

step : Imports & Reading Data

```
In [1]: pip install pandas sqlite3 matplotlib seaborn
```

Requirement already satisfied: pandas in c:\Users\Wseoin\Anaconda\lib\site-packages (2.2.2)

Note: you may need to restart the kernel to use updated packages.

ERROR: Could not find a version that satisfies the requirement sqlite3 (from version s: none)

ERROR: No matching distribution found for sqlite3

```
In [2]: import pandas as pd
import sqlite3
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
from scipy import stats
pd.set_option('display.max_columns', None)
```

C:\Users\Wseoin\Anaconda\lib\site-packages\pandas\core\arrays\masked.py:60: UserWarning: Pandas requires version '1.3.6' or newer of 'bottleneck' (version '1.3.5' currently installed).

```
from pandas.core import (
```

```
In [3]: # Connect to SQLite database
conn = sqlite3.connect('cruise_data.db')
cursor = conn.cursor()

# Load CSV data into a DataFrame
df = pd.read_csv('C:/Users/seoin/Desktop/Tui/task_data/data.csv')

# Create a table in the SQLite database
df.to_sql('cruise_data', conn, if_exists='replace', index=False)

# Verify the data is loaded
query = "SELECT * FROM cruise_data"
df_sql = pd.read_sql_query(query, conn)
print(df_sql.head())
```

| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | W |
|---|---------------------|---------------------|-------------|---------------------|---|
| 0 | 2023-01-01T00:00:00 | 2023-01-01T00:05:00 | Vessel 1 | 0.0946 | |
| 1 | 2023-01-01T00:05:00 | 2023-01-01T00:10:00 | Vessel 1 | 0.0540 | |
| 2 | 2023-01-01T00:10:00 | 2023-01-01T00:15:00 | Vessel 1 | 0.0439 | |
| 3 | 2023-01-01T00:15:00 | 2023-01-01T00:20:00 | Vessel 1 | 0.0733 | |
| 4 | 2023-01-01T00:20:00 | 2023-01-01T00:25:00 | Vessel 1 | 0.0780 | |

| | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | W |
|---|---------------------|--------------------|---------------------------|---|
| 0 | 0.1384 | 5.4654 | 0.5074 | |
| 1 | 0.1370 | 5.4387 | 0.5158 | |
| 2 | 0.1785 | 5.5265 | 0.5117 | |
| 3 | 0.1725 | 5.5257 | 0.5177 | |
| 4 | 0.1397 | 5.4634 | 0.5169 | |

| | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | W |
|---|---------------------------|---------------------------|---------------------|---|
| 0 | 0.0 | 0.4979 | 0.4191 | |
| 1 | 0.0 | 0.4982 | 0.4204 | |
| 2 | 0.0 | 0.5032 | 0.4199 | |
| 3 | 0.0 | 0.5103 | 0.4188 | |
| 4 | 0.0 | 0.5100 | 0.4203 | |

| | Sea Temperature (Celsius) | Boiler 1 Fuel Flow Rate (L/h) | W |
|---|---------------------------|-------------------------------|---|
| 0 | 27.3000 | 0.0000 | |
| 1 | 27.3000 | 47.7695 | |
| 2 | 27.3000 | 77.2034 | |
| 3 | 27.3076 | 60.6369 | |
| 4 | 27.3518 | 55.2184 | |

| | Boiler 2 Fuel Flow Rate (L/h) | Incinerator 1 Fuel Flow Rate (L/h) | W |
|---|-------------------------------|------------------------------------|---|
| 0 | 0.0 | 19.0090 | |
| 1 | 0.0 | 216.3180 | |
| 2 | 0.0 | 439.4300 | |
| 3 | 0.0 | 218.2797 | |
| 4 | 0.0 | 0.0000 | |

| | Diesel Generator 1 Power (MW) | Diesel Generator 2 Power (MW) | W |
|---|-------------------------------|-------------------------------|---|
| 0 | 0.0 | 0.0 | |
| 1 | 0.0 | 0.0 | |
| 2 | 0.0 | 0.0 | |
| 3 | 0.0 | 0.0 | |
| 4 | 0.0 | 0.0 | |

| | Diesel Generator 3 Power (MW) | Diesel Generator 4 Power (MW) | W |
|---|-------------------------------|-------------------------------|---|
| 0 | 0.0 | 7.3349 | |
| 1 | 0.0 | 7.3011 | |
| 2 | 0.0 | 7.3299 | |
| 3 | 0.0 | 7.3712 | |
| 4 | 0.0 | 7.3032 | |

| | Latitude (Degrees) | Longitude (Degrees) | Relative Wind Angle (Degrees) | W |
|---|--------------------|---------------------|-------------------------------|---|
| 0 | 17.72523 | -65.45738 | 8.4428 | |
| 1 | 17.73088 | -65.44803 | 41.3100 | |
| 2 | 17.73655 | -65.43887 | 23.9997 | |
| 3 | 17.74202 | -65.42980 | 14.5540 | |
| 4 | 17.74713 | -65.42042 | 14.5632 | |

| | True Wind Angle (Degrees) | Depth (m) | Relative Wind Direction (Degrees) | W |
|---|---------------------------|-----------|-----------------------------------|---|
| 0 | 10.9049 | NaN | 64.3112 | |
| 1 | 78.7817 | NaN | 62.8161 | |
| 2 | 33.6216 | NaN | 80.7356 | |
| 3 | 20.0348 | NaN | 75.9723 | |
| 4 | 20.0328 | NaN | 74.6509 | |

| | True Wind Direction (Degrees) | Draft (m) | Speed Over Ground (knots) | W |
|--|-------------------------------|-----------|---------------------------|---|
|--|-------------------------------|-----------|---------------------------|---|

| | | | |
|---|---------|--------|--------|
| 0 | 66.7735 | 7.8721 | 7.6300 |
| 1 | 64.3452 | 7.8713 | 7.5800 |
| 2 | 90.3574 | 7.8718 | 7.4379 |
| 3 | 81.4529 | 7.8710 | 7.3979 |
| 4 | 80.1204 | 7.8707 | 7.4343 |

| | True Wind Speed (knots) | Relative Wind Speed (knots) | W |
|---|-------------------------|-----------------------------|---|
| 0 | 19.5050 | 27.0579 | |
| 1 | 19.2968 | 26.8067 | |
| 2 | 19.4491 | 25.8380 | |
| 3 | 20.6231 | 27.6498 | |
| 4 | 20.4554 | 27.5341 | |

| | Speed Through Water (knots) | Local Time (h) | Trim (m) | W |
|---|-----------------------------|----------------|----------|---|
| 0 | 7.8881 | 19.67367 | -0.1425 | |
| 1 | 7.7438 | 19.75763 | -0.1405 | |
| 2 | 7.6320 | 19.84158 | -0.1450 | |
| 3 | 7.5080 | 19.92551 | -0.1308 | |
| 4 | 7.5521 | 20.00947 | -0.1269 | |

| | Propulsion Power (MW) | Port Side Propulsion Power (MW) | W |
|---|-----------------------|---------------------------------|---|
| 0 | 1.8691 | 0.8854 | |
| 1 | 1.8622 | 0.8737 | |
| 2 | 1.8036 | 0.8441 | |
| 3 | 1.8457 | 0.8543 | |
| 4 | 1.8399 | 0.8467 | |

| | Starboard Side Propulsion Power (MW) | Bow Thruster 1 Power (MW) | W |
|---|--------------------------------------|---------------------------|---|
| 0 | 0.9837 | 0.0 | |
| 1 | 0.9885 | 0.0 | |
| 2 | 0.9595 | 0.0 | |
| 3 | 0.9914 | 0.0 | |
| 4 | 0.9932 | 0.0 | |

| | Bow Thruster 2 Power (MW) | Bow Thruster 3 Power (MW) | W |
|---|---------------------------|---------------------------|---|
| 0 | 0.0 | 0.0 | |
| 1 | 0.0 | 0.0 | |
| 2 | 0.0 | 0.0 | |
| 3 | 0.0 | 0.0 | |
| 4 | 0.0 | 0.0 | |

| | Stern Thruster 1 Power (MW) | Stern Thruster 2 Power (MW) | W |
|---|-----------------------------|-----------------------------|---|
| 0 | 0.0 | 0.0 | |
| 1 | 0.0 | 0.0 | |
| 2 | 0.0 | 0.0 | |
| 3 | 0.0 | 0.0 | |
| 4 | 0.0 | 0.0 | |

| | Main Engine 1 Fuel Flow Rate (kg/h) | Main Engine 2 Fuel Flow Rate (kg/h) | W |
|---|-------------------------------------|-------------------------------------|---|
| 0 | 0.0 | 0.0 | |
| 1 | 0.0 | 0.0 | |
| 2 | 0.0 | 0.0 | |
| 3 | 0.0 | 0.0 | |
| 4 | 0.0 | 0.0 | |

| | Main Engine 3 Fuel Flow Rate (kg/h) | Main Engine 4 Fuel Flow Rate (kg/h) |
|---|-------------------------------------|-------------------------------------|
| 0 | 0.0 | 1645.82000 |
| 1 | 0.0 | 1643.78999 |
| 2 | 0.0 | 1642.07000 |
| 3 | 0.0 | 1650.71000 |
| 4 | 0.0 | 1644.54000 |

Step : Data Understanding

```
In [4]: df.shape
```

```
Out[4]: (210240, 44)
```

```
In [5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 210240 entries, 0 to 210239
Data columns (total 44 columns):
#   Column                                          Non-Null Count  Dtype
---  -
0   Start Time                                    210240 non-null object
1   End Time                                      210240 non-null object
2   Vessel Name                                  210240 non-null object
3   Power Galley 1 (MW)                         210224 non-null float64
4   Power Galley 2 (MW)                         210224 non-null float64
5   Power Service (MW)                         210222 non-null float64
6   HVAC Chiller 1 Power (MW)                  210033 non-null float64
7   HVAC Chiller 2 Power (MW)                  210033 non-null float64
8   HVAC Chiller 3 Power (MW)                  210033 non-null float64
9   Scrubber Power (MW)                        210224 non-null float64
10  Sea Temperature (Celsius)                   210224 non-null float64
11  Boiler 1 Fuel Flow Rate (L/h)               210224 non-null float64
12  Boiler 2 Fuel Flow Rate (L/h)               210224 non-null float64
13  Incinerator 1 Fuel Flow Rate (L/h)          210224 non-null float64
14  Diesel Generator 1 Power (MW)               210224 non-null float64
15  Diesel Generator 2 Power (MW)               210224 non-null float64
16  Diesel Generator 3 Power (MW)               210224 non-null float64
17  Diesel Generator 4 Power (MW)               210224 non-null float64
18  Latitude (Degrees)                          209900 non-null float64
19  Longitude (Degrees)                        209900 non-null float64
20  Relative Wind Angle (Degrees)               210226 non-null float64
21  True Wind Angle (Degrees)                   210166 non-null float64
22  Depth (m)                                   152746 non-null float64
23  Relative Wind Direction (Degrees)           210185 non-null float64
24  True Wind Direction (Degrees)               210166 non-null float64
25  Draft (m)                                   209097 non-null float64
26  Speed Over Ground (knots)                   209340 non-null float64
27  True Wind Speed (knots)                     210166 non-null float64
28  Relative Wind Speed (knots)                 210226 non-null float64
29  Speed Through Water (knots)                 209299 non-null float64
30  Local Time (h)                              209900 non-null float64
31  Trim (m)                                    209161 non-null float64
32  Propulsion Power (MW)                       210224 non-null float64
33  Port Side Propulsion Power (MW)             210224 non-null float64
34  Starboard Side Propulsion Power (MW)        210224 non-null float64
35  Bow Thruster 1 Power (MW)                  210224 non-null float64
36  Bow Thruster 2 Power (MW)                  210224 non-null float64
37  Bow Thruster 3 Power (MW)                  210224 non-null float64
38  Stern Thruster 1 Power (MW)                210224 non-null float64
39  Stern Thruster 2 Power (MW)                210224 non-null float64
40  Main Engine 1 Fuel Flow Rate (kg/h)         210224 non-null float64
41  Main Engine 2 Fuel Flow Rate (kg/h)         210224 non-null float64
42  Main Engine 3 Fuel Flow Rate (kg/h)         210224 non-null float64
43  Main Engine 4 Fuel Flow Rate (kg/h)         210224 non-null float64
dtypes: float64(41), object(3)
memory usage: 70.6+ MB
```

```
In [6]: df.head()
```

Out[6]:

| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | Tem |
|---|---------------------|---------------------|-------------|---------------------|---------------------|--------------------|---------------------------|---------------------------|---------------------------|---------------------|-----|
| 0 | 2023-01-01T00:00:00 | 2023-01-01T00:05:00 | Vessel 1 | 0.0946 | 0.1384 | 5.4654 | 0.5074 | 0.0 | 0.4979 | 0.4191 | |
| 1 | 2023-01-01T00:05:00 | 2023-01-01T00:10:00 | Vessel 1 | 0.0540 | 0.1370 | 5.4387 | 0.5158 | 0.0 | 0.4982 | 0.4204 | |
| 2 | 2023-01-01T00:10:00 | 2023-01-01T00:15:00 | Vessel 1 | 0.0439 | 0.1785 | 5.5265 | 0.5117 | 0.0 | 0.5032 | 0.4199 | |
| 3 | 2023-01-01T00:15:00 | 2023-01-01T00:20:00 | Vessel 1 | 0.0733 | 0.1725 | 5.5257 | 0.5177 | 0.0 | 0.5103 | 0.4188 | |
| 4 | 2023-01-01T00:20:00 | 2023-01-01T00:25:00 | Vessel 1 | 0.0780 | 0.1397 | 5.4634 | 0.5169 | 0.0 | 0.5100 | 0.4203 | |

