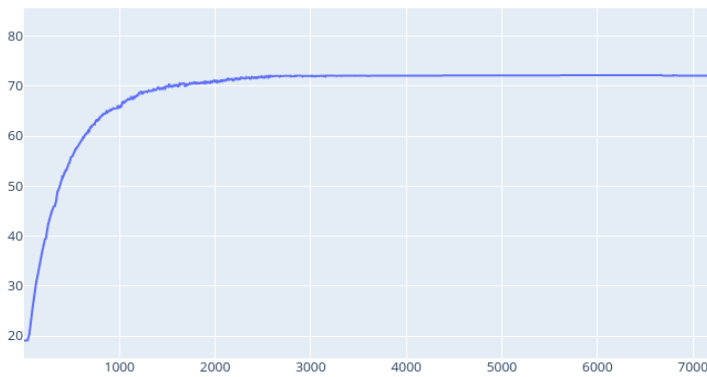
Ramp Error curve: Rising



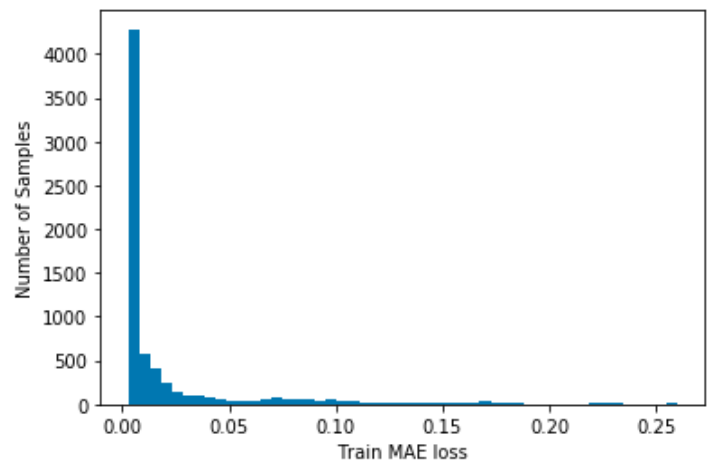
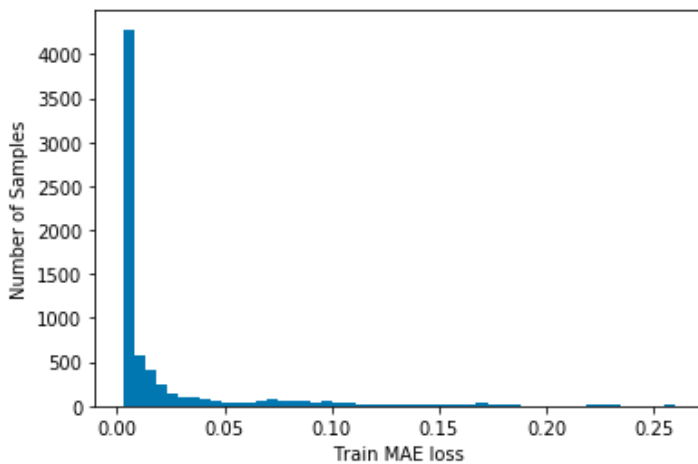
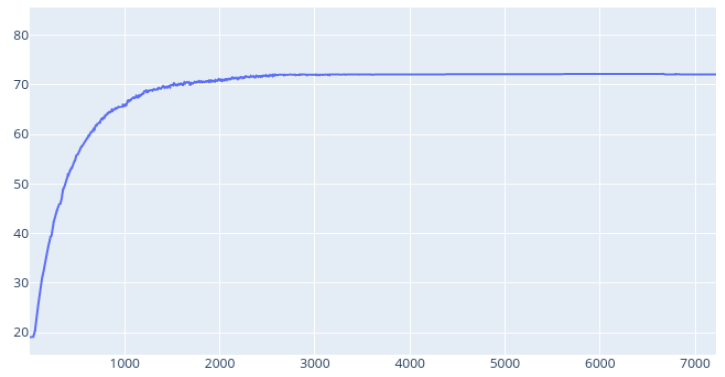
Ramp Error curve: Falling

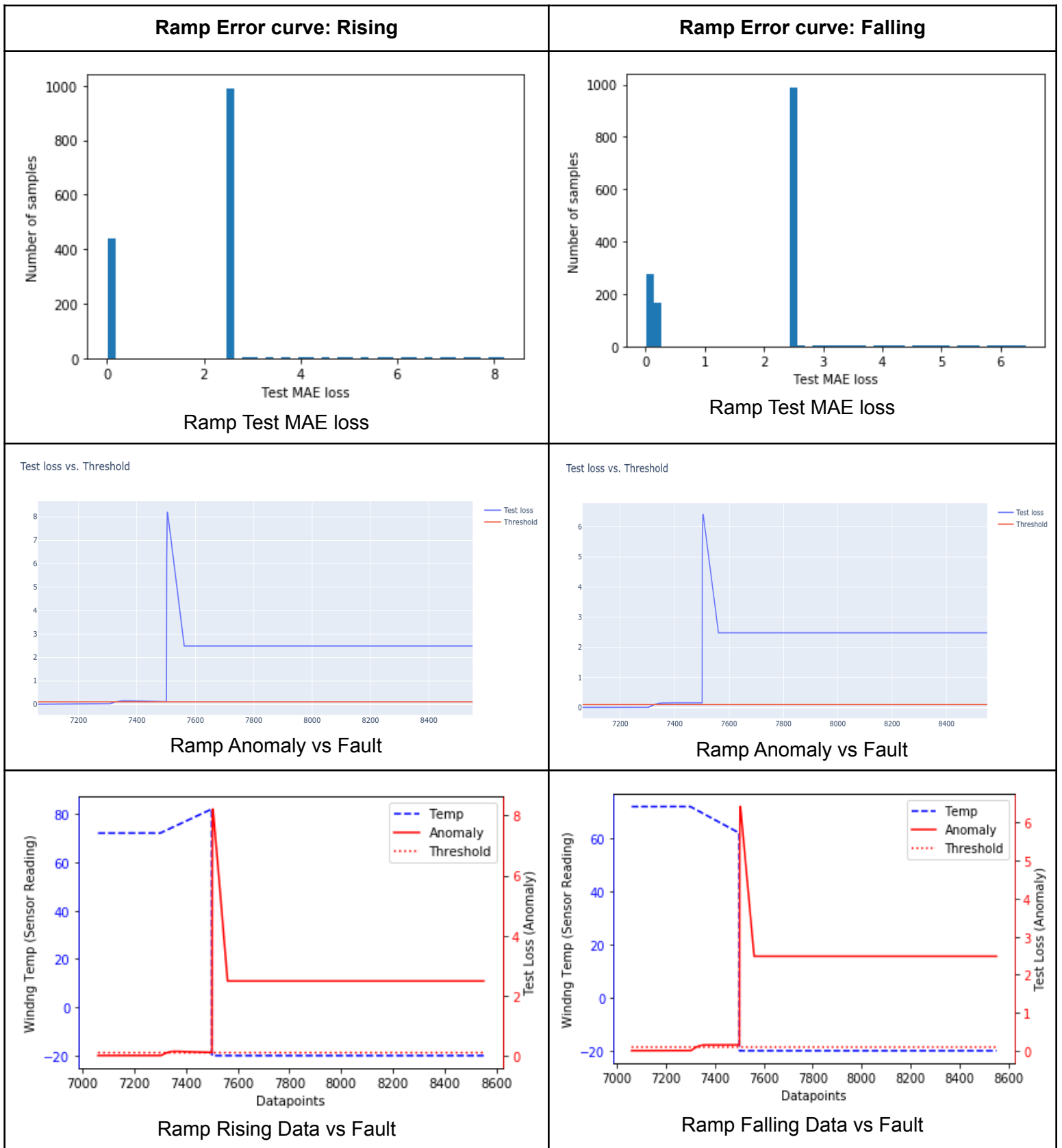


stator\_winding Temp



stator\_winding Temp





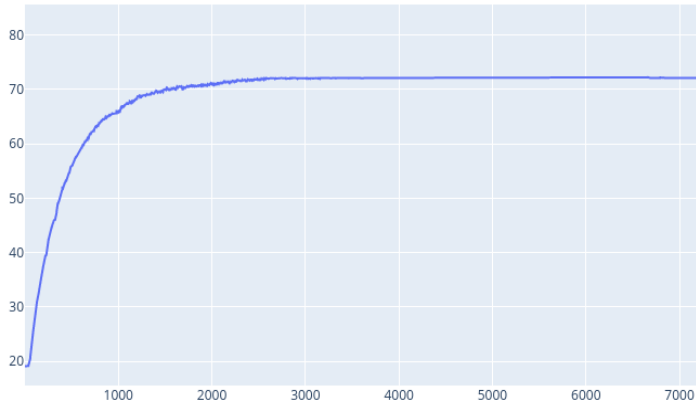
## Part B (LSTM>Error Window)

### 3. Data Vs Anomaly Plots Configuration:

- Dataset: EV stator winding Temp. vs Time (0.5 sec per datapoint)
- LSTM window: 600 data points
- Error window: 200 data points
- Error injected as Sine, Square & Ramp wave
- Error injected as both Positive and Negative values in +Y axis

