MEASURE ENERGY CONSUMPTION

Phase 4

```
import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

import warnings


warnings.filterwarnings("ignore", category=UserWarning)


from sklearn.model_selection import train_test_split

from sklearn.preprocessing import StandardScaler

from sklearn.svm import SVR

from sklearn.metrics import mean_squared_error, r2_score


RED = "\033[91m"

GREEN = "\033[92m"

YELLOW = "\033[93m"

BLUE = "\033[94m"

RESET = "\033[0m"


df = pd.read_csv("/kaggle/input/hourly-energy-consumption/AEP_hourly.csv")
```

```
df["Datetime"] = pd.to_datetime(df["Datetime"])
```

DATA CLEANING

```
print(BLUE + "\nDATA CLEANING" + RESET)
missing_values = df.isnull().sum()
print(GREEN + "Missing Values : " + RESET)
print(missing_values)
df.dropna(inplace=True)
duplicate_values = df.duplicated().sum()
print(GREEN + "Duplicate Values : " + RESET)
print(duplicate_values)
df.drop_duplicates(inplace=True)
```

DATA ANALYSIS

```
print(BLUE + "\nDATA ANALYSIS" + RESET)
summary_stats = df.describe()
print(GREEN + "Summary Statistics : " + RESET)
print(summary_stats)

print(BLUE + "\nMODELLING" + RESET)
df = df.sample(frac=0.2, random_state=42)
X = df[["Datetime"]]
y = df["AEP_MW"]
X_train, X_test, y_train, y_test = train_test_split(
        X, y, test_size=0.2, random_state=42
)
```

```
X_train["DayOfYear"] = X_train["Datetime"].dt.dayofyear

X_test["DayOfYear"] = X_test["Datetime"].dt.dayofyear

X_train = X_train["DayOfYear"].values.reshape(-1, 1)

X_test = X_test["DayOfYear"].values.reshape(-1, 1)

scaler = StandardScaler()

X_train_scaled = scaler.fit_transform(X_train)

X_test_scaled = scaler.transform(X_test)

svr = SVR(kernel="linear", C=1.0)

svr.fit(X_train_scaled, y_train)

y_pred = svr.predict(X_test_scaled)

mse = mean_squared_error(y_test, y_pred)

r2 = r2_score(y_test, y_pred)

print(f"Mean Squared Error: {mse}")

print(f"R-squared: {r2}")

plt.figure(figsize=(10, 6))

plt.scatter(X_test, y_test, color="b", label="Actual")

plt.scatter(X_test, y_pred, color="r", label="Predicted")

plt.xlabel("Day of the Year")

plt.ylabel("Energy Consumption (MW)")

plt.title("SVR Model: Actual vs. Predicted")

plt.legend()

plt.grid()

plt.show()
```

DATA VISUALIZATION

```
print(BLUE + "\nDATA VISUALIZATION" + RESET)
print(GREEN + "LinePlot : " + RESET)
plt.figure(figsize=(10, 6))
sns.lineplot(data=df, x="Datetime", y="AEP_MW")
plt.xlabel("Datetime")
plt.ylabel("Energy Consumption (MW)")
plt.title("Energy Consumption Over Year")
plt.grid()
plt.show()
print(GREEN + "Histogram : " + RESET)
plt.figure(figsize=(10, 6))
plt.hist(
    df["AEP_MW"],
    bins=100,
    histtype="barstacked",
    edgecolor="white",
)
plt.xlabel("AEPMW")
plt.ylabel("Frequency")
plt.title("Histogram of MEGAWATT USAGE")
plt.show()

df.to_csv("/kaggle/working/cleaned_AEP_hourly.csv", index=False)
print(BLUE + "\nDATA ANALYSIS" + RESET)
print(GREEN + "Data Cleaned and Saved !" + RESET)
```

OUTPUT:

DATA CLEANING

Missing Values :

Datetime 0

AEP_MW 0

dtype: int64

Duplicate Values :

0

DATA ANALYSIS

Summary Statistics :

	Datetime	AEP_MW
count	121273	121273.000000
mean	2011-09-02 03:17:01.553025024	15499.513717
min	2004-10-01 01:00:00	9581.000000
25%	2008-03-17 15:00:00	13630.000000
50%	2011-09-02 04:00:00	15310.000000
75%	2015-02-16 17:00:00	17200.000000
max	2018-08-03 00:00:00	25695.000000
std	NaN	2591.399065