In [7]: `df.columns`

Out[7]:

```
Index(['Start Time', 'End Time', 'Vessel Name', 'Power Galley 1 (MW)',  
      'Power Galley 2 (MW)', 'Power Service (MW)',  
      'HVAC Chiller 1 Power (MW)', 'HVAC Chiller 2 Power (MW)',  
      'HVAC Chiller 3 Power (MW)', 'Scrubber Power (MW)',  
      'Sea Temperature (Celsius)', 'Boiler 1 Fuel Flow Rate (L/h)',  
      'Boiler 2 Fuel Flow Rate (L/h)', 'Incinerator 1 Fuel Flow Rate (L/h)',  
      'Diesel Generator 1 Power (MW)', 'Diesel Generator 2 Power (MW)',  
      'Diesel Generator 3 Power (MW)', 'Diesel Generator 4 Power (MW)',  
      'Latitude (Degrees)', 'Longitude (Degrees)',  
      'Relative Wind Angle (Degrees)', 'True Wind Angle (Degrees)',  
      'Depth (m)', 'Relative Wind Direction (Degrees)',  
      'True Wind Direction (Degrees)', 'Draft (m)',  
      'Speed Over Ground (knots)', 'True Wind Speed (knots)',  
      'Relative Wind Speed (knots)', 'Speed Through Water (knots)',  
      'Local Time (h)', 'Trim (m)', 'Propulsion Power (MW)',  
      'Port Side Propulsion Power (MW)',  
      'Starboard Side Propulsion Power (MW)', 'Bow Thruster 1 Power (MW)',  
      'Bow Thruster 2 Power (MW)', 'Bow Thruster 3 Power (MW)',  
      'Stern Thruster 1 Power (MW)', 'Stern Thruster 2 Power (MW)',  
      'Main Engine 1 Fuel Flow Rate (kg/h)',  
      'Main Engine 2 Fuel Flow Rate (kg/h)',  
      'Main Engine 3 Fuel Flow Rate (kg/h)',  
      'Main Engine 4 Fuel Flow Rate (kg/h)'],  
      dtype='object')
```

Step : Data Preperation

In [8]:

```
# Convert time columns to datetime dtype  
df['Start Time'] = pd.to_datetime(df['Start Time'])  
df['End Time'] = pd.to_datetime(df['End Time'])  
  
# Display info for only 'Start Time' and 'End Time' columns  
df[['Start Time', 'End Time']].info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 210240 entries, 0 to 210239
Data columns (total 2 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Start Time  210240 non-null  datetime64[ns]
1   End Time    210240 non-null  datetime64[ns]
dtypes: datetime64[ns](2)
memory usage: 3.2 MB
```

Missing value check

```
In [9]: df.isnull().sum()
```

```
Out[9]: Start Time                                0
End Time                                          0
Vessel Name                                      0
Power Galley 1 (MW)                             16
Power Galley 2 (MW)                             16
Power Service (MW)                              18
HVAC Chiller 1 Power (MW)                       207
HVAC Chiller 2 Power (MW)                       207
HVAC Chiller 3 Power (MW)                       207
Scrubber Power (MW)                             16
Sea Temperature (Celsius)                       16
Boiler 1 Fuel Flow Rate (L/h)                   16
Boiler 2 Fuel Flow Rate (L/h)                   16
Incinerator 1 Fuel Flow Rate (L/h)              16
Diesel Generator 1 Power (MW)                   16
Diesel Generator 2 Power (MW)                   16
Diesel Generator 3 Power (MW)                   16
Diesel Generator 4 Power (MW)                   16
Latitude (Degrees)                             340
Longitude (Degrees)                            340
Relative Wind Angle (Degrees)                   14
True Wind Angle (Degrees)                       74
Depth (m)                                       57494
Relative Wind Direction (Degrees)               55
True Wind Direction (Degrees)                   74
Draft (m)                                       1143
Speed Over Ground (knots)                       900
True Wind Speed (knots)                         74
Relative Wind Speed (knots)                     14
Speed Through Water (knots)                     941
Local Time (h)                                  340
Trim (m)                                        1079
Propulsion Power (MW)                           16
Port Side Propulsion Power (MW)                 16
Starboard Side Propulsion Power (MW)            16
Bow Thruster 1 Power (MW)                       16
Bow Thruster 2 Power (MW)                       16
Bow Thruster 3 Power (MW)                       16
Stern Thruster 1 Power (MW)                     16
Stern Thruster 2 Power (MW)                     16
Main Engine 1 Fuel Flow Rate (kg/h)             16
Main Engine 2 Fuel Flow Rate (kg/h)             16
Main Engine 3 Fuel Flow Rate (kg/h)             16
Main Engine 4 Fuel Flow Rate (kg/h)             16
dtype: int64
```

```
In [10]: df.duplicated().sum()
```

Out[10]: 0

In [11]: `df[df['Power Galley 1 (MW)'].isna()]`

Out[11]:

| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | Temp |
|--------|---------------------|---------------------|-------------|---------------------|---------------------|--------------------|---------------------------|---------------------------|---------------------------|---------------------|------|
| 88578 | 2023-11-04 13:30:00 | 2023-11-04 13:35:00 | Vessel 1 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 140167 | 2023-05-02 16:35:00 | 2023-05-02 16:40:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 145233 | 2023-05-20 06:45:00 | 2023-05-20 06:50:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 161071 | 2023-07-14 06:35:00 | 2023-07-14 06:40:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 163498 | 2023-07-22 16:50:00 | 2023-07-22 16:55:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 163499 | 2023-07-22 16:55:00 | 2023-07-22 17:00:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 163500 | 2023-07-22 17:00:00 | 2023-07-22 17:05:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 165191 | 2023-07-28 13:55:00 | 2023-07-28 14:00:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 165767 | 2023-07-30 13:55:00 | 2023-07-30 14:00:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 165768 | 2023-07-30 14:00:00 | 2023-07-30 14:05:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 168077 | 2023-08-07 14:25:00 | 2023-08-07 14:30:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 168078 | 2023-08-07 14:30:00 | 2023-08-07 14:35:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 179568 | 2023-09-16 12:00:00 | 2023-09-16 12:05:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 179569 | 2023-09-16 12:05:00 | 2023-09-16 12:10:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |
| 179572 | 2023-09-16 12:20:00 | 2023-09-16 12:25:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |

| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | Temp |
|--------|---------------------|---------------------|----------------|------------------------------|------------------------------|--------------------------|---------------------------------------|---------------------------------------|---------------------------------------|---------------------------|------|
| 209699 | 2023-12-30 02:55:00 | 2023-12-30 03:00:00 | Vessel 2 | NaN | NaN | NaN | NaN | NaN | NaN | NaN | |

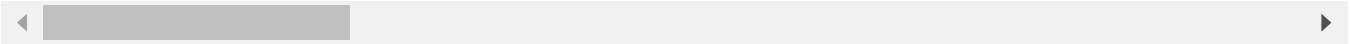
```
In [12]: # Drop rows where 'Power Galley 1 (MW)' has missing values because most of the data is missing
df = df.dropna(subset=['Power Galley 1 (MW)'])
```

```
In [13]: df[df['HVAC Chiller 1 Power (MW)'].isna()]
```

Out[13]:

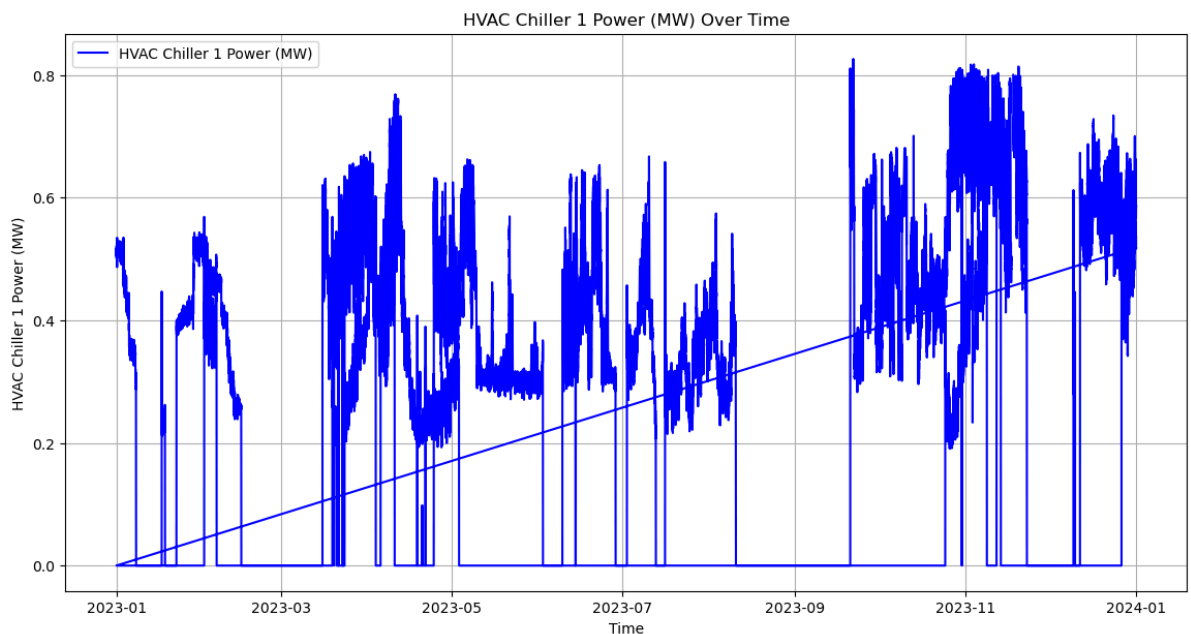
| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | Temp |
|--------|---------------------|---------------------|-------------|---------------------|---------------------|--------------------|---------------------------|---------------------------|---------------------------|---------------------|------|
| 197377 | 2023-11-17 08:05:00 | 2023-11-17 08:10:00 | Vessel 2 | 0.0000 | 0.0918 | 4.4755 | NaN | NaN | NaN | 0.1543 | |
| 197378 | 2023-11-17 08:10:00 | 2023-11-17 08:15:00 | Vessel 2 | 0.0000 | 0.0716 | 4.5178 | NaN | NaN | NaN | 0.1534 | |
| 197379 | 2023-11-17 08:15:00 | 2023-11-17 08:20:00 | Vessel 2 | 0.0000 | 0.0638 | 4.4713 | NaN | NaN | NaN | 0.1530 | |
| 197380 | 2023-11-17 08:20:00 | 2023-11-17 08:25:00 | Vessel 2 | 0.0071 | 0.0620 | 4.5521 | NaN | NaN | NaN | 0.1545 | |
| 197381 | 2023-11-17 08:25:00 | 2023-11-17 08:30:00 | Vessel 2 | 0.0012 | 0.0420 | 4.4380 | NaN | NaN | NaN | 0.1539 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | |
| 197563 | 2023-11-17 23:35:00 | 2023-11-17 23:40:00 | Vessel 2 | 0.0218 | 0.1772 | 5.5094 | NaN | NaN | NaN | 0.1577 | |
| 197564 | 2023-11-17 23:40:00 | 2023-11-17 23:45:00 | Vessel 2 | 0.0487 | 0.1671 | 5.5004 | NaN | NaN | NaN | 0.1569 | |
| 197565 | 2023-11-17 23:45:00 | 2023-11-17 23:50:00 | Vessel 2 | 0.0308 | 0.1678 | 5.5444 | NaN | NaN | NaN | 0.1581 | |
| 197566 | 2023-11-17 23:50:00 | 2023-11-17 23:55:00 | Vessel 2 | 0.0497 | 0.1516 | 5.5213 | NaN | NaN | NaN | 0.1584 | |
| 197567 | 2023-11-17 23:55:00 | 2023-11-18 00:00:00 | Vessel 2 | 0.0697 | 0.2012 | 5.6888 | NaN | NaN | NaN | 0.1585 | |

191 rows × 44 columns



```
In [14]: # Ensure 'Start Time' is in datetime format
df['Start Time'] = pd.to_datetime(df['Start Time'])

# Plot the time series for 'HVAC Chiller 1 Power (MW)'
plt.figure(figsize=(14, 7))
plt.plot(df['Start Time'], df['HVAC Chiller 1 Power (MW)'], label='HVAC Chiller 1 Power (MW)')
plt.xlabel('Time')
plt.ylabel('HVAC Chiller 1 Power (MW)')
plt.title('HVAC Chiller 1 Power (MW) Over Time')
plt.legend()
plt.grid(True)
plt.show()
```



```
In [15]: # Sort the dataframe by 'Start Time' to ensure proper filling
df = df.sort_values(by='Start Time')

# Set 'Start Time' as the index
df.set_index('Start Time', inplace=True)

# Interpolate to fill missing values for chiller columns
#interpolation is a process of determining the unknown values that lie in between the
df['HVAC Chiller 1 Power (MW)'] = df['HVAC Chiller 1 Power (MW)'].interpolate()
df['HVAC Chiller 2 Power (MW)'] = df['HVAC Chiller 2 Power (MW)'].interpolate()
df['HVAC Chiller 3 Power (MW)'] = df['HVAC Chiller 3 Power (MW)'].interpolate()

# If you need to reset the index back to columns
df.reset_index(inplace=True)
```

```
In [16]: df[df['Power Service (MW)'].isna()]
```

