

```
In [1]: import pandas as pd
from statsmodels.tsa.seasonal import STL
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
import statsmodels.api as sm
from statsmodels.tsa.seasonal import seasonal_decompose
```

1. Wczytanie danych

```
In [3]: data = pd.read_excel('Pytybydgoszcz.xlsx', names = ['date', 'PM 2.5 [ug/m3]', skiprows = 5, index_col = 'date')
data.index = pd.to_datetime(data.index)
```

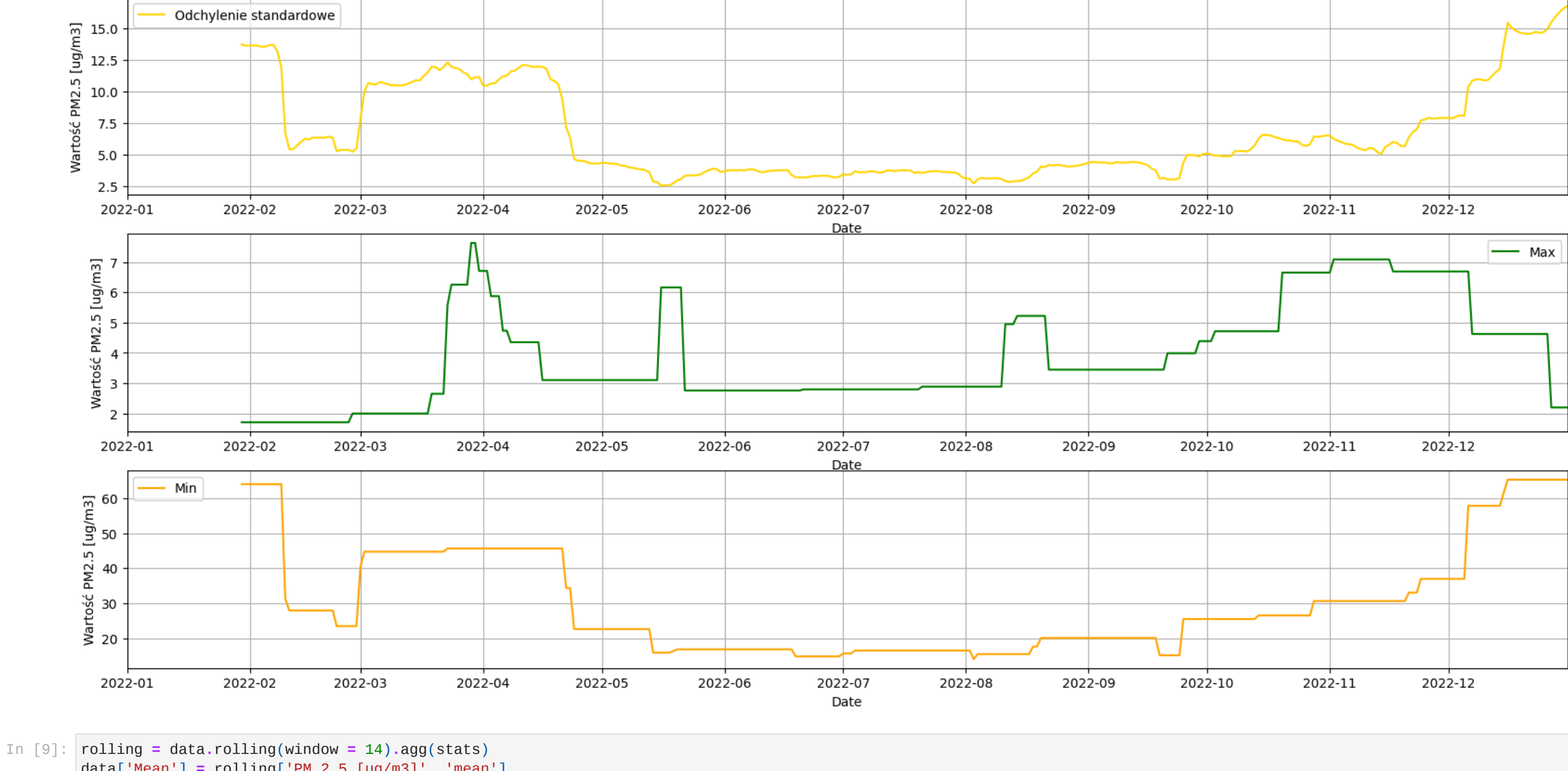
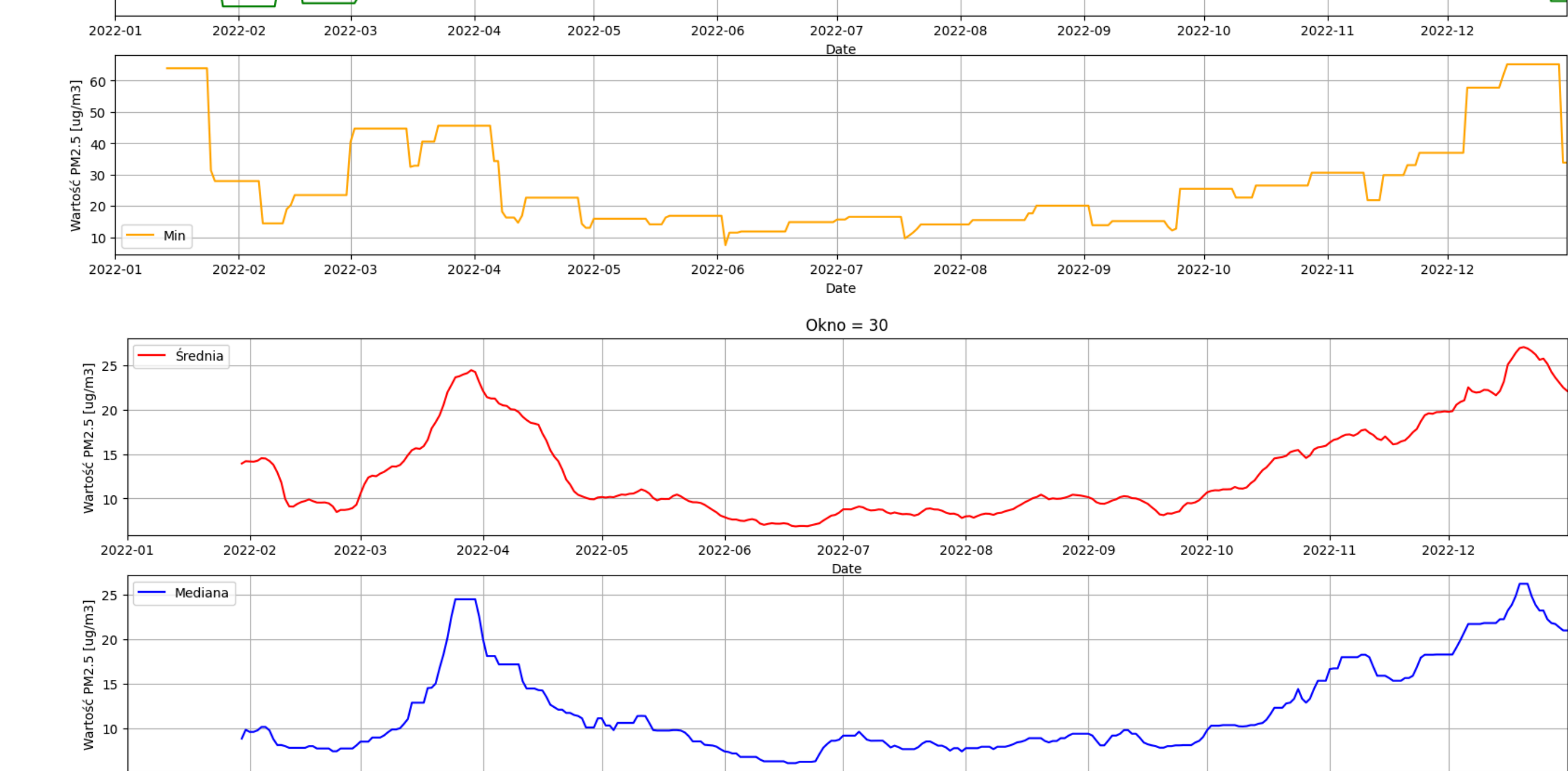
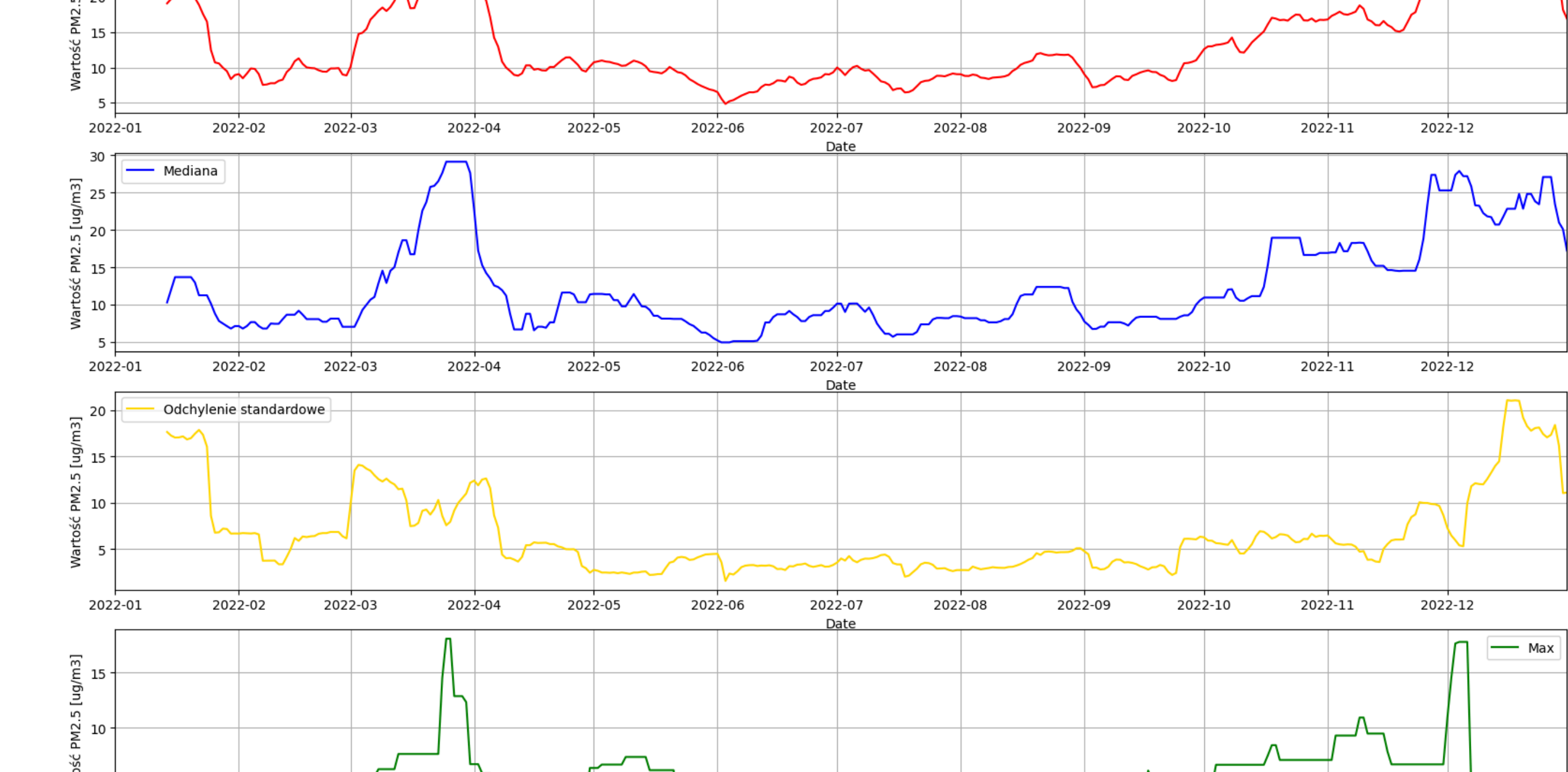
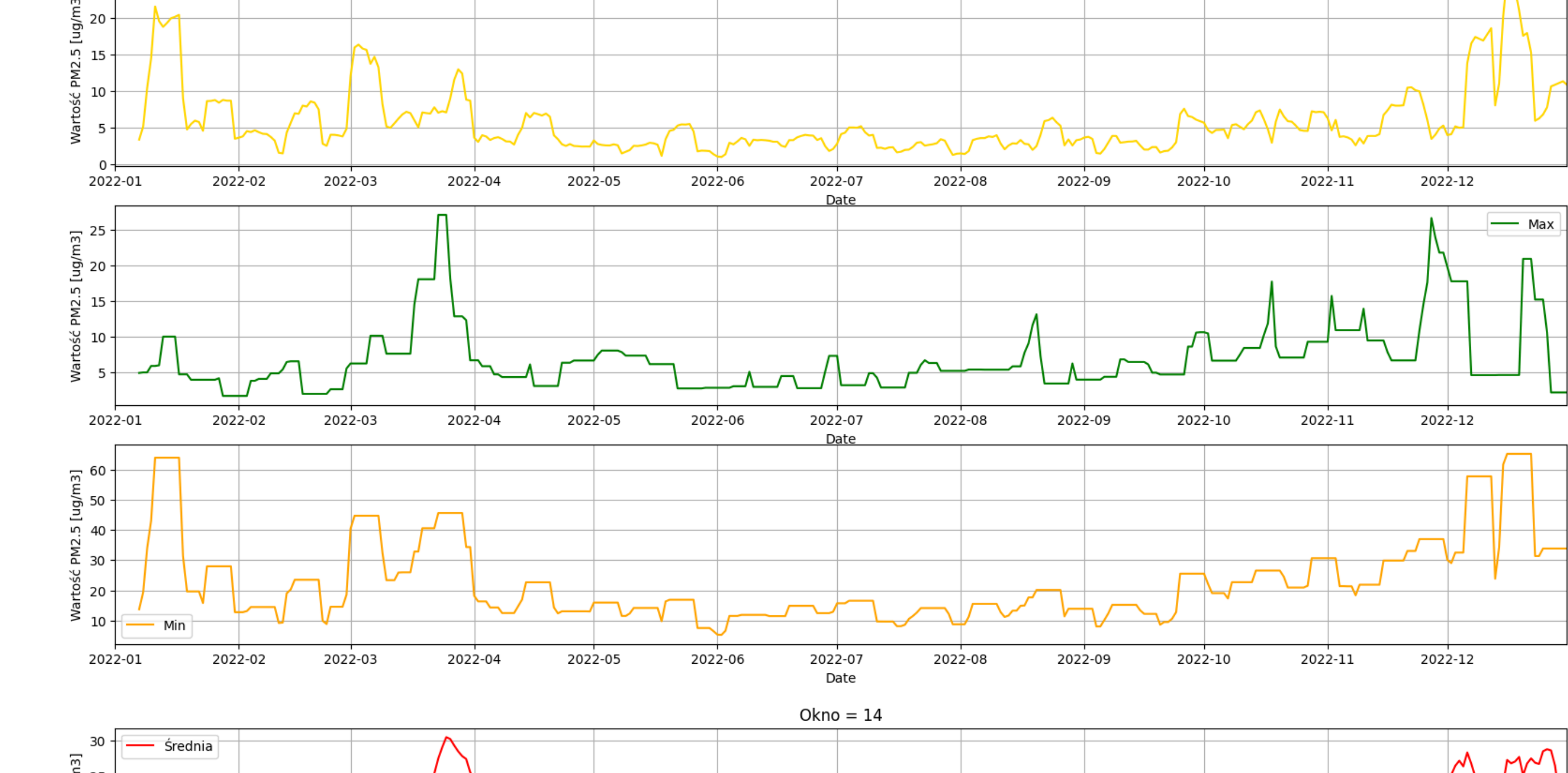


