

```
In [77]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [78]: df = pd.read_csv("ICNSA.csv")
```

```
In [79]: df.head()
```

Out[79]:

	DATE	ICNSA
0	1967-01-07	346000
1	1967-01-14	334000
2	1967-01-21	277000
3	1967-01-28	252000
4	1967-02-04	274000

```
In [80]: df.shape
```

Out[80]: (2981, 2)

```
In [81]: df.describe
```

Out[81]: <bound method NDFrame.describe of

	DATE	ICNSA
0	1967-01-07	346000
1	1967-01-14	334000
2	1967-01-21	277000
3	1967-01-28	252000
4	1967-02-04	274000
...
2976	2024-01-20	249947
2977	2024-01-27	263919
2978	2024-02-03	234729
2979	2024-02-10	223985
2980	2024-02-17	197932

[2981 rows x 2 columns]>

```
In [82]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2981 entries, 0 to 2980
Data columns (total 2 columns):
#   Column  Non-Null Count  Dtype
---  -
0    DATE      2981 non-null    object
1    ICNSA     2981 non-null    int64
dtypes: int64(1), object(1)
memory usage: 46.7+ KB
```

```
In [83]: df.isnull().sum()
```

Out[83]: DATE 0
ICNSA 0
dtype: int64



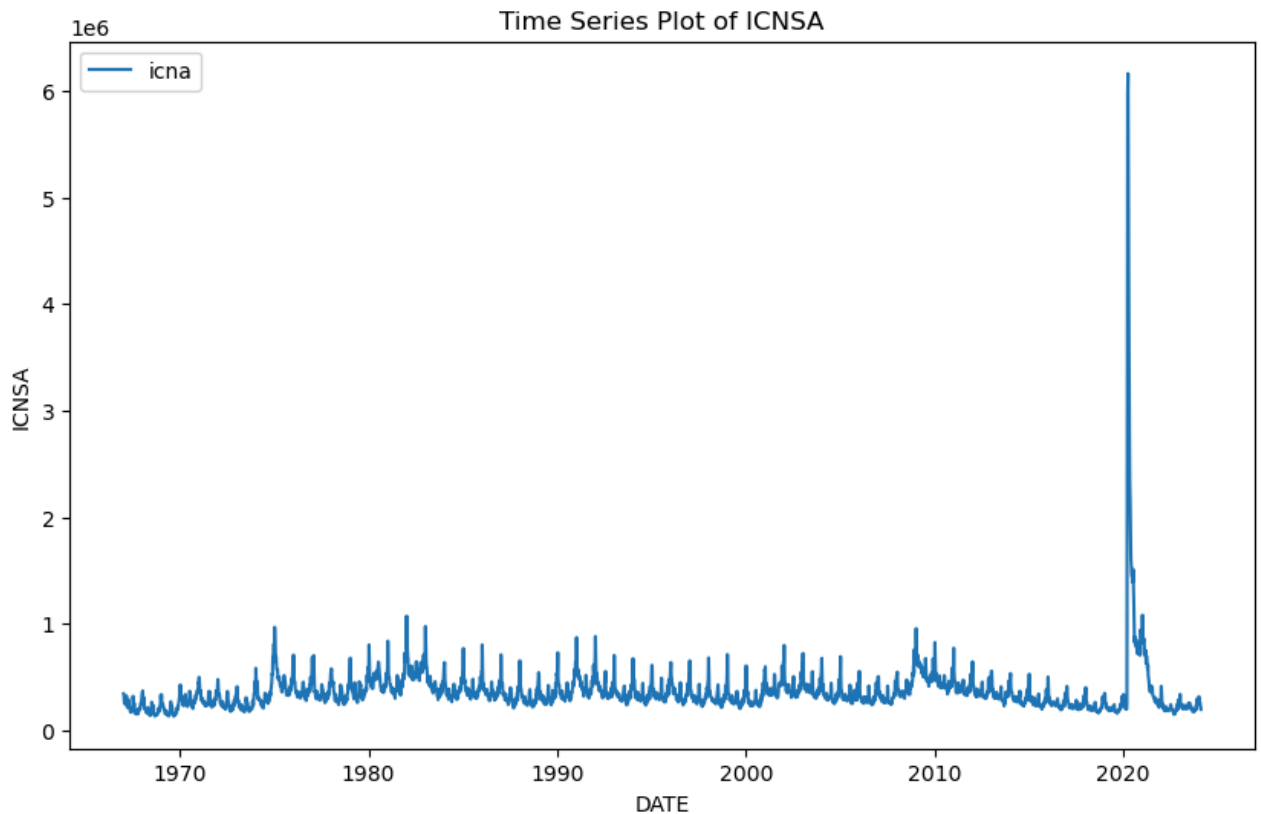
```
In [84]: df['DATE'] = pd.to_datetime(df['DATE'])
df.corr()
```

Out[84]:

	DATE	ICNSA
DATE	1.000000	0.082476
ICNSA	0.082476	1.000000

```
In [85]: plt.figure(figsize=(10, 6))
plt.plot(df['DATE'], df['ICNSA'], label='icna')

plt.xlabel('DATE')
plt.ylabel('ICNSA')
plt.title('Time Series Plot of ICNSA')
plt.legend()
plt.show()
```

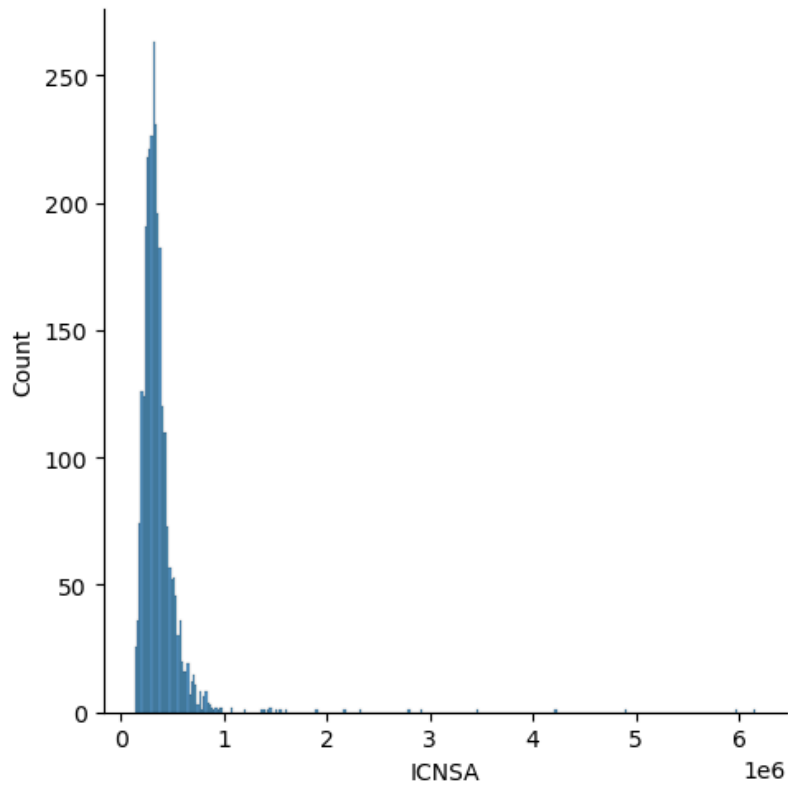


```
In [86]: sns.displot(df['ICNSA'])
```

C:\ProgramData\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning: The figure layout has changed to tight

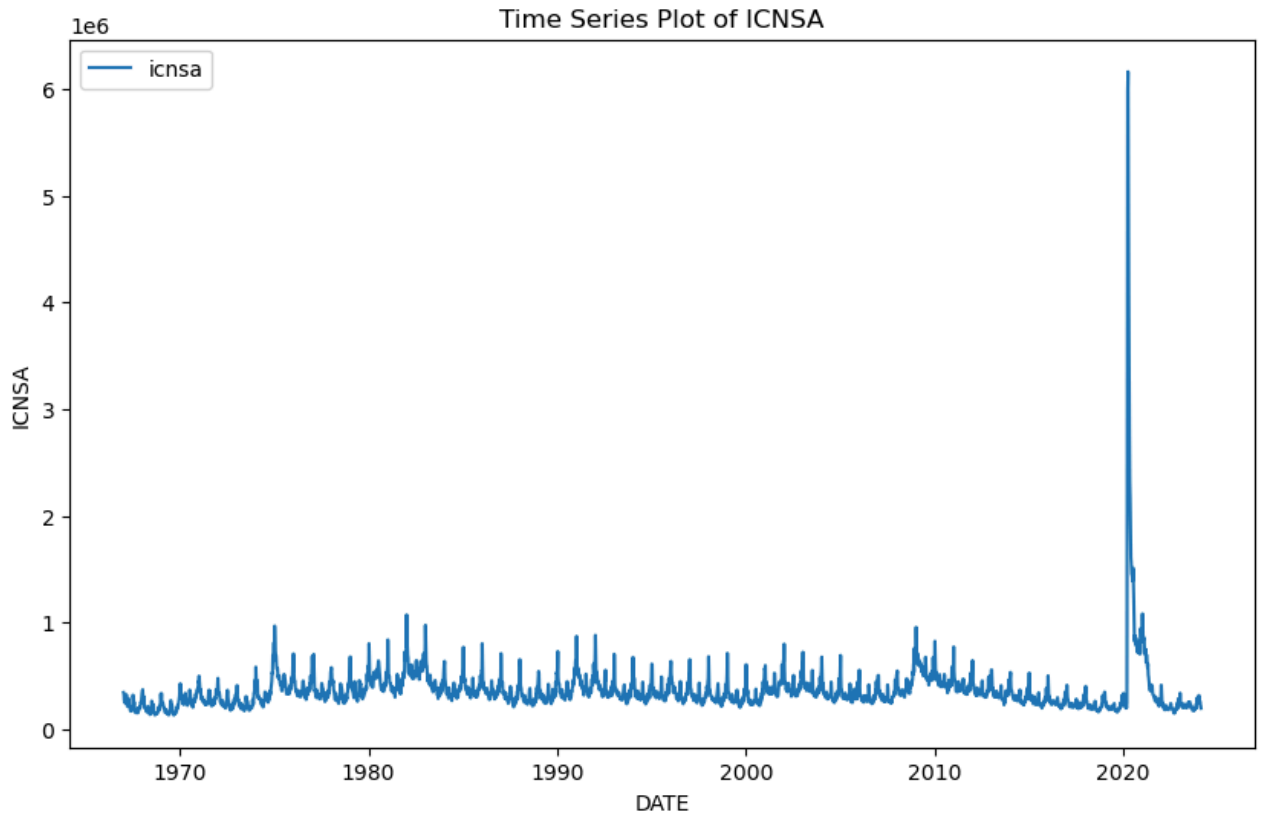
```
self._figure.tight_layout(*args, **kwargs)
```

```
Out[86]: <seaborn.axisgrid.FacetGrid at 0x2b42d072d90>
```



```
In [87]: plt.figure(figsize=(10, 6))
plt.plot(df['DATE'], df['ICNSA'], label='icnsa')

plt.xlabel('DATE')
plt.ylabel('ICNSA')
plt.title('Time Series Plot of ICNSA')
plt.legend()
plt.show()
```

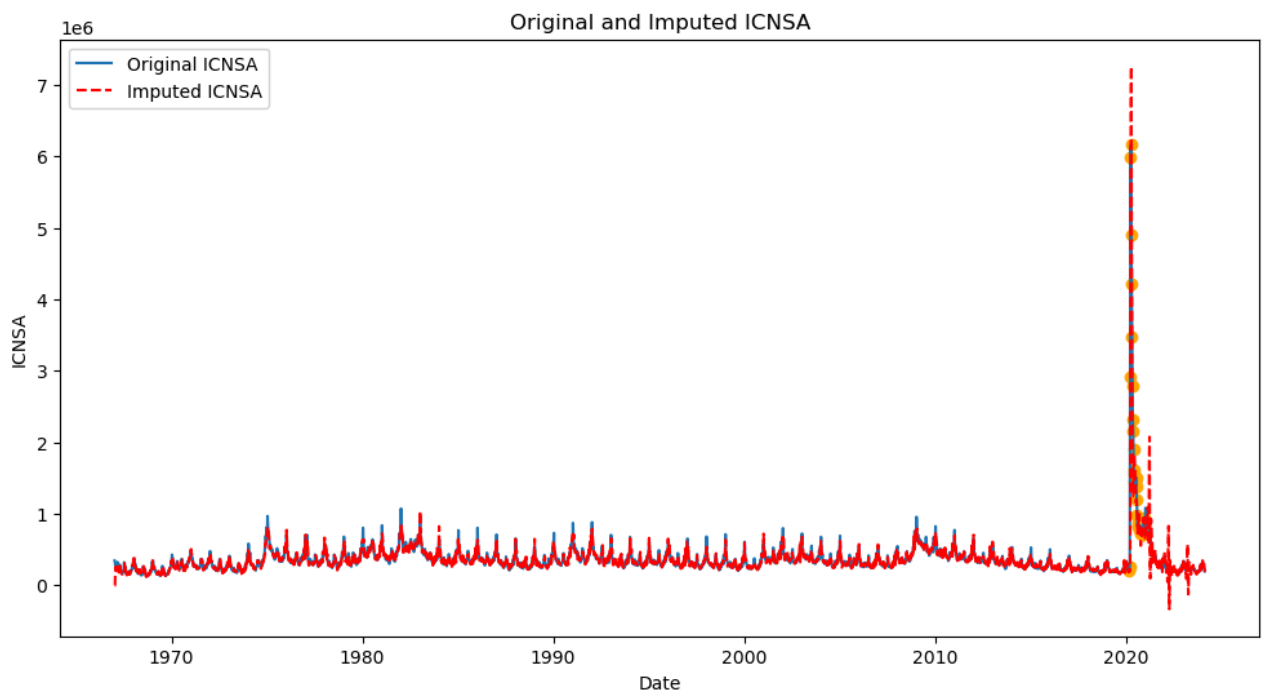


```
In [88]: from statsmodels.tsa.statespace.structural import UnobservedComponents
import matplotlib.pyplot as plt

start_date = '2020-03-01'
end_date = '2020-12-31'
covid_period_data = df[(df['DATE'] >= start_date) & (df['DATE'] <= end_date)]

df['ICNSA_imputed'] = UnobservedComponents(df['ICNSA'],freq_seasonal=[{'period': 52, 'harmonics': 4}])

plt.figure(figsize=(12, 6))
plt.plot(df['DATE'], df['ICNSA'], label='Original ICNSA')
plt.plot(df['DATE'], df['ICNSA_imputed'], label='Imputed ICNSA', linestyle='--',color = 'red')
plt.scatter(covid_period_data['DATE'], covid_period_data['ICNSA'], color='orange')