Out[16]:

| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | Temp |
|--------|---------------------|---------------------|-------------|---------------------|---------------------|--------------------|---------------------------|---------------------------|---------------------------|---------------------|------|
| 148890 | 2023-09-16 12:15:00 | 2023-09-16 12:20:00 | Vessel 2 | 0.03 | 0.080 | NaN | 0.000 | 0.384 | 0.0 | 0.773 | |
| 209145 | 2023-12-30 03:00:00 | 2023-12-30 03:05:00 | Vessel 2 | 0.03 | 0.087 | NaN | 0.508 | 0.528 | 0.0 | 0.346 | |

```
In [17]: # First forward fill, then backward fill
# the next available value after the missing data point replaces the missing value be
df['Power Service (MW)'] = df['Power Service (MW)'].ffill().bfill()

print(df[['Start Time', 'Power Service (MW)']].head())
```

| | Start Time | Power Service (MW) |
|---|---------------------|--------------------|
| 0 | 2023-01-01 00:00:00 | 5.4654 |
| 1 | 2023-01-01 00:00:00 | 5.3725 |
| 2 | 2023-01-01 00:05:00 | 5.4387 |
| 3 | 2023-01-01 00:05:00 | 5.9642 |
| 4 | 2023-01-01 00:10:00 | 5.5265 |

In [18]: `df[df['Speed Over Ground (knots)'].isna()]`

Out[18]:

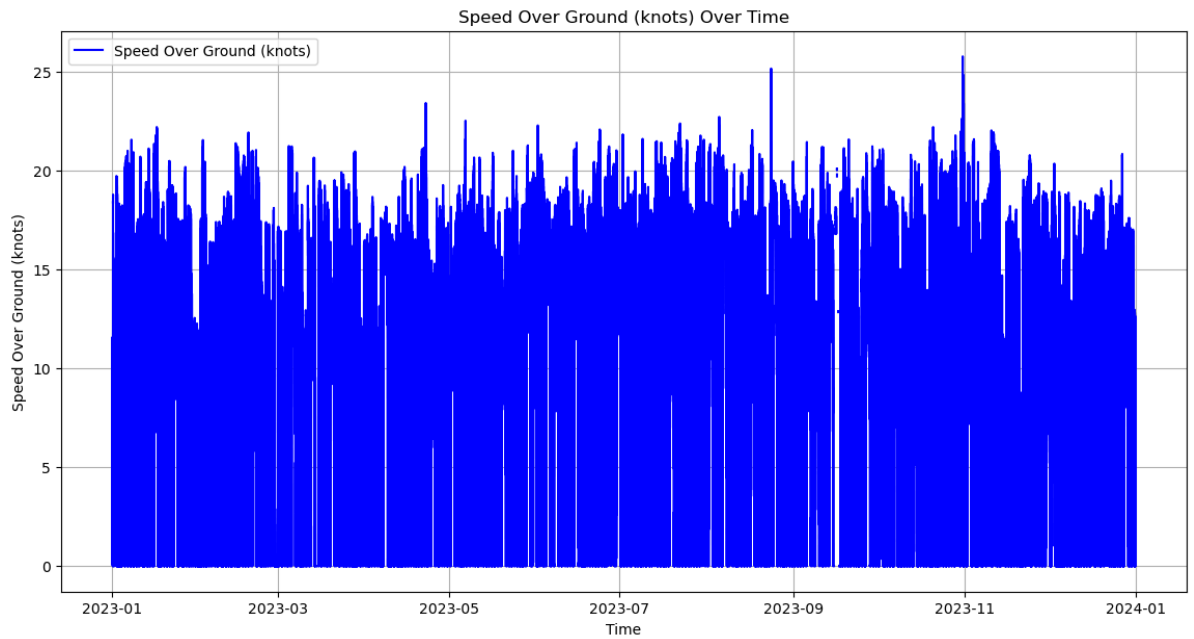
| | Start Time | End Time | Vessel Name | Power Galley 1 (MW) | Power Galley 2 (MW) | Power Service (MW) | HVAC Chiller 1 Power (MW) | HVAC Chiller 2 Power (MW) | HVAC Chiller 3 Power (MW) | Scrubber Power (MW) | Temp |
|---------------|---------------------|---------------------|-------------|---------------------|---------------------|--------------------|---------------------------|---------------------------|---------------------------|---------------------|------|
| 137881 | 2023-08-28 09:30:00 | 2023-08-28 09:35:00 | Vessel 2 | 0.1025 | 0.1235 | -0.0400 | 0.0000 | 0.0 | 0.4086 | 0.0000 | |
| 137884 | 2023-08-28 09:35:00 | 2023-08-28 09:40:00 | Vessel 2 | 0.1113 | 0.1194 | -0.0400 | 0.0000 | 0.0 | 0.4061 | 0.0000 | |
| 137885 | 2023-08-28 09:40:00 | 2023-08-28 09:45:00 | Vessel 2 | 0.0895 | 0.0986 | -0.0400 | 0.0000 | 0.0 | 0.4006 | 0.0000 | |
| 137887 | 2023-08-28 09:45:00 | 2023-08-28 09:50:00 | Vessel 2 | 0.1265 | 0.1295 | -0.0400 | 0.0000 | 0.0 | 0.3959 | 0.0000 | |
| 137889 | 2023-08-28 09:50:00 | 2023-08-28 09:55:00 | Vessel 2 | 0.1095 | 0.1345 | -0.0400 | 0.0000 | 0.0 | 0.3969 | 0.0000 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | |
| 153941 | 2023-09-25 06:45:00 | 2023-09-25 06:50:00 | Vessel 2 | 0.0281 | 0.1347 | 5.8139 | 0.4105 | 0.0 | 0.0000 | 0.7854 | |
| 153943 | 2023-09-25 06:50:00 | 2023-09-25 06:55:00 | Vessel 2 | 0.0195 | 0.1565 | 5.8844 | 0.4183 | 0.0 | 0.0000 | 0.7832 | |
| 153945 | 2023-09-25 06:55:00 | 2023-09-25 07:00:00 | Vessel 2 | 0.0317 | 0.1817 | 5.8343 | 0.4083 | 0.0 | 0.0000 | 0.7865 | |
| 153947 | 2023-09-25 07:00:00 | 2023-09-25 07:05:00 | Vessel 2 | 0.0166 | 0.2210 | 5.9644 | 0.4213 | 0.0 | 0.0000 | 0.7857 | |
| 153949 | 2023-09-25 07:05:00 | 2023-09-25 07:10:00 | Vessel 2 | 0.0486 | 0.2097 | 5.9657 | 0.4175 | 0.0 | 0.0000 | 0.7886 | |

886 rows × 44 columns

In [19]:

```
# Plot the time series for 'Speed Over Ground (knots)'
plt.figure(figsize=(14, 7))
plt.plot(df['Start Time'], df['Speed Over Ground (knots)'], label='Speed Over Ground')
plt.xlabel('Time')
plt.ylabel('Speed Over Ground (knots)')
```

```
plt.title('Speed Over Ground (knots) Over Time')
plt.legend()
plt.grid(True)
plt.show()
```



```
In [20]: # Sort the dataframe by 'Start Time' to ensure proper filling
df = df.sort_values(by='Start Time')

# Set 'Start Time' as the index
df.set_index('Start Time', inplace=True)

# Interpolate to fill missing values
df['Speed Over Ground (knots)'] = df['Speed Over Ground (knots)'].interpolate()

# If you need to reset the index back to columns
df.reset_index(inplace=True)
```

```
In [21]: # # missing value check
df.isnull().sum()
```

```

Out[21]: Start Time                                0
        End Time                                    0
        Vessel Name                                0
        Power Galley 1 (MW)                        0
        Power Galley 2 (MW)                        0
        Power Service (MW)                         0
        HVAC Chiller 1 Power (MW)                  0
        HVAC Chiller 2 Power (MW)                  0
        HVAC Chiller 3 Power (MW)                  0
        Scrubber Power (MW)                        0
        Sea Temperature (Celsius)                  0
        Boiler 1 Fuel Flow Rate (L/h)               0
        Boiler 2 Fuel Flow Rate (L/h)               0
        Incinerator 1 Fuel Flow Rate (L/h)          0
        Diesel Generator 1 Power (MW)               0
        Diesel Generator 2 Power (MW)               0
        Diesel Generator 3 Power (MW)               0
        Diesel Generator 4 Power (MW)               0
        Latitude (Degrees)                          326
        Longitude (Degrees)                         326
        Relative Wind Angle (Degrees)                0
        True Wind Angle (Degrees)                   60
        Depth (m)                                   57479
        Relative Wind Direction (Degrees)            41
        True Wind Direction (Degrees)                60
        Draft (m)                                    1127
        Speed Over Ground (knots)                   0
        True Wind Speed (knots)                     60
        Relative Wind Speed (knots)                  0
        Speed Through Water (knots)                  927
        Local Time (h)                              326
        Trim (m)                                    1063
        Propulsion Power (MW)                       0
        Port Side Propulsion Power (MW)              0
        Starboard Side Propulsion Power (MW)         0
        Bow Thruster 1 Power (MW)                   0
        Bow Thruster 2 Power (MW)                   0
        Bow Thruster 3 Power (MW)                   0
        Stern Thruster 1 Power (MW)                 0
        Stern Thruster 2 Power (MW)                 0
        Main Engine 1 Fuel Flow Rate (kg/h)          0
        Main Engine 2 Fuel Flow Rate (kg/h)          0
        Main Engine 3 Fuel Flow Rate (kg/h)          0
        Main Engine 4 Fuel Flow Rate (kg/h)          0
        dtype: int64

```

```

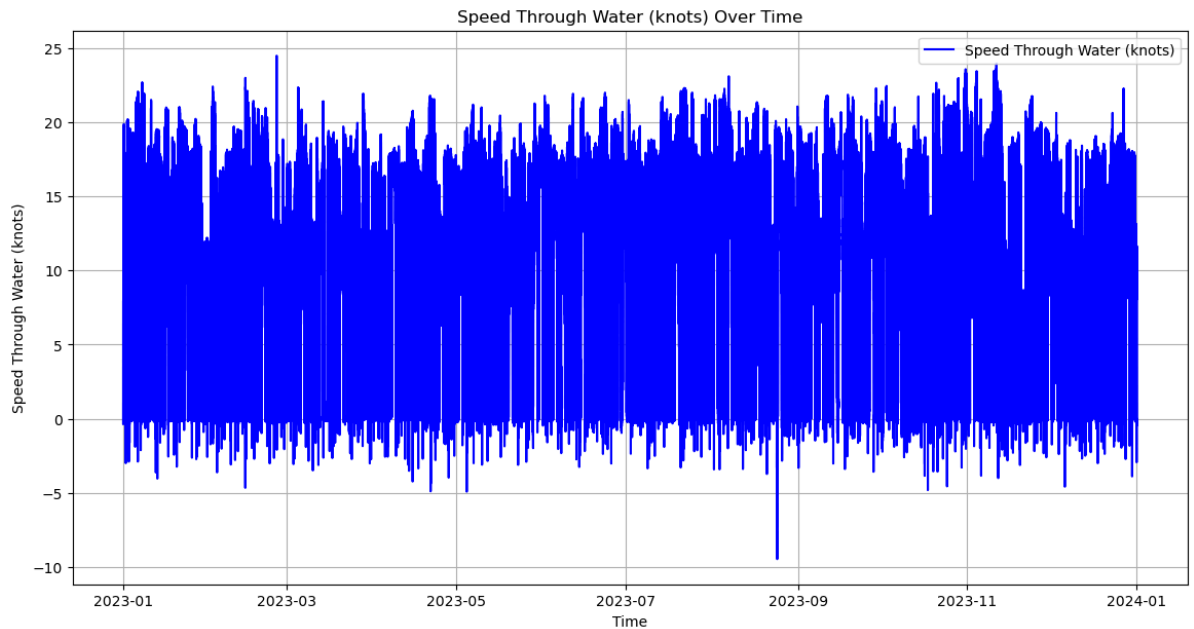
In [22]: # Set 'Start Time' as the index
df.set_index('Start Time', inplace=True)

# Interpolate to fill missing values
df['Speed Through Water (knots)'] = df['Speed Through Water (knots)'].interpolate()

# If you need to reset the index back to columns
df.reset_index(inplace=True)

# Plot the time series for 'Speed Through Water (knots)'
plt.figure(figsize=(14, 7))
plt.plot(df['Start Time'], df['Speed Through Water (knots)'], label='Speed Through W
plt.xlabel('Time')
plt.ylabel('Speed Through Water (knots)')
plt.title('Speed Through Water (knots) Over Time')
plt.legend()
plt.grid(True)
plt.show()