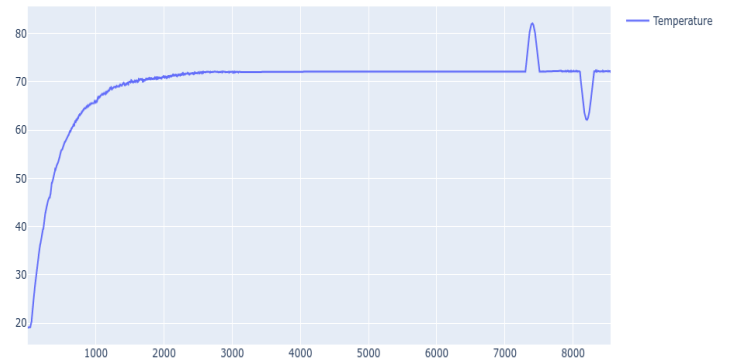
### 3.1 Sine Waveform error

stator\_winding Temp

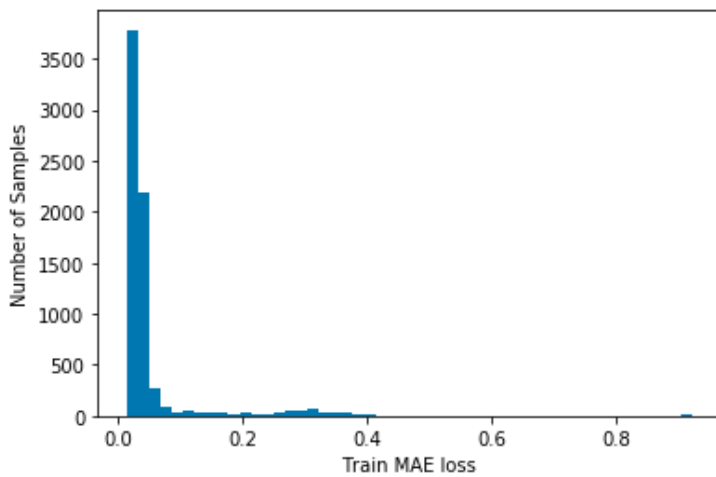


Training Dataset

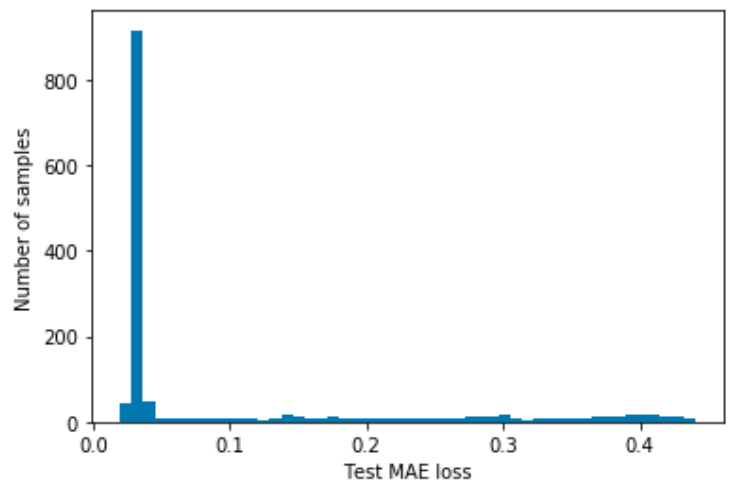
stator\_winding Temp



Sine Dataset

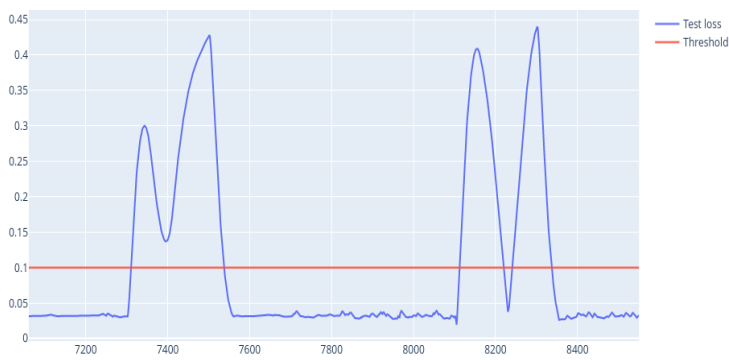


Sine Train MAE loss

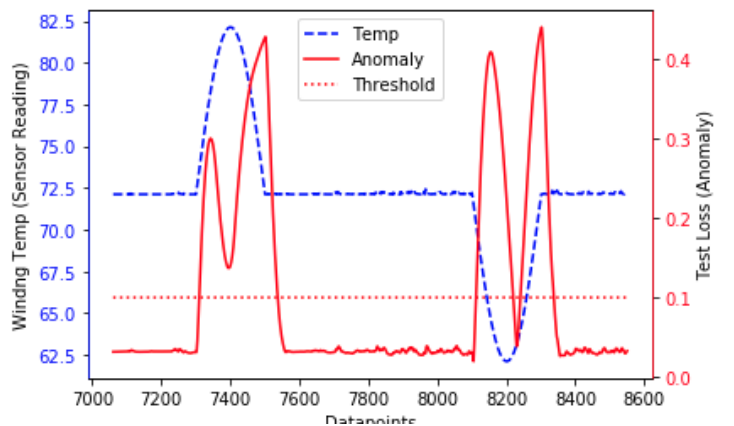


Sine Test MAE loss

Test loss vs. Threshold



Sine Anomaly vs Threshold

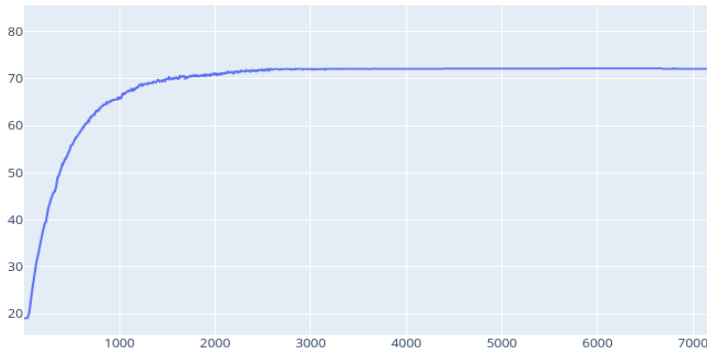


Sine Data vs Anomaly

Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

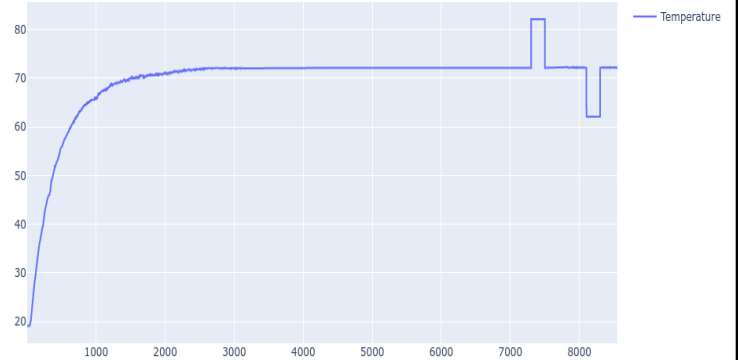
### 3.2 Square Waveform error

stator\_winding Temp

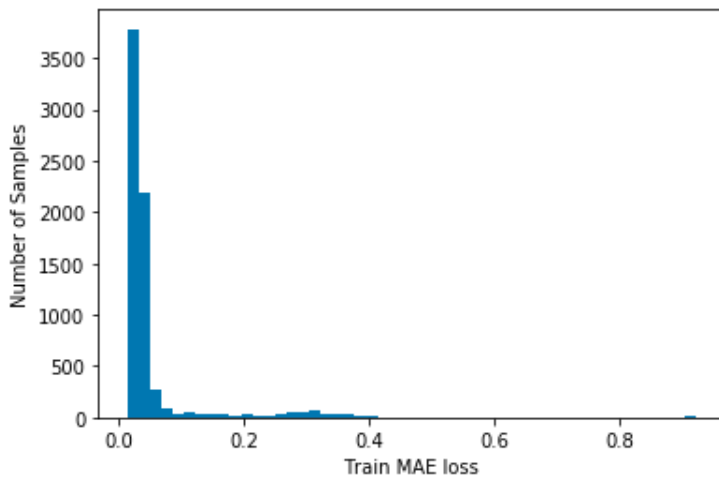


Training Dataset

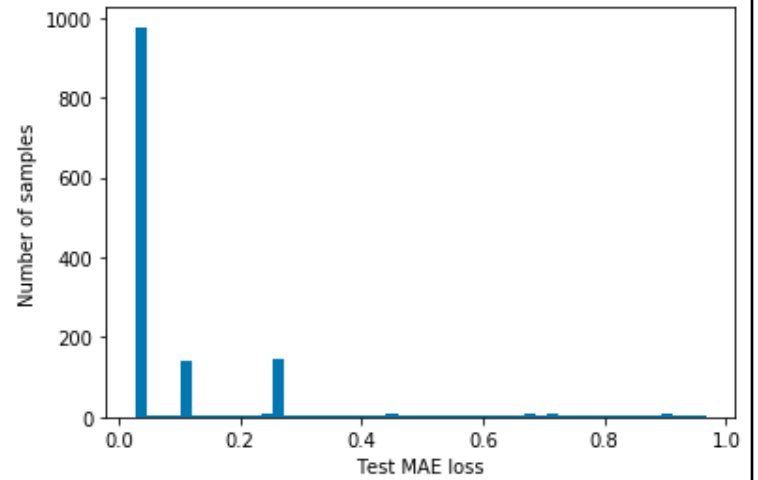
stator\_winding Temp



Square Dataset

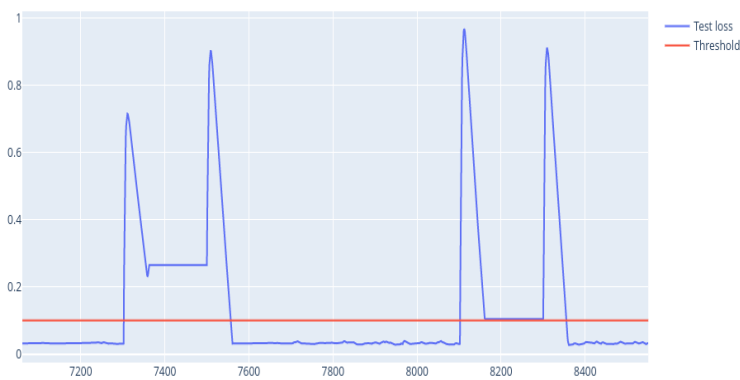


Square Train MAE loss

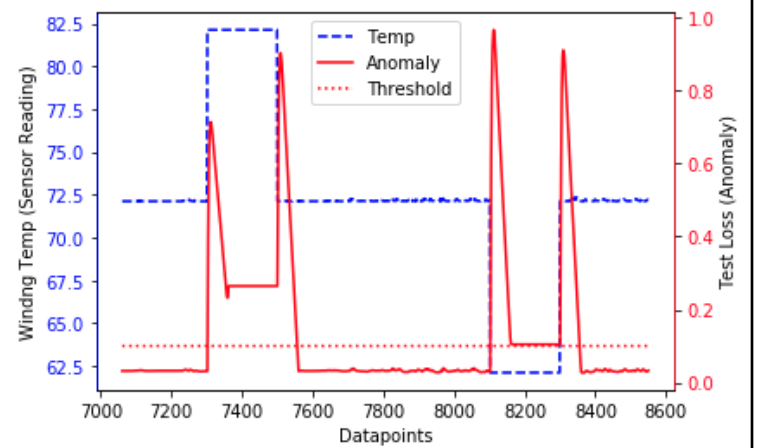


Square Test MAE loss

Test loss vs. Threshold



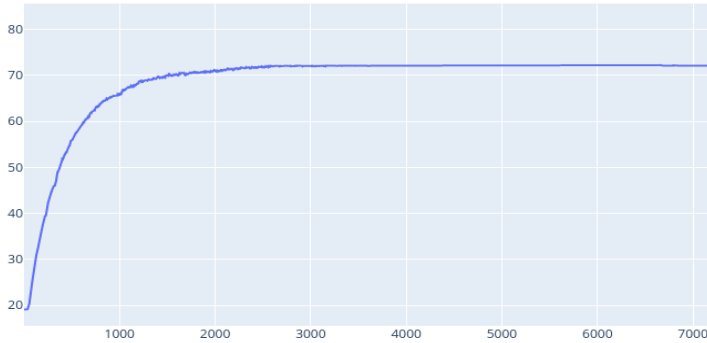