2. Zbiór cech

```
In [6]: stats = ["mean", "median", "std", "max", "min"]
```

2.1 Okna czasowe typu rolling

```
In [8]: windows = [7, 14, 30]
for window in windows:
    fig, axes = plt.subplots(5, 1, figsize = [20, 16])
    axes[0].set_title(f'Okno {window}')
    sns.lineplot(data = rolling_df, x = rolling_df.index, y = ('PM 2.5 [ug/m3]', 'mean'), ax = axes[0], color = 'red', label = 'Średnia')
    axes[0].grid()
    axes[1].set_xlabel('wartość PM2.5 [ug/m3]')
    axes[1].set_ylabel('wartość PM2.5 [ug/m3]')
    sns.lineplot(data = rolling_df, x = rolling_df.index, y = ('PM 2.5 [ug/m3]', 'median'), ax = axes[1], color = 'blue', label = 'Mediana')
    axes[1].grid()
    axes[2].set_xlabel('wartość PM2.5 [ug/m3]')
    axes[2].set_ylabel('wartość PM2.5 [ug/m3]')
    sns.lineplot(data = rolling_df, x = rolling_df.index, y = ('PM 2.5 [ug/m3]', 'std'), ax = axes[2], color = 'gold', label = 'Odchylenie standardowe')
    axes[2].grid()
    axes[3].set_xlabel('wartość PM2.5 [ug/m3]')
    axes[3].set_ylabel('wartość PM2.5 [ug/m3]')
    sns.lineplot(data = rolling_df, x = rolling_df.index, y = ('PM 2.5 [ug/m3]', 'min'), ax = axes[3], color = 'green', label = 'Max')
    axes[3].grid()
    axes[4].set_xlabel('wartość PM2.5 [ug/m3]')
    axes[4].set_ylabel('wartość PM2.5 [ug/m3]')
    sns.lineplot(data = rolling_df, x = rolling_df.index, y = ('PM 2.5 [ug/m3]', 'max'), ax = axes[4], color = 'orange', label = 'Min')
    axes[4].grid()
    axes[4].set_xlabel('wartość PM2.5 [ug/m3]')
    axes[4].set_ylabel('wartość PM2.5 [ug/m3]')
```



```
In [9]: rolling = data.rolling(window = 14).agg(stats)
data['mean'] = rolling['PM 2.5 [ug/m3]', 'mean']
data['median'] = rolling['PM 2.5 [ug/m3]', 'median']
data['std'] = rolling['PM 2.5 [ug/m3]', 'std']
data['min'] = rolling['PM 2.5 [ug/m3]', 'min']
data['max'] = rolling['PM 2.5 [ug/m3]', 'max']
```