```



```
In [23]: # Set 'Start Time' as the index
df.set_index('Start Time', inplace=True)

# Interpolate to fill missing values
df['Speed Through Water (knots)'] = df['Speed Through Water (knots)'].interpolate()

# If you need to reset the index back to columns
df.reset_index(inplace=True)

print(df[['Start Time', 'Speed Through Water (knots)']].head())
```

| | Start Time | Speed Through Water (knots) |
|---|---------------------|-----------------------------|
| 0 | 2023-01-01 00:00:00 | 7.8881 |
| 1 | 2023-01-01 00:00:00 | -0.1337 |
| 2 | 2023-01-01 00:05:00 | 7.7438 |
| 3 | 2023-01-01 00:05:00 | -0.3794 |
| 4 | 2023-01-01 00:10:00 | 7.6320 |

```
In [24]: df.isnull().sum()
```

```

Out[24]: Start Time                                0
End Time                                           0
Vessel Name                                       0
Power Galley 1 (MW)                             0
Power Galley 2 (MW)                             0
Power Service (MW)                              0
HVAC Chiller 1 Power (MW)                       0
HVAC Chiller 2 Power (MW)                       0
HVAC Chiller 3 Power (MW)                       0
Scrubber Power (MW)                             0
Sea Temperature (Celsius)                       0
Boiler 1 Fuel Flow Rate (L/h)                   0
Boiler 2 Fuel Flow Rate (L/h)                   0
Incinerator 1 Fuel Flow Rate (L/h)              0
Diesel Generator 1 Power (MW)                   0
Diesel Generator 2 Power (MW)                   0
Diesel Generator 3 Power (MW)                   0
Diesel Generator 4 Power (MW)                   0
Latitude (Degrees)                             326
Longitude (Degrees)                            326
Relative Wind Angle (Degrees)                   0
True Wind Angle (Degrees)                       60
Depth (m)                                       57479
Relative Wind Direction (Degrees)               41
True Wind Direction (Degrees)                   60
Draft (m)                                       1127
Speed Over Ground (knots)                       0
True Wind Speed (knots)                         60
Relative Wind Speed (knots)                     0
Speed Through Water (knots)                     0
Local Time (h)                                 326
Trim (m)                                       1063
Propulsion Power (MW)                           0
Port Side Propulsion Power (MW)                 0
Starboard Side Propulsion Power (MW)            0
Bow Thruster 1 Power (MW)                       0
Bow Thruster 2 Power (MW)                       0
Bow Thruster 3 Power (MW)                       0
Stern Thruster 1 Power (MW)                     0
Stern Thruster 2 Power (MW)                     0
Main Engine 1 Fuel Flow Rate (kg/h)             0
Main Engine 2 Fuel Flow Rate (kg/h)             0
Main Engine 3 Fuel Flow Rate (kg/h)             0
Main Engine 4 Fuel Flow Rate (kg/h)             0
dtype: int64

```

Sorting Vessels data to Vessel 1 & Vessel 2

```

In [25]: # Assuming 'Vessel Name' is the column indicating the vessel
vessel_1_data = df[df['Vessel Name'] == 'Vessel 1']
vessel_2_data = df[df['Vessel Name'] == 'Vessel 2']

# Verify the split
print(vessel_1_data.shape)
print(vessel_2_data.shape)

(105119, 44)
(105105, 44)

```

```

In [26]: df.columns

```



```
Out[26]: Index(['Start Time', 'End Time', 'Vessel Name', 'Power Galley 1 (MW)',
        'Power Galley 2 (MW)', 'Power Service (MW)',
        'HVAC Chiller 1 Power (MW)', 'HVAC Chiller 2 Power (MW)',
        'HVAC Chiller 3 Power (MW)', 'Scrubber Power (MW)',
        'Sea Temperature (Celsius)', 'Boiler 1 Fuel Flow Rate (L/h)',
        'Boiler 2 Fuel Flow Rate (L/h)', 'Incinerator 1 Fuel Flow Rate (L/h)',
        'Diesel Generator 1 Power (MW)', 'Diesel Generator 2 Power (MW)',
        'Diesel Generator 3 Power (MW)', 'Diesel Generator 4 Power (MW)',
        'Latitude (Degrees)', 'Longitude (Degrees)',
        'Relative Wind Angle (Degrees)', 'True Wind Angle (Degrees)',
        'Depth (m)', 'Relative Wind Direction (Degrees)',
        'True Wind Direction (Degrees)', 'Draft (m)',
        'Speed Over Ground (knots)', 'True Wind Speed (knots)',
        'Relative Wind Speed (knots)', 'Speed Through Water (knots)',
        'Local Time (h)', 'Trim (m)', 'Propulsion Power (MW)',
        'Port Side Propulsion Power (MW)',
        'Starboard Side Propulsion Power (MW)', 'Bow Thruster 1 Power (MW)',
        'Bow Thruster 2 Power (MW)', 'Bow Thruster 3 Power (MW)',
        'Stern Thruster 1 Power (MW)', 'Stern Thruster 2 Power (MW)',
        'Main Engine 1 Fuel Flow Rate (kg/h)',
        'Main Engine 2 Fuel Flow Rate (kg/h)',
        'Main Engine 3 Fuel Flow Rate (kg/h)',
        'Main Engine 4 Fuel Flow Rate (kg/h)'],
        dtype='object')
```

Outlier check of Power Generation

```
In [27]: power_generation_columns = [
        'Diesel Generator 1 Power (MW)', 'Diesel Generator 2 Power (MW)',
        'Diesel Generator 3 Power (MW)', 'Diesel Generator 4 Power (MW)',
        'HVAC Chiller 1 Power (MW)', 'HVAC Chiller 2 Power (MW)',
        'HVAC Chiller 3 Power (MW)', 'Scrubber Power (MW)',
        'Power Galley 1 (MW)', 'Power Galley 2 (MW)', 'Power Service (MW)'
    ]
```

```
# Assuming 'Vessel Name' is the column indicating the vessel
vessel_1_data = df[df['Vessel Name'] == 'Vessel 1'].copy()
vessel_2_data = df[df['Vessel Name'] == 'Vessel 2'].copy()
```

```
In [28]: # Function to create box plots for outliers visualization
def plot_outliers(df, columns, vessel_name):
    plt.figure(figsize=(20, 10))
    df[columns].boxplot()
    plt.title(f'Outlier Visualization for {vessel_name}', fontsize=20)
    plt.ylabel('Value', fontsize=20)
    plt.xticks(rotation=15, fontsize=20) # Change rotation to 0 for horizontal labels
    plt.yticks(fontsize=20)
    plt.grid(True)
    plt.tight_layout() # Adjust layout to make room for the labels
    plt.show()

# Plot outliers for Vessel 1
plot_outliers(vessel_1_data, power_generation_columns, 'Vessel 1')

# Plot outliers for Vessel 2
plot_outliers(vessel_2_data, power_generation_columns, 'Vessel 2')
```



```

plt.legend()
plt.grid(True)
plt.show()

# Function to calculate and print statistical summaries
def print_summaries(df, column, outliers):
    data_without_outliers = df[~df.index.isin(outliers.index)]
    summary_with_outliers = df[column].describe()
    summary_without_outliers = data_without_outliers[column].describe()

    print(f"Summary with Outliers for {column}:Wn", summary_with_outliers)
    print(f"WnSummary without Outliers for {column}:Wn", summary_without_outliers)

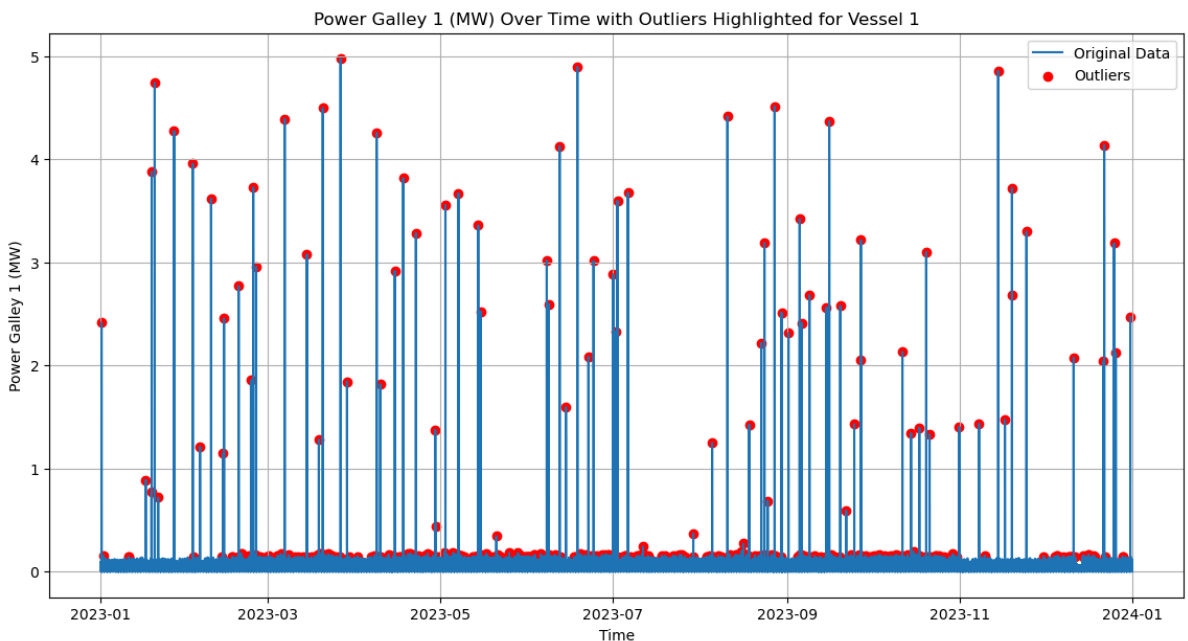
# Columns to check
columns_to_check = ['Power Galley 1 (MW)', 'Power Galley 2 (MW)', 'Power Service (MW)']

# Vessel 1 Analysis
print("Vessel 1 Analysis:Wn")
for column in columns_to_check:
    outliers_vessel_1 = detect_outliers(vessel_1_data, column)
    plot_with_outliers(vessel_1_data, column, outliers_vessel_1, 'Vessel 1')
    print_summaries(vessel_1_data, column, outliers_vessel_1)

# Vessel 2 Analysis
print("WnVessel 2 Analysis:Wn")
for column in columns_to_check:
    outliers_vessel_2 = detect_outliers(vessel_2_data, column)
    plot_with_outliers(vessel_2_data, column, outliers_vessel_2, 'Vessel 2')
    print_summaries(vessel_2_data, column, outliers_vessel_2)

```

Vessel 1 Analysis:



Summary with Outliers for Power Galley 1 (MW):

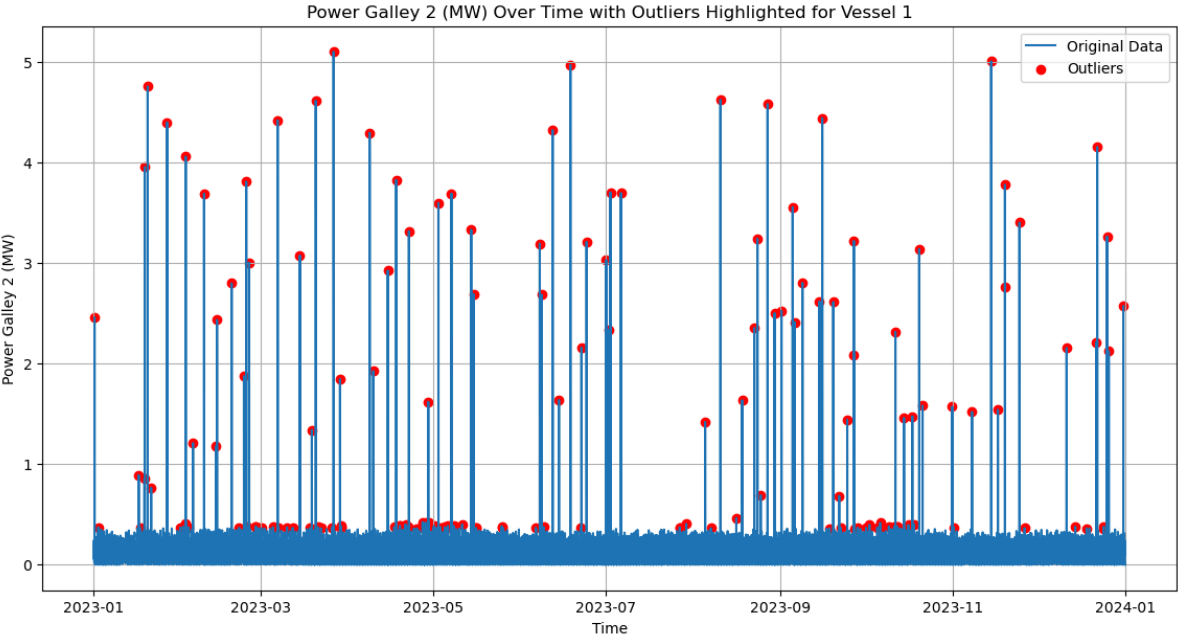
| | |
|-------|---------------|
| count | 105119.000000 |
| mean | 0.041498 |
| std | 0.088231 |
| min | 0.000000 |
| 25% | 0.008800 |
| 50% | 0.030600 |
| 75% | 0.061400 |
| max | 4.973600 |

Name: Power Galley 1 (MW), dtype: float64

Summary without Outliers for Power Galley 1 (MW):

| | |
|-------|---------------|
| count | 104441.000000 |
| mean | 0.038812 |
| std | 0.035026 |
| min | 0.000000 |
| 25% | 0.008700 |
| 50% | 0.030300 |
| 75% | 0.060500 |
| max | 0.140300 |