Square Anomaly vs Threshold



Square Data vs Anomaly

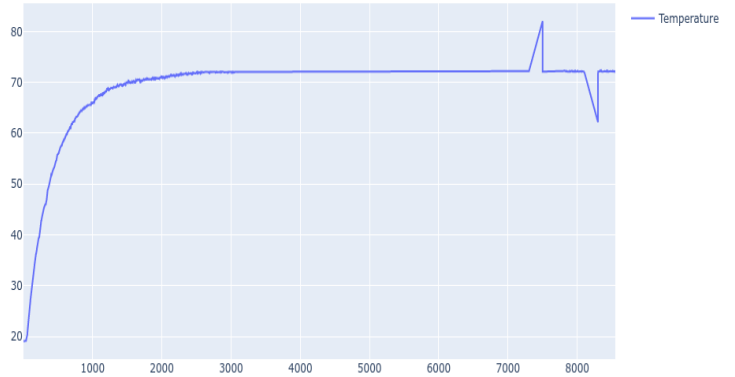
### 3.3 Ramp Waveform error

stator\_winding Temp

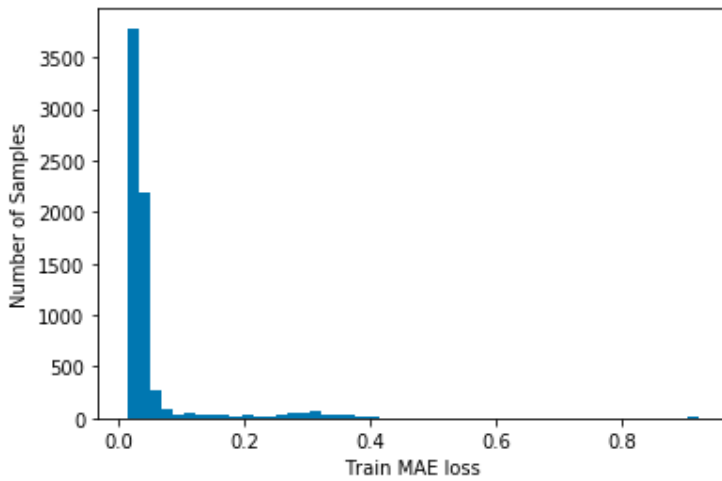


Training Dataset

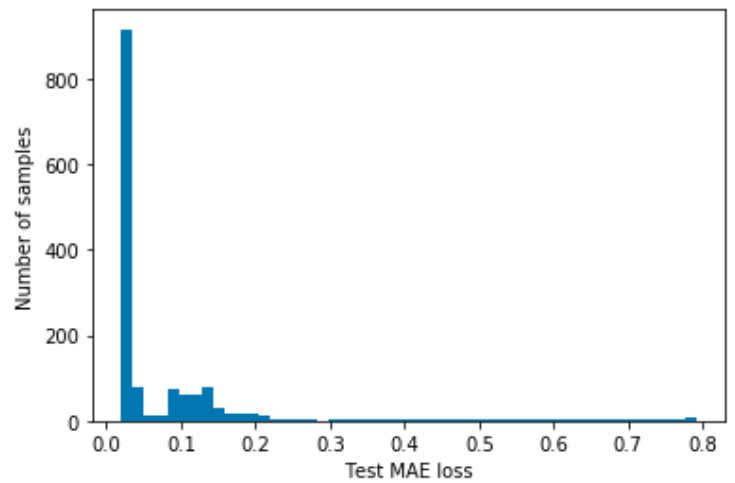
stator\_winding Temp



Ramp Dataset

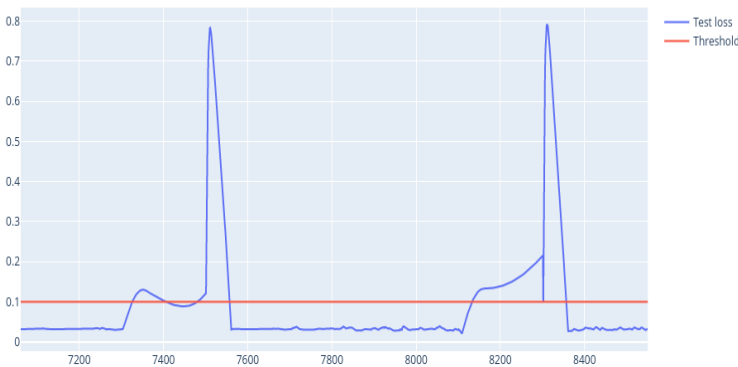


Ramp Train MAE loss

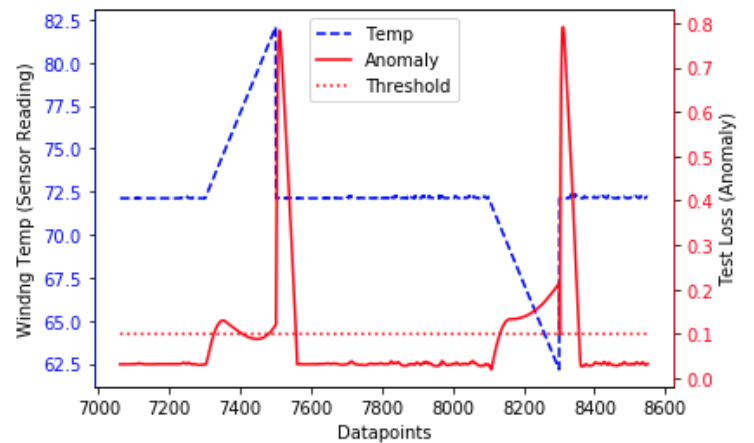


Ramp Test MAE loss

Test loss vs. Threshold



Ramp Anomaly vs Threshold



Ramp Data vs Anomaly

## 4. Data Vs Fault Plots

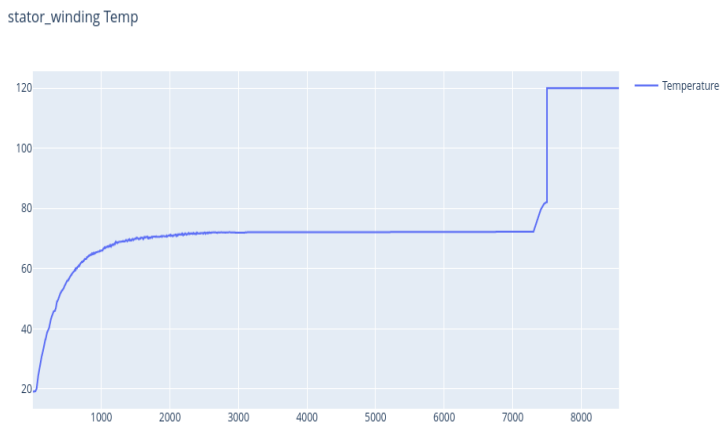
Configuration:

- Dataset: EV stator winding Temp. vs Time (0.5 sec per datapoint)
- LSTM window: 600 data points
- Error window: 200 data points
- Error injected as Sine, Square & Ramp wave
- Error injected as Positive and Negative values in +Y axis separately
- Operating range of the sensor is assumed from 0°C to 100°C,
- When the sensor is faulty, it will send the temp. that is beyond the operating range, i.e. -20°C and 120°C



