



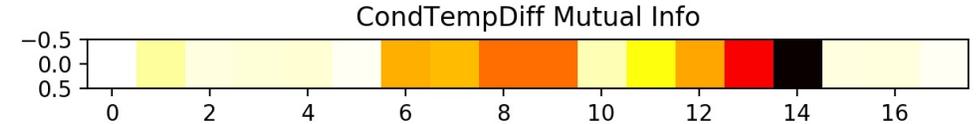
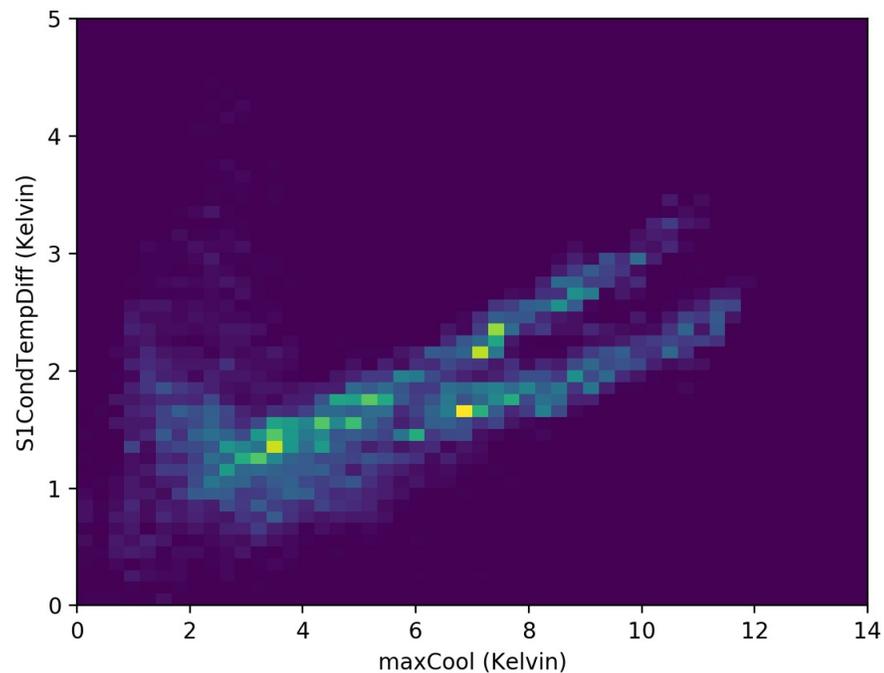
# Anomaly Detection in a Cooling Tower System

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# Data extraction and Pre-Processing

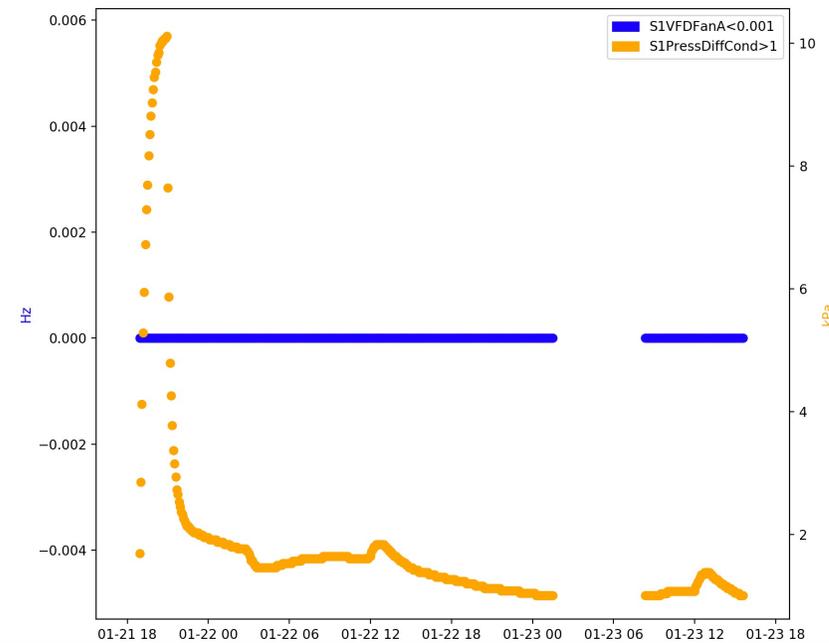
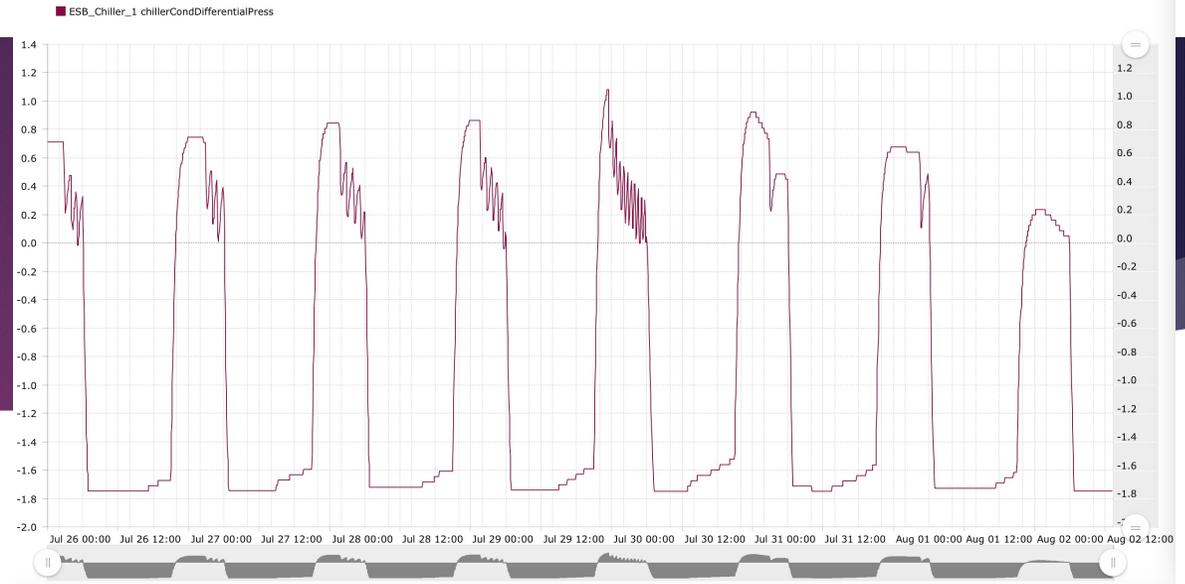
- ▶ Bad data, Noise data, and irrelevant data removed
- ▶ Analyzed how Different variables correlated and affected each other.