Name: Power Galley 1 (MW), dtype: float64



Summary with Outliers for Power Galley 2 (MW):

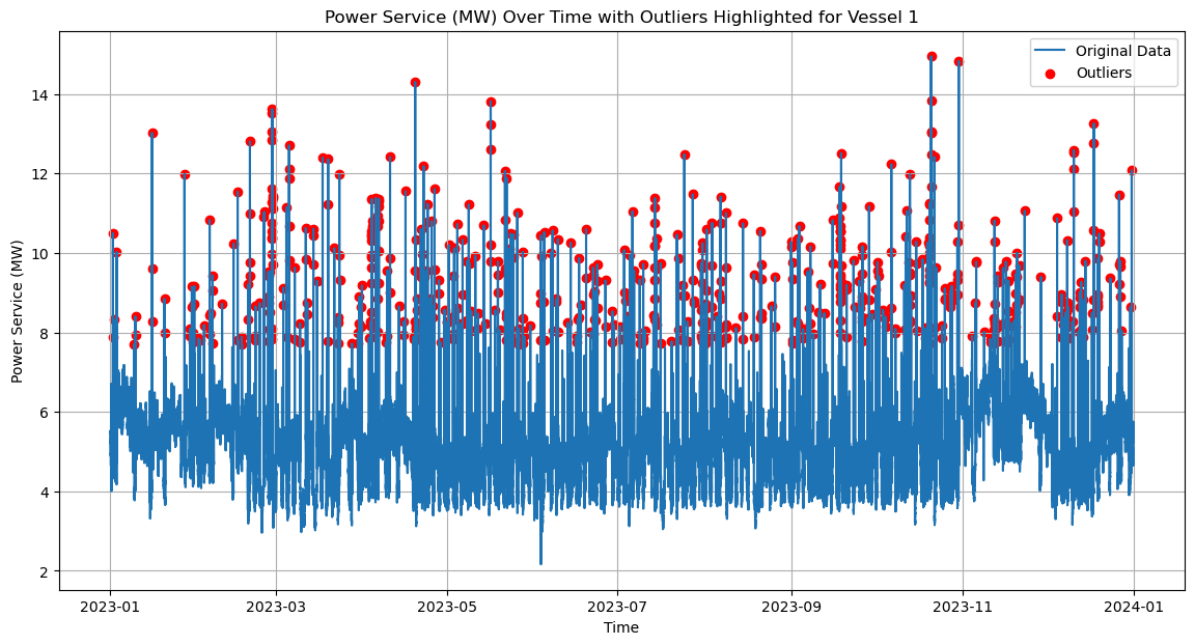
| | |
|-------|---------------|
| count | 105119.000000 |
| mean | 0.119118 |
| std | 0.112306 |
| min | 0.000000 |
| 25% | 0.053400 |
| 50% | 0.106600 |
| 75% | 0.176000 |
| max | 5.097700 |

Name: Power Galley 2 (MW), dtype: float64

Summary without Outliers for Power Galley 2 (MW):

| | |
|-------|---------------|
| count | 104942.000000 |
| mean | 0.116837 |
| std | 0.077931 |
| min | 0.000000 |
| 25% | 0.053300 |
| 50% | 0.106400 |
| 75% | 0.175700 |
| max | 0.359900 |

Name: Power Galley 2 (MW), dtype: float64



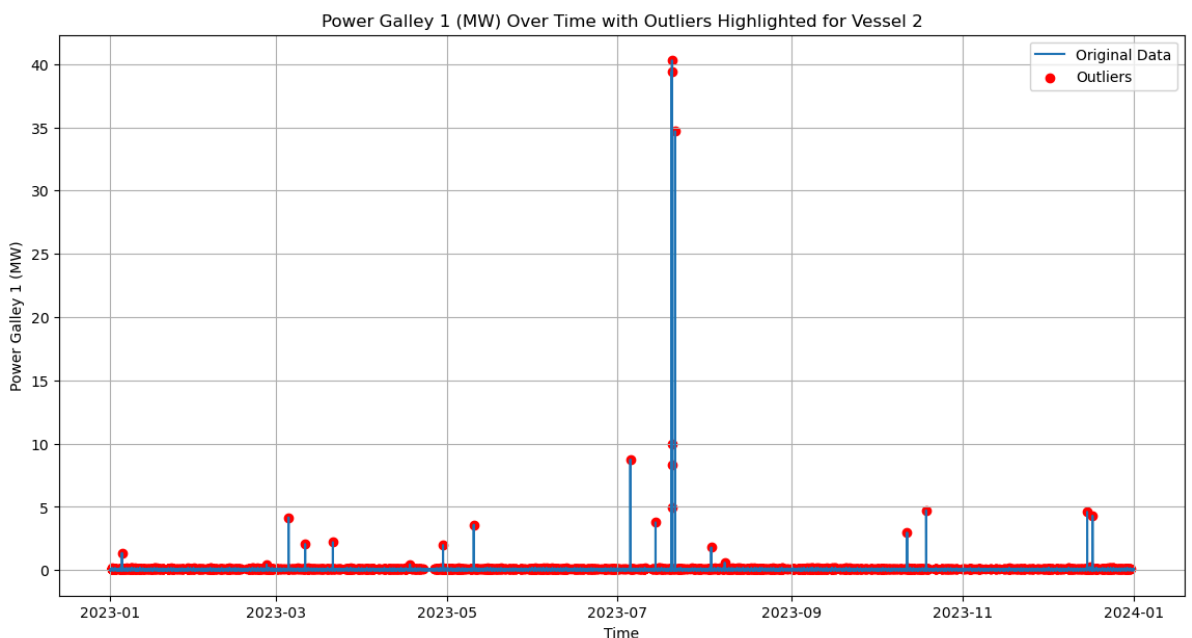
Summary with Outliers for Power Service (MW):

```
count    105119.000000
mean      4.930506
std       0.924900
min       2.162600
25%      4.141750
50%      4.917700
75%      5.567000
max       14.950500
Name: Power Service (MW), dtype: float64
```

Summary without Outliers for Power Service (MW):

```
count    104478.000000
mean      4.903686
std       0.855240
min       2.162600
25%      4.137200
50%      4.910300
75%      5.553100
max       7.700200
Name: Power Service (MW), dtype: float64
```

Vessel 2 Analysis:



Summary with Outliers for Power Galley 1 (MW):

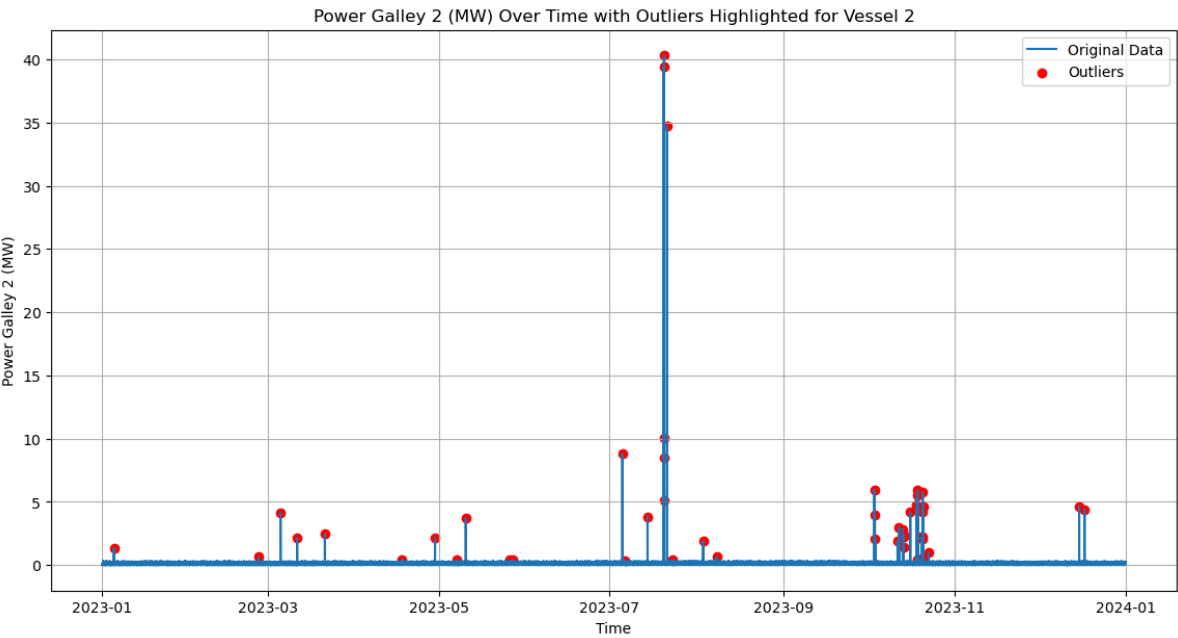
| | |
|-------|---------------|
| count | 105105.000000 |
| mean | 0.034159 |
| std | 0.215483 |
| min | 0.000000 |
| 25% | 0.006400 |
| 50% | 0.021300 |
| 75% | 0.048600 |
| max | 40.285400 |

Name: Power Galley 1 (MW), dtype: float64

Summary without Outliers for Power Galley 1 (MW):

| | |
|-------|---------------|
| count | 103065.000000 |
| mean | 0.030652 |
| std | 0.030372 |
| min | 0.000000 |
| 25% | 0.006200 |
| 50% | 0.020700 |
| 75% | 0.046000 |
| max | 0.111900 |

Name: Power Galley 1 (MW), dtype: float64



Summary with Outliers for Power Galley 2 (MW):

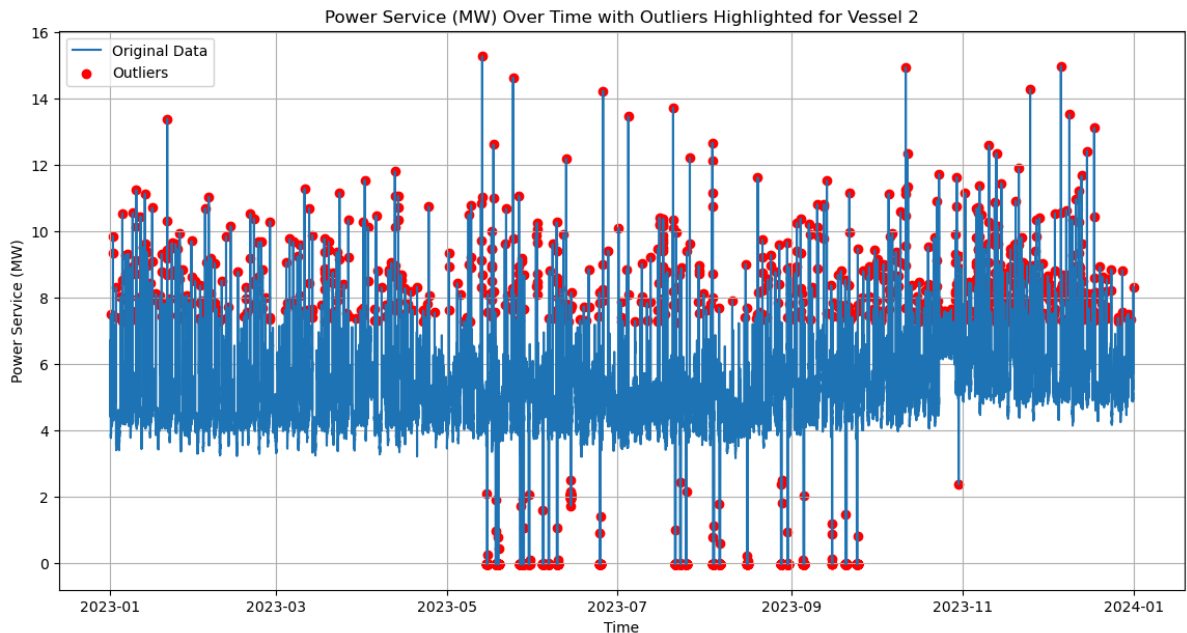
| | |
|-------|---------------|
| count | 105105.000000 |
| mean | 0.118563 |
| std | 0.232160 |
| min | 0.000000 |
| 25% | 0.050000 |
| 50% | 0.101600 |
| 75% | 0.181300 |
| max | 40.305400 |

Name: Power Galley 2 (MW), dtype: float64

Summary without Outliers for Power Galley 2 (MW):

| | |
|-------|---------------|
| count | 105055.000000 |
| mean | 0.116177 |
| std | 0.078353 |
| min | 0.000000 |
| 25% | 0.050000 |
| 50% | 0.101500 |
| 75% | 0.181200 |
| max | 0.371700 |

Name: Power Galley 2 (MW), dtype: float64



Summary with Outliers for Power Service (MW):

```
count    105105.000000
mean      4.916077
std       1.097217
min       -0.040000
25%       4.307700
50%       4.915700
75%       5.486500
max       15.264000
Name: Power Service (MW), dtype: float64
```

Summary without Outliers for Power Service (MW):

```
count    101910.000000
mean      4.968452
std       0.788179
min       2.644600
25%       4.328000
50%       4.925250
75%       5.475600
max       7.254300
Name: Power Service (MW), dtype: float64
```