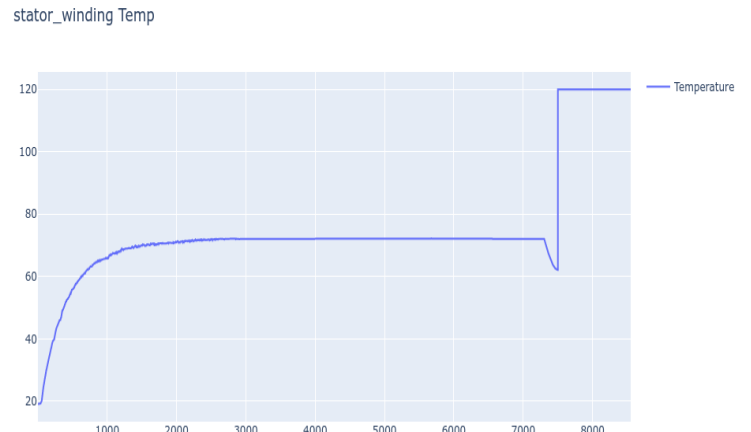
#### 4.1.1 Sine, Fault HIGH

Sine Error curve: Rising



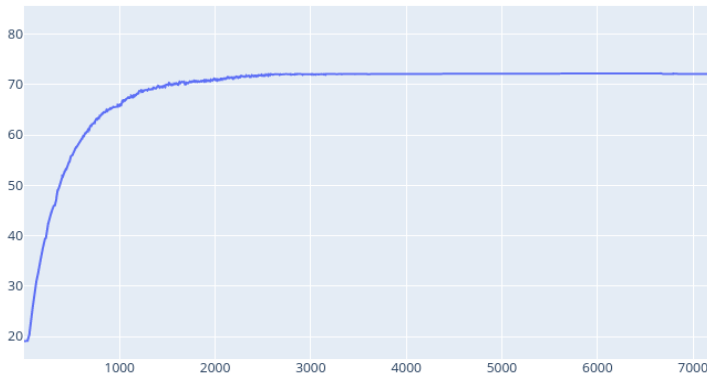
Sine Dataset

Sine Error curve: Falling



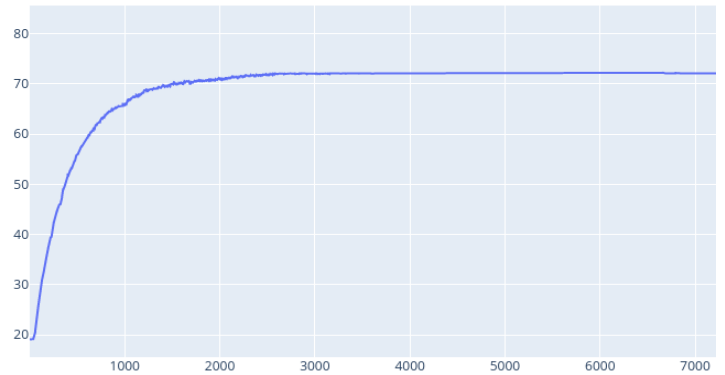
Sine Dataset

stator\_winding Temp

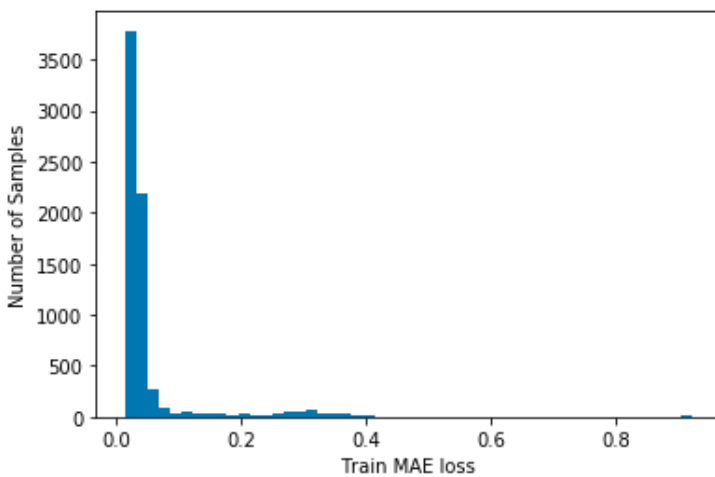


Training Dataset

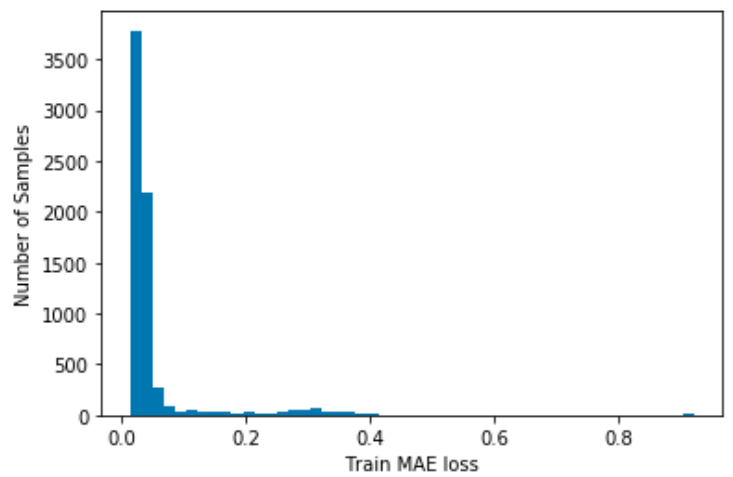
stator\_winding Temp



Training Dataset

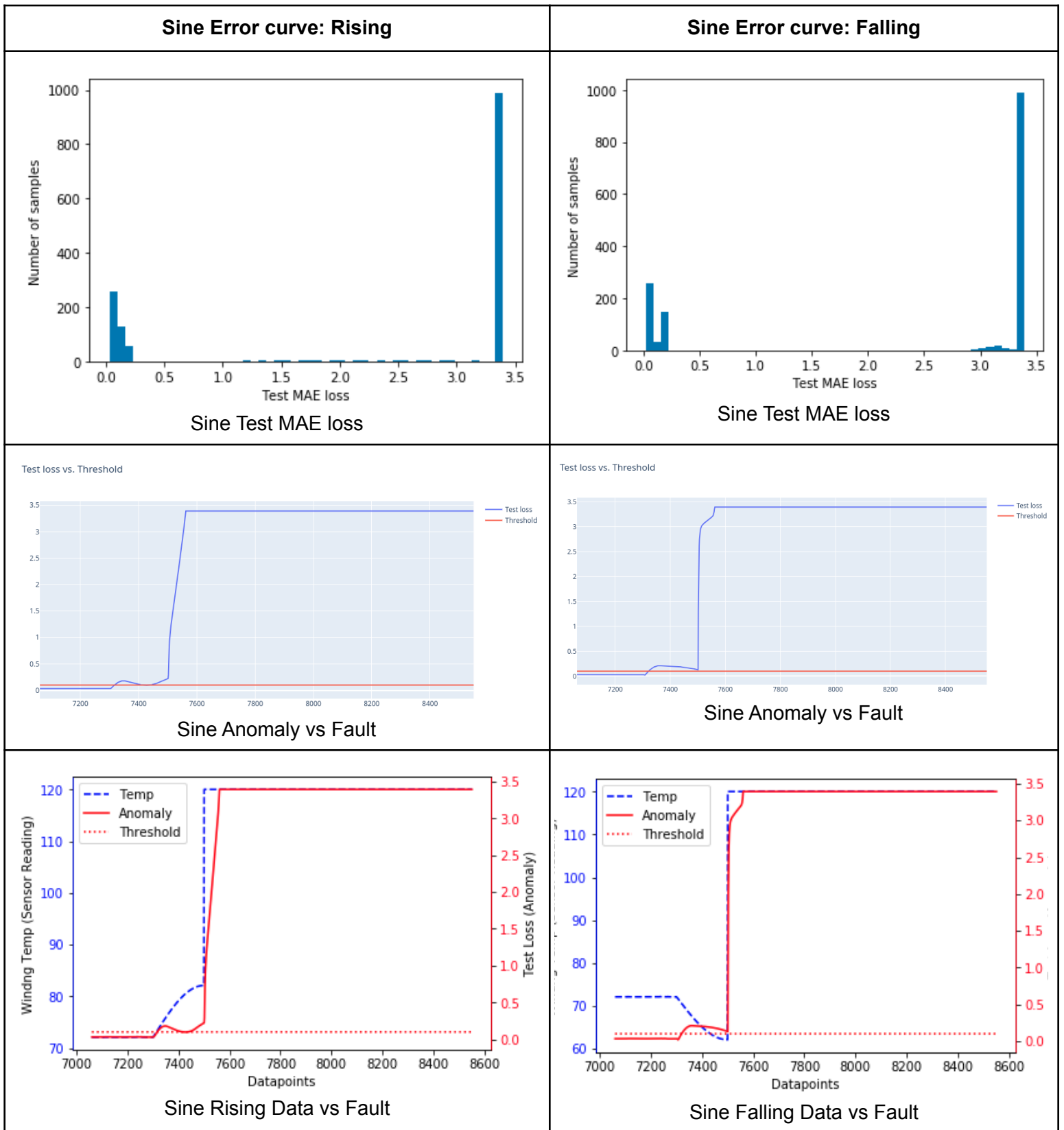


Train MAE loss



Train MAE loss

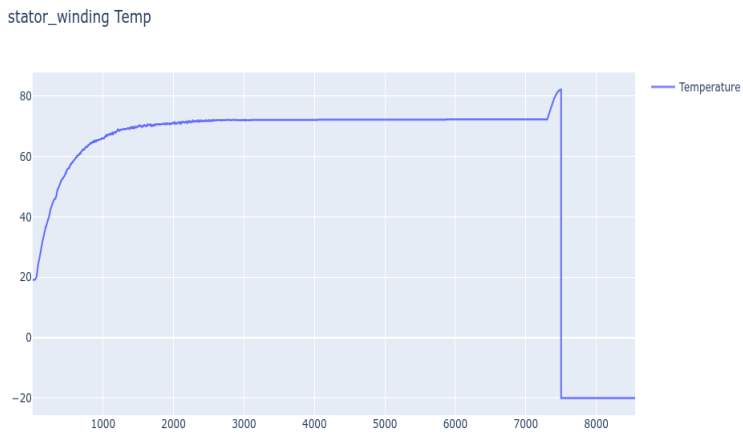
Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>



Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

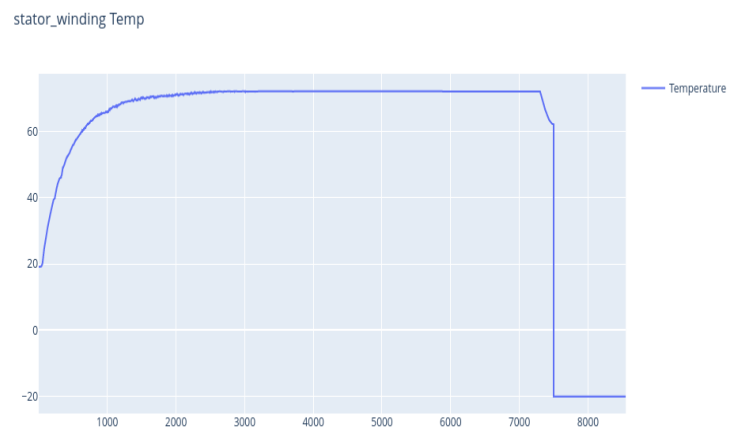
#### 4.1.2 Sine, Fault LOW

Sine Error curve: Rising



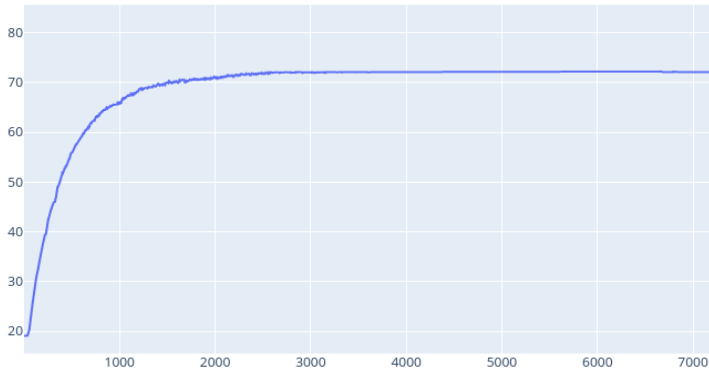
Sine Dataset

Sine Error curve: Falling



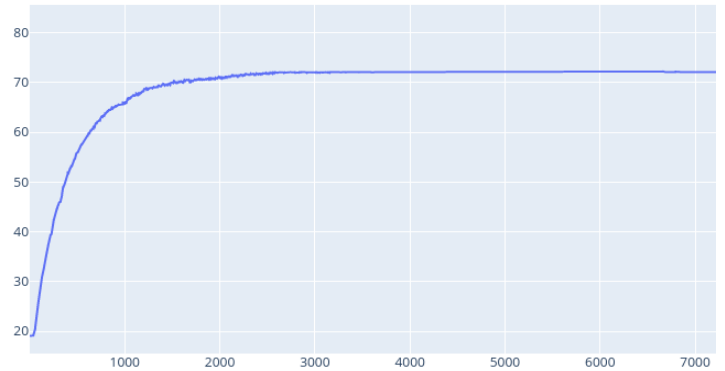
Sine Dataset

stator\_winding Temp

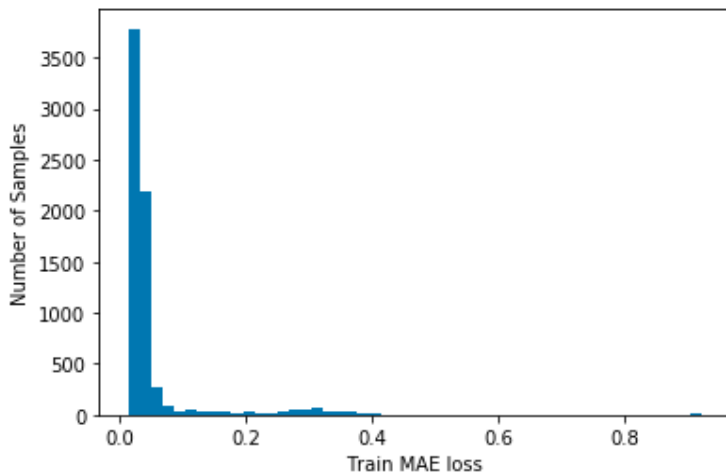


Training Dataset

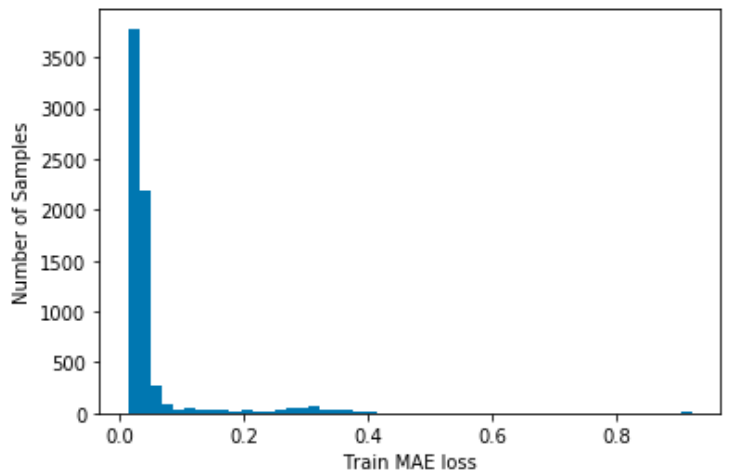
stator\_winding Temp



Training Dataset

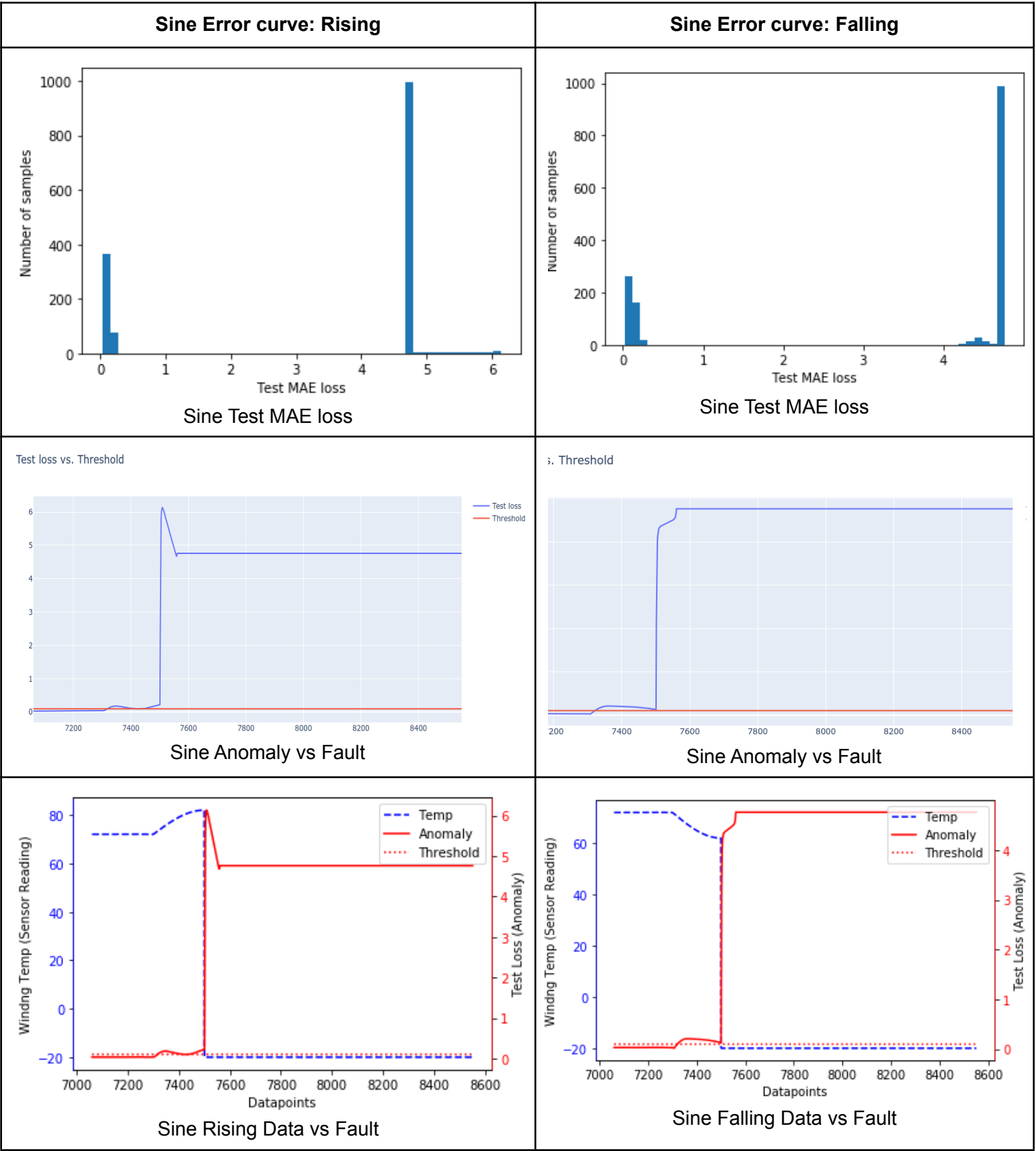


Train MAE loss



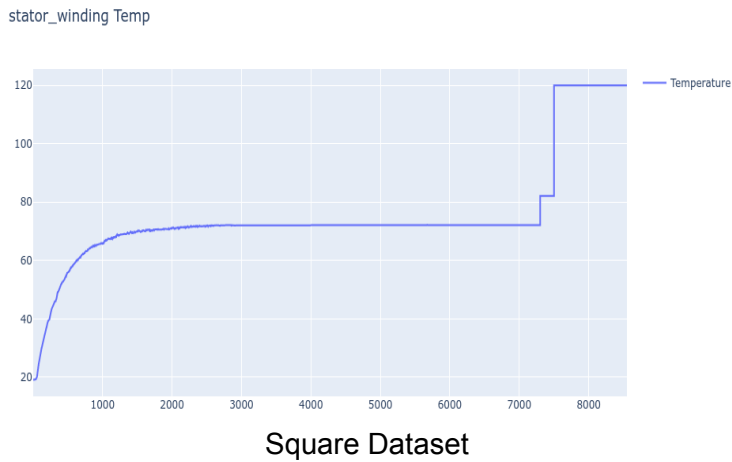
Train MAE loss

Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

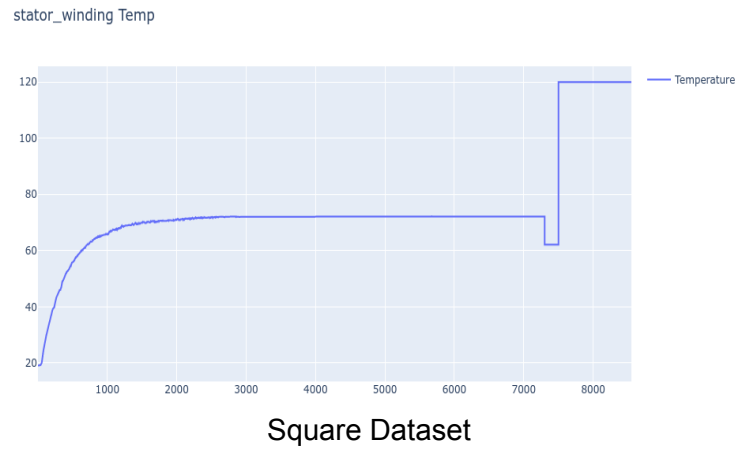


### 4.2.1 Square, Fault HIGH

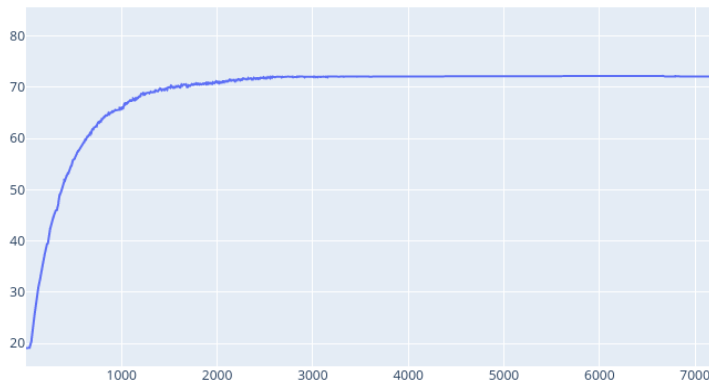
**Square Error curve: Rising**



**Square Error curve: Falling**

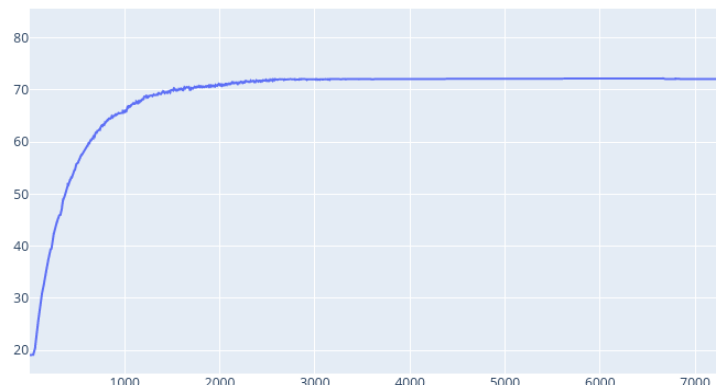


stator\_winding Temp

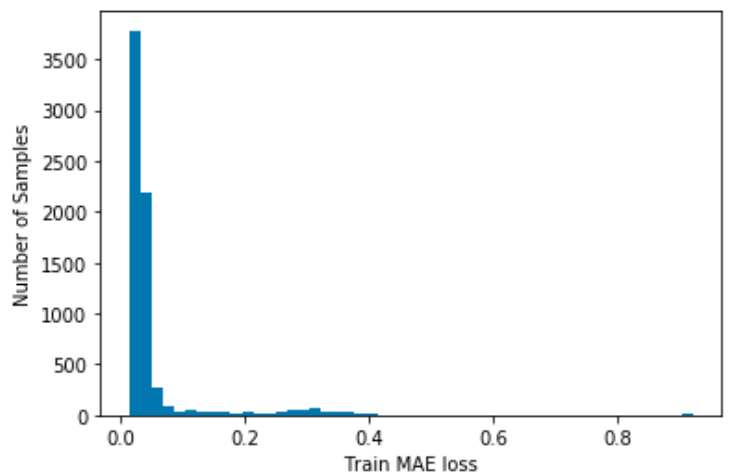
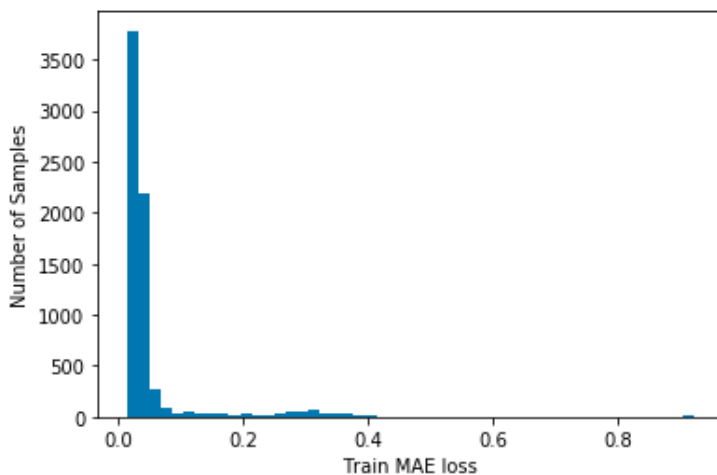