2.2 Okna czasowe typu expanding

```
In [11]: expanding_df = data.expanding().agg(stats)
fig, axes = plt.subplots(5, 1, figsize = [20, 16])
axes[0].set_title(f'Okno {window}')
sns.lineplot(data = expanding_df, x = expanding_df.index, y = ('PM 2.5 [ug/m3]', 'mean'), ax = axes[0], color = 'red', label = 'Średnia')
axes[0].grid()
axes[1].set_xlabel('wartość PM2.5 [ug/m3]')
axes[1].set_ylabel('wartość PM2.5 [ug/m3]')
sns.lineplot(data = expanding_df, x = expanding_df.index, y = ('PM 2.5 [ug/m3]', 'median'), ax = axes[1], color = 'blue', label = 'Mediana')
axes[1].grid()
axes[2].set_xlabel('wartość PM2.5 [ug/m3]')
axes[2].set_ylabel('wartość PM2.5 [ug/m3]')
sns.lineplot(data = expanding_df, x = expanding_df.index, y = ('PM 2.5 [ug/m3]', 'std'), ax = axes[2], color = 'gold', label = 'Odchylenie standardowe')
axes[2].grid()
axes[3].set_xlabel('wartość PM2.5 [ug/m3]')
axes[3].set_ylabel('wartość PM2.5 [ug/m3]')
sns.lineplot(data = expanding_df, x = expanding_df.index, y = ('PM 2.5 [ug/m3]', 'min'), ax = axes[3], color = 'green', label = 'Max')
axes[3].grid()
axes[4].set_xlabel('wartość PM2.5 [ug/m3]')
axes[4].set_ylabel('wartość PM2.5 [ug/m3]')
sns.lineplot(data = expanding_df, x = expanding_df.index, y = ('PM 2.5 [ug/m3]', 'max'), ax = axes[4], color = 'orange', label = 'Min')
axes[4].grid()
axes[4].set_xlabel('wartość PM2.5 [ug/m3]')
axes[4].set_ylabel('wartość PM2.5 [ug/m3]')
```

```
Out[11]: Text(0, 0.5, 'wartość PM2.5 [ug/m3]')
```



2.3 Okna czasowe typu nested

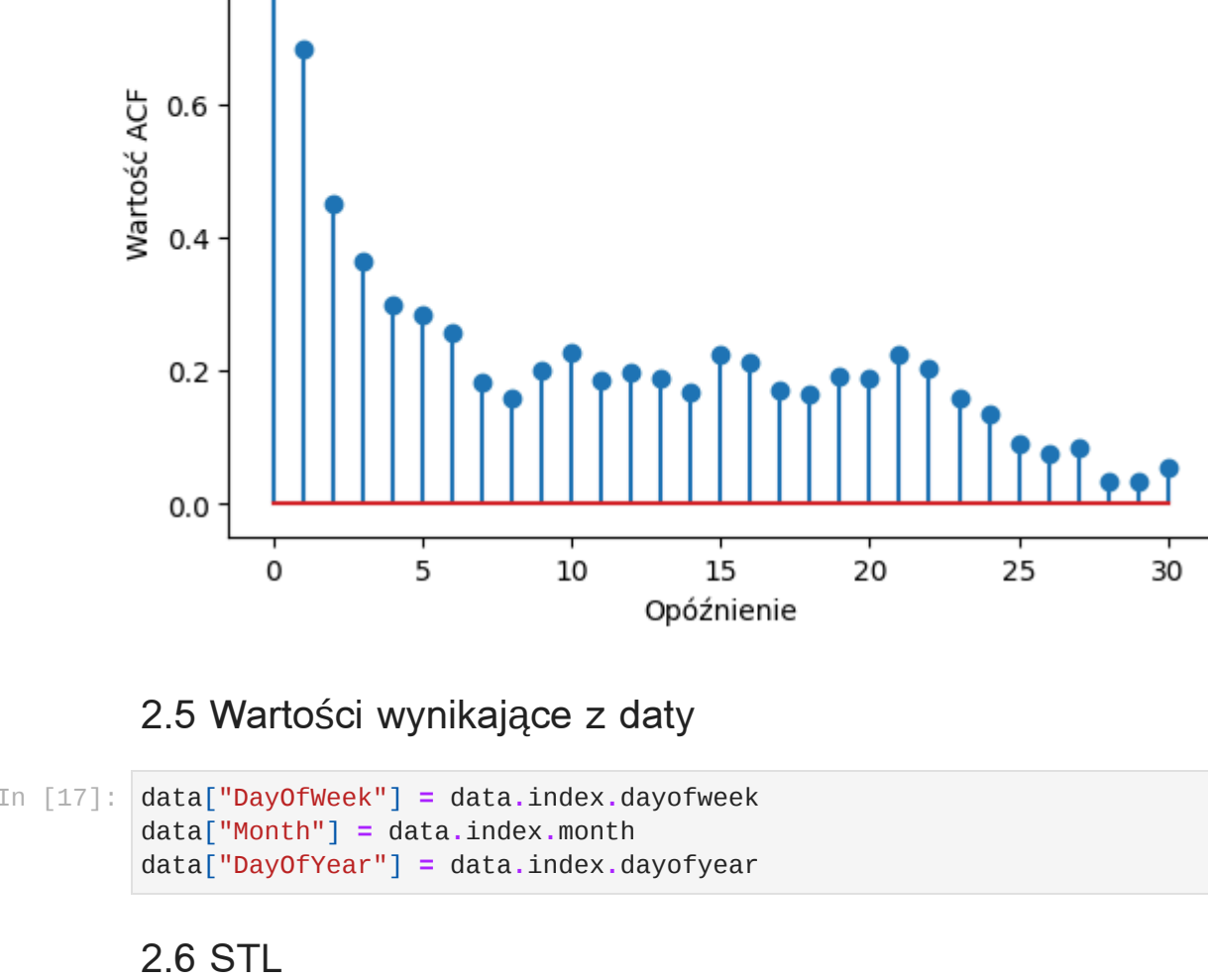
```
In [13]: nested_windows = [(7, 14), (7, 30), (14, 30)]
for outer_window, inner_window in nested_windows:
    outer_stats = data.rolling(window=outer_window).agg(stats)
    nested_df = outer_stats.rolling(window=inner_window).agg(stats)
```