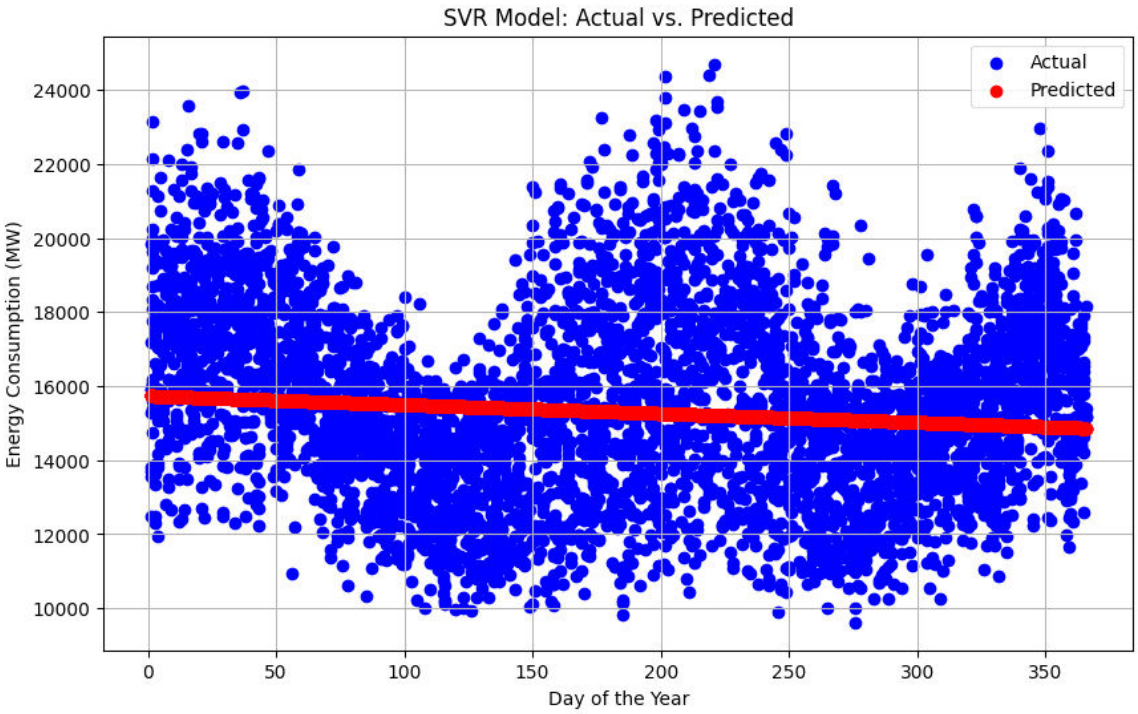
MODELLING

Mean Squared Error: 6758395.805638685

R-squared: 0.00270160624748228

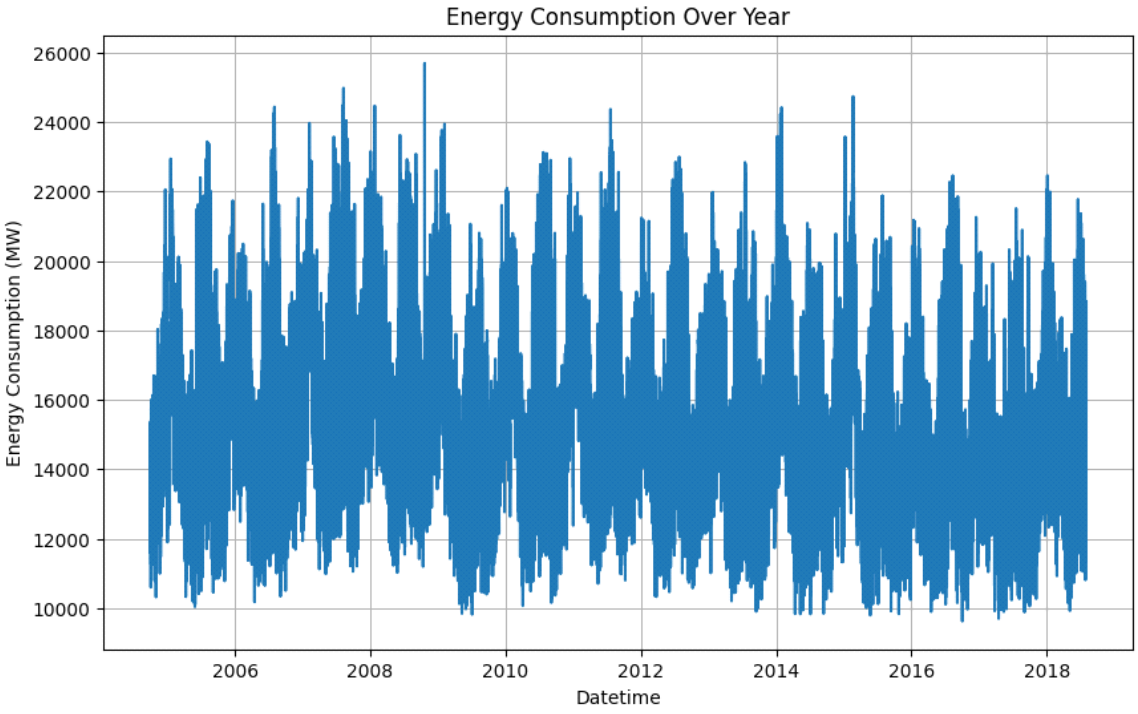
DATA VISUALIZATION

LinePlot :

