



Utilizing Machine Learning for Anomaly Detection

By: Ryan Park

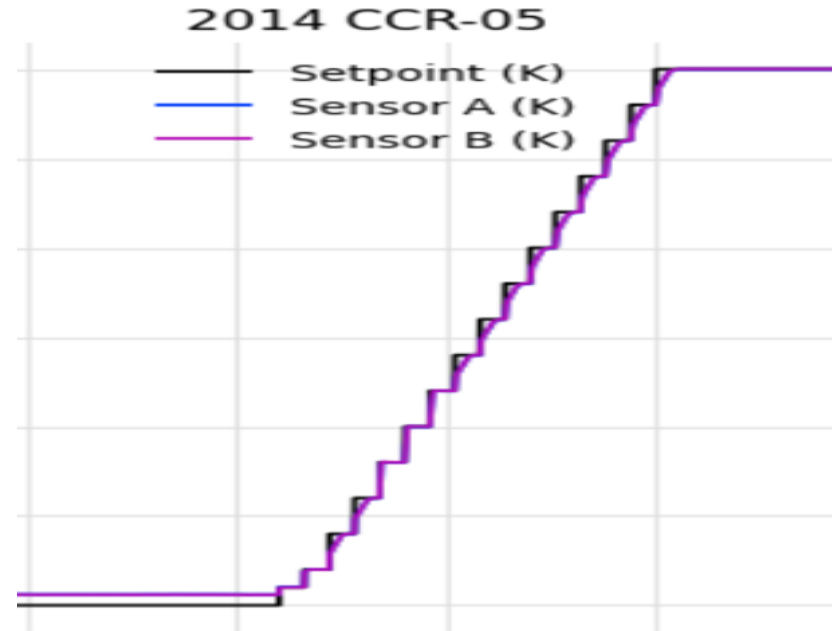
Thomas Jefferson High School for Science and Technology

Mentor: Paul Kienzle



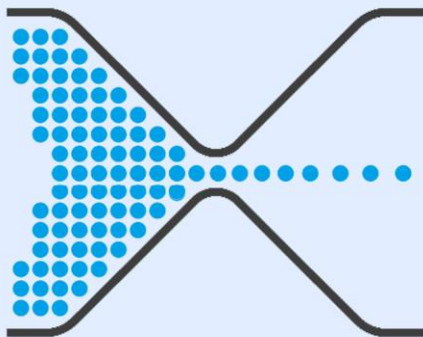
Problem

- Closed Cycle Refrigerator (CCR)
- How does it currently work?
 - Control System
 - Setpoint
 - Sensor A, B, C, D
 - Manual troubleshooting (no notification or pinpoint system)