- 0 = S1CondTempDiff
- 1 = TempAmbient
- 2 = TempWetBulb
- 3 = PerHumidity
- 4 = S1TempCondIn
- 5 = S1TempCondOut
- 6 = S1PowFanA
- 7 = S1PowFanB
- 8 = S1VDFFanA
- 9 = S1VDFFanB
- 10 = S1PressDiffCond
- 11 = S1PowChil
- 12 = S1PowPump
- 13 = DaySin
- 14 = MonthSin
- 15 = effCool
- 16 = effCoolNext
- 17 = maxCool

# Analysis

- ▶ Certain variables that displayed unusual behavior in certain time periods were analyzed.
- ▶ Certain behaviors were deemed to be normal behavior or human-caused behavior, while others were possible anomalies.



# Prediction

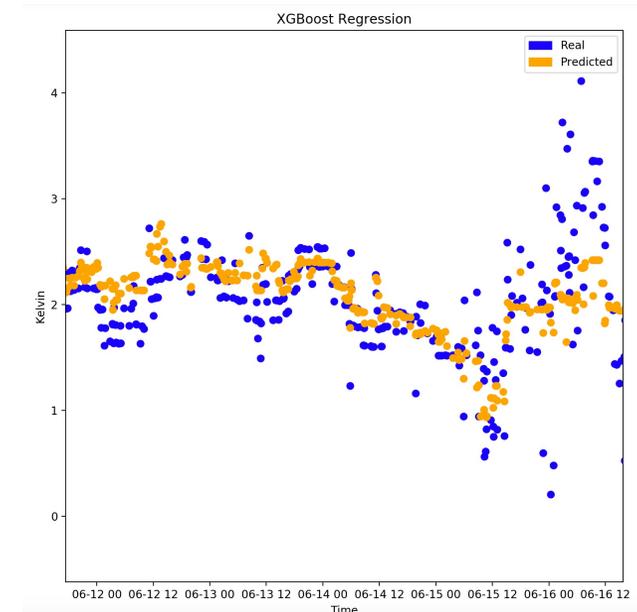
- ▶ Algorithms including SVM, MLP Regressor, Neural Networks, and XGBoost were used to create prediction algorithms for the temperature difference across the cooling tower.
- ▶ Algorithms were tested on data from different years, as well as modified equalized data sets.

## Regression: MLPRegressor

- Hidden layers: (32,32)
- Learning\_rate: 1e-3
- Random state = 1
  - Regular R<sup>2</sup>: 0.617
  - Equalized R<sup>2</sup>: 0.727
- Run 50 times, random state not set
  - Regular R<sup>2</sup>: 0.746
  - Equalized R<sup>2</sup>: 0.450

## Regression: SVM

- Kernel: rbf
- Regular R<sup>2</sup>: 0.873
- Equalized R<sup>2</sup>: 0.705



# Next Steps

- ▶ Try to obtain better data for certain variables, such as those which should vary but do not.
- ▶ Improve the prediction algorithm.
- ▶ Use clustering algorithms to enhance predictions and detect anomalies.