2.4 Wartości opóźnione bazując na ACF

```
In [15]: acf = sm.tsa.acf(data["PM 2.5 [ug/m3]"], nlags = 30)
```

```
plt.stem(len(acf), acf)
plt.title('Funkcja Autokorelacji (ACF)')
plt.xlabel('Opóźnienie')
plt.ylabel('Wartość ACF')
```

```
Out[15]: Text(0, 0.5, 'Wartość ACF')
```



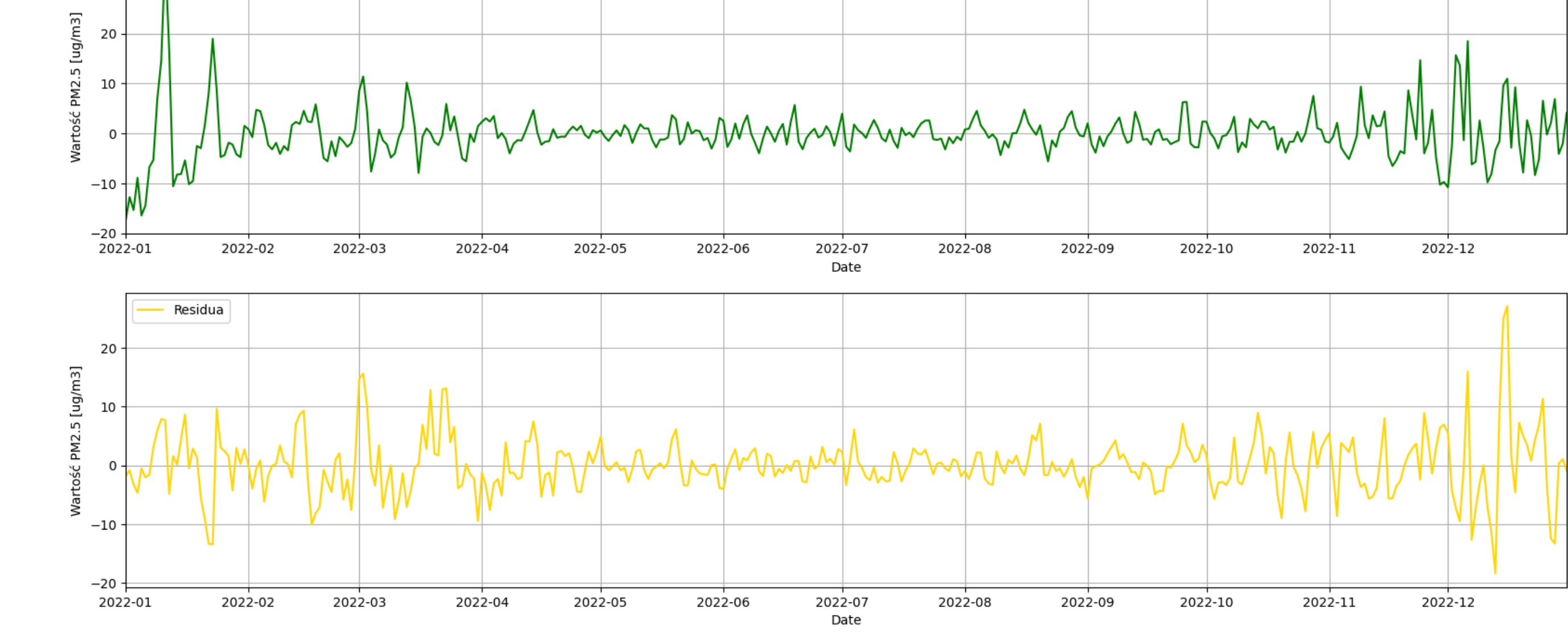
2.5 Wartości wynikające z daty

```
In [17]: data['DayOfWeek'] = data.index.dayofweek
data['Month'] = data.index.month
data['DayOfYear'] = data.index.dayofyear
```