Outliers Delete or keep it depends on the status

```
In [30]: # Ensure 'Start Time' is in datetime format and set as index
df['Start Time'] = pd.to_datetime(df['Start Time'])
vessel_2_data = df[df['Vessel Name'] == 'Vessel 2'].copy()
vessel_2_data.set_index('Start Time', inplace=True)

# Function to calculate Q1 and Q3
def calculate_q1_q3(df, column):
    Q1 = df[column].quantile(0.25)
    Q3 = df[column].quantile(0.75)
    return Q1, Q3

# Calculate Q1 and Q3 for the relevant columns
q1_pg1, q3_pg1 = calculate_q1_q3(vessel_2_data, 'Power Galley 1 (MW)')
q1_pg2, q3_pg2 = calculate_q1_q3(vessel_2_data, 'Power Galley 2 (MW)')
q1_ps, q3_ps = calculate_q1_q3(vessel_2_data, 'Power Service (MW)')

# Remove outliers
vessel_2_data_cleaned = vessel_2_data[
    (vessel_2_data['Power Galley 1 (MW)'] <= q3_pg1) &
```

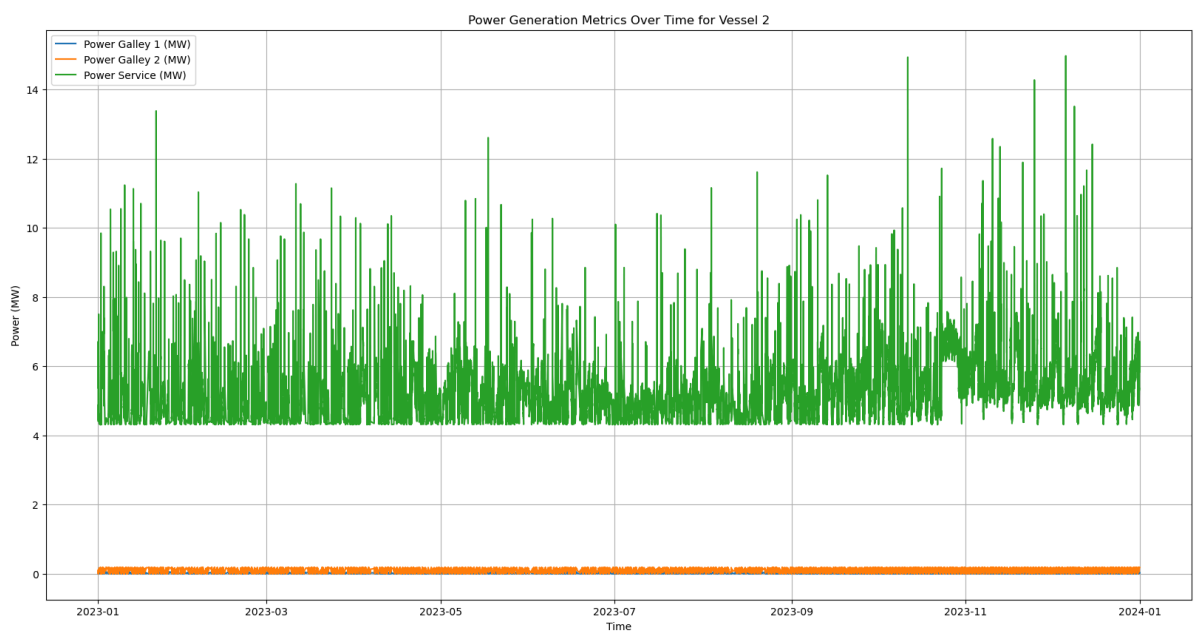
```

(vessel_2_data['Power Galley 2 (MW)'] <= q3_pg2) &
(vessel_2_data['Power Service (MW)'] >= q1_ps)
]

# Plotting function to visualize the cleaned data (Optional)
def plot_cleaned_data(df, columns, vessel_name):
    plt.figure(figsize=(20, 10))
    for column in columns:
        plt.plot(df.index, df[column], label=column)
    plt.xlabel('Time')
    plt.ylabel('Power (MW)')
    plt.title(f'Power Generation Metrics Over Time for {vessel_name}')
    plt.legend()
    plt.grid(True)
    plt.show()

# Plot the cleaned data for the relevant columns
plot_cleaned_data(vessel_2_data_cleaned, ['Power Galley 1 (MW)', 'Power Galley 2 (MW)']

```



Data Analysis

```

In [31]: # Define categories for Power Generation Analysis
diesel_generators = ['Diesel Generator 1 Power (MW)', 'Diesel Generator 2 Power (MW)',
                    'Diesel Generator 3 Power (MW)', 'Diesel Generator 4 Power (MW)']
hvac_chillers = ['HVAC Chiller 1 Power (MW)', 'HVAC Chiller 2 Power (MW)', 'HVAC Chiller 3 Power (MW)']
scrubber_power = ['Scrubber Power (MW)']
power_galley = ['Power Galley 1 (MW)', 'Power Galley 2 (MW)']
power_service = ['Power Service (MW)']

# Combine all columns into a single list
power_columns = diesel_generators + hvac_chillers + scrubber_power + power_galley + power_service

# Ensure all columns are numeric
df[power_columns] = df[power_columns].apply(pd.to_numeric, errors='coerce')

# Assuming 'Vessel Name' is the column indicating the vessel
vessel_1_data = df[df['Vessel Name'] == 'Vessel 1'].copy()
vessel_2_data = df[df['Vessel Name'] == 'Vessel 2'].copy()

# Calculate monthly averages for Vessel 1