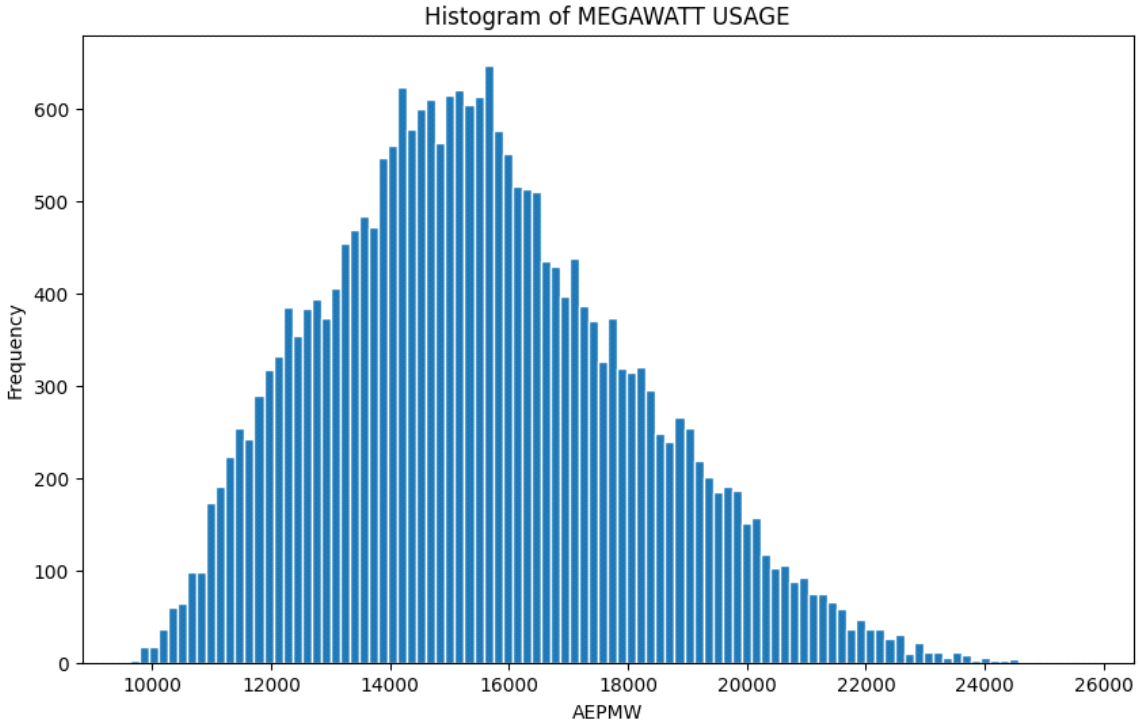


DATA VISUALIZATION

LinePlot :



HISTOGRAM


Histogram of MEGAWATT USAGE
Frequency
AEPMW
0
100
200
300
400
500
600
10000
12000
14000
16000
18000
20000
22000
24000
26000