2.6 STL

```
In [19]: result = STL(data["PM 2.5 [ug/m3]"], period = 12, seasonal = 5).fit()
data['Trend'] = data["Seasonal"], data['Resid'] = result.trend, result.resid
fig, axes = plt.subplots(3, 1, figsize = [20, 14])
sns.lineplot(data = data.index, y = data["PM 2.5 [ug/m3]"], ax = axes[0], color = 'blue', label = "PM2.5 [ug/m3]")
sns.lineplot(data = data.index, y = data["Trend"], ax = axes[0], color = 'red', label = "Trend")
axes[0].grid()
axes[0].set_xlabel('wartość PM2.5 [ug/m3]')
axes[0].set_ylabel('wartość PM2.5 [ug/m3]')
sns.lineplot(data = data.index, y = data["Seasonal"], ax = axes[1], color = 'green', label = "Sezonowość")
axes[1].set_xlabel('wartość PM2.5 [ug/m3]')
axes[1].set_ylabel('wartość PM2.5 [ug/m3]')
sns.lineplot(data = data.index, y = data["Resid"], ax = axes[2], color = 'gold', label = "Residua")
axes[2].grid()
axes[2].set_xlabel('wartość PM2.5 [ug/m3]')
axes[2].set_ylabel('wartość PM2.5 [ug/m3]')
```

```
Out[19]: (18995.8, 19357.0)
```



In [28]: data

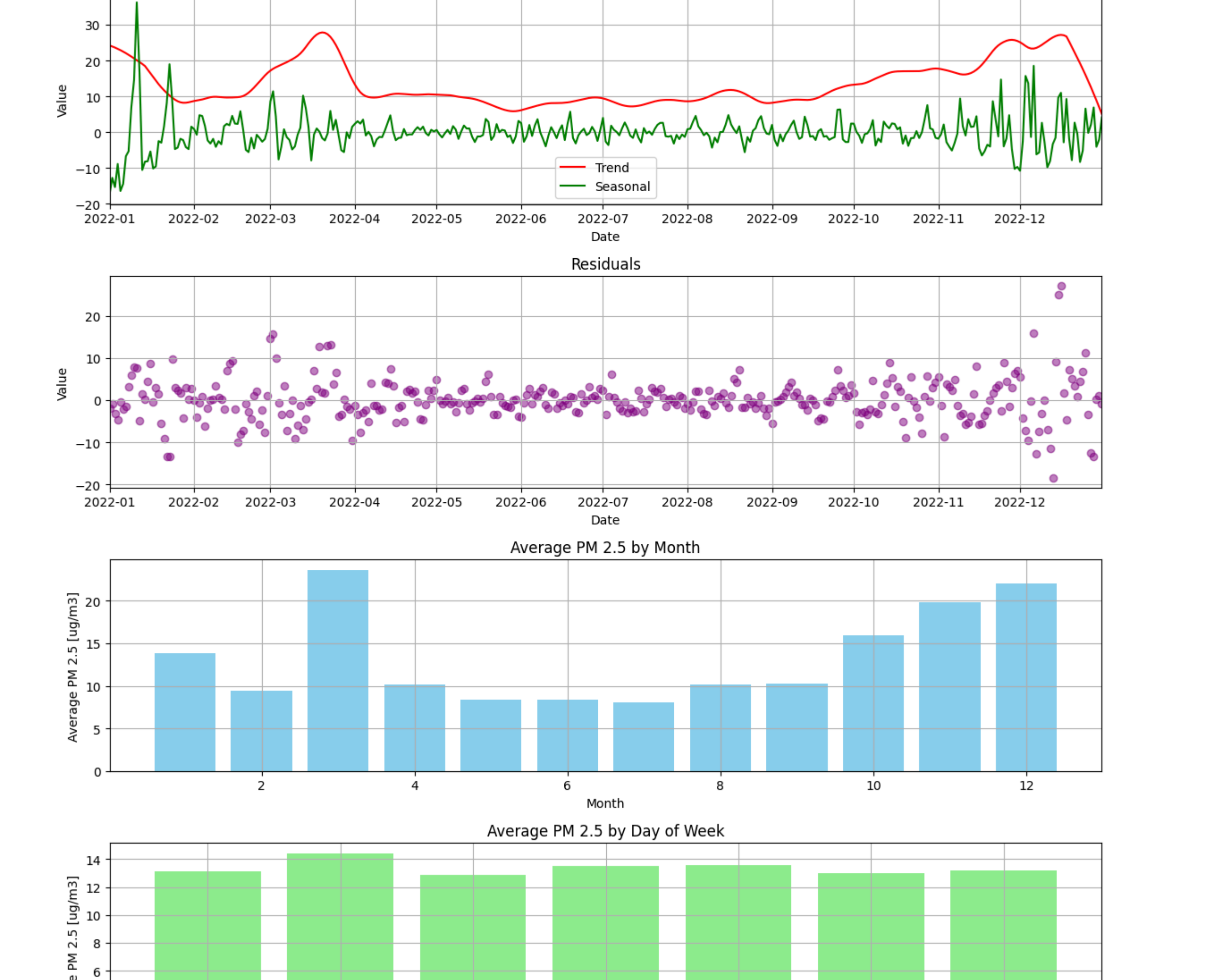
Out[28]:

