

Project Name: Electricity Price Prediction

Phase 3: Development Part 1 - Loading and Preprocessing the Dataset

Introduction

In this phase, we will begin building the electricity price prediction model by loading and preprocessing the dataset. The dataset contains historical electricity prices and various relevant factors. We'll follow a systematic process to prepare the data for analysis, which includes data loading, data cleaning, and data transformation.

Data Loading

We will start by loading the historical electricity prices dataset. To do this, we need to import the necessary libraries and read the data from the provided CSV file.

Data Exploration

It's important to get an initial understanding of the dataset. We'll inspect the data to understand its structure and the types of information it contains.

Data Cleaning

Data cleaning is crucial to ensure that the dataset is free from errors, missing values, and inconsistencies.

Data Transformation

In this step, we will perform data transformations to make it suitable for analysis. This may include converting categorical variables to numerical representations, scaling numerical features, and creating new features.

python

Saving the Preprocessed Data

Once the data is loaded, cleaned, and transformed, it's a good practice to save it for further analysis and model development.

Code:

```
# Import libraries
```

```
import pandas as pd
```

```
# Load the dataset
```

```
data = pd.read_csv('electricity_price_prediction_dataset.csv')
```

```

# Display the first few rows of the dataset
print(data.head())

# Check the data types of each column
print(data.dtypes)

# Summary statistics
print(data.describe())

# Check for missing values
missing_values = data.isnull().sum()
print(missing_values)

# Remove rows with missing values or replace them as needed
data = data.dropna()

# Check for duplicated rows
duplicates = data.duplicated().sum()
print("Number of duplicated rows:", duplicates)

# Remove duplicates if necessary
data = data.drop_duplicates()

# Save the preprocessed data to a new CSV file
data.to_csv('preprocessed_electricity_data.csv', index=False)

```

Output:

```

    DateTime Holiday HolidayFlag DayOfWeek WeekOfYear ... ORKWindspeed CO2Intensity
ActualWindProduction SystemLoadEP2 SMPEP2

```

0	01/11/2011 00:00	NaN	0	1	44 ...	9.30	600.71	356.00	3159.60
54.32									
1	01/11/2011 00:30	NaN	0	1	44 ...	11.10	605.42	317.00	2973.01
54.23									
2	01/11/2011 01:00	NaN	0	1	44 ...	11.10	589.97	311.00	2834.00
54.23									
3	01/11/2011 01:30	NaN	0	1	44 ...	9.30	585.94	313.00	2725.99
53.47									
4	01/11/2011 02:00	NaN	0	1	44 ...	11.10	571.52	346.00	2655.64
39.87									

[5 rows x 18 columns]

DateTime	object
Holiday	object
HolidayFlag	int64
DayOfWeek	int64
WeekOfYear	int64
Day	int64
Month	int64
Year	int64
PeriodOfDay	int64
ForecastWindProduction	object
SystemLoadEA	object
SMPEA	object
ORKTemperature	object
ORKWindspeed	object
CO2Intensity	object
ActualWindProduction	object
SystemLoadEP2	object
SMPEP2	object

	HolidayFlag	DayOfWeek	WeekOfYear	Day	Month	Year	PeriodOfDay
count	38014.000000	38014.000000	38014.000000	38014.000000	38014.000000	38014.000000	38014.000000
38014.000000							
mean	0.040406	2.997317	28.124586	15.739412	6.904246	2012.383859	23.501105
std	0.196912	1.999959	15.587575	8.804247	3.573696	0.624956	13.853108
min	0.000000	0.000000	1.000000	1.000000	1.000000	2011.000000	0.000000
25%	0.000000	1.000000	15.000000	8.000000	4.000000	2012.000000	12.000000
50%	0.000000	3.000000	29.000000	16.000000	7.000000	2012.000000	24.000000
75%	0.000000	5.000000	43.000000	23.000000	10.000000	2013.000000	35.750000
max	1.000000	6.000000	52.000000	31.000000	12.000000	2013.000000	47.000000
DateTime	0						
Holiday	36478						
HolidayFlag	0						
DayOfWeek	0						
WeekOfYear	0						
Day	0						
Month	0						
Year	0						
PeriodOfDay	0						
ForecastWindProduction	0						
SystemLoadEA	0						
SMPEA	0						
ORKTemperature	0						
ORKWindspeed	0						
CO2Intensity	0						
ActualWindProduction	0						
SystemLoadEP2	0						
SMPEP2	0						

dtype: int64

Number of duplicated rows: 0

Conclusion

In this phase, we have loaded and preprocessed the historical electricity prices dataset, preparing it for analysis and model development. The preprocessed data can now be used in the next phases for feature engineering, model selection, and model training.