```



```

vessel_1_data.set_index('Start Time', inplace=True)
vessel_1_monthly_avg = vessel_1_data[power_columns].resample('M').mean()

# Calculate monthly averages for Vessel 2
vessel_2_data.set_index('Start Time', inplace=True)
vessel_2_monthly_avg = vessel_2_data[power_columns].resample('M').mean()

# Function to plot monthly averages for Vessel 1
def plot_monthly_averages_vessel_1(columns, title):
    plt.figure(figsize=(14, 7))
    for column in columns:
        plt.plot(vessel_1_monthly_avg.index, vessel_1_monthly_avg[column], label=f'V
    plt.xlabel('Time')
    plt.ylabel('Values')
    plt.title(f'Monthly Average {title} Over Time for Vessel 1')
    plt.legend()
    plt.grid(True)
    plt.show()

# Function to plot monthly averages for Vessel 2
def plot_monthly_averages_vessel_2(columns, title):
    plt.figure(figsize=(14, 7))
    for column in columns:
        plt.plot(vessel_2_monthly_avg.index, vessel_2_monthly_avg[column], label=f'V
    plt.xlabel('Time')
    plt.ylabel('Values')
    plt.title(f'Monthly Average {title} Over Time for Vessel 2')
    plt.legend()
    plt.grid(True)
    plt.show()

# Plotting for each category for Vessel 1
plot_monthly_averages_vessel_1(diesel_generators, 'Diesel Generator Power')
plot_monthly_averages_vessel_1(hvac_chillers, 'HVAC Chiller Power')
plot_monthly_averages_vessel_1(scrubber_power, 'Scrubber Power')
plot_monthly_averages_vessel_1(power_galley, 'Power Galley')
plot_monthly_averages_vessel_1(power_service, 'Power Service')

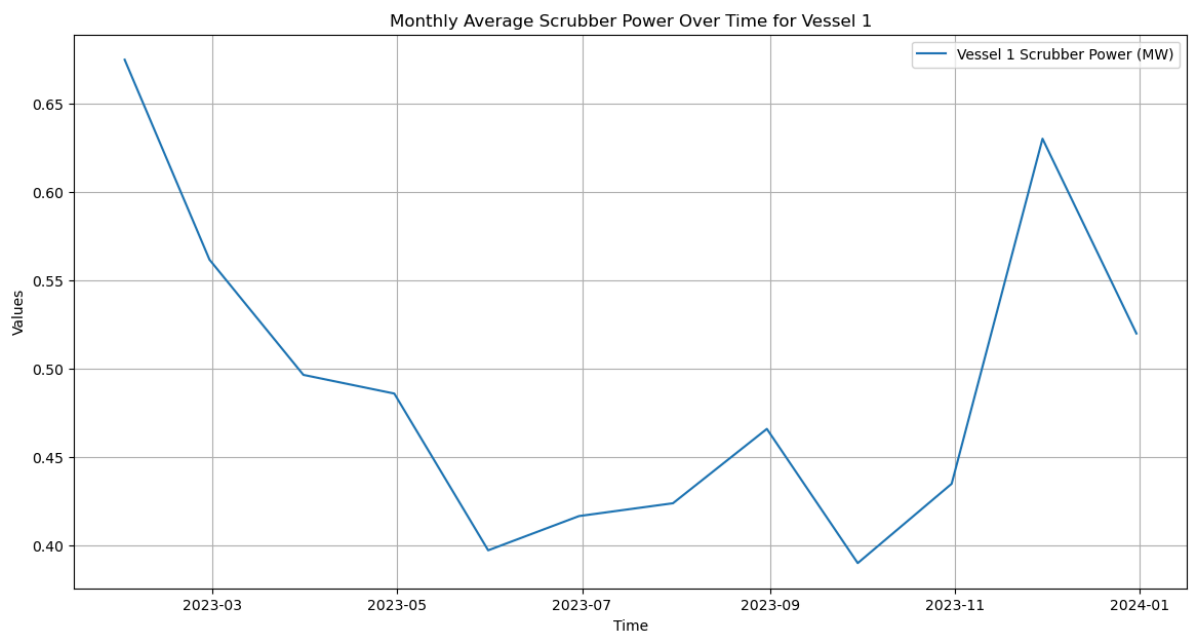
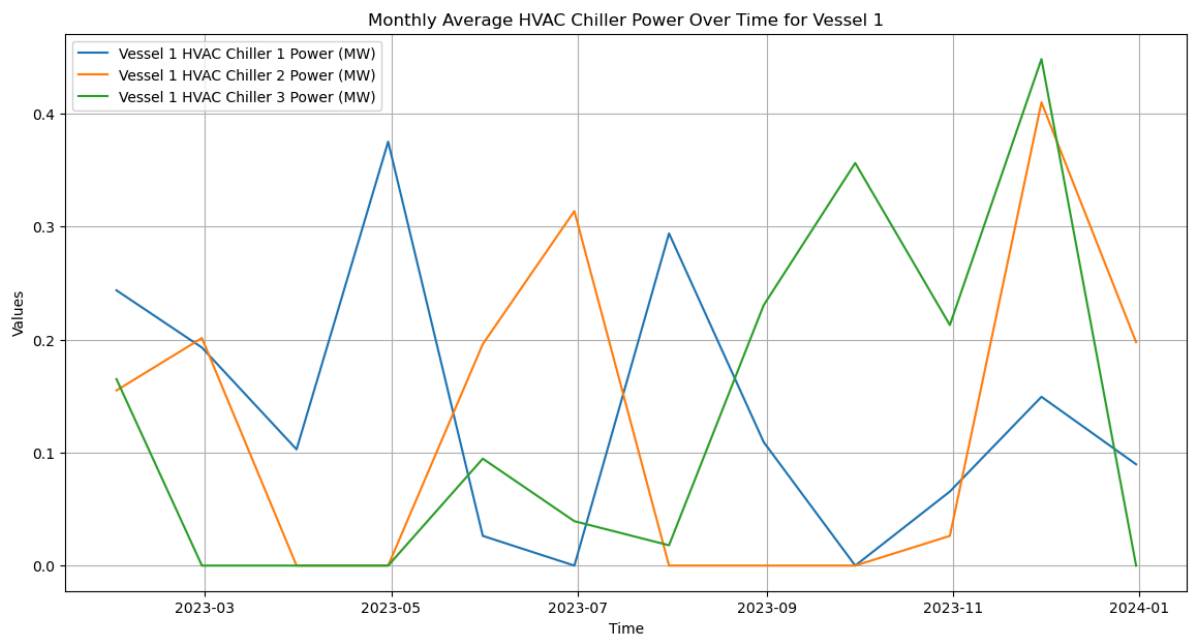
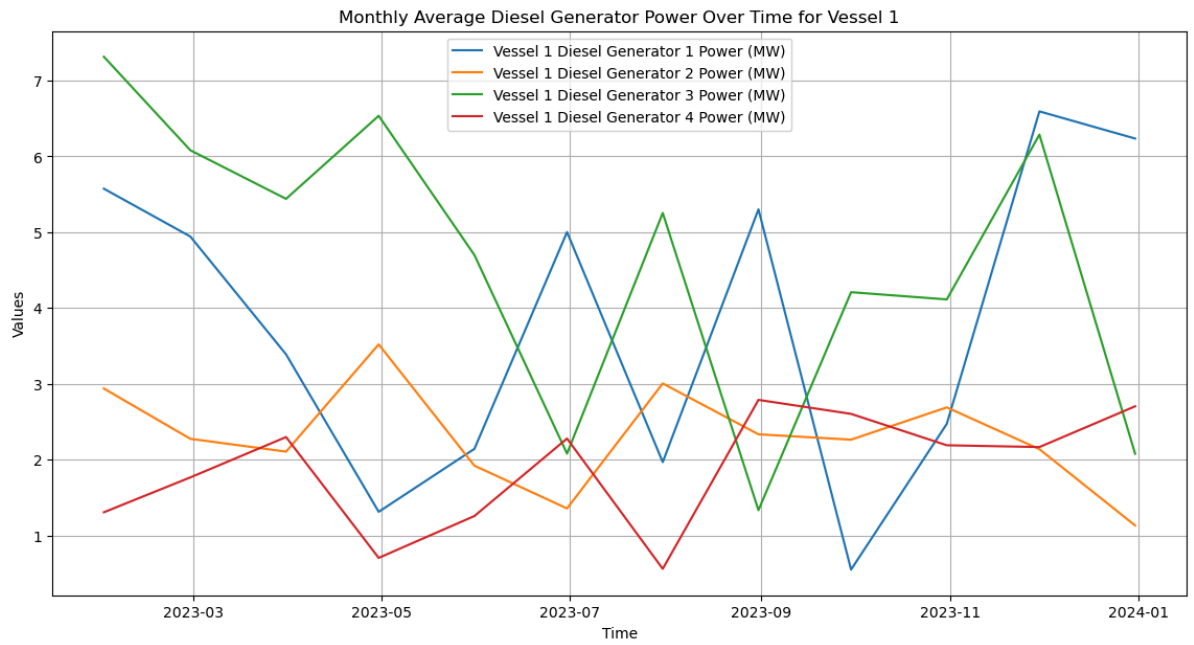
# Plotting for each category for Vessel 2
plot_monthly_averages_vessel_2(diesel_generators, 'Diesel Generator Power')
plot_monthly_averages_vessel_2(hvac_chillers, 'HVAC Chiller Power')
plot_monthly_averages_vessel_2(scrubber_power, 'Scrubber Power')
plot_monthly_averages_vessel_2(power_galley, 'Power Galley')
plot_monthly_averages_vessel_2(power_service, 'Power Service')

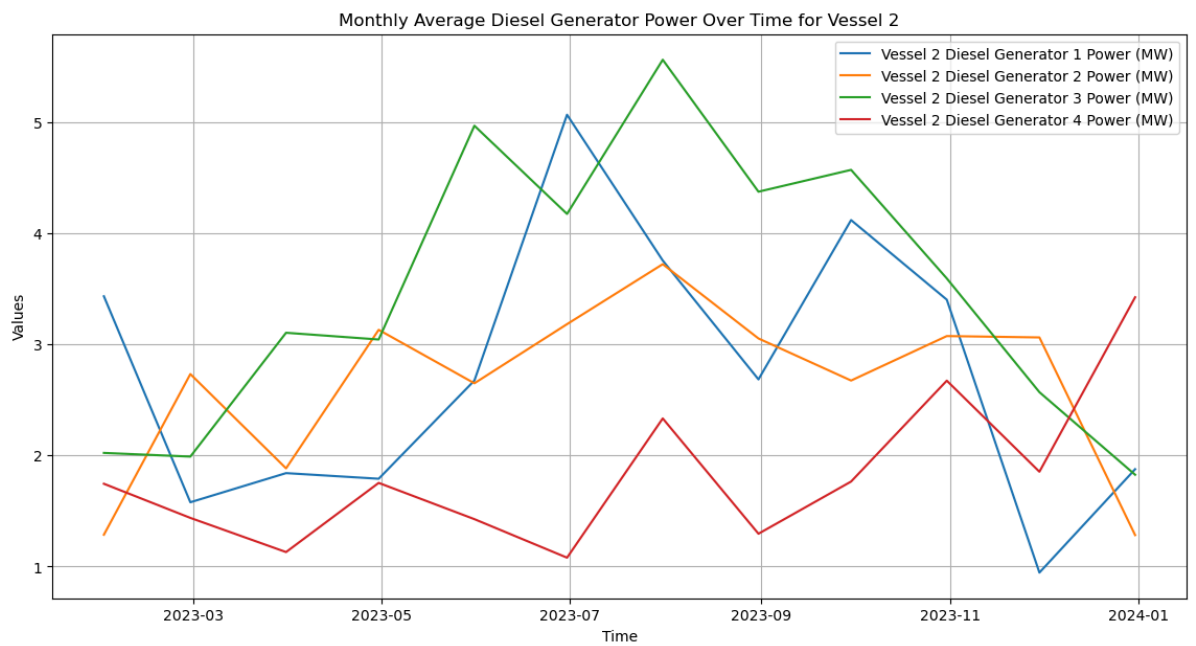
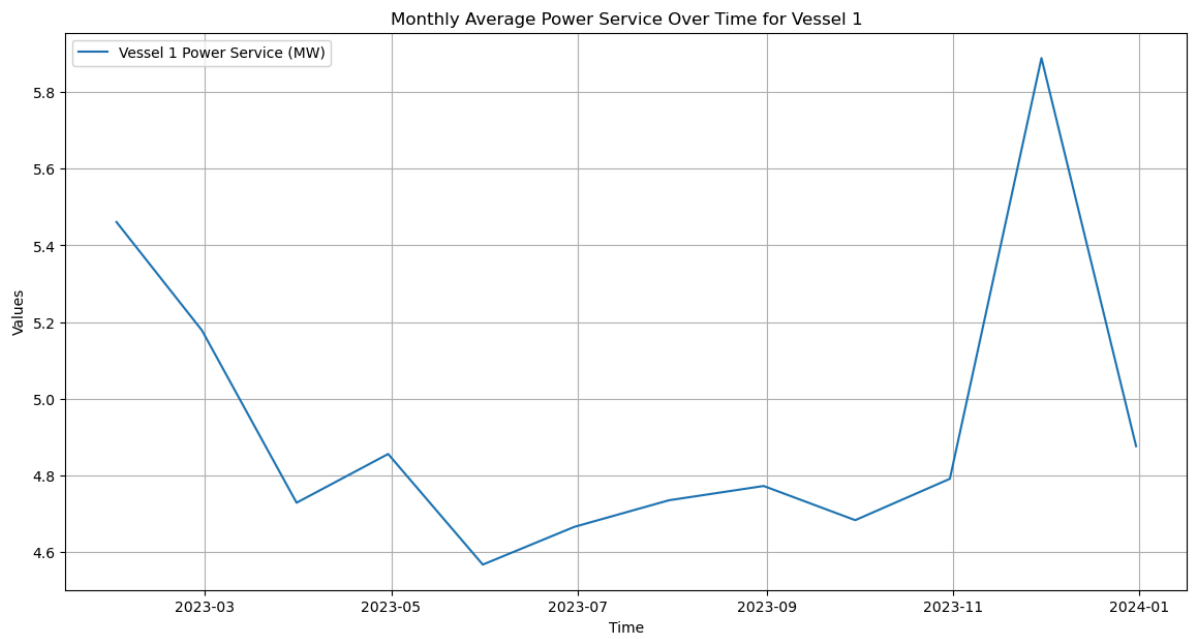
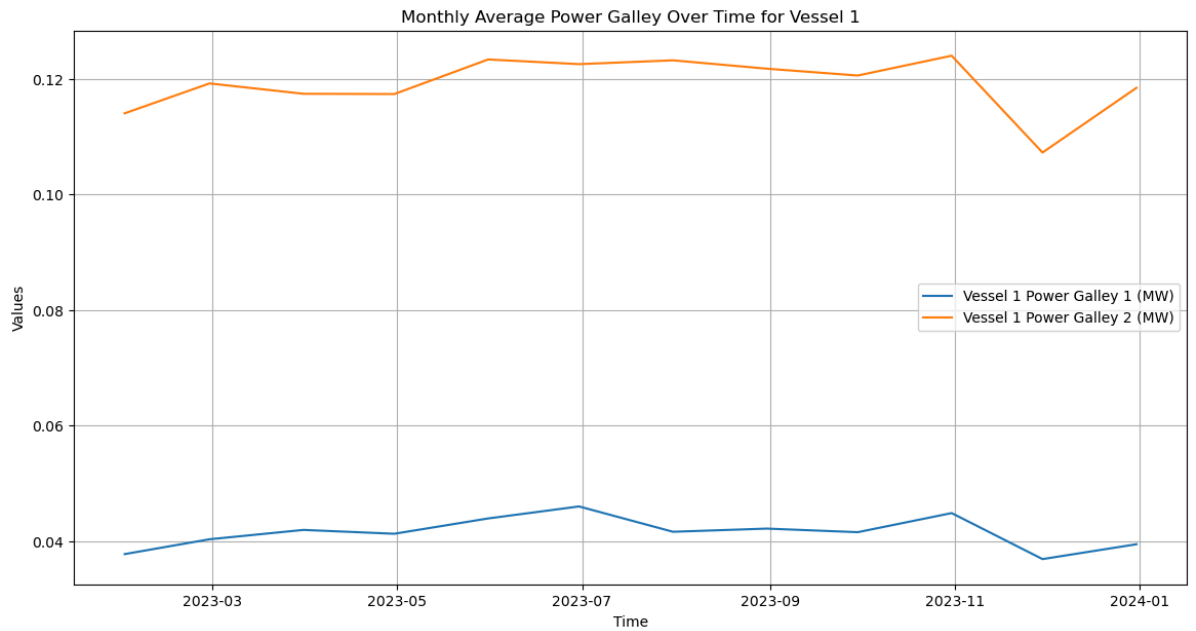
```

