

Project 4: Electricity Prices Prediction

Phase 3: Development Part - 1

Team Members:

Ajita Fairen J - 715521106003 - B.E. ECE - 3rd year
M. Mekanya - 715521106028 - B.E. ECE - 3rd year
Shanmugapriya M - 715521106043 - B.E. ECE - 3rd year
K Shree Harini - 715521106044 - B.E. ECE - 3rd year
Vasanth L - 715521106055 - B.E. ECE - 3rd year
G. R. Tharunika - 715521106310 - B.E. ECE - 3rd year

Problem Statement:

To create a predictive model that utilizes electricity prices and relevant factors to forecast future electricity prices, assisting energy providers and consumers in making informed decisions regarding consumption and investment.

Phase 3 Task: To begin building the electricity prices prediction model by loading and preprocessing the dataset. Load the historical electricity prices dataset and preprocess the data analysis.

Dataset:

Source:

<https://www.kaggle.com/datasets/chakradharmattapalli/electricity-price-prediction/>

This dataset contains information related to electricity markets and factors that can influence electricity prices.

Source Code:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
```

#Data Loading

```
print("DATA LOADING\n\n")
data = pd.read_csv("Electricity.csv",low_memory=False)
```

```

print("Head of the dataset\n")
print(data.head())
print("\nInfo of the dataset\n")
print(data.info())
print("\nDescription of the dataset\n")
print(data.describe())

```

#Data Preprocessing

```

print("\n\n\nDATA TRANSFORMATION\n\n")
#Changing the type of data in the dataset to numerical values
data["ForecastWindProduction"] = pd.to_numeric(data["ForecastWindProduction"],errors
= 'coerce')
data["SystemLoadEA"] = pd.to_numeric(data["SystemLoadEA"],errors = 'coerce')
data["SMPEA"] = pd.to_numeric(data["SMPEA"],errors = 'coerce')
data["ORKTemperature"] = pd.to_numeric(data["ORKTemperature"],errors = 'coerce')
data["ORKWindspeed"] = pd.to_numeric(data["ORKWindspeed"],errors = 'coerce')
data["CO2Intensity"] = pd.to_numeric(data["CO2Intensity"],errors = 'coerce')
data["ActualWindProduction"] = pd.to_numeric(data["ActualWindProduction"],errors =
'coerce')
data["SystemLoadEP2"] = pd.to_numeric(data["SystemLoadEP2"],errors = 'coerce')
data["SMPEP2"] = pd.to_numeric(data["SMPEP2"],errors = 'coerce')
print(data.info())

```

```

print("\n\n\nDATA CLEANING\n\n")

```

#Data Cleaning

#Displaying the no. of data which has null values in it

```

print("With Null Values\n\n")
print(data.isnull().sum())

```

#Dropping or cleaning the null values

```

data = data.dropna()

```

#When displayed again there are no null values

```

print("\n\nAfter Dropping Null Values\n\n")
print(data.isnull().sum())

```

#Data Splitting

#Data is split into training and test tests

```

x = data[["Day", "Month", "ForecastWindProduction", "SystemLoadEA", "SMPEA",
"ORKTemperature", "ORKWindspeed", "CO2Intensity", "ActualWindProduction",
"SystemLoadEP2"]]
y = data["SMPEP2"]

```

```
xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2, random_state=42)
```

```
print("\n\n\nDATA SPLITTING\n\n")
print("x train\n\n")
print(xtrain)
print("\n\nx test\n")
print(xtest)
print("\n\ny train\n")
print(ytrain)
print("\n\ny test \n")
print(ytest)
```

Output:

Data Loading:

```
DATA LOADING

Head of the dataset

   DateTime Holiday HolidayFlag DayOfWeek WeekOfYear Day  ... ORKTemperature ORK Windspeed CO2Intensity ActualWindProduction SystemLoadEP2 SMPEP2
0  01/11/2011 00:00   NaN         0         1         44   1  ...         6.00         9.30         600.71           356.00         3159.60      54.32
1  01/11/2011 00:30   NaN         0         1         44   1  ...         6.00        11.10         605.42           317.00         2973.01      54.23
2  01/11/2011 01:00   NaN         0         1         44   1  ...         5.00        11.10         589.97           311.00         2834.00      54.23
3  01/11/2011 01:30   NaN         0         1         44   1  ...         6.00         9.30         585.94           313.00         2725.99      53.47
4  01/11/2011 02:00   NaN         0         1         44   1  ...         6.00        11.10         571.52           346.00         2655.64      39.87