Date	PM 2.5 [ug/m3]	Mean	Median	Std	Min	Max	DayOfWeek	Month	DayOfYear	Trend	Seasonal	Resid
2022-01-01	4.934921	NaN	NaN	NaN	NaN	NaN	5	1	1	24.212075	-17.424580	-1.852074
2022-01-02	10.341562	NaN	NaN	NaN	NaN	NaN	6	1	2	23.900994	-12.747473	-0.831959
2022-01-03	5.007348	NaN	NaN	NaN	NaN	NaN	0	1	3	23.8003979	-15.345300	-3.247331
2022-01-04	9.688094	NaN	NaN	NaN	NaN	NaN	1	1	4	23.258334	-8.882773	-4.687467
2022-01-05	5.914476	NaN	NaN	NaN	NaN	NaN	2	1	5	22.862371	-16.434513	-0.533382
...
2022-12-27	2.195314	28.624122	27.105343	17.306268	2.195314	65.242437	1	12	361	12.521737	2.096572	-12.424995
2022-12-28	4.317930	28.482004	23.458849	18.420872	2.195314	65.242437	2	12	362	10.721940	6.884442	-13.284452
2022-12-29	5.061757	26.4237507	21.018141	16.176562	2.195314	65.242437	3	12	363	8.870863	4.102361	-0.687467
2022-12-30	5.950761	18.202387	20.129194	11.080319	2.195314	33.855008	4	12	364	6.975471	-2.080389	1.055679
2022-12-31	8.273009	16.947308	17.271721	11.123749	2.195314	33.855008	5	12	365	5.042661	4.126937	-0.896589

365 rows x 12 columns

3. Przedstaw uzyskany zbiór cech na w przejrzysty sposób na wykresach i opis co można na ich podstawie zinterpretować.

```
In [22]: import matplotlib.pyplot as plt
fig, axes = plt.subplots(5, 1, figsize=(12, 16))
axes[0].plot(data.index, data["PM 2.5 [ug/m3]"], color="blue")
axes[0].set_title("Average PM 2.5 by Month")
axes[0].set_xlabel("Month")
axes[0].set_ylabel("Average PM 2.5 [ug/m3]")
axes[0].set_xlim(np.min(data.index), np.max(data.index))
axes[0].grid()
axes[1].plot(data.index, data["Trend"], color="red", label="Trend")
axes[1].plot(data.index, data["Seasonal"], color="green", label="Seasonal")
axes[1].set_title("Trend and Seasonality")
axes[1].set_xlabel("Date")
axes[1].set_ylabel("Value")
axes[1].grid()
axes[2].scatter(data.index, data["Resid"], color="purple", alpha=0.5)
axes[2].set_title("Residuals")
axes[2].set_xlabel("Date")
axes[2].set_ylabel("Value")
axes[2].grid()
monthly_avg = data.groupby(data.index.month)["PM 2.5 [ug/m3]"].mean()
axes[3].bar(monthly_avg.index, monthly_avg, color="skyblue")
axes[3].set_title("Average PM 2.5 by Month")
axes[3].set_xlabel("Month")
axes[3].set_ylabel("Average PM 2.5 [ug/m3]")
axes[3].grid()
dayofweek_avg = data.groupby(data.index.dayofweek)["PM 2.5 [ug/m3]"].mean()
axes[4].bar(dayofweek_avg.index, dayofweek_avg, color="lightgreen")
axes[4].set_title("Average PM 2.5 by Day of Week")
axes[4].set_xlabel("Day of Week")
axes[4].set_ylabel("Average PM 2.5 [ug/m3]")
axes[4].grid()
plt.tight_layout()
```



Przedstawione powyżej wykresy zbierają otrzymane informacje o cechach stworzonych dla zbioru, dzięki nim widzimy jak rozkładają się wartości PM2.5 na przestrzeni roku. Dodatkowo ukazana jest sezonowość oraz trend. Ostatnie dwa wykresy to rozkład średniej wartości PM2.5 w poszczególnym miesiącu - pokazują nam to ciekawe wnioski o zmianie średniej w miesiącu - dużych wartościach w okresie marca oraz październik - grudzień oraz wykres średniej wartości PM2.5 z podziałem na dzień tygodnia - w miarę równomierny rozkład.

```
In [24]: data.to_csv('pm25_fe.csv')
```