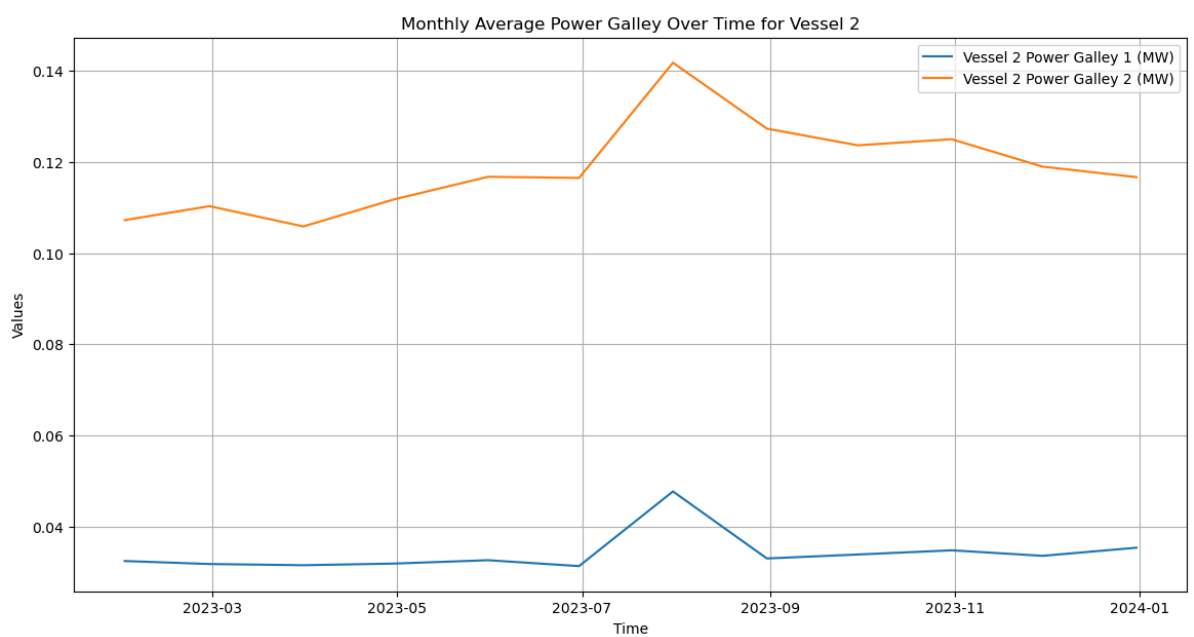
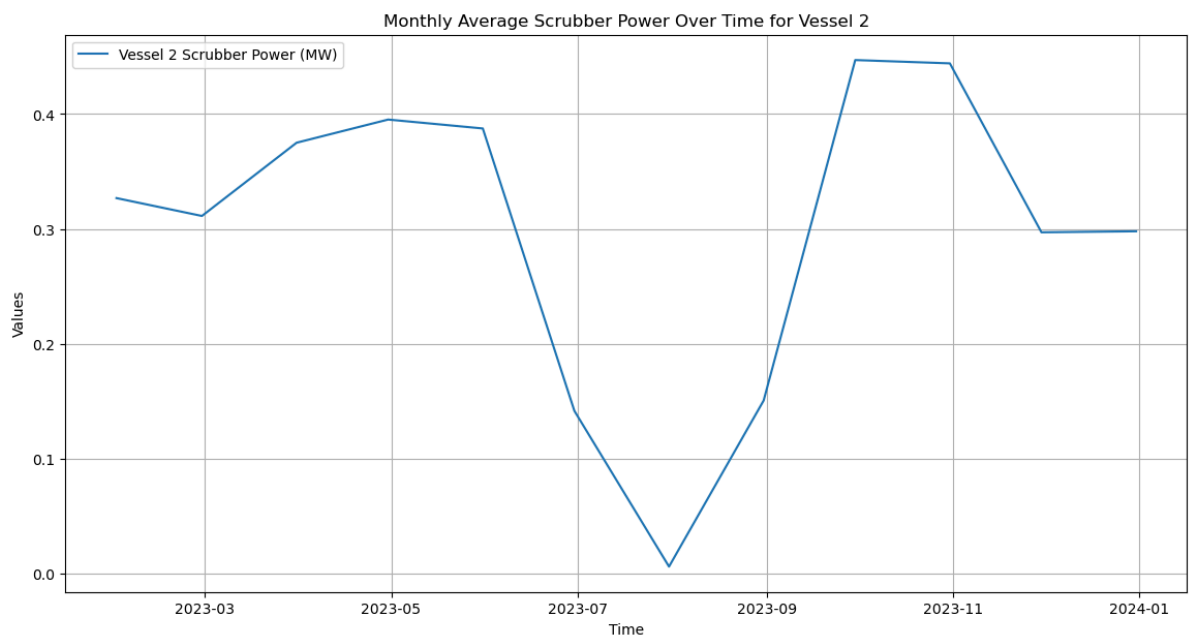
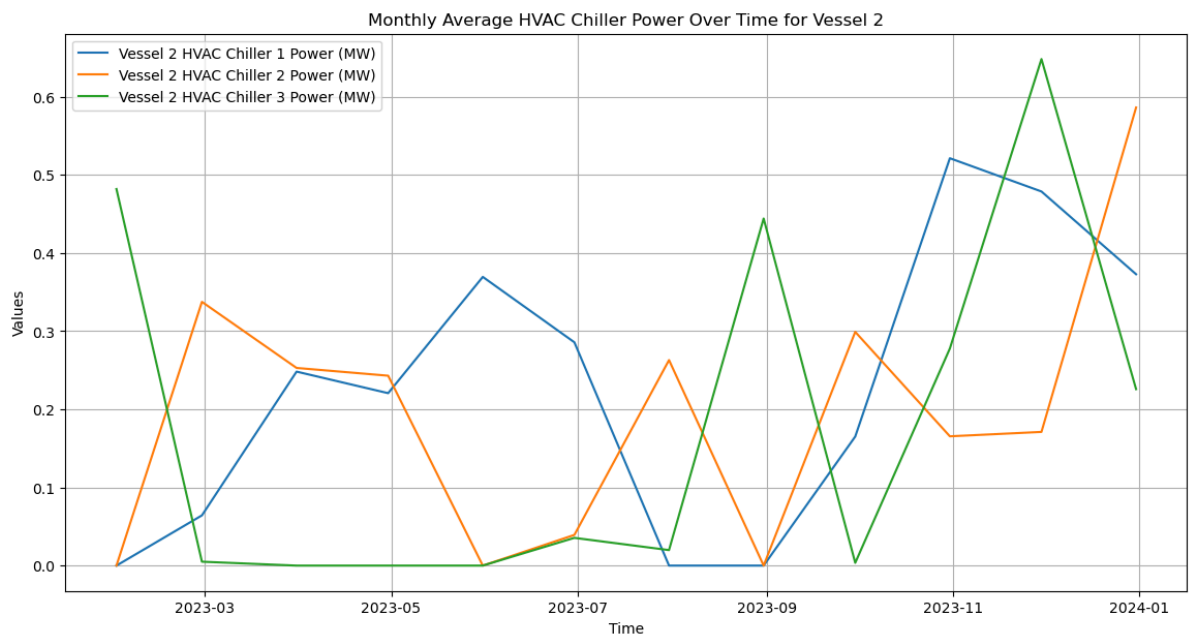
```

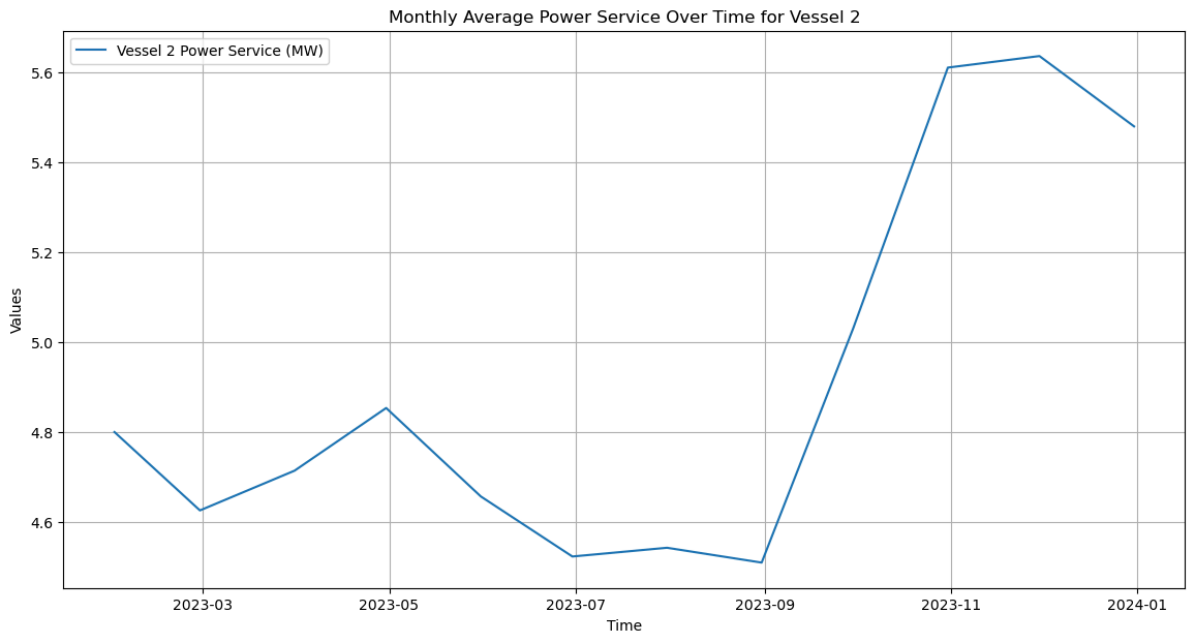
C:\Users\Wseoin\AppData\Local\Temp\ipykernel_2904\W3416136731.py:21: FutureWarning:
'M' is deprecated and will be removed in a future version, please use 'ME' instead.
    vessel_1_monthly_avg = vessel_1_data[power_columns].resample('M').mean()
C:\Users\Wseoin\AppData\Local\Temp\ipykernel_2904\W3416136731.py:25: FutureWarning:
'M' is deprecated and will be removed in a future version, please use 'ME' instead.
    vessel_2_monthly_avg = vessel_2_data[power_columns].resample('M').mean()

```









```
In [32]: # Combine all columns into a single list
power_columns = diesel_generators + hvac_chillers + scrubber_power + power_galley +

# Ensure all columns are numeric
df[power_columns] = df[power_columns].apply(pd.to_numeric, errors='coerce')

# Ensure 'Start Time' is in datetime format and set as index
df['Start Time'] = pd.to_datetime(df['Start Time'])
vessel_1_data = df[df['Vessel Name'] == 'Vessel 1'].copy()
vessel_2_data = df[df['Vessel Name'] == 'Vessel 2'].copy()
vessel_1_data.set_index('Start Time', inplace=True)
vessel_2_data.set_index('Start Time', inplace=True)

# Calculate monthly averages for Vessel 1
vessel_1_monthly_avg = vessel_1_data[power_columns].resample('M').mean()

# Calculate monthly averages for Vessel 2
vessel_2_monthly_avg = vessel_2_data[power_columns].resample('M').mean()

# Prepare DataFrames for SQL storage
power_vessel_1 = vessel_1_monthly_avg.copy()
power_vessel_1['Vessel'] = 'Vessel 1'

power_vessel_2 = vessel_2_monthly_avg.copy()
power_vessel_2['Vessel'] = 'Vessel 2'

# Combine all results
combined_power_results = pd.concat([power_vessel_1, power_vessel_2])

# Store combined results into SQLite database
combined_power_results.to_sql('power_generation_analysis', conn, if_exists='replace')

# Verify data is stored correctly
query = "SELECT * FROM power_generation_analysis"
df_sql = pd.read_sql_query(query, conn)
print(df_sql.head())

# Close the connection
conn.close()
```

| | Start Time | Diesel Generator 1 Power (MW) | | W |
|---|---------------------|-------------------------------|--|---|
| 0 | 2023-01-31 00:00:00 | 5.573617 | | |
| 1 | 2023-02-28 00:00:00 | 4.943044 | | |
| 2 | 2023-03-31 00:00:00 | 3.389795 | | |
| 3 | 2023-04-30 00:00:00 | 1.313696 | | |
| 4 | 2023-05-31 00:00:00 | 2.143690 | | |

| | Diesel Generator 2 Power (MW) | | Diesel Generator 3 Power (MW) | | W |
|---|-------------------------------|--|-------------------------------|--|---|
| 0 | 2.939083 | | 7.313376 | | |
| 1 | 2.276261 | | 6.080360 | | |
| 2 | 2.107315 | | 5.439922 | | |
| 3 | 3.521354 | | 6.534617 | | |
| 4 | 1.922345 | | 4.699523 | | |

| | Diesel Generator 4 Power (MW) | | HVAC Chiller 1 Power (MW) | | W |
|---|-------------------------------|--|---------------------------|--|---|
| 0 | 1.307893 | | 0.243542 | | |
| 1 | 1.767643 | | 0.192895 | | |
| 2 | 2.299480 | | 0.102834 | | |
| 3 | 0.706250 | | 0.375020 | | |
| 4 | 1.258140 | | 0.026296 | | |

| | HVAC Chiller 2 Power (MW) | | HVAC Chiller 3 Power (MW) | | Scrubber Power (MW) | W |
|---|---------------------------|--|---------------------------|--|---------------------|---|
| 0 | 0.154953 | | 0.165033 | | 0.675067 | |
| 1 | 0.201288 | | 0.000000 | | 0.561772 | |
| 2 | 0.000000 | | 0.000000 | | 0.496456 | |
| 3 | 0.000000 | | 0.000000 | | 0.485872 | |
| 4 | 0.195975 | | 0.094598 | | 0.397032 | |

| | Power Galley 1 (MW) | | Power Galley 2 (MW) | | Power Service (MW) | | Vessel |
|---|---------------------|--|---------------------|--|--------------------|--|----------|
| 0 | 0.037764 | | 0.114036 | | 5.460985 | | Vessel 1 |
| 1 | 0.040328 | | 0.119204 | | 5.178252 | | Vessel 1 |
| 2 | 0.041948 | | 0.117415 | | 4.728527 | | Vessel 1 |
| 3 | 0.041283 | | 0.117366 | | 4.855669 | | Vessel 1 |
| 4 | 0.043934 | | 0.123354 | | 4.567114 | | Vessel 1 |

```
C:\Users\Wseoin\AppData\Local\Temp\ipykernel_2904\510739112.py:15: FutureWarning: 'M'
is deprecated and will be removed in a future version, please use 'ME' instead.
    vessel_1_monthly_avg = vessel_1_data[power_columns].resample('M').mean()
C:\Users\Wseoin\AppData\Local\Temp\ipykernel_2904\510739112.py:18: FutureWarning: 'M'
is deprecated and will be removed in a future version, please use 'ME' instead.
    vessel_2_monthly_avg = vessel_2_data[power_columns].resample('M').mean()
```

Power Generation Performance Trend Analysis Report

Introduction

The power generation performance of vessels is critical to ensure operational efficiency and compliance with environmental standards. This report analyzes the trends in power generation for two vessels over a year. The analysis covers various components such as diesel generators, HVAC chillers, scrubbers, galley power, and power service.

- Diesel Generators Power Analysis

Vessel 1:

The monthly average power generated by Diesel Generators 1-4 shows significant fluctuations throughout the year. Diesel Generator 1 saw a gradual decrease until mid-year and then an increase towards the end of the year. Diesel Generator 2 had relatively stable

power generation with minor fluctuations. Diesel Generator 3 showed high variability with peaks during mid-year and at the end. Diesel Generator 4 had lower but more consistent power generation.

Vessel 2:

Diesel Generators on Vessel 2 also exhibited variability, with Diesel Generator 3 showing the highest fluctuations. Diesel Generator 2 displayed a gradual increase over the year. Diesel Generators 1 and 4 showed relatively stable trends with minor fluctuations.

- HVAC Chillers Power Analysis

Vessel 1:

The HVAC Chiller power consumption showed notable variability. HVAC Chiller 1 had a few peaks during the mid-year but generally maintained lower power consumption. HVAC Chiller 2 and 3 had irregular peaks throughout the year, indicating sporadic high usage periods.

Vessel 2:

HVAC Chiller 1 showed significant fluctuations, particularly towards the end of the year. HVAC Chiller 2 and 3 had erratic usage patterns with notable peaks at various times.

- Scrubber Power Analysis

Vessel 1:

The scrubber power consumption trend decreased during the first half of the year but increased significantly towards the year-end, indicating more extensive use possibly due to stricter emission control periods.

Vessel 2:

Scrubber power usage was relatively low but spiked dramatically in the mid and later parts of the year, suggesting intermittent but intense usage periods.

- Power Galley Analysis

Vessel 1:

Power Galley 1 maintained low and stable power consumption. Power Galley 2 had higher but consistent power consumption throughout the year.

Vessel 2:

Power Galley 1 also showed stable and low power consumption. Power Galley 2 experienced higher power usage, with a peak towards the end of the year.

- Power Service Analysis

Vessel 1:

Power service consumption decreased during the first half but increased towards the end of the year, suggesting variations in operational demand or seasonal changes.

Vessel 2:

Power service consumption had a steady trend with a significant increase at the end of the year.

Conclusion

The power generation trends for both vessels indicate variability across different components. Diesel generators, HVAC chillers, and scrubbers show significant fluctuations, likely due to operational demands and environmental compliance requirements. Stable components like power galley indicate consistent energy usage patterns. Monitoring these trends helps in optimizing power management strategies and ensuring efficient vessel operations.

In []: