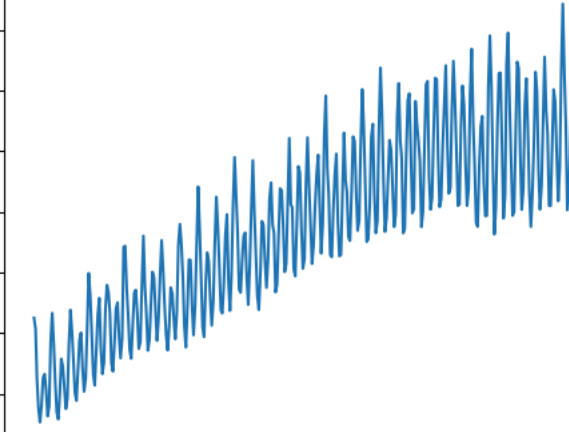


```
1 df.head()
```

	DATE	IPG2211A2N
0	1/1/1985	72.5052
1	2/1/1985	70.6720
2	3/1/1985	62.4502
3	4/1/1985	57.4714
4	5/1/1985	55.3151

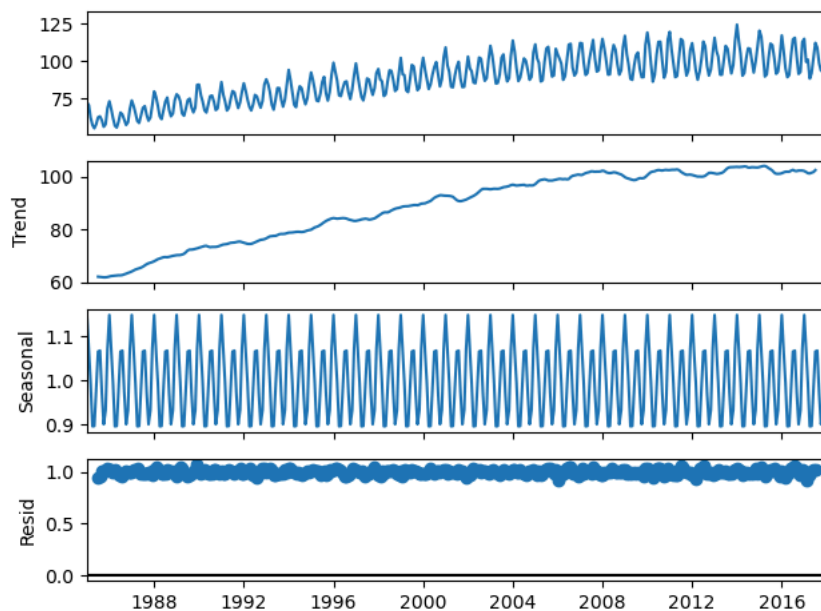
```
1 df.head()
```

IPG2211A2N	
DATE	
1985-01-01	72.5052
1985-02-01	70.6720
1985-03-01	62.4502
1985-04-01	57.4714
1985-05-01	55.3151

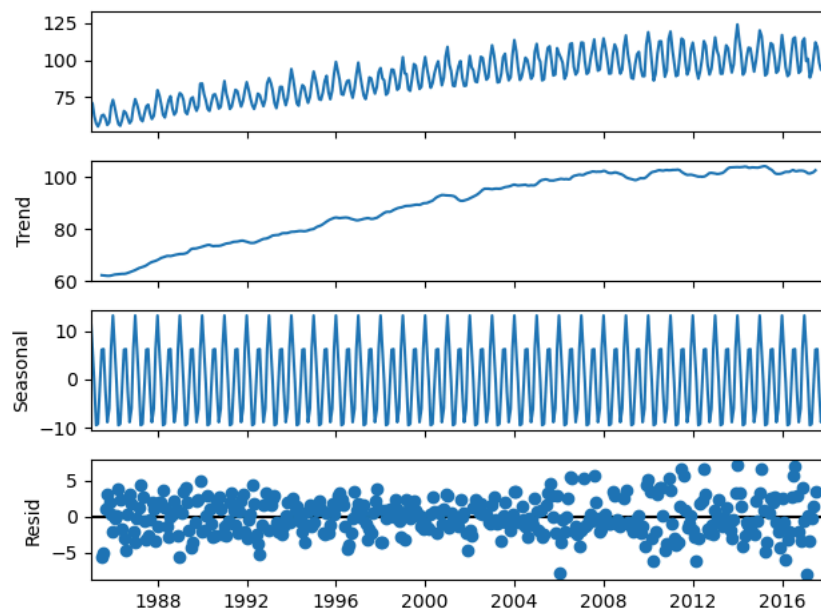


The line plot shows the time series of IPG2211A2N from 1984 to 2017. The x-axis is labeled 'DATE' and ranges from 1984 to 2017 with major ticks every four years. The y-axis is labeled 'IPG2211A2N' and ranges from 60 to 130 with major ticks every 10 units. The plot features a single blue line representing the data. It exhibits a clear upward trend with significant seasonal fluctuations. The values start around 65 in 1984 and rise to approximately 130 by 2017. The seasonal pattern is consistent, with peaks occurring roughly every four years, reaching values between 110 and 125, and troughs occurring roughly every four years, reaching values between 65 and 85.

1/15



```
1 result=seasonal_decompose(df[['IPG2211A2N']],model='additive')
2 result.plot()
3 plt.show()
```



```
1 pip install pymannkendall
```



Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: pymannkendall in c:\users\nmims.student\appdata\roaming\python\python311\site-packages (1.4.3)
Requirement already satisfied: numpy in c:\programdata\anaconda3\lib\site-packages (from pymannkendall) (1.26.4)
Requirement already satisfied: scipy in c:\programdata\anaconda3\lib\site-packages (from pymannkendall) (1.11.4)
Note: you may need to restart the kernel to use updated packages.

```
1 import pymannkendall as mk
2 mk.original_test(df[['IPG2211A2N']])
```



Mann_Kendall_Test(trend='increasing', h=True, p=0.0, z=19.125339232804578, Tau=0.6427499173091112, s=50524.0, var_s=6978466.0, slope=0.11486610139860137, intercept=67.03601192307693)

```
1 train_df=df[:int(df.shape[0]*0.7)]
2 test_df=df[int(df.shape[0]*0.7):]
```

```
1 from statsmodels.tsa.api import ExponentialSmoothing
2 model_triple = ExponentialSmoothing(train_df, seasonal_periods=12, trend='add', seasonal='add')
3 model_triple_fit =model_triple.fit()
```