[5 rows x 18 columns]
```

Info of the dataset

```
Info of the dataset

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 38014 entries, 0 to 38013
Data columns (total 18 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   DateTime                             38014 non-null  object
 1   Holiday                             1536 non-null   object
 2   HolidayFlag                         38014 non-null  int64
 3   DayOfWeek                           38014 non-null  int64
 4   WeekOfYear                          38014 non-null  int64
 5   Day                                 38014 non-null  int64
 6   Month                               38014 non-null  int64
 7   Year                                38014 non-null  int64
 8   PeriodOfDay                         38014 non-null  int64
 9   ForecastWindProduction              38014 non-null  object
10   SystemLoadEA                        38014 non-null  object
11   SMPEA                               38014 non-null  object
12   ORKTemperature                      38014 non-null  object
13   ORK Windspeed                       38014 non-null  object
14   CO2Intensity                        38014 non-null  object
15   ActualWindProduction                38014 non-null  object
16   SystemLoadEP2                      38014 non-null  object
17   SMPEP2                              38014 non-null  object
dtypes: int64(7), object(11)
memory usage: 5.2+ MB
None
```

Description of the dataset:

Description of the dataset

	HolidayFlag	DayOfWeek	WeekOfYear	Day	Month	Year	PeriodOfDay
count	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

Data Transformation

DATA TRANSFORMATION

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 38014 entries, 0 to 38013
Data columns (total 18 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   DateTime                             38014 non-null  object
1   Holiday                             1536 non-null   object
2   HolidayFlag                         38014 non-null  int64
3   DayOfWeek                          38014 non-null  int64
4   WeekOfYear                         38014 non-null  int64
5   Day                                38014 non-null  int64
6   Month                              38014 non-null  int64
7   Year                                38014 non-null  int64
8   PeriodOfDay                        38014 non-null  int64
9   ForecastWindProduction             38009 non-null  float64
10  SystemLoadEA                       38012 non-null  float64
11  SMPEA                              38012 non-null  float64
12  ORKTemperature                     37719 non-null  float64
13  ORKWindspeed                      37715 non-null  float64
14  CO2Intensity                       38007 non-null  float64
15  ActualWindProduction              38009 non-null  float64
16  SystemLoadEP2                     38012 non-null  float64
17  SMPEP2                             38012 non-null  float64
dtypes: float64(9), int64(7), object(2)
memory usage: 5.2+ MB
None
```

Data Cleaning:

DATA CLEANING

With Null Values

```
DateTime                             0
Holiday                             36478
HolidayFlag                         0
DayOfWeek                          0
WeekOfYear                         0
Day                                0
Month                              0
Year                                0
PeriodOfDay                        0
ForecastWindProduction             5
SystemLoadEA                       2
SMPEA                              2
ORKTemperature                     295
ORKWindspeed                      299
CO2Intensity                       7
ActualWindProduction              5
SystemLoadEP2                     2
SMPEP2                             2
dtype: int64
```

After Dropping Null Values:

After Dropping Null Values

```
DateTime          0
Holiday           0
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
```

Data Splitting:

DATA SPLITTING

x_train

	Day	Month	ForecastWindProduction	SystemLoadEA	SMPEA	ORKTemperature	ORKWindspeed	CO2Intensity	ActualWindProduction	SystemLoadEP2
20485	31	12	937.42	4709.48	84.13	5.0	20.4	432.98	735.0	4454.59
10376	4	6	34.79	2359.78	49.89	7.0	9.3	641.22	53.0	2260.84
2592	25	12	1225.85	4266.13	44.38	9.0	25.2	367.90	1227.0	3280.81
2990	2	1	1264.26	2980.84	25.05	2.0	20.4	402.11	750.0	2663.97
17434	29	10	205.36	2865.76	38.90	5.0	18.5	730.15	185.0	2737.51
...
26529	6	5	258.80	4073.50	90.69	11.0	16.7	438.98	178.0	3912.03
27860	3	6	66.10	3933.23	75.24	15.0	13.0	464.02	30.0	3927.27
37673	24	12	715.87	4525.94	76.91	3.0	7.4	362.45	613.0	3918.75
24178	18	3	297.39	4712.93	74.24	5.0	22.2	490.32	355.0	4648.78
27856	3	6	56.80	3545.50	70.82	14.0	13.0	533.35	22.0	3554.09

[1132 rows x 10 columns]

x test

	Day	Month	ForecastWindProduction	SystemLoadEA	SMPEA	ORKTemperature	ORKWindspeed	CO2Intensity	ActualWindProduction	SystemLoadEP2
20530	1	1	499.60	4893.88	98.95	6.0	13.0	410.33	421.0	4524.85
24714	29	3	887.40	4640.69	74.03	3.0	22.2	493.69	576.0	4086.59
7637	8	4	333.31	2640.94	49.60	8.0	11.1	422.36	397.0	2453.58
6675	19	3	245.40	3198.43	54.10	3.0	5.6	619.30	226.0	2871.33
6686	19	3	498.70	3376.08	51.83	5.0	11.1	530.52	529.0	2775.95
...
38006	31	12	1160.57	4188.85	66.08	5.0	18.5	262.97	1143.0	4207.57
3023	2	1	1456.80	4214.78	42.57	10.0	48.2	373.80	1274.0	3493.14
7608	7	4	290.70	4096.35	55.35	11.0	25.9	543.80	533.0	3871.78
24746	30	3	454.40	4370.79	66.08	3.0	25.9	503.30	591.0	4024.49
20120	24	12	740.20	3195.72	47.81	7.0	14.8	469.45	532.0	2452.40

[284 rows x 10 columns]

y train

20485	87.47
10376	39.75
2592	43.96
2990	38.35
17434	51.26

...	
26529	74.81
27860	73.33
37673	74.74
24178	98.09
27856	73.10

Name: SMPEP2, Length: 1132, dtype: float64

y test

20530	255.04
24714	99.50
7637	47.12
7637	47.12
6675	45.79
6686	45.88

...	
38006	62.05
3023	41.38
7608	59.85
24746	66.08
20120	45.45

Name: SMPEP2, Length: 284, dtype: float64