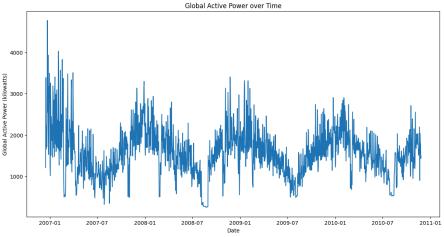
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from statsmodels.tsa.arima.model import ARIMA
from \ sklearn.metrics \ import \ mean\_squared\_error
# Load the text file
df = pd.read_csv('household_power_consumption.txt', sep=';',parse_dates={'datetime': ['Date', 'Time']}, infer_datetime_format=True,low_memory
# Fill missing values with the mean
df['Global_active_power'] = pd.to_numeric(df['Global_active_power'], errors='coerce')
df['Global_active_power'].fillna(df['Global_active_power'].mean(), inplace=True)
# Set the datetime as the index
df.set_index('datetime', inplace=True)
# Resample the data to daily frequency
df_daily = df.resample('D').sum()
# Basic statistics
print(df_daily.describe())
# Check for missing values
print(df_daily.isnull().sum())
# Plotting the data
plt.figure(figsize=(14, 7))
plt.plot(df_daily.index, df_daily['Global_active_power'])
plt.title('Global Active Power over Time')
plt.xlabel('Date')
plt.ylabel('Global Active Power (kilowatts)')
plt.show()
```

```
<ipython-input-15-3570c6b41cfa>:9: FutureWarning: The argument 'infer_datetime_format'
      df = pd.read_csv('household_power_consumption.txt', sep=';',parse_dates={'datetime':
    <ipython-input-15-3570c6b41cfa>:9: UserWarning: Parsing dates in %d/%m/%Y %H:%M:%S form
      df = pd.read_csv('household_power_consumption.txt', sep=';',parse_dates={'datetime':
           Global_active_power Global_reactive_power
                                                              Voltage
                   1442.000000
                                           1442.000000
                                                          1442.000000
    count
                   1571.001338
                                           175.815258
                                                        342266,507732
    mean
    std
                    595.405647
                                             51.998109
                                                         36707.752471
                    250.298000
                                              0.000000
    min
                                                             0.000000
    25%
                   1191,182615
                                            140,911500
                                                        345621.842500
    50%
                   1559.085000
                                            170.290000
                                                        346934.710000
    75%
                   1889.859500
                                            202.372500
                                                        348251.527500
                   4773.386000
                                            417.834000
                                                        356306.410000
    max
           Global_intensity
                             Sub_metering_1
                                             Sub_metering_2 Sub_metering_3
                                                                 1442.000000
                1442.000000
                                 1442.000000
                                                 1442.000000
    count
                6576.681415
                                 1594.407074
                                                 1845.375173
                                                                 9178.340499
    mean
                                                                 3787.898093
    std
                2559.505974
                                 1587.840580
                                                 2089.590342
                   0.000000
                                   0.000000
                                                    0.000000
                                                                    0.000000
    min
    25%
                4988.800000
                                  555.500000
                                                  424.250000
                                                                 6604.250000
                6510.300000
                                 1109,000000
                                                  678,500000
                                                                 9251,000000
    50%
    75%
                7953.350000
                                 2196.750000
                                                 2712.750000
                                                                11708.500000
                                11178.000000
                                                12109.000000
                                                                23743.000000
    max
               20200.400000
    Global_active_power
    Global_reactive_power
                             0
                              0
    Voltage
    Global intensity
                             0
    Sub_metering_1
                             a
    Sub_metering_2
                             0
    Sub_metering_3
    dtype: int64
```



```
# Define the training and testing periods
# Adjust these dates to have some overlap and avoid a gap
train_end_date = '2009-12-31'
test_start_date = '2009-01-01'

# Split the data
train = df_daily[:train_end_date]
test = df_daily[test_start_date:]

# Check if the test set is empty and handle the case
if test.empty:
    print("Warning: Test set is empty. Adjust the date range.")
else:
    # Fit an ARIMA model
    # You might need to experiment with different (p, d, q) orders
```

```
model = ARIMA(train['Global_active_power'], order=(5, 1, 0))
   model_fit = model.fit()
   # Forecasting
   forecast = model_fit.forecast(steps=len(test))
   test['Forecast'] = forecast.values
   # Calculate MSE
   mse = mean_squared_error(test['Global_active_power'], test['Forecast'])
   print(f'Mean Squared Error: {mse}')
   # Plot the results
   plt.figure(figsize=(14, 7))
   plt.plot(train.index, train['Global_active_power'], label='Training Data')
   plt.plot(test.index, test['Global_active_power'], label='Test Data')
   plt.plot(test.index, test['Forecast'], label='Forecast')
   plt.legend(loc='upper left')
   plt.title('Global Active Power Forecast')
   plt.xlabel('Date')
   plt.ylabel('Global Active Power (kilowatts)')
   plt.show()
<ipython-input-19-40b94adc05a4>:21: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row_indexer,col_indexer] = value instead
    See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user">https://pandas.pydata.org/pandas-docs/stable/user</a>
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user test['Forecast'] = forecast.values

Mean Squared Error: 689430.7539605086

