



# Solar Power Generation Prediction

## Week 1 Report – Data Loading, Cleaning, Merging & EDA

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**Duration:** Week 1 (Project Theme: Energy)

### **1** Objective of Week 1

The goal of Week 1 was to perform **data understanding, cleaning, and exploration** on the provided datasets to prepare them for the prediction model in Week 2.

The tasks included:

- Loading and merging power generation and weather sensor datasets for both solar plants.
  - Performing data cleaning, handling missing values, and feature engineering.
  - Conducting **Exploratory Data Analysis (EDA)** to understand patterns and relationships between weather conditions and power generation.
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### **2** Datasets Used

The project uses four CSV files:

1. Plant\_1\_Generation\_Data.csv
2. Plant\_1\_Weather\_Sensor\_Data.csv
3. Plant\_2\_Generation\_Data.csv
4. Plant\_2\_Weather\_Sensor\_Data.csv

Each generation file contains **power output readings** from multiple inverters, while the weather data files include **solar irradiance, temperature, and wind speed** readings from sensors.

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### **3** Data Preprocessing Steps

- **Datetime Conversion:** Converted DATE\_TIME columns to proper datetime format and extracted DATE, HOUR, and MONTH.
- **Merging:** Combined each plant's generation and weather datasets using the DATE\_TIME column.

- **Missing Value Handling:** Filled missing data using **forward fill (ffill)** to ensure time-series continuity.
  - **Feature Engineering:**
    - Created `TOTAL_POWER` as the sum of DC and AC power.
    - Extracted new columns for `MONTH` and `WEEKDAY`.
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## **Exploratory Data Analysis (EDA)**

### **a. Basic Statistics**

- Generated descriptive statistics for power and weather features.
- Identified consistent ranges for `DC_POWER` and `AC_POWER` across both plants.

### **b. Correlation Heatmap**

- Created a heatmap showing correlations between numeric features.
- Observed that **solar irradiance and DC power** had a strong positive correlation.
- Ambient temperature showed a moderate negative effect on DC power (higher temperature, slightly reduced efficiency).

### **c. Power Generation Over Time**

- Plotted both **DC Power** and **AC Power** over time.
- Observed clear daily patterns—power generation increases during the day and drops at night.

### **d. Temperature vs Power**

- Scatterplot showed that as **ambient temperature** rises, DC power increases up to a point, then stabilizes or slightly drops (indicating optimal operating range).

### **e. Daily Energy Comparison**

- Grouped data by `DATE` and summed daily DC power for each plant.
  - Found that **Plant 2** generally produced slightly more energy than **Plant 1**, possibly due to better location or irradiance conditions.
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## **Insights from Week 1**

- ✓ Solar power generation is highly dependent on **irradiance** and **time of day**.
- ✓ Temperature impacts power output efficiency.
- ✓ Both plants exhibit similar daily generation patterns but differ in total output.
- ✓ The datasets are now **clean, merged, and feature-enriched**, ready for model building.

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## 6 Plan for Week 2

In Week 2, we will:

- Prepare final feature sets for training and testing.
- Build and compare multiple regression models:
  - **Linear Regression**
  - **Random Forest**
  - **XGBoost**
- Evaluate models using **MAE**, **RMSE**, and **R<sup>2</sup> score**.
- Visualize prediction performance and deploy a small **Streamlit app** to demonstrate real-time prediction

## 7 Summary

Step	Task	Status
1	Data Loading & Inspection	Completed
2	Data Cleaning & Merging	Completed
3	Feature Engineering	Completed
4	Exploratory Data Analysis	Completed
5	Insights Documentation	Completed