KINGS ENGINEERING COLLEGE ELECTRICITY PRICE PREDICTION

Department: B.Tech. Information Technology

Batch No: 07

Domain: Applied Data Science

Topic: Electricity Prices - Prediction

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PHASE 3

Building an electricity price prediction model involves several steps, including loading and preprocessing the dataset. Here's a step-by-step guide

Import Necessary Libraries:

First, you'll need to import the necessary Python libraries for data manipulation, visualization, and modeling. Common libraries include pandas, numpy, matplotlib, and scikit-learn.

PYTHON CODE:

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

1. Load the Dataset:

Assuming that your dataset is stored in a CSV file (let's call it your_dataset.csv'), you can load it using the `read_csv` function in Pandas:

#data loading
df=pd.read_csv("/content/sample_data/electricity.gui (1).zip")
df.head()

	DateTime	Holiday	HolidayFlag	DayOfWeek	WeekOfYear	Day	Month	Year	PeriodOfDay	ForecastWindProduction	SystemLoadEA	SMPEA	ORKTemperature	01
0	01/11/2011 00:00	None	0	1	44	1	11	2011	0	315.31	3388.77	49.26	6.00	
1	01/11/2011 00:30	None	0	1	44	1	11	2011	1	321.80	3196.66	49.26	6.00	
2	01/11/2011 01:00	None	0	1	44	1	11	2011	2	328.57	3060.71	49.10	5.00	
3	01/11/2011 01:30	None	0	1	44	1	11	2011	3	335.60	2945.56	48.04	6.00	
4	01/11/2011 02:00	None	0	1	44	1	11	2011	4	342.90	2849.34	33.75	6.00	

2. Explore the Dataset:

Now that you've loaded the dataset, you should explore it to understand its structure and contents. You can start by looking at the first few rows of the dataset:

```
Python Code:
df. shape
(38014, 18)
df. tail()
```



3. Preprocess the Data:

Preprocessing involves tasks like handling missing values, encoding categorical variables, and scaling numerical features.

Handle Missing Values:

missing value query

df.isna().sum()

WeekOfYear		0				
Day		0				
Month						
Year		0				
PeriodOfDay						
ForecastWindProd	ecastWindProduction 0					
SystemLoadEA		0				
SMPEA	0					
ORKTemperature	0					
ORKWindspeed 0						

```
CO2Intensity 0

ActualWindProduction 0

SystemLoadEP2 0

SMPEP2 0

dtype: int64
```

Encode Categorical Variables:

Encoding categorical variables is an essential step in preparing data for machine learning models, as most algorithms require numerical inputs.

#create a list for numeric and categorical values

```
cat_list=[]
num_list=[]
for i in df.columns:
  unique_val=len(df[i].unique())
  if unique_val<40:
    cat_list.append(i)
  else:
    num_list.append(i)
cat_list.append("WeekOfYear")
cat list
OUTPUT:
['Holiday',
'HolidayFlag',
'DayOfWeek',
'Day',
'Month',
'Year',
```

```
'ORKTemperature',

'WeekOfYear']

# distributions of numeric attributes

# distributions of numeric attributes

num_list.remove("DateTime")

num_list
```

OUTPUT:

```
['WeekOfYear',
'PeriodOfDay',
'ForecastWindProduction',
'SystemLoadEA',
'SMPEA',
'ORKWindspeed',
'CO2Intensity',
'ActualWindProduction',
'SystemLoadEP2',
'SMPEP2']
```

Scale Numerical Features:

Scaling numerical features is an important preprocessing step in many machine learning algorithms. It helps ensure that all features contribute equally to the model's training process, preventing features with larger scales from dominating the learning process.

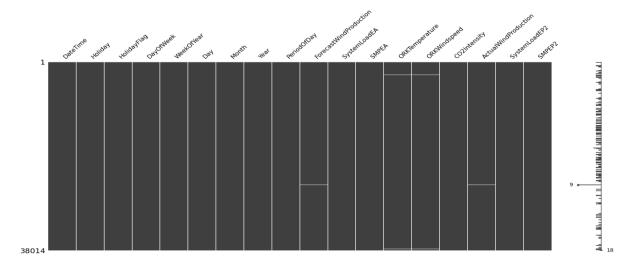
```
num_list.append("ORKTemperature")
k=1
plt.figure(figsize=(12,12))
```

```
plt.suptitle("distribution of numerical values")
for i in df.loc[:,num_list]:
    plt.subplot(6,2,k)
    sns.distplot(df[i])
    plt.title(i)
 k+=1
    plt.tight_layout()
                                                       distribution of numerical values
                                  WeekOfYear
      0.03
     0.02
                                                                         0.02
0.01
     0.00
                                                                           0.00
                                  20 30
WeekOfYear
                                                                                                        20 :
PeriodOfDay
                            ForecastWindProduction
                                                                                                       SystemLoadEA
   0.0015
                                                                         0.0006
                                                                      0.0004
0.0002
   0.0010
                                                                         0.0000
   0.0000
                             500 750 1000 1250
ForecastWindProduction
                                                                                                        4000 5000
SystemLoadEA
                                                                                                                              6000
                                    SMPEA
                                                                                                      ORKWindspeed
     0.02
                                                                         0.04
0.02
     0.01
      0.00
                                                                           0.00
                                     300
SMPEA
                                                        500
                                                                                                       ORKWindspeed
                                  CO2Intensity
                                                                                                   ActualWindProduction
                                                                         0.0015
    0.004
                                                                       0.00010
0.0005
    0.002
    0.000
                                                                         0.0000
                                  400
CO2Intensity
                                                             800
                                                                                                                            1500
                                                                                                                                  1750
                                SystemLoadEP2
                                                                                                          SMPEP2
                                                                           0.02
   0.0004
                                                                         0.01
   0.0002
   0.0000
                                 4000
SystemLoadEP2
                                                          6000
                                                                                                                                      1000
                                                                                                         400
SMPEP2
                                ORKTemperature
     0.15
   O 10
```

Visualization of missing values:

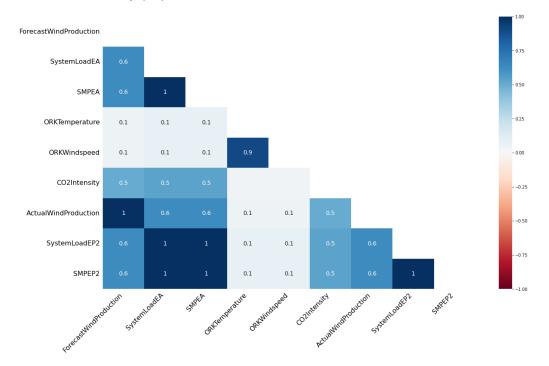
Visualizing missing data is crucial for understanding the extent and patterns of missingness in a dataset. It helps in making informed decisions about how to handle missing values.

import missingno as msno
msno.matrix(df);



let's visualize whether there is a relationship between the missing values

msno.heatmap(df);



Different Analysis:

Time Series Analysis:

Time series analysis is a statistical technique used to analyze and extract meaningful information from time-ordered data points.

```
from datetime import datetime

df["DateTime"] = pd.to_datetime(df.DateTime)

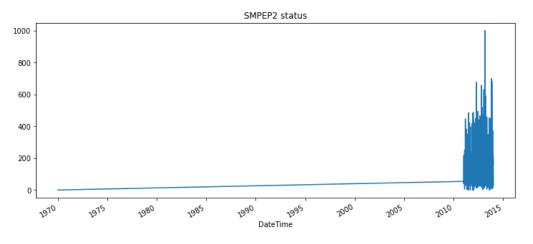
df['year'] = df['DateTime'].dt.year

df['month'] = df['DateTime'].dt.month

df["day"]=df["DateTime"].dt.day
```

We have created 3 new columns# we can start our time series analysis# change of real price of consumed electricity with time

custgroup=df.groupby('DateTime').mean()
plt.figure(figsize=(12,5))
custgroup['SMPEP2'].plot(x=df.DateTime)
plt.figure(figsize=(12,5))
custgroup['SMPEP2'].plot(x=df.DateTime)
plt.title("SMPEP2 status")
plt.show()



Data Visualize:

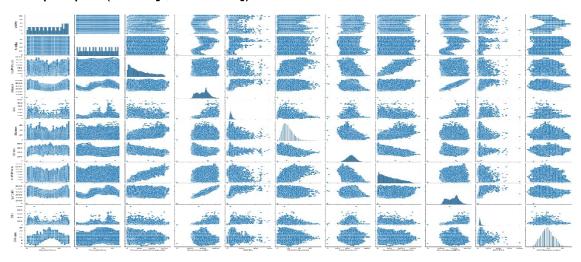
Data visualization is a critical part of the data analysis process. It helps in understanding the underlying patterns, trends, and relationships in the data.

```
# Categorical Analysis
for i in cat_list:
  plt.figure(figsize=(13,13))
  sns.countplot(x=i,data=df.loc[:,cat_list])
  plt.title(i)
```

numerical analysis

Numerical analysis is a branch of mathematics and computer science that deals with the development and application of computational algorithms to solve mathematical problems. It involves techniques for approximating solutions to mathematical problems that may be too complex to solve analytically.

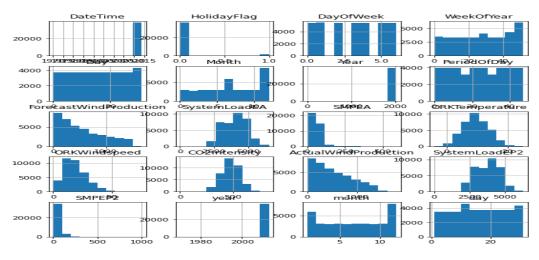
sns.pairplot(df.loc[:,num_list]);



histogram

A histogram is a graphical representation of the distribution of a dataset. It provides a visual summary of the underlying frequency distribution of a set of continuous or discrete data.

df.hist(figsize=(9,9));



Conclus	sion:		
	e model utilizing h		
	g electricity prices. els, and a more su		g, reduced relian