Training Dataset

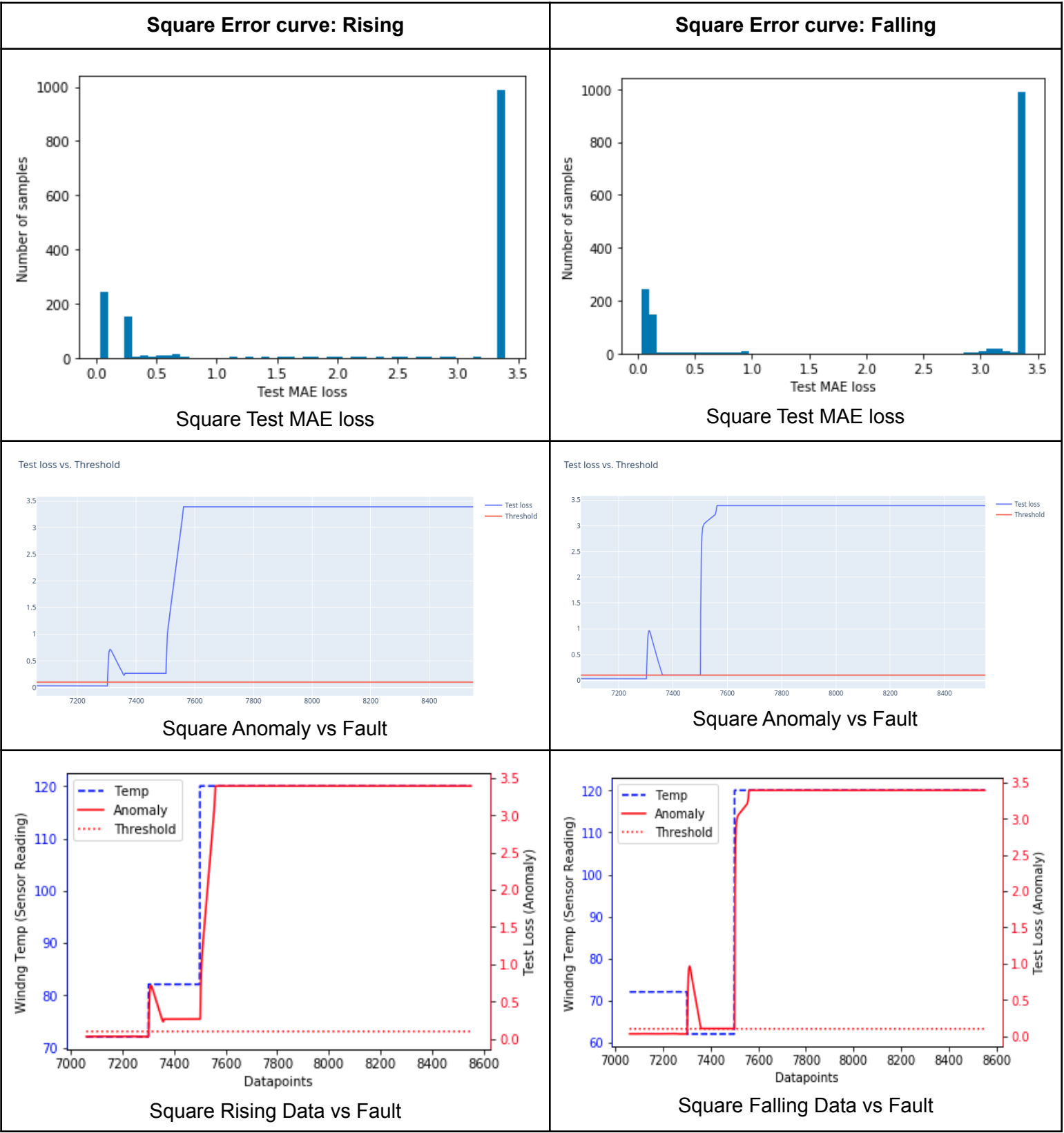
stator\_winding Temp



Training Dataset

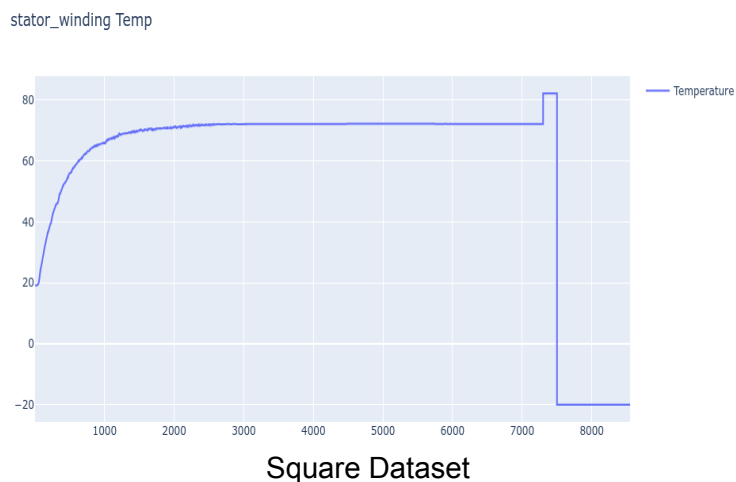


Images: <https://github.com/biplabro/Anomaly-Detection-LSTM-AutoEncoder/tree/master/Summary-Images>

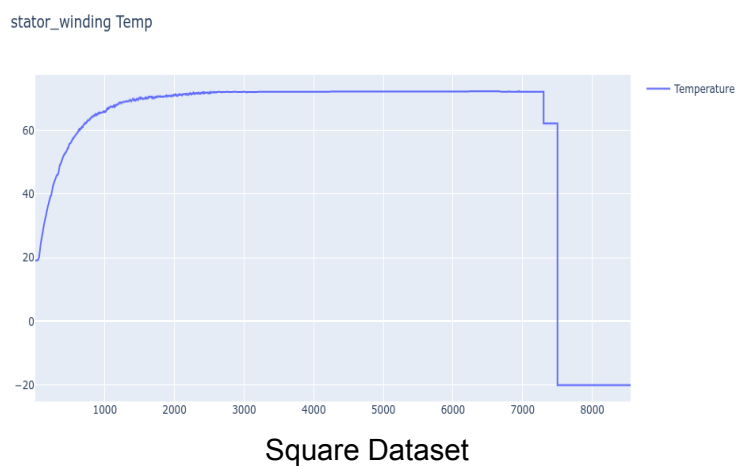


## 4.2.2 Square, Fault LOW

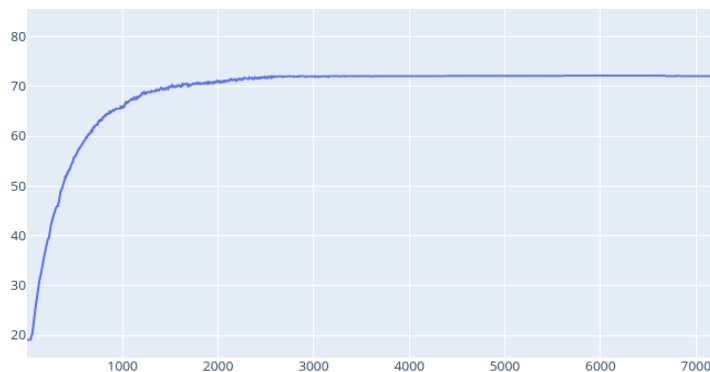
**Square Error curve: Rising**



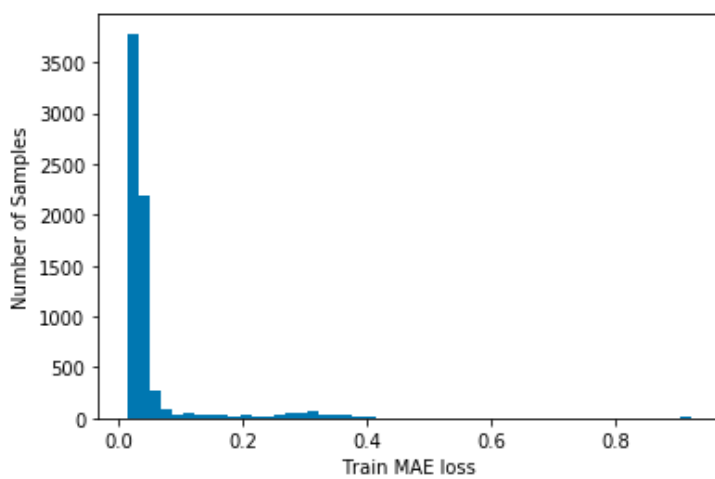
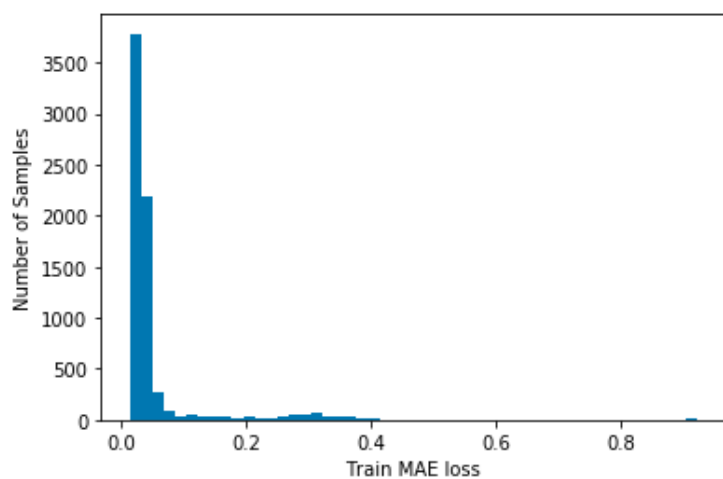
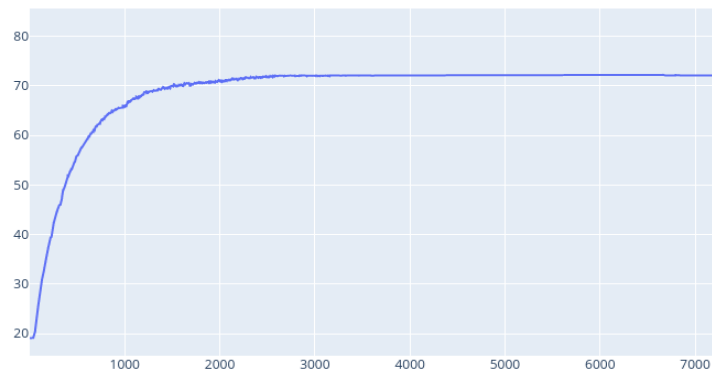
**Square Error curve: Falling**

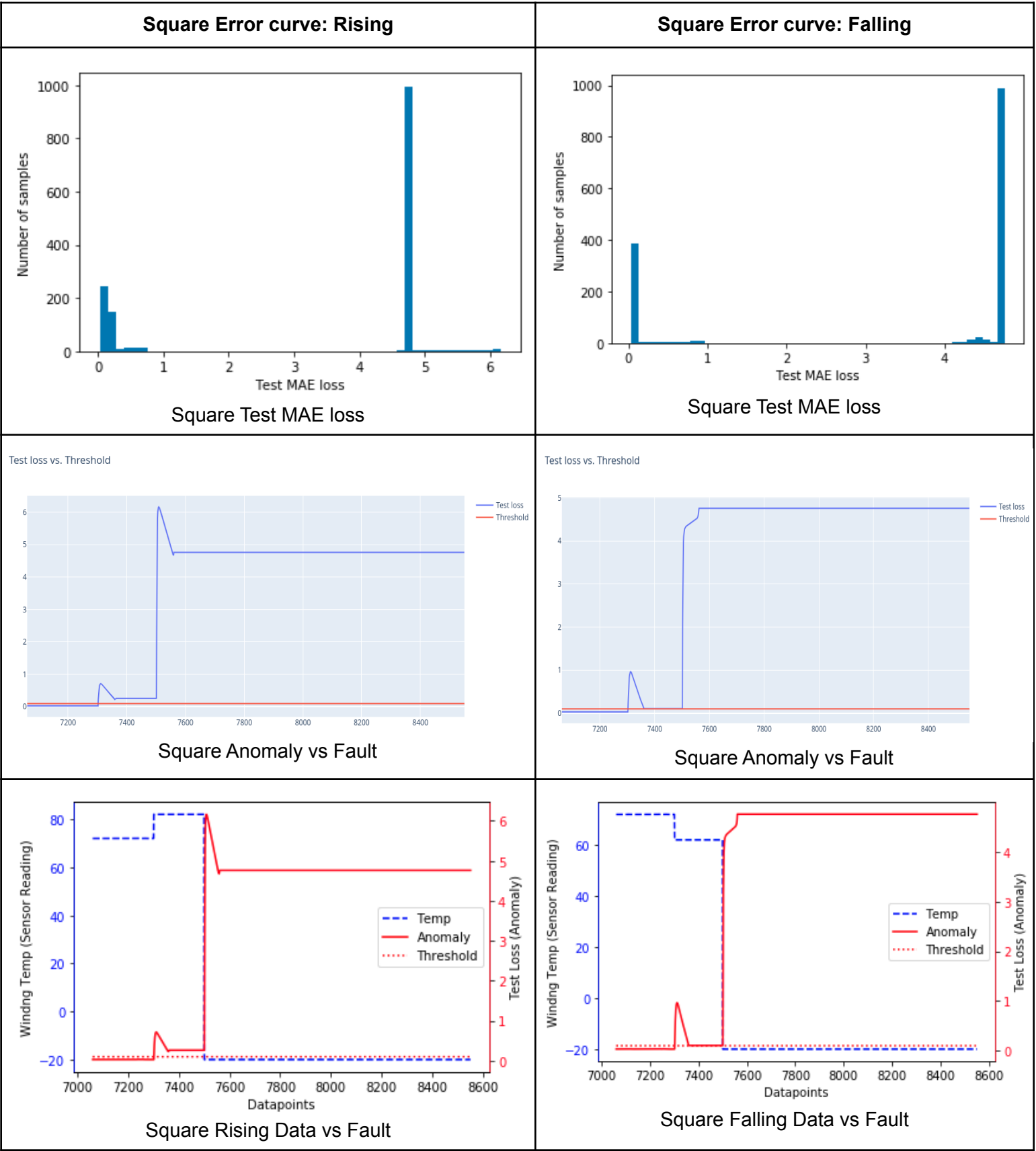


stator\_winding Temp



stator\_winding Temp

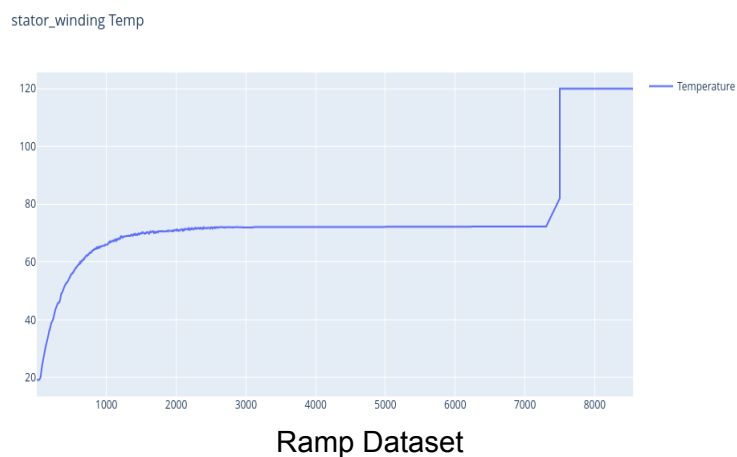




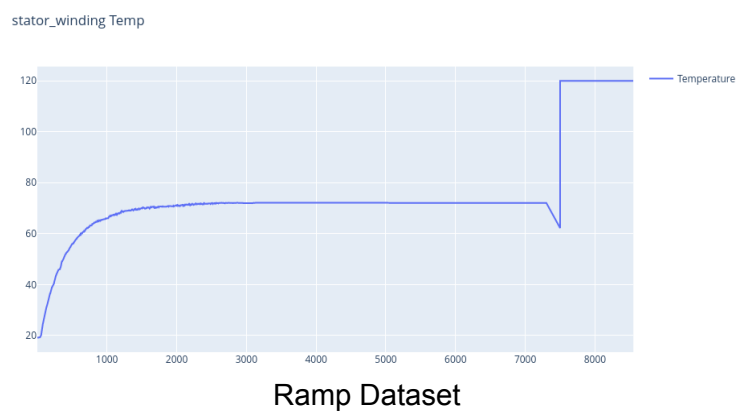


### 4.3.1 Ramp, Fault HIGH

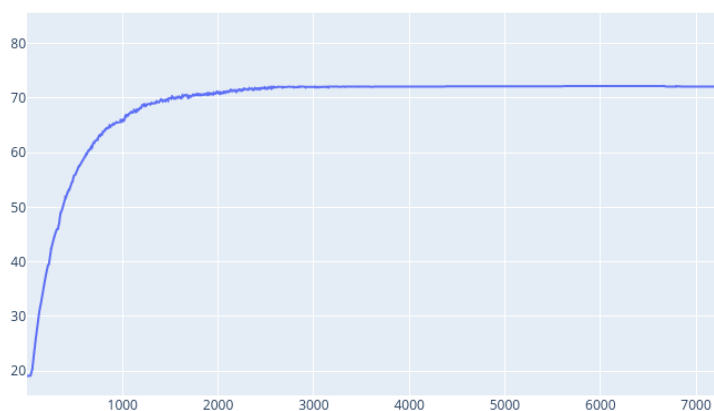
Ramp Error curve: Rising



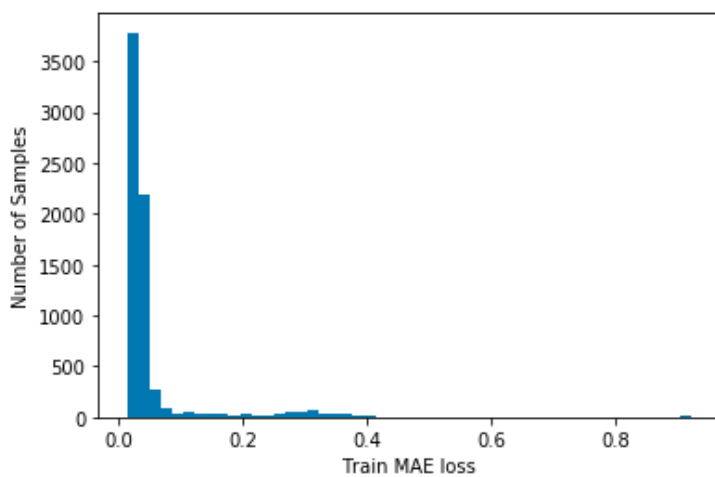
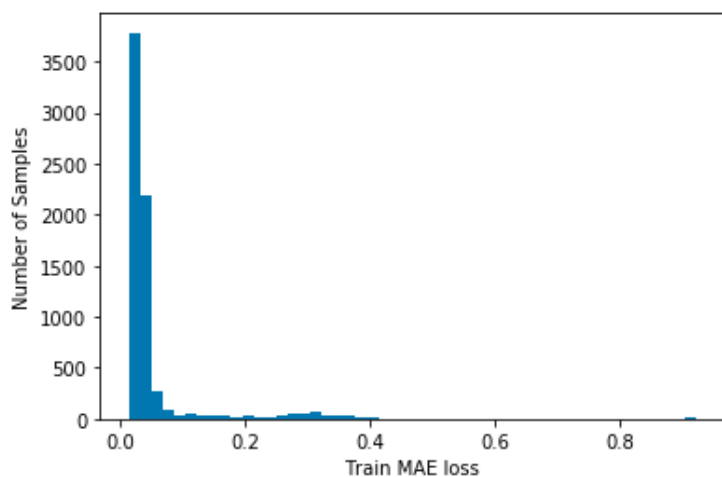
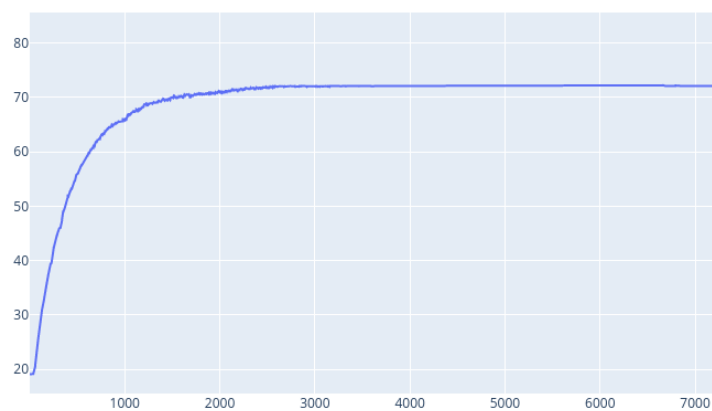
Ramp Error curve: Falling

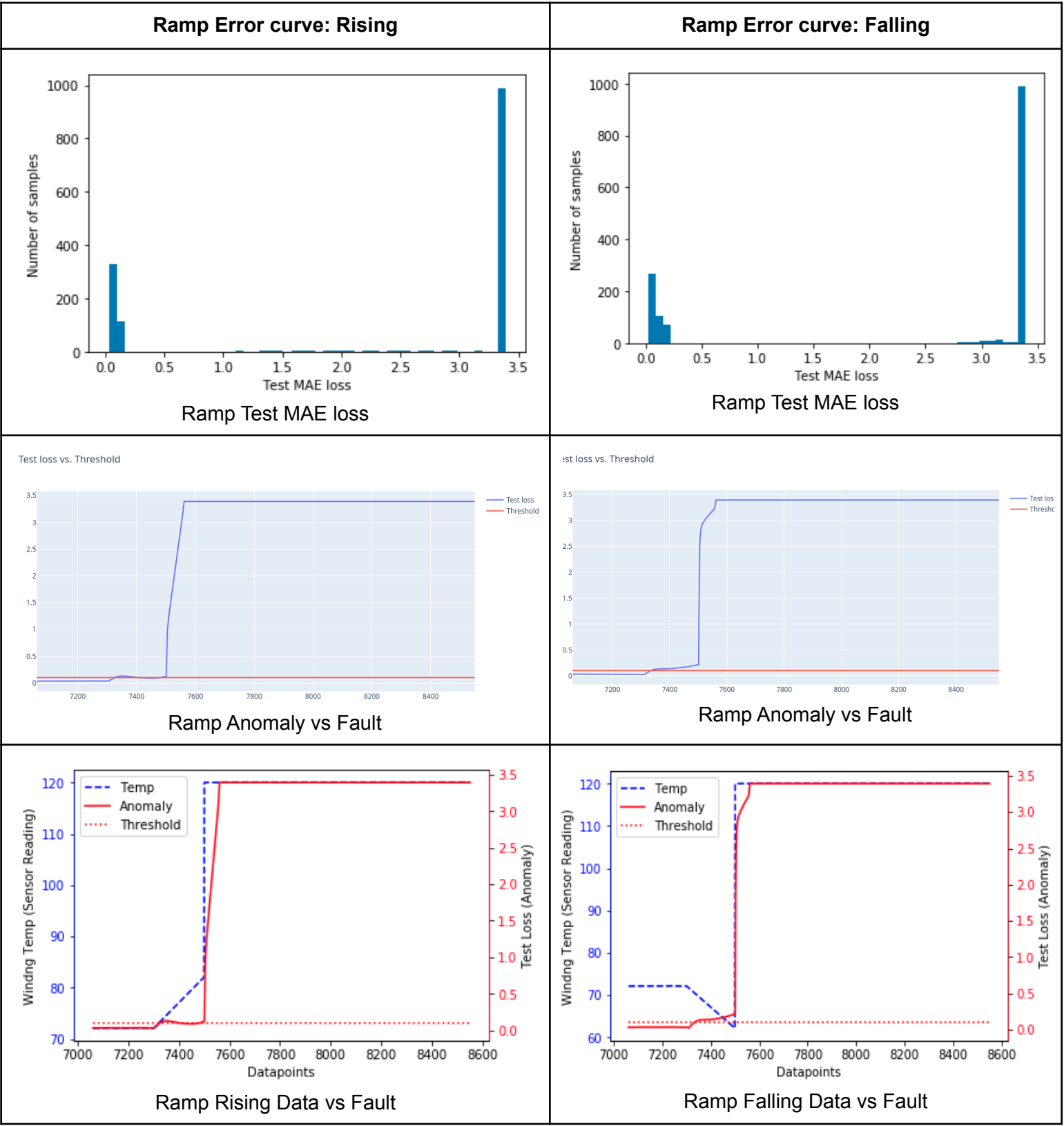


stator\_winding Temp



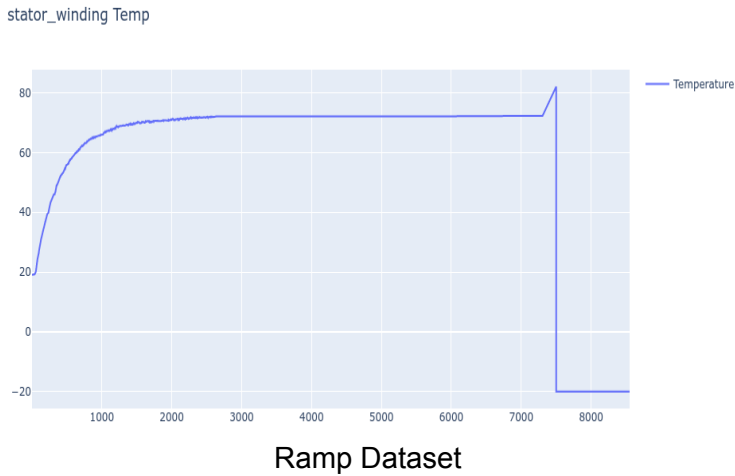
stator\_winding Temp



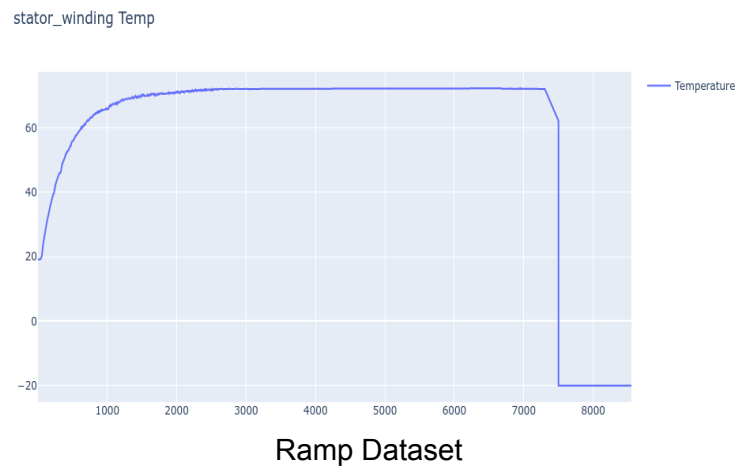


### 4.3.2 Ramp, Fault LOW

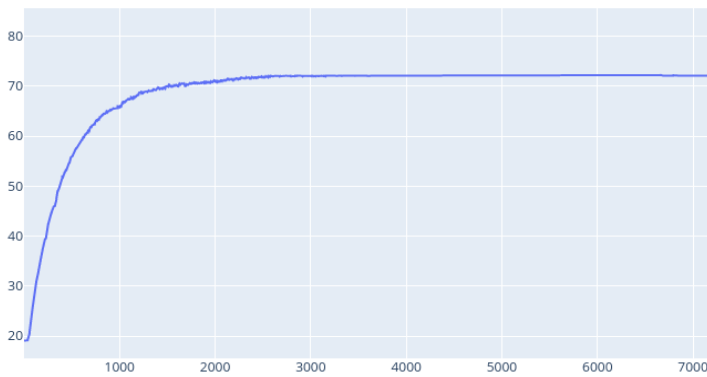
Ramp Error curve: Rising



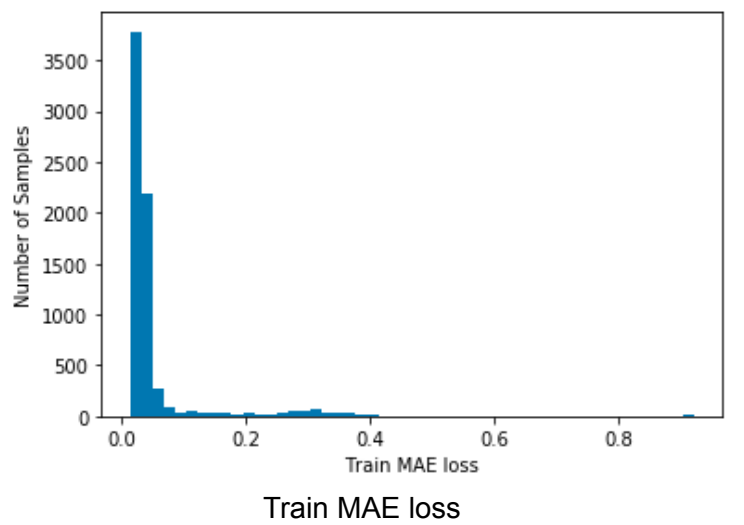
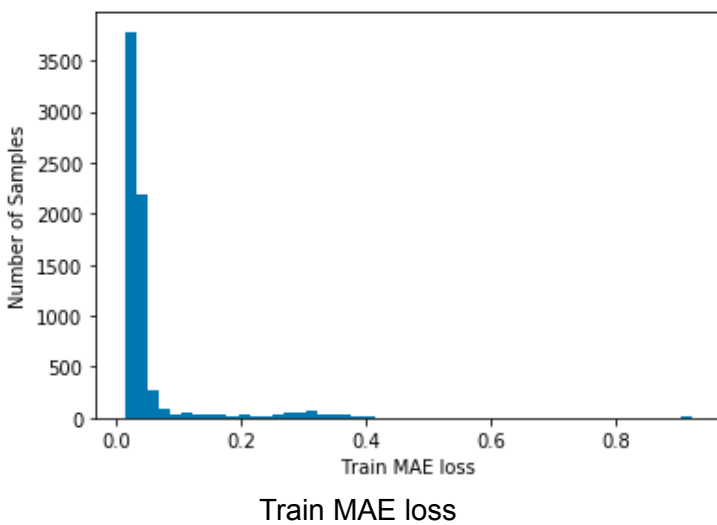
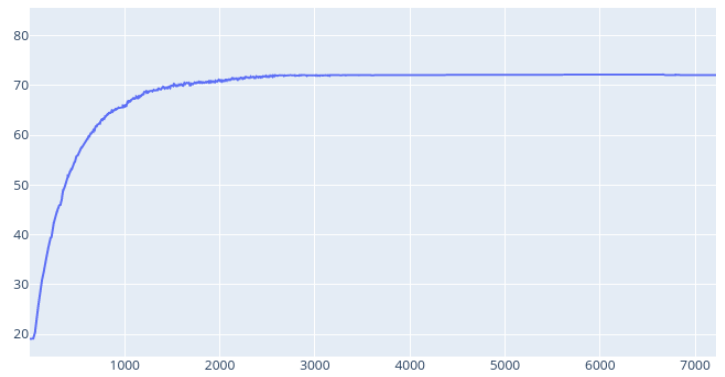
Ramp Error curve: Falling

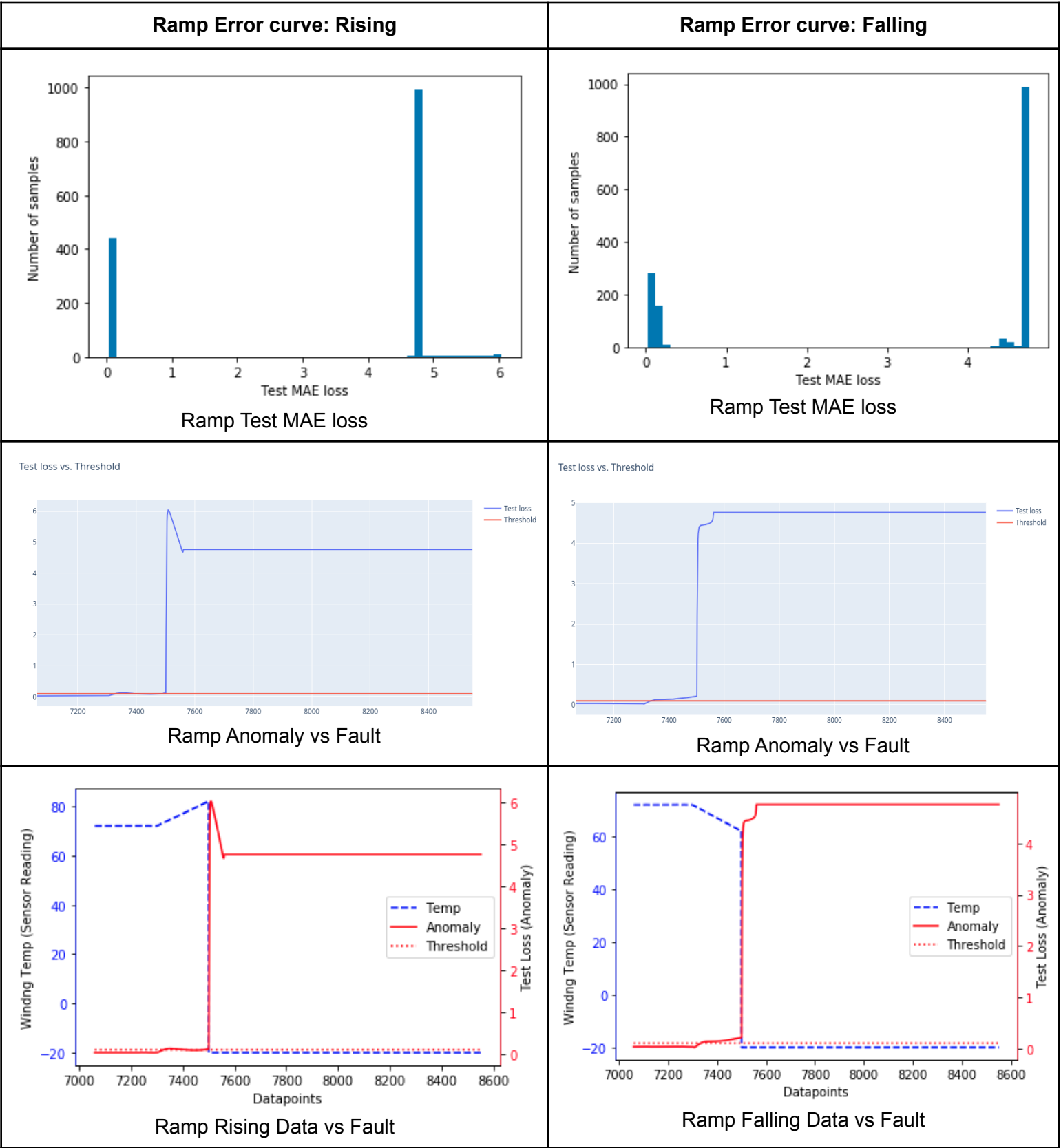


stator\_winding Temp



stator\_winding Temp





## Observations:

- Increased LSTM window leads to decreased training time
- Ramp anomaly Notch at datapoint 8302 (table 3.3)
- LSTM window detects the anomaly in the rising or falling trend of the error pattern.
- Most useful for detecting and confirming if there is an anomaly started or stopped
- Not much useful for calculating the degree of anomaly or how severe the anomaly is.
- LSTM window is specially useful & most effective in detecting the peak or sudden reversal of an error pattern while dealing with time series values.