Unsupervised Machine Learning

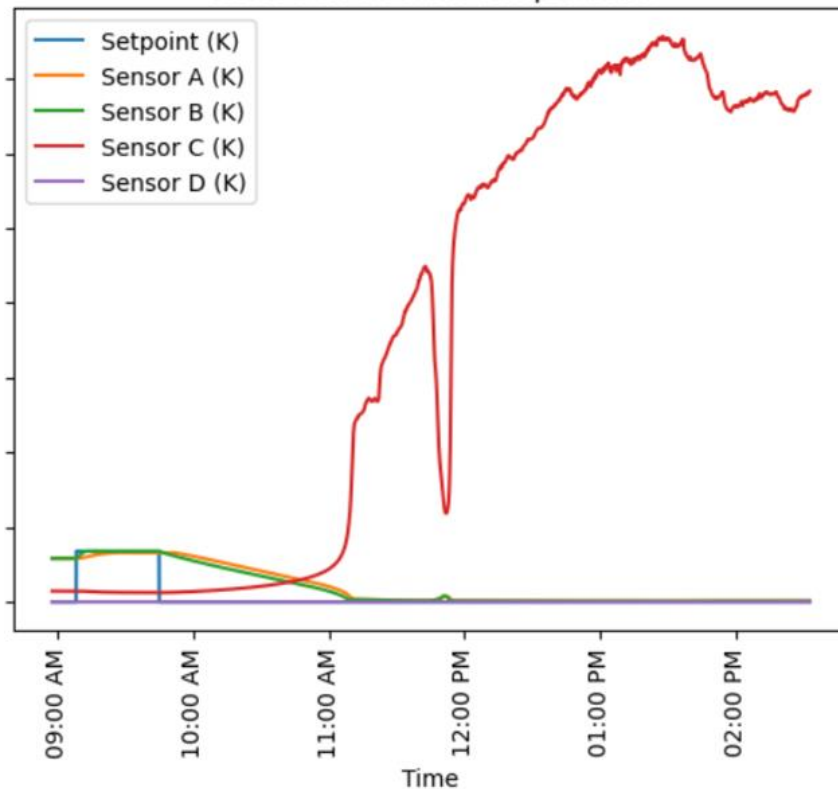


- Building an internal representation of “normal behavior”
- Identify things that are completely new and never before seen as anomalous behavior

Objective

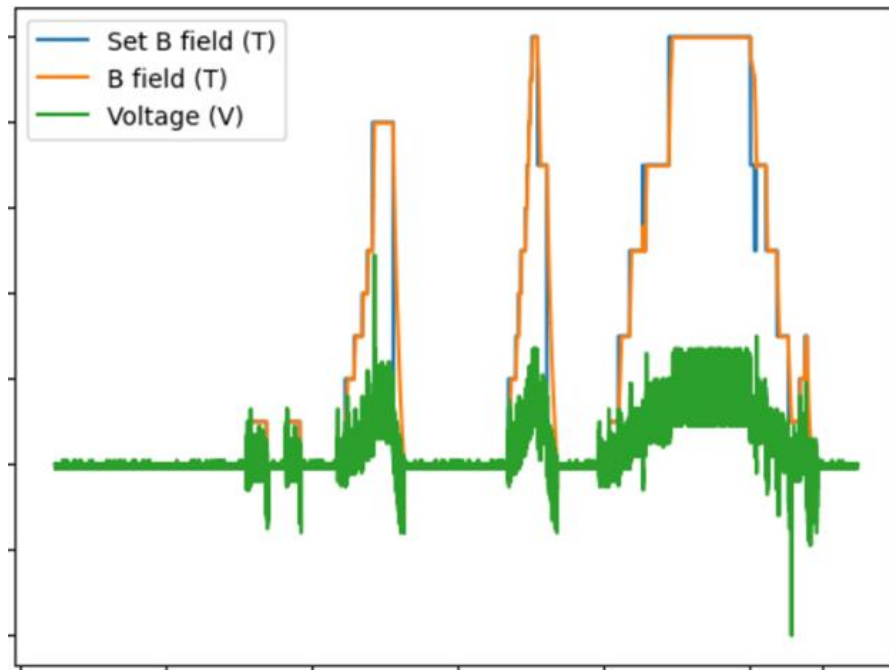
- Build and train a transformer model for sequence prediction
- Long Term: Tracks and predicts every available control parameter and sensor across the facility for anomalies

2021 CCR-08 Base Temperature

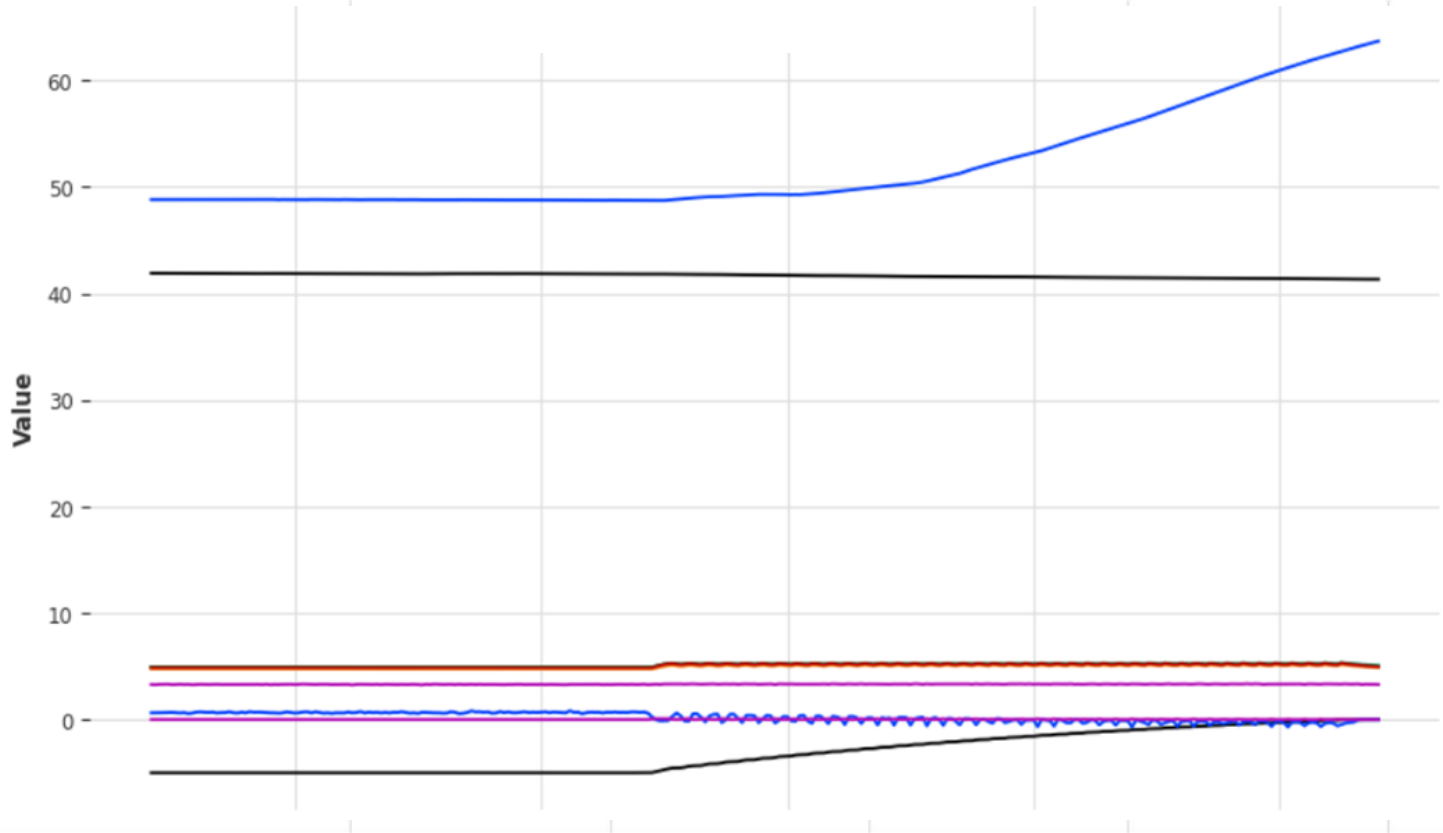


Transfer Learning

- A machine learning technique that uses knowledge learned from one task to improve performance on a related task
- Magnetism Dataset: comprised of exclusively normal CCR data
 - 30 individual experiments

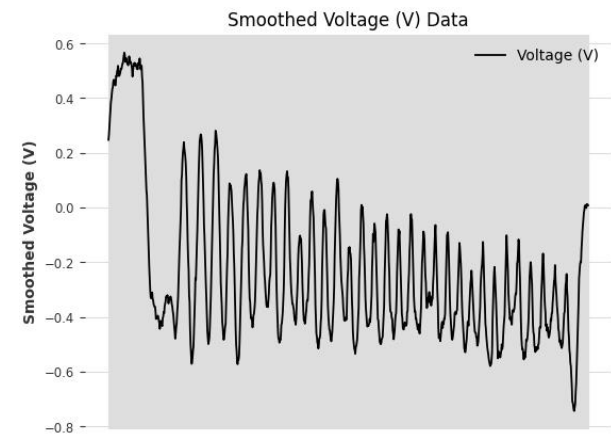
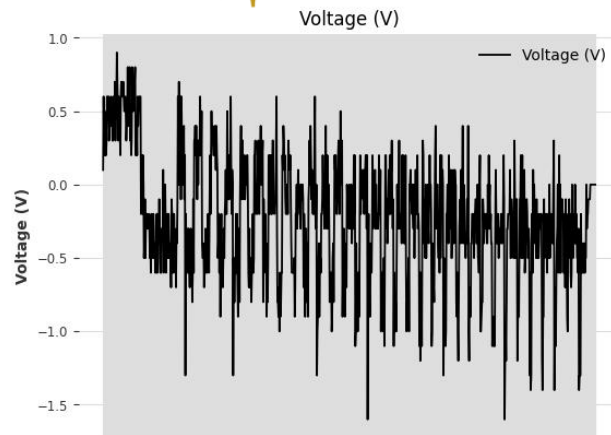


Actual



Feature Engineering

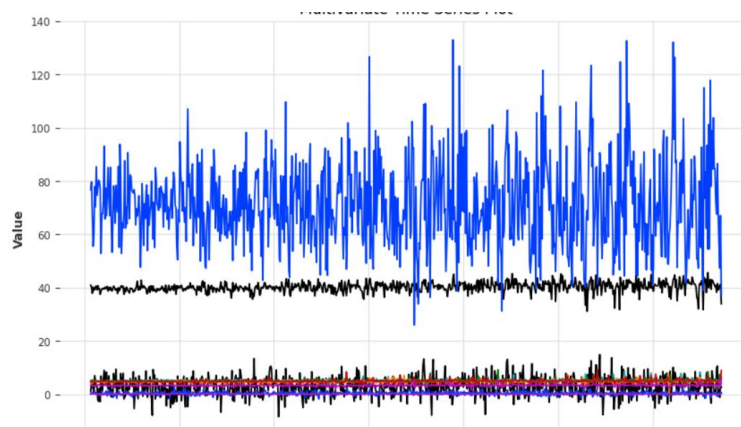
- Normalization (respective to each experiment)
- Upsample (1 minute)
 - Slicing into 2000 timestamps per time series
- Savitzky-Golay Filter
 - Smoothes Data



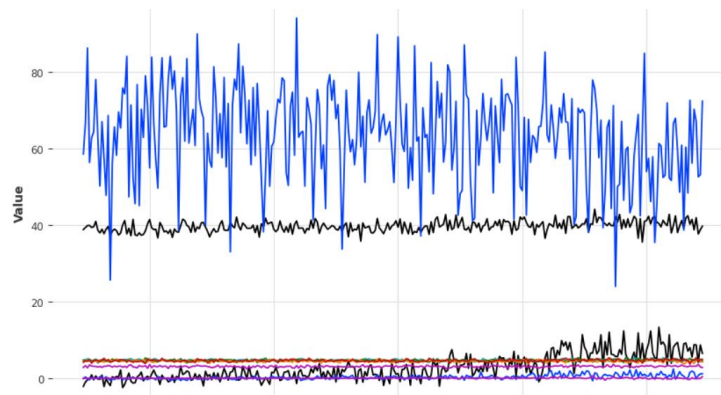


Results

197% SMAPE Score



45% SMAPE Score





Future Work

- Data Augmentation
 - Introduce Gaussian noise in conjunction with Savitzky-Golay Filter
- Known Unknown: trained signal used for labelled anomaly detection
- Next Steps:
 - Create a supervised machine learning model to detect anomalies
 - Identify frequent anomalies

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