

# Electric Motor Temperature Prediction

## Project Overview

This project aims to predict the **temperature of a permanent magnet (pm) motor** based on various operational parameters. By analyzing features such as voltages, currents, and coolant temperature, the goal is to develop a regression model that can accurately forecast the motor's internal temperature. This is crucial for condition monitoring, preventing overheating, and extending the lifespan of electric motors.

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## Technical Highlights

- **Dataset:** The project uses a dataset related to electric motor temperature. The specific dataset is titled `measures_v2.csv` and is likely from a Kaggle source related to electric motor temperature prediction, though the specific link is missing from the provided code block.
- **Size:** 20,000 entries, 13 columns.
- **Key Features:**
  - **Motor Electrical Signals:** `u_q`, `u_d`, `i_d`, `i_q`.
  - **Temperatures:** coolant, stator\_winding, stator\_tooth, stator\_yoke, ambient.
  - **Performance Metrics:** `motor_speed`, `torque`.
- **Approach:**
  - **Data Cleaning:** The dataset appears to be clean, with no missing values or duplicates in the sample used.
  - **Exploratory Data Analysis:** Histograms, boxplots, and a heatmap were used for visualization to understand data distributions and correlations. The heatmap reveals several strong correlations between the features.
  - **Regression Task:** The target variable is `pm`, which likely represents the permanent magnet temperature.
  - **Models Used:**
    - A suite of regression models were trained, including Ridge Regression, XGBoost, Random Forest, AdaBoost, Gradient Boosting, Bagging, Decision Tree, SVR, and K-Nearest Neighbors (KNN).
- **Best R<sup>2</sup> Score:**
  - **0.99992** with Random Forest Regressor.
  - **0.99984** with KNN Regressor.
  - **0.99983** with XGBoost Regressor.

- The extremely high  $R^2$  scores across multiple models indicate that the input features are highly predictive of the motor's temperature.
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## Purpose and Applications

- Enable **predictive maintenance** by forecasting motor temperatures to prevent overheating and component failure.
  - Optimize the operation of electric motors in industrial settings to improve efficiency and reduce energy consumption.
  - Support the design of more effective cooling systems for motors.
  - Provide a foundational model for developing real-time condition monitoring and fault diagnosis systems.
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## Installation

Clone the repository and download the dataset.

Install the necessary libraries:

```
pip install pandas numpy seaborn matplotlib scikit-learn xgboost
```

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## Collaboration

We welcome contributions to improve the project. You can help by:

- Performing a deeper analysis of the relationships between the input features and the target variable to understand the underlying physics.
- Investigating the impact of the highly correlated features on the models.
- Exploring more advanced time-series forecasting techniques, as temperature prediction is often a time-dependent problem.
- Adding explainability (e.g., SHAP or LIME) to understand which operational parameters are the most significant drivers of motor temperature.

## GITHUB LINK:

<https://github.com/manyam-vamsi/Electric-motor-temperature-prediction>

## Video Link:

<https://drive.google.com/file/d/1mWtbZcGJHYywfbDpy9sVU8WzfnpwCbFK/view?usp=sharing>

## PHOTOS

The screenshot shows a code editor interface with a dark theme. The top menu bar includes 'FILE', 'EDITOR', 'VIEW', 'GO', 'RUN', and '...'. The left sidebar, titled 'EXPLORER', shows a project structure for 'ELECTRIC MOTOR TEMPERATURE PREDICTION'. It contains a 'templates' folder with 'index.html', 'app.py', 'c11.PNG', 'c12.PNG', 'c13.PNG', 'model.pkl', and 'motor\_temperature.csv'. The main workspace displays the 'app.py' file content:

```
Electric motor temperature prediction
app.py
...
38     </form>
39     (% if result is not none %)
40         <div class="result">Predicted Temperature: {{ result }} </div>
41     (% endif %)
42     </div>
43 </body>
44 </html>
45 """
46
47 @app.route("/", methods=["GET", "POST"])
48 def index():
49     result = None
50     if request.method == "POST":
51         # Get form values
52         values = [float(request.form.get(col)) for col in
53                   ["ambient_temp", "coolant_temp", "voltage_d", "voltage_q", "motor_speed", "current_d", "current_q"]]
54         prediction = model.predict([np.array(values)])
55         result = round(prediction, 2)
56     return render_template_string(html_template, result=result)
57
58 if __name__ == "__main__":
59     app.run(debug=True)
60 
```

The bottom navigation bar includes 'PROBLEMS', 'OUTPUT', 'DEBUG CONSOLE', 'TERMINAL', and 'PORTS'.

The screenshot shows a web application titled 'Electric Motor Temperature Prediction'. The page has a light gray background with a white input form. The form fields are labeled 'Ambient Temp', 'Coolant Temp', 'Voltage d', 'Voltage q', 'Motor Speed', 'Current d', and 'Current q', each followed by a text input field. At the bottom of the form is a blue horizontal button.