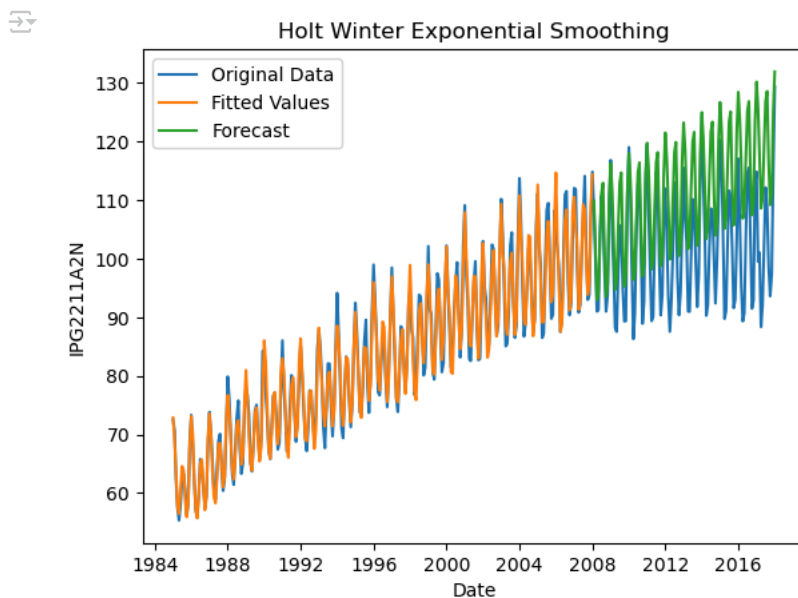


C:\ProgramData\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provide
self._init_dates(dates, freq)

```
1 forecast_triple= model_triple_fit.forecast(120)
2 print(forecast_triple)
```

```
2008-02-01    109.875492
2008-03-01    102.218329
2008-04-01     92.951407
2008-05-01     94.533240
2008-06-01    103.490456
...
2017-09-01    116.297379
2017-10-01    109.190245
2017-11-01    111.197022
2017-12-01    125.175611
2018-01-01    131.934824
Freq: MS, Length: 120, dtype: float64
```

```
1 plt.plot(df, label='Original Data')
2 plt.plot(model_triple_fit.fittedvalues, label='Fitted Values')
3 plt.plot(forecast_triple, label='Forecast')
4 plt.xlabel('Date')
5 plt.ylabel('IPG2211A2N')
6 plt.title('Holt Winter Exponential Smoothing')
7 plt.legend()
8 plt.show()
```



```
1 from sklearn.metrics import mean_squared_error, mean_absolute_percentage_error
2 mape_train = mean_absolute_percentage_error(train_df['IPG2211A2N'], model_triple_fit.fittedvalues)
3 print('Mape_Train', mape_train)
```

```
Mape_Train 0.01985344246329941
```

```
1 test_df.shape
```

```
(120, 1)
```

```
1 mape_test = mean_absolute_percentage_error(test_df['IPG2211A2N'], forecast_triple)
2 print('Mape_Test', mape_test)
```

```
Mape_Test 0.09448179758628884
```

```
1 # ADF Test
2 # Ho : Series is non-stationary, or series has a unit root
3 # HA : Series is Stationary, or series has no unit root
4 # p-value < 0.05 - Reject Null Hypothesis (Ho)
```

```
1 from statsmodels.tsa.stattools import adfuller
2 print("Result of Dickey Fuller Test")
3 dfctest = adfuller(df['IPG2211A2N'])
4 dfcoutput = pd.Series(dfctest[0:4], index=["Test Statistic", 'P-value', '#Lags Used', 'Number of Observations'])
5 for key,value in dfctest[4].items():
```

```
6 dfoutput['Critical Value (%s)' %key]=value
7 print(dfoutput)
```

```
Result of Dickey Fuller Test
Test Statistic      -2.256990
P-value             0.186215
#Lags Used          15.000000
Number of Observations 381.000000
Critical Value (1%)  -3.447631
Critical Value (5%)  -2.869156
Critical Value (10%) -2.570827
dtype: float64
```

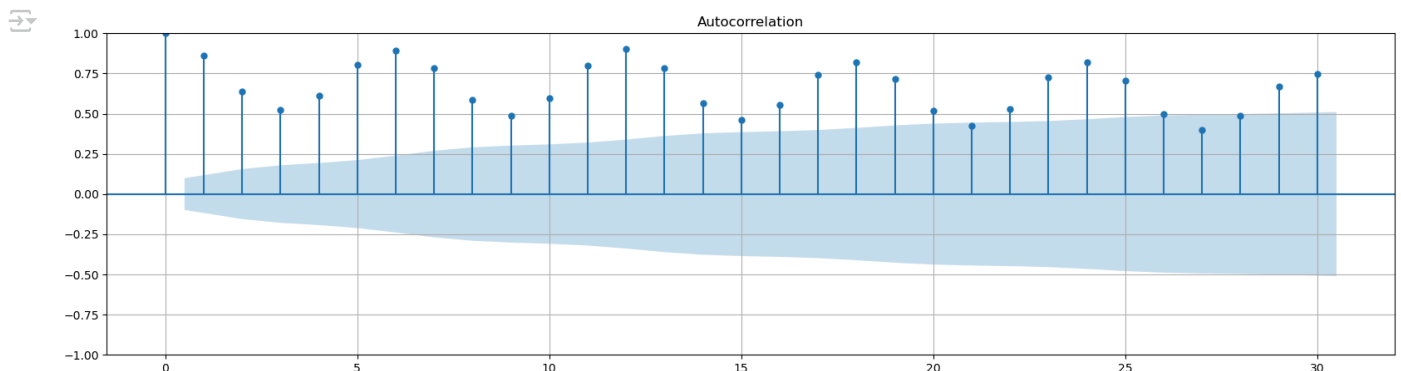
```
1 # KPSS Test
2 # Ho : Series is Trend Stationary, or series has no unit root
3 # HA : Series is non-stationary, or series has a unit root
4 # p-value < 0.05 - Reject Null Hypothesis (Ho)
```

```
1 from statsmodels.tsa.stattools import kpss
2 kp = kpss(df['IPG2211A2N'])
3 p = kp[1]
4
5 print("p-value for KPSS Test (untransformed) = ", p)
```

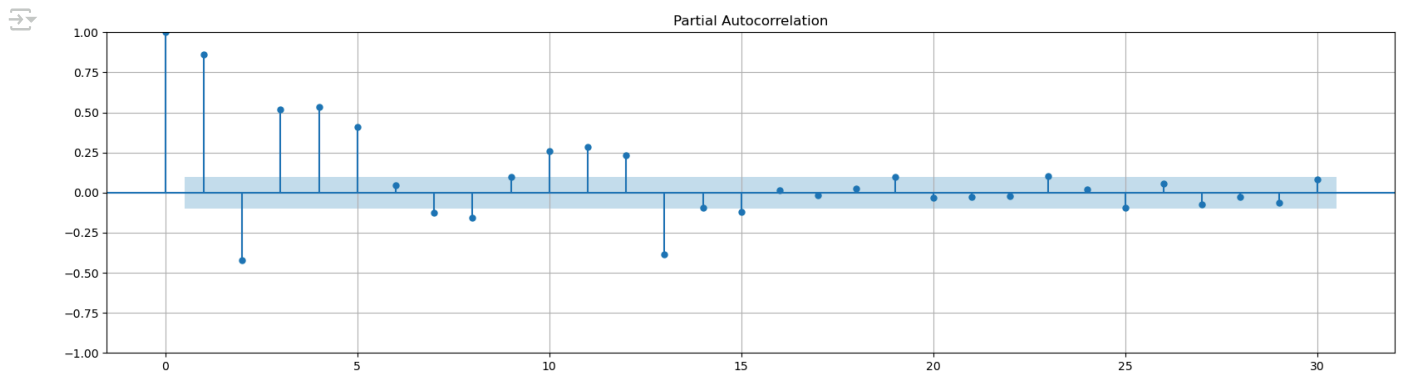
```
p-value for KPSS Test (untransformed) = 0.01
C:\Users\nmims.student\AppData\Local\Temp\ipykernel_8372\1245037153.py:2: InterpolationWarning: The test statistic is outside of the
look-up table. The actual p-value is smaller than the p-value returned.
```

```
kp = kpss(df['IPG2211A2N'])
```

```
1 from statsmodels.tsa.stattools import acf,pacf
2 from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
3 plt.figure(figsize=(20,5))
4 plt.grid()
5 plot_acf(df['IPG2211A2N'], ax = plt.gca(), lags=30)
6 plt.show()
```



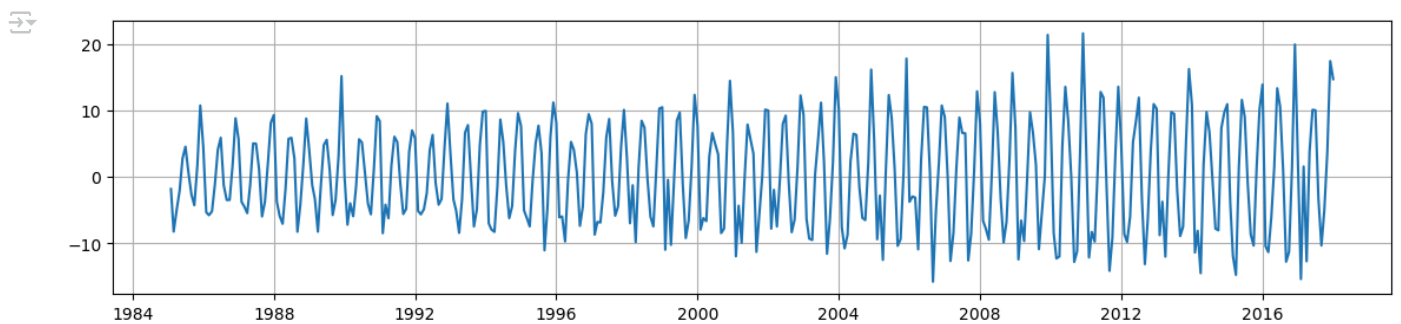
```
1 plt.figure(figsize=(20,5))
2 plt.grid()
3 plot_pacf(df['IPG2211A2N'], ax = plt.gca(), lags=30)
4 plt.show()
```



```

1 # First Order Non Seasonal Differencing
2 diff = df['IPG2211A2N'].diff().dropna()      # Trying to make data Trend Stationary
3 plt.figure(figsize=(14,3))
4 plt.grid()
5 plt.plot(diff)
6 plt.show()

```



```

1 # ADF Test
2 from statsmodels.tsa.stattools import adfuller
3 print("Result of Dickey Fuller Test")
4 dfctest = adfuller(diff)
5 dfcoutput = pd.Series(dfctest[0:4], index=["Test Statistic", 'P-value', '#Lags Used', 'Number of Observations'])
6 for key,value in dfctest[4].items():
7     dfcoutput['Critical Value (%s)' %key]=value
8 print(dfcoutput)

```

```

Result of Dickey Fuller Test
Test Statistic      -7.104891e+00
P-value              4.077787e-10
#Lags Used           1.400000e+01
Number of Observations  3.810000e+02
Critical Value (1%)   -3.447631e+00
Critical Value (5%)   -2.869156e+00
Critical Value (10%)  -2.570827e+00
dtype: float64

```

```

1 # KPSS Test
2 from statsmodels.tsa.stattools import kpss
3 kp = kpss(diff)
4 p = kp[1]
5
6 print("p-value for KPSS Test (untransformed) = ", p)

```

```

p-value for KPSS Test (untransformed) = 0.1
C:\Users\nmims.student\AppData\Local\Temp\ipykernel_8372\528972782.py:3: InterpolationWarning: The test statistic is outside of the
look-up table. The actual p-value is greater than the p-value returned.

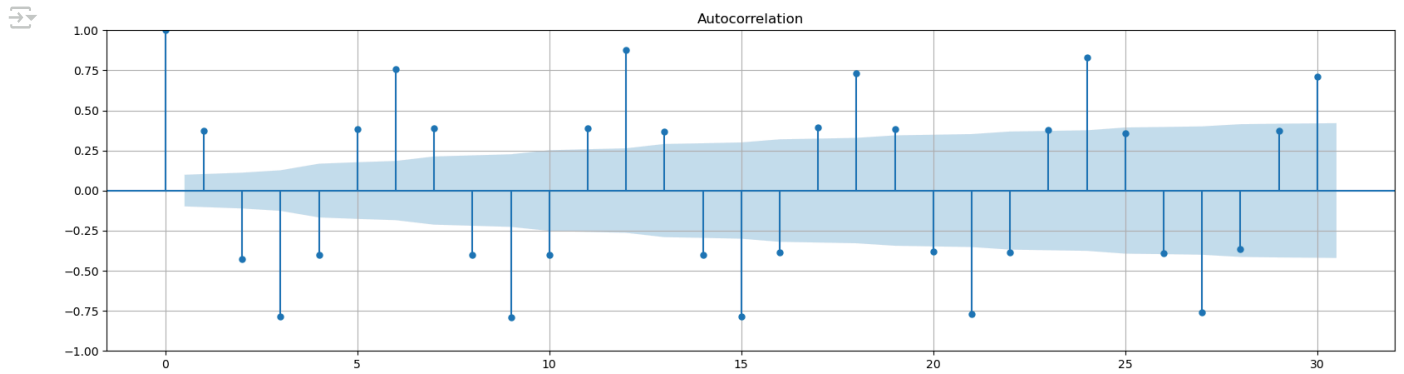
```

```
kp = kpss(diff)
```

```

1 # ACF Plot
2 from statsmodels.tsa.stattools import acf,pacf
3 from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
4 plt.figure(figsize=(20,5))
5 plt.grid()
6 plot_acf(diff, ax = plt.gca(), lags=30)
7 plt.show()

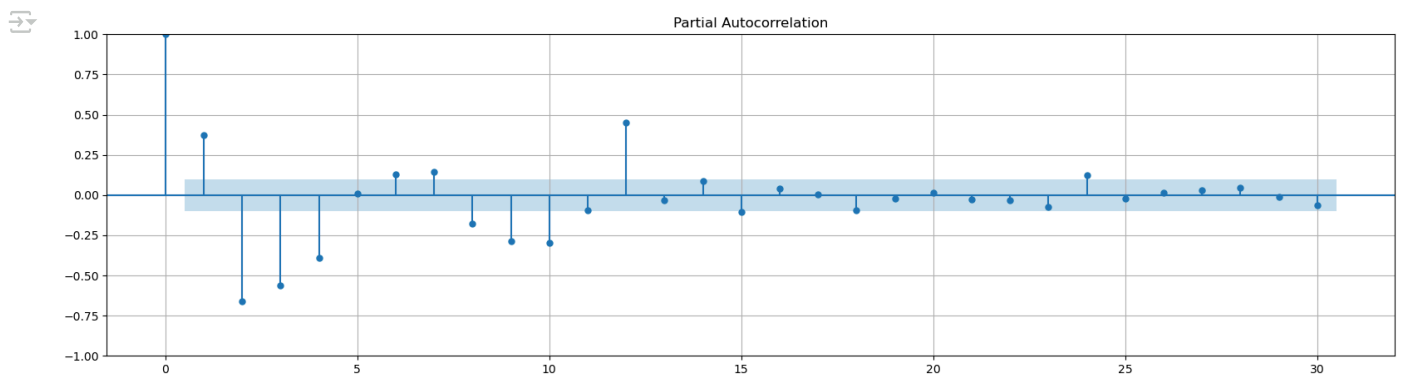
```



```

1 # PACF Plot
2 plt.figure(figsize=(20,5))
3 plt.grid()
4 plot_pacf(diff, ax = plt.gca(), lags=30)
5 plt.show()

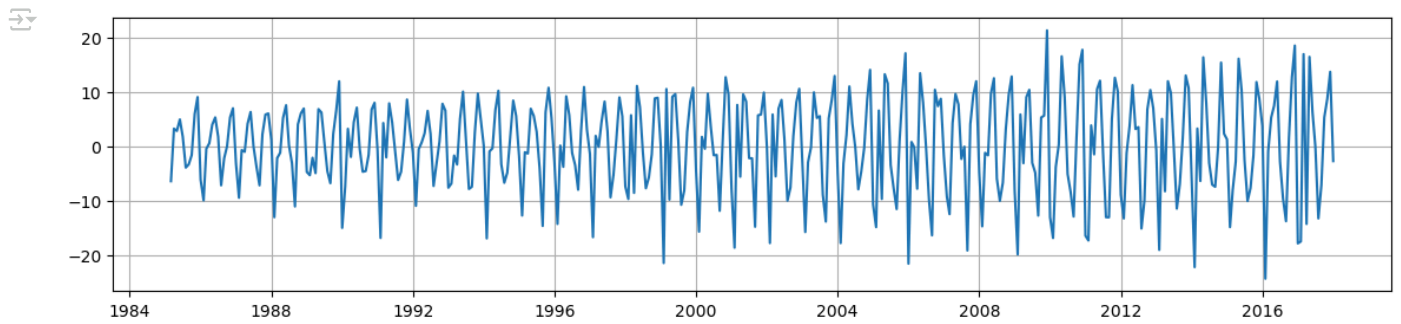
```



```

1 # Second Order Non Seasonal Differencing
2 diff2 = df['IPG2211A2N'].diff().diff().dropna() # Trying to make data Trend Stationary
3 plt.figure(figsize=(14,3))
4 plt.grid()
5 plt.plot(diff2)
6 plt.show()

```



```

1 # ADF Test
2 from statsmodels.tsa.stattools import adfuller
3 print("Result of Dickey Fuller Test")
4 dfctest = adfuller(diff2)
5 dfoutput = pd.Series(dfctest[0:4], index=["Test Statistic", 'P-value', '#Lags Used', 'Number of Observations'])
6 for key,value in dfctest[4].items():
7     dfoutput['Critical Value (%s)' %key]=value
8 print(dfoutput)

```

```

Result of Dickey Fuller Test
Test Statistic      -9.863042e+00
P-value             4.183694e-17
#Lags Used          1.700000e+01
Number of Observations  3.770000e+02
Critical Value (1%)   -3.447815e+00
Critical Value (5%)   -2.869237e+00
Critical Value (10%)  -2.570870e+00
dtype: float64

```

```

1 # KPSS Test
2 from statsmodels.tsa.stattools import kpss
3 kp = kpss(diff2)
4 p = kp[1]
5
6 print("p-value for KPSS Test (untransformed) = ", p)

```

```

p-value for KPSS Test (untransformed) = 0.1
C:\Users\nmims.student\AppData\Local\Temp\ipykernel_8372\1400720088.py:3: InterpolationWarning: The test statistic is outside of the
look-up table. The actual p-value is greater than the p-value returned.

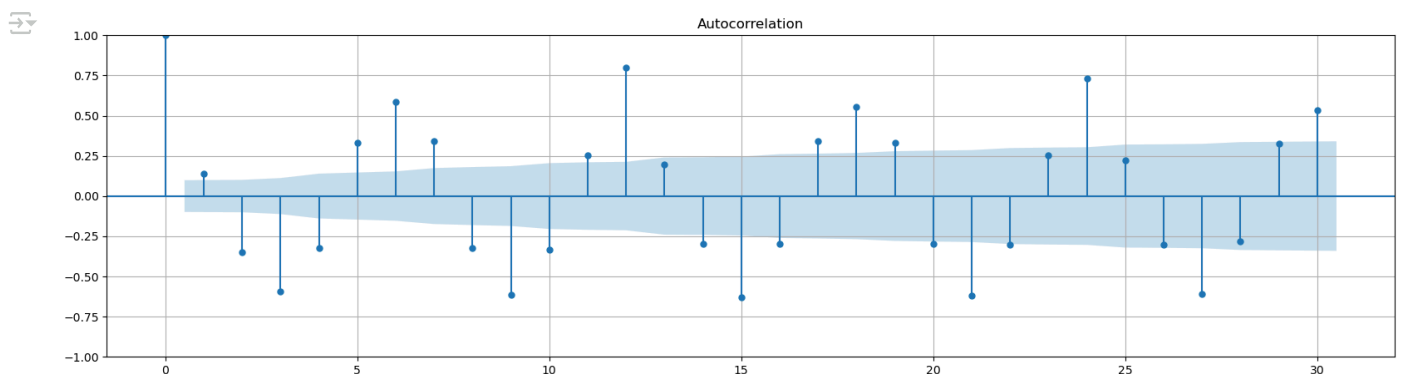
```

```
kp = kpss(diff2)
```

```

1 # ACF Plot
2 from statsmodels.tsa.stattools import acf, pacf
3 from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
4 plt.figure(figsize=(20,5))
5 plt.grid()
6 plot_acf(diff2, ax = plt.gca(), lags=30)
7 plt.show()

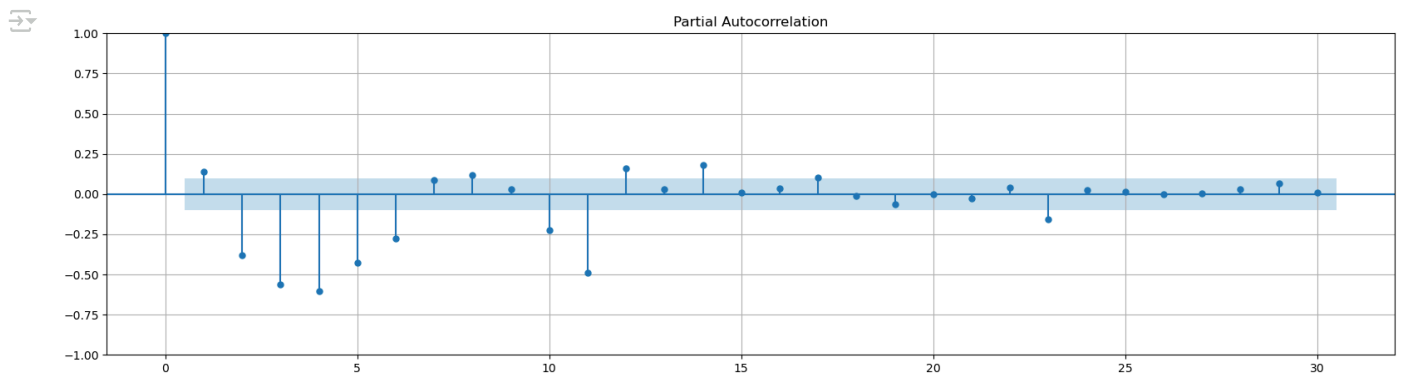
```



```

1 # PACF Plot
2 plt.figure(figsize=(20,5))
3 plt.grid()
4 plot_pacf(diff2, ax = plt.gca(), lags=30)
5 plt.show()

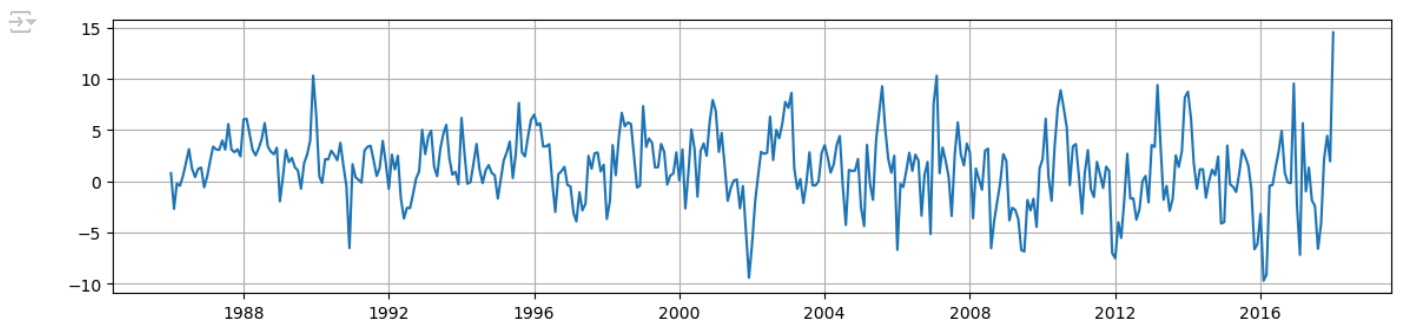
```



```

1 # First Order Seasonal Differencing
2 sdiff = df['IPG2211A2N'].diff(periods=12).dropna() # Trying to make data Trend Stationary
3 plt.figure(figsize=(14,3))
4 plt.grid()
5 plt.plot(sdiff)
6 plt.show()

```



```

1 # ADF Test
2 from statsmodels.tsa.stattools import adfuller
3 print("Result of Dickey Fuller Test")
4 dfctest = adfuller(sdiff)
5 dfctest_output = pd.Series(dfctest[0:4], index=["Test Statistic", 'P-value', '#Lags Used', 'Number of Observations'])
6 for key,value in dfctest_output.items():
7     dfctest_output['Critical Value (%s)' %key]=value
8 print(dfctest_output)

```

```

Result of Dickey Fuller Test
Test Statistic      -5.673482e+00
P-value              8.812645e-07
#Lags Used           1.200000e+01
Number of Observations  3.720000e+02
Critical Value (1%)   -3.448052e+00
Critical Value (5%)   -2.869341e+00
Critical Value (10%)  -2.570926e+00
dtype: float64

```



```

1 # KPSS Test
2 from statsmodels.tsa.stattools import kpss
3 kp = kpss(sdiff)
4 p = kp[1]
5
6 print("p-value for KPSS Test (untransformed) = ", p)

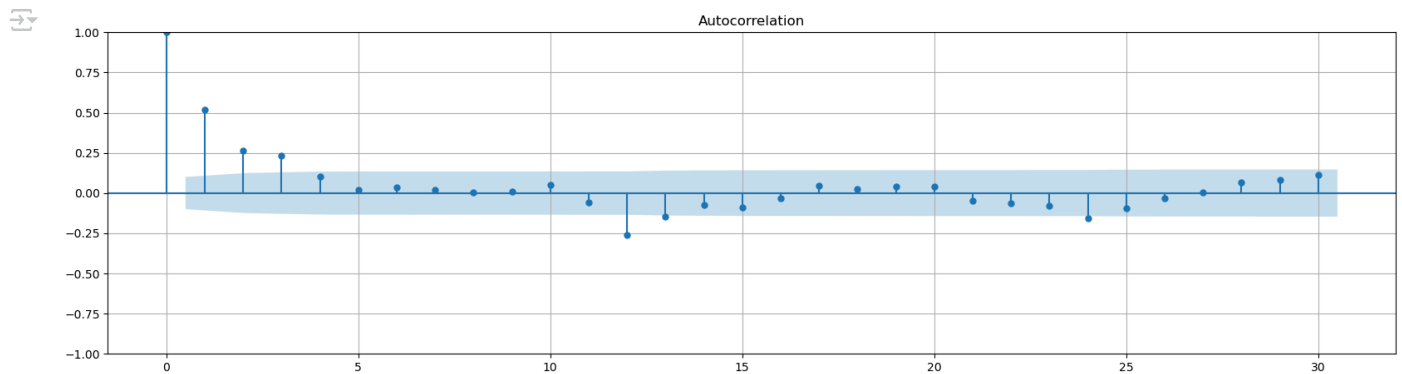
```

↗ p-value for KPSS Test (untransformed) = 0.024345634791440068

```

1 # ACF Plot
2 from statsmodels.tsa.stattools import acf, pacf
3 from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
4 plt.figure(figsize=(20,5))
5 plt.grid()
6 plot_acf(sdiff, ax = plt.gca(), lags=30)
7 plt.show()

```



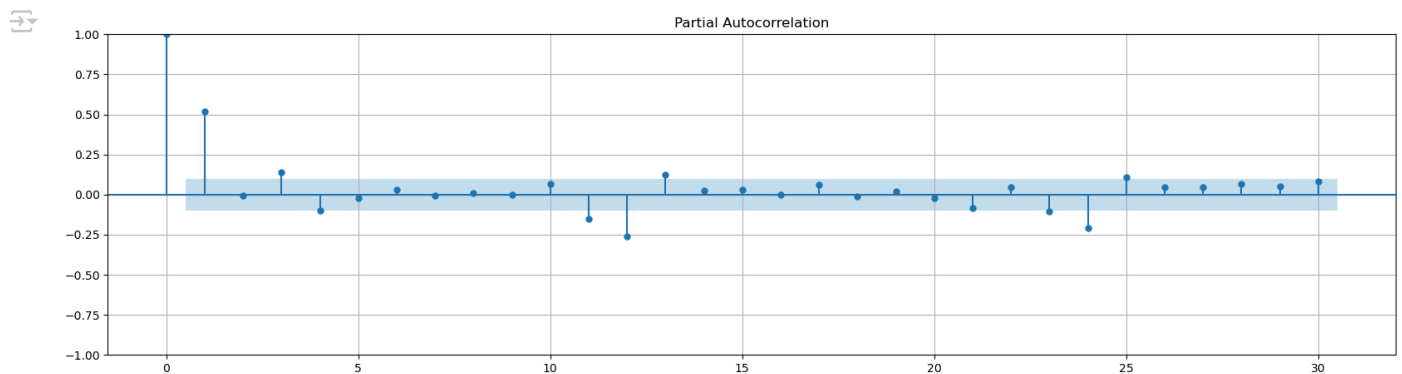
+ Code

+ Text

```

1 # PACF Plot
2 plt.figure(figsize=(20,5))
3 plt.grid()
4 plot_pacf(sdiff, ax = plt.gca(), lags=30)
5 plt.show()

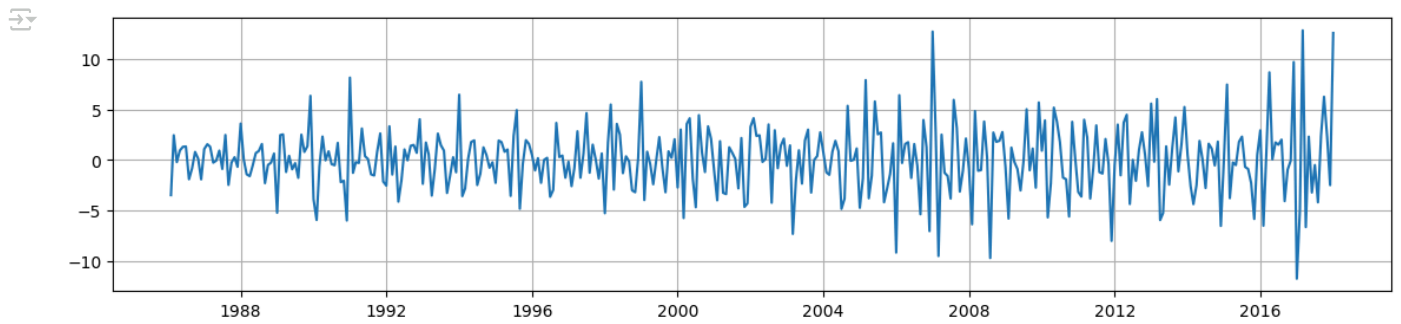
```



```

1 # First Order Seasonal & Non-Seasonal Differencing
2 sdiff = df['IPG2211A2N'].diff(periods=12).dropna() # Seasonal Differencing
3 sddiff = sdiff.diff().dropna() # Non-Seasonal Differencing
4 plt.figure(figsize=(14,3))
5 plt.grid()
6 plt.plot(sddiff)
7 plt.show()

```



```

1 # ADF Test
2 from statsmodels.tsa.stattools import adfuller
3 print("Result of Dickey Fuller Test")
4 dfctest = adfuller(sddiff)
5 dfctest = pd.Series(dfctest[0:4], index=["Test Statistic", 'P-value', '#Lags Used', 'Number of Observations'])
6 for key,value in dfctest[4].items():
7     dfctest['Critical Value (%)' %key]=value
8 print(dfctest)

```

```

Result of Dickey Fuller Test
Test Statistic      -8.022039e+00
P-value             2.063925e-12
#Lags Used          1.500000e+01
Number of Observations 3.680000e+02
Critical Value (1%)  -3.448245e+00
Critical Value (5%)  -2.869426e+00
Critical Value (10%) -2.570971e+00
dtype: float64

```

```

1 # KPSS Test
2 from statsmodels.tsa.stattools import kpss
3 kp = kpss(sddiff)
4 p = kp[1]
5
6 print("p-value for KPSS Test (untransformed) = ", p)

```

```

p-value for KPSS Test (untransformed) = 0.1
C:\Users\nmims.student\AppData\Local\Temp\ipykernel_8372\546374317.py:3: InterpolationWarning: The test statistic is outside of the
look-up table. The actual p-value is greater than the p-value returned.

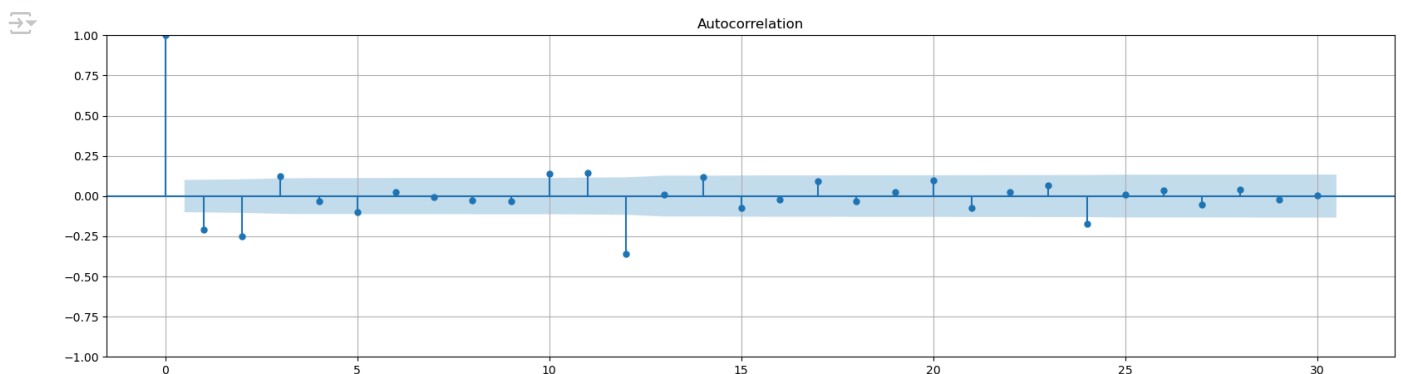
```

```
kp = kpss(sddiff)
```

```

1 # ACF Plot
2 from statsmodels.tsa.stattools import acf, pacf
3 from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
4 plt.figure(figsize=(20,5))
5 plt.grid()
6 plot_acf(sddiff, ax = plt.gca(), lags=30)
7 plt.show()

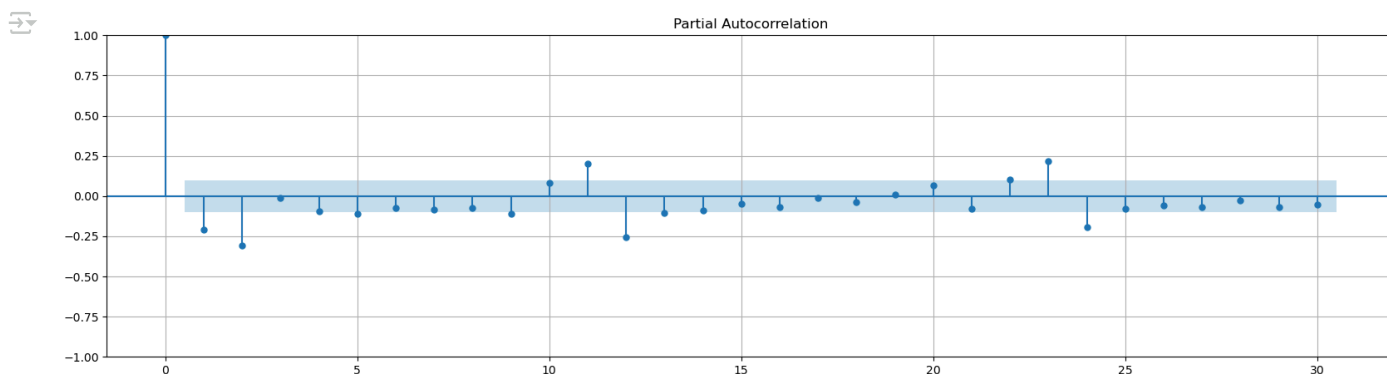
```



```

1 # PACF Plot
2 plt.figure(figsize=(20,5))
3 plt.grid()
4 plot_pacf(sddiff, ax = plt.gca(), lags=30)
5 plt.show()

```



```
1 !pip install pmdarima
```

Defaulting to user installation because normal site-packages is not writeable

Requirement already satisfied: pmdarima in c:\users\nmims.student\appdata\roaming\python\python311\site-packages (2.0.4)

Requirement already satisfied: joblib>=0.11 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (1.2.0)

Requirement already satisfied: Cython!=0.29.18,!0.29.31,>=0.29 in c:\users\nmims.student\appdata\roaming\python\python311\site-packages (from pmdarima) (3.0.12)

Requirement already satisfied: numpy>=1.21.2 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (1.26.4)

Requirement already satisfied: pandas>=0.19 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (2.1.4)

Requirement already satisfied: scikit-learn>=0.22 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (1.2.2)

Requirement already satisfied: scipy>=1.3.2 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (1.11.4)

Requirement already satisfied: statsmodels>=0.13.2 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (0.14.0)

Requirement already satisfied: urllib3 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (2.0.7)

Requirement already satisfied: setuptools!=50.0.0,>=38.6.0 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (68.2.2)

Requirement already satisfied: packaging>=17.1 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (23.1)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\programdata\anaconda3\lib\site-packages (from pandas>=0.19->pmdarima) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in c:\programdata\anaconda3\lib\site-packages (from pandas>=0.19->pmdarima) (2023.3.post1)

Requirement already satisfied: tzdata>=2022.1 in c:\programdata\anaconda3\lib\site-packages (from pandas>=0.19->pmdarima) (2023.3)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\programdata\anaconda3\lib\site-packages (from scikit-learn>=0.22->pmdarima) (3.1.0)

Requirement already satisfied: patsy>=0.5.2 in c:\programdata\anaconda3\lib\site-packages (from statsmodels>=0.13.2->pmdarima) (0.5.2)

Requirement already satisfied: six in c:\programdata\anaconda3\lib\site-packages (from patsy>=0.5.2->statsmodels>=0.13.2->pmdarima) (1.16.0)

```
1 pip install pymannkendall pmdarima
```

Defaulting to user installation because normal site-packages is not writeable

Requirement already satisfied: pymannkendall in c:\users\nmims.student\appdata\roaming\python\python311\site-packages (1.4.3)

Requirement already satisfied: pmdarima in c:\users\nmims.student\appdata\roaming\python\python311\site-packages (2.0.4)

Requirement already satisfied: numpy in c:\programdata\anaconda3\lib\site-packages (from pymannkendall) (1.26.4)

Requirement already satisfied: scipy in c:\programdata\anaconda3\lib\site-packages (from pymannkendall) (1.11.4)

Requirement already satisfied: joblib>=0.11 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (1.2.0)

Requirement already satisfied: Cython!=0.29.18,!0.29.31,>=0.29 in c:\users\nmims.student\appdata\roaming\python\python311\site-packages (from pmdarima) (3.0.12)

Requirement already satisfied: pandas>=0.19 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (2.1.4)

Requirement already satisfied: scikit-learn>=0.22 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (1.2.2)

Requirement already satisfied: statsmodels>=0.13.2 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (0.14.0)

Requirement already satisfied: urllib3 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (2.0.7)

Requirement already satisfied: setuptools!=50.0.0,>=38.6.0 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (68.2.2)

Requirement already satisfied: packaging>=17.1 in c:\programdata\anaconda3\lib\site-packages (from pmdarima) (23.1)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\programdata\anaconda3\lib\site-packages (from pandas>=0.19->pmdarima) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in c:\programdata\anaconda3\lib\site-packages (from pandas>=0.19->pmdarima) (2023.3.post1)

Requirement already satisfied: tzdata>=2022.1 in c:\programdata\anaconda3\lib\site-packages (from pandas>=0.19->pmdarima) (2023.3)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\programdata\anaconda3\lib\site-packages (from scikit-learn>=0.22->pmdarima) (3.1.0)

Requirement already satisfied: patsy>=0.5.2 in c:\programdata\anaconda3\lib\site-packages (from statsmodels>=0.13.2->pmdarima) (0.5.2)

Requirement already satisfied: six in c:\programdata\anaconda3\lib\site-packages (from patsy>=0.5.2->statsmodels>=0.13.2->pmdarima) (1.16.0)

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

```

1 import statsmodels.api as sm
2 model_1 = sm.tsa.statespace.SARIMAX(train_df['IPG2211A2N'], order = (1,1,1), seasonal_order = (1,1,1,12))
3 model_1_fit = model_1.fit()
4
5 print(model_1_fit.summary())

```

C:\ProgramData\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provide
self._init_dates(dates, freq)
C:\ProgramData\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provide
self._init_dates(dates, freq)

SARIMAX Results

```

=====
Dep. Variable:          IPG2211A2N      No. Observations:          277
Model:                SARIMAX(1, 1, 1)x(1, 1, 1, 12)  Log Likelihood          -575.211
Date:                  Mon, 03 Feb 2025      AIC              1160.422
Time:                  14:00:50              BIC              1178.302
Sample:                01-01-1985          HQIC             1167.607
                   - 01-01-2008
Covariance Type:                opg
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
ar.L1           0.4861      0.056       8.618      0.000       0.376       0.597
ma.L1          -0.9810      0.019     -52.984      0.000      -1.017      -0.945
ar.S.L12         0.0288      0.092       0.313      0.755      -0.152       0.209
ma.S.L12        -0.7279      0.080     -9.100      0.000      -0.885      -0.571
sigma2          4.3600      0.313     13.944      0.000       3.747       4.973
=====
Ljung-Box (L1) (Q):                0.02      Jarque-Bera (JB):          55.34
Prob(Q):                          0.88      Prob(JB):              0.00
Heteroskedasticity (H):            2.25      Skew:                  -0.29
Prob(H) (two-sided):              0.00      Kurtosis:              5.17
=====

```

Warnings:

[1] Covariance matrix calculated using the outer product of gradients (complex-step).

```

1 from pmdarima.arima import auto_arima
2 model_auto_arima = auto_arima(df['IPG2211A2N'], seasonal = True, trace = True,
m=12)

