

# Forecasting Household Electricity Usage

This project combines time series forecasting and pattern recognition to analyze residential electricity use. Using machine learning, we develop models for short-term consumption prediction and detect daily usage patterns.

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# Problem Overview & Opportunity



Smart meters produce massive datasets daily from household consumption.



Most households lack visibility into consumption patterns and usage drivers.



Energy dynamics vary due to weather, appliances, and seasonal behaviors.

Data science offers predictive insights to help save costs by understanding these complex factors.

# Data Science Vision



## Data Integration

Combining minute-level electricity data with hourly weather metrics



## Pattern Exploration

Spotting inefficiencies and unusual consumption trends



## Advanced Analytics

Clustering similar usage days and forecasting with time and weather variables

Detecting spikes and anomalies in consumption

Unlock the power of data - uncover hidden patterns, spot inefficiencies, and find fresh ways to slash household electricity waste and costs.

### 1 Time Series Forecasting

Develop accurate short-term usage models using ARIMA

### 2 Seasonal Pattern Recognition

Identify seasonal consumption trends through Prophet

### 3 Clustering

Categorize days by consumption patterns for weekdays, weekends, through KMeans

### 4 Sub-metering Analysis

Analyze appliance-level usage to pinpoint high-consumption metrics and weather influence through Regression

# What Can This Solution Achieve?

## Households

Save 10–20% on energy bills through behavior change.

Gain visibility on which appliances and conditions drive high usage.

## Utilities

Better demand forecasting reduces grid strain significantly.

More effective time-of-use pricing and strategic planning capabilities.

## Environmental

Reduce unnecessary electricity consumption across residential sectors.

Support sustainable, low-carbon lifestyles through data-driven insights.





# Understanding the Data

## UCI Dataset

Individual household electric power consumption (France, 2006–2010)

- 2M+ rows of 1-minute resolution data, with 6 features
- 3 Sub-metering power features: kitchen, laundry, HVAC systems
- Average Voltage, Average Current and Total Power usage (Target) used for the House

## Weather Dataset

Historical hourly weather data from Open-Meteo API which has features such as Temperature, Relative Humidity, Pressure, Cloud Cover, Wind Speed etc. A total of 12 features

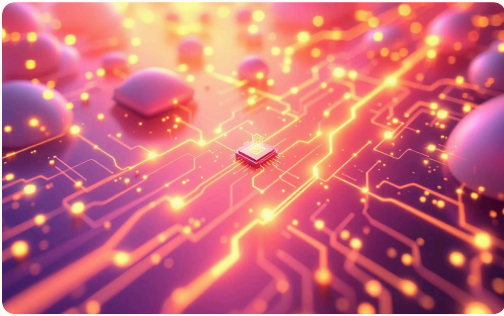
## Data Quality Notes

- Missing values and timestamps (e.g., blackout periods)
- Sub-meterings do not fully cover all consumption(eg. lights)
- Weather data has lower time resolution (hourly vs minute)
- Interdependencies like  $\text{Power} = \text{Current} * \text{Voltage}$ , Active vs Reactive Power

## Early Basic EDA

- Clear daily/weekly usage patterns for Power consumption
- High Power use in colder months — likely heating-related
- Voltage, Current dips/highs correspond to power usage spikes

# Next Steps for Data Processing, Feature Engineering, and Modeling



## Data Processing

Normalize and scale features  
weatherdataset features with  
UCI dataset features for  
consistent model input.



## Feature Engineering

Create time-based features,  
rolling averages, and lag  
variables to boost model  
accuracy.



## Modeling

- ARIMA & Prophet for seasonality forecasting
- KMeans clustering to identify usage patterns
- Baseline prediction via Linear Regression on weather and appliances metrics



## Evaluation

Measure regression accuracy  
using MAE and RMSE metrics.

# Thank You!

Comments or Suggestions?