plt.xlabel('Date')
plt.ylabel('ICNSA')
plt.title('Original and Imputed ICNSA')
plt.legend()
plt.show()
```



```
In [89]: import matplotlib.pyplot as plt

decomposition_result = seasonal_decompose(df['ICNSA'], period=seasonal_period)

fig, axes = plt.subplots(4, 1, figsize=(15, 12))

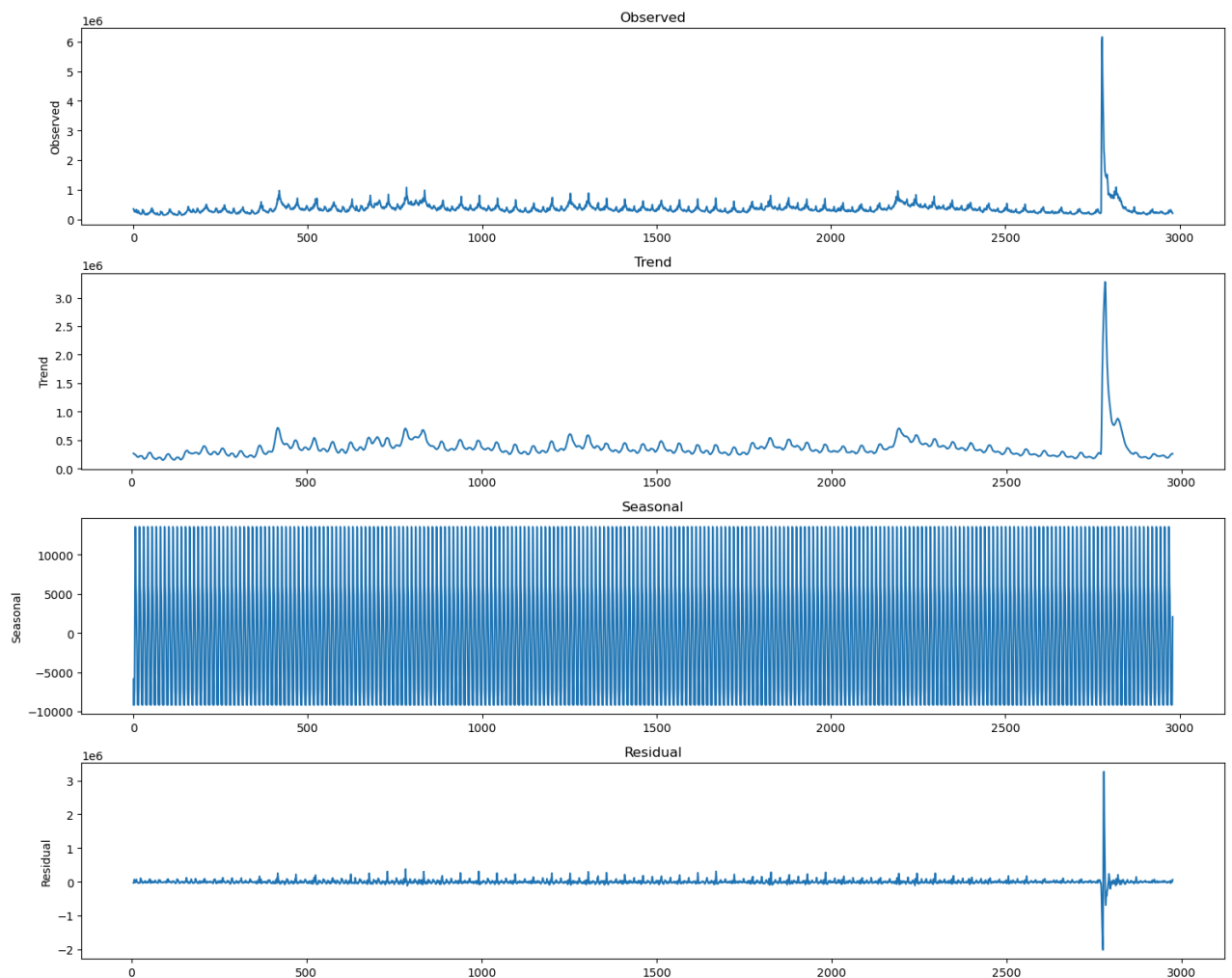
decomposition_result.observed.plot(ax=axes[0], title='Observed')
axes[0].set_ylabel('Observed')

decomposition_result.trend.plot(ax=axes[1], title='Trend')
axes[1].set_ylabel('Trend')

decomposition_result.seasonal.plot(ax=axes[2], title='Seasonal')
axes[2