```

Performing stepwise search to minimize aic

```

ARIMA(2,1,2)(1,0,1)[12] intercept : AIC=1953.424, Time=1.09 sec
ARIMA(0,1,0)(0,0,0)[12] intercept : AIC=2750.050, Time=0.02 sec
ARIMA(1,1,0)(1,0,0)[12] intercept : AIC=inf, Time=0.24 sec
ARIMA(0,1,1)(0,0,1)[12] intercept : AIC=2401.309, Time=0.22 sec
ARIMA(0,1,0)(0,0,0)[12] intercept : AIC=2748.185, Time=0.01 sec
ARIMA(2,1,2)(0,0,1)[12] intercept : AIC=inf, Time=0.91 sec
ARIMA(2,1,2)(1,0,0)[12] intercept : AIC=inf, Time=0.95 sec
ARIMA(2,1,2)(2,0,1)[12] intercept : AIC=1955.636, Time=2.02 sec
ARIMA(2,1,2)(1,0,2)[12] intercept : AIC=inf, Time=2.13 sec
ARIMA(2,1,2)(0,0,0)[12] intercept : AIC=2105.437, Time=0.41 sec
ARIMA(2,1,2)(0,0,2)[12] intercept : AIC=2016.179, Time=1.69 sec
ARIMA(2,1,2)(2,0,0)[12] intercept : AIC=inf, Time=1.97 sec
ARIMA(2,1,2)(2,0,2)[12] intercept : AIC=inf, Time=3.66 sec
ARIMA(1,1,2)(1,0,1)[12] intercept : AIC=1866.763, Time=1.12 sec
ARIMA(1,1,2)(0,0,1)[12] intercept : AIC=2265.648, Time=0.42 sec
ARIMA(1,1,2)(1,0,0)[12] intercept : AIC=1985.835, Time=0.56 sec
ARIMA(1,1,2)(2,0,1)[12] intercept : AIC=1868.527, Time=2.71 sec
ARIMA(1,1,2)(1,0,2)[12] intercept : AIC=1868.082, Time=1.98 sec
ARIMA(1,1,2)(0,0,0)[12] intercept : AIC=2493.118, Time=0.16 sec
ARIMA(1,1,2)(0,0,2)[12] intercept : AIC=2179.668, Time=0.82 sec
ARIMA(1,1,2)(2,0,0)[12] intercept : AIC=1945.389, Time=1.69 sec
ARIMA(1,1,2)(2,0,2)[12] intercept : AIC=inf, Time=3.51 sec
ARIMA(0,1,2)(1,0,1)[12] intercept : AIC=1872.483, Time=0.87 sec
ARIMA(1,1,1)(1,0,1)[12] intercept : AIC=1869.005, Time=0.94 sec
ARIMA(1,1,3)(1,0,1)[12] intercept : AIC=1870.118, Time=1.30 sec
ARIMA(0,1,1)(1,0,1)[12] intercept : AIC=1923.996, Time=0.66 sec
ARIMA(0,1,3)(1,0,1)[12] intercept : AIC=1869.749, Time=1.10 sec
ARIMA(2,1,1)(1,0,1)[12] intercept : AIC=1870.511, Time=0.91 sec
ARIMA(2,1,3)(1,0,1)[12] intercept : AIC=1929.168, Time=1.29 sec
ARIMA(1,1,2)(1,0,1)[12] intercept : AIC=inf, Time=1.27 sec

```

Best model: ARIMA(1,1,2)(1,0,1)[12] intercept
Total fit time: 36.651 seconds

```

1 import statsmodels.api as sm
2 model_2 = sm.tsa.statespace.SARIMAX(train_df['IPG2211A2N'], order = (1,1,2), seasonal_order = (1,0,1,12))
3 model_2_fit = model_2.fit()
4
5 print(model_2_fit.summary())

```



```
-----
NameError                                Traceback (most recent call last)
Cell In[1], line 2
      1 import statsmodels.api as sm
----> 2 model_2 = sm.tsa.statespace.SARIMAX(train_df['IPG2211A2N'], order = (1,1,2), seasonal_order = (1,0,1,12))
      3 model_2_fit = model_2.fit()
      5 print(model_2_fit.summary())

NameError: name 'train_df' is not defined
```

```
1 test_predictions1 = model_1_fit.forecast(len(test_df))
2 forecast_further1 = model_1_fit.forecast(len(test_df['IPG2211A2N'])+90)
3 forecast_further1 = forecast_further1[test_df.index.max():]
4 forecast_further1
```



```
2018-01-01    131.233972
2018-02-01    126.187633
2018-03-01    119.036998
2018-04-01    109.462493
2018-05-01    110.648377
...
2025-03-01    130.734526
2025-04-01    121.160021
2025-05-01    122.345905
2025-06-01    131.234333
2025-07-01    139.346525
Freq: MS, Name: predicted_mean, Length: 91, dtype: float64
```

```
1 test_predictions2 = model_2_fit.forecast(len(test_df))
2 forecast_further2 = model_2_fit.forecast(len(test_df['IPG2211A2N'])+90)
3 forecast_further2 = forecast_further2[test_df.index.max():]
4 forecast_further2
```



```
2018-01-01    129.028578
2018-02-01    124.299441
2018-03-01    117.562077
2018-04-01    108.577938
2018-05-01    109.695474
...
2025-03-01    127.920577
2025-04-01    119.245520
2025-05-01    120.324609
2025-06-01    128.362982
2025-07-01    135.700235
Freq: MS, Name: predicted_mean, Length: 91, dtype: float64
```

```
1 mape_test1 = mean_absolute_percentage_error(test_df['IPG2211A2N'], test_predictions1)
2 print('MAPE for TEST Data:', mape_test1)
```



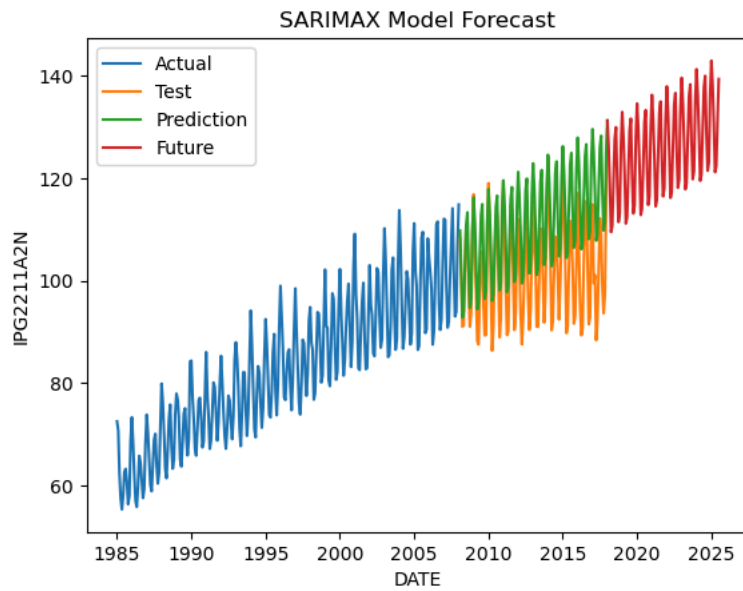
```
MAPE for TEST Data: 0.09303131713801367
```

```
1 mape_test2 = mean_absolute_percentage_error(test_df['IPG2211A2N'], test_predictions2)
2 print('MAPE for TEST Data:', mape_test2)
```



```
MAPE for TEST Data: 0.0862379804230265
```

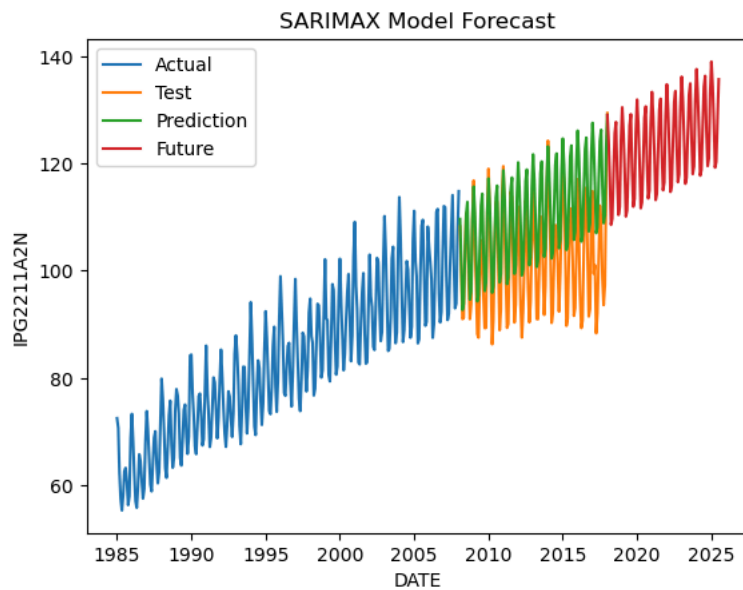
```
1 plt.plot(train_df['IPG2211A2N'], label = 'Actual')
2 plt.plot(test_df['IPG2211A2N'], label = 'Test')
3 plt.plot(test_predictions1, label = 'Prediction')
4 plt.plot(forecast_further1, label = 'Future')
5
6 plt.title('SARIMAX Model Forecast')
7 plt.xlabel('DATE')
8 plt.ylabel('IPG2211A2N')
9 plt.legend()
10
11 plt.show()
```



```

1 plt.plot(train_df['IPG2211A2N'], label = 'Actual')
2 plt.plot(test_df['IPG2211A2N'], label = 'Test')
3 plt.plot(test_predictions2, label = 'Prediction')
4 plt.plot(forecast_further2, label = 'Future')
5
6 plt.title('SARIMAX Model Forecast')
7 plt.xlabel('DATE')
8 plt.ylabel('IPG2211A2N')
9 plt.legend()
10
11 plt.show()

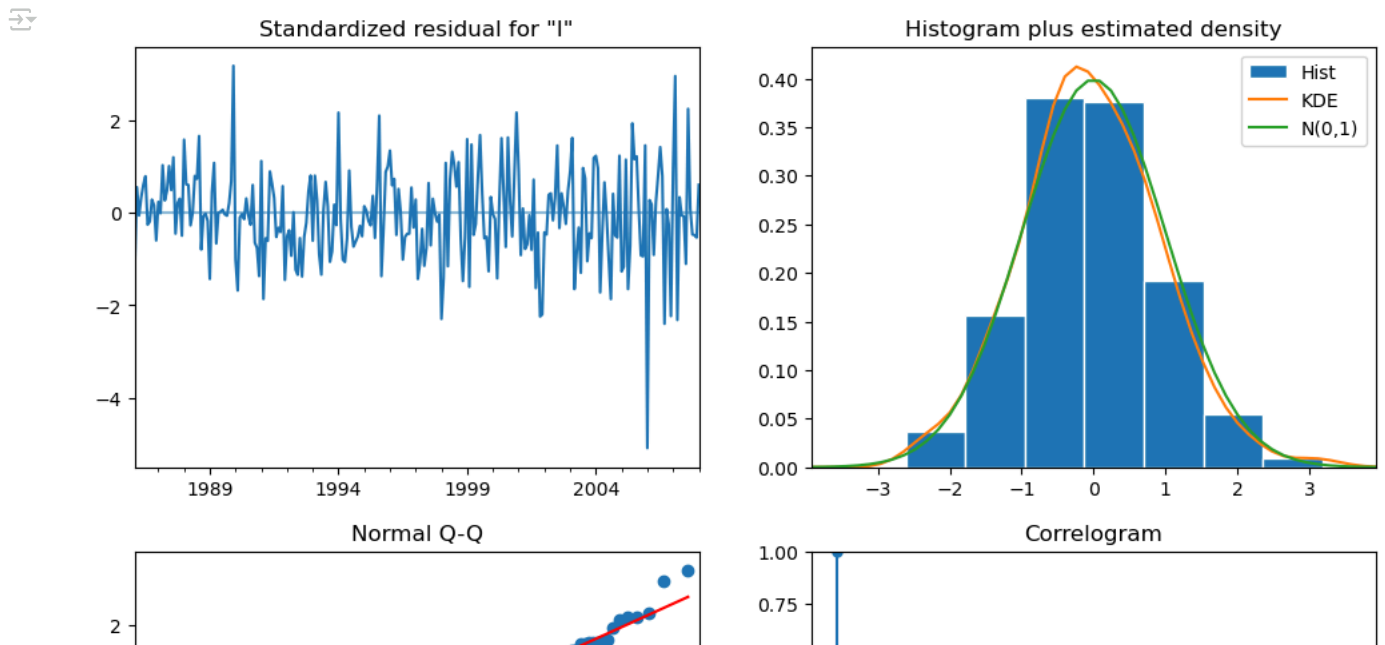
```



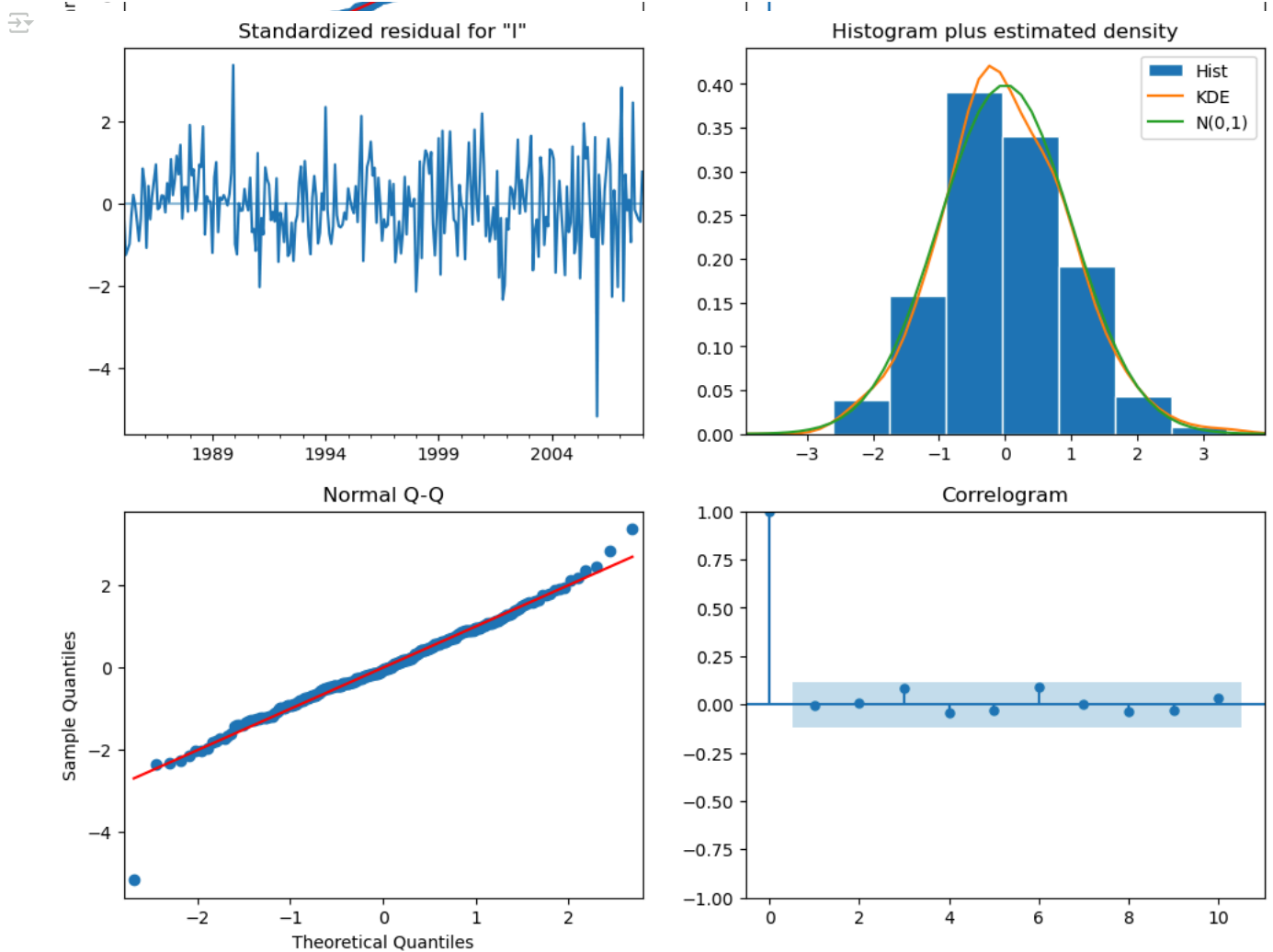
```

1 model_1_fit.plot_diagnostics(figsize=(12,9))
2 plt.show()

```



```
1 model_2_fit.plot_diagnostics(figsize=(12,9))
2 plt.show()
```



```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import seaborn as sns
4
5 # Create some data for the table
6 models = ['Exponential Smoothing', 'SARIMAX (1,1,1)x(1,1,1,12)', 'SARIMAX (1,1,2)x(1,0,1,12)']
7 mane test = [0.09448179758628884, 0.09303131713801367, 0.08623798042302651]
```