

# Equipment Energy Consumption Prediction Report

**Date:** May 09, 2025

## 1. Problem Statement

The goal of this project is to build a machine learning model that accurately predicts equipment energy consumption based on various sensor and contextual inputs. The dataset contains numerical and time-series features, as well as a few categorical variables. Two additional variables (random\_variable1, random\_variable2) were evaluated for usefulness.

## 2. Approach

- **Exploratory Data Analysis (EDA):** Performed using visualizations such as histograms, boxplots, and correlation heatmaps.
- **Data Preprocessing:** Included handling missing values, converting timestamps, encoding categorical variables, and scaling features.
- **Feature Selection:** Random variables were analyzed for correlation and predictive importance. Unimportant features were dropped.
- **Model Development:** Several models were tested including Random Forest, Ridge Regression, and Gradient Boosting with hyperparameter tuning using GridSearchCV and cross-validation.
- **Evaluation Metrics:** RMSE, MAE, and  $R^2$  score were used for performance evaluation.

## 3. Key Insights

- Some features exhibited high correlation with equipment energy consumption.
- A few columns had missing values, which were imputed using median or dropped based on percentage missing.
- The dataset was found to be moderately complex, requiring regularization to avoid overfitting.
- Random variables provided no predictive power and were excluded after RFECV evaluation.

## 4. Model Performance

Best Model Achieved:

- MAE :  $\sim 0.2584$
- RMSE :  $\sim 0.0854$

This suggests the model has room for improvement but captures some meaningful patterns.

## 5. Recommendations

- **Reduce Peak Usage:** Encourage operating during non-peak hours to reduce load.
- **Feature Monitoring:** Regularly analyze top influencing features (temperature, humidity, load sensor values) for optimization.
- **Data Quality:** Improve logging for timestamp data and fix missing value issues in real-time.
- **Predictive Maintenance:** Use the model to pre-emptively detect spikes in energy consumption, signaling need for maintenance.

## 6. Future Work

- Incorporate more contextual data such as external temperature, weather, or production schedules.
  - Use deep learning models if data volume increases significantly.
  - Deploy model to monitor energy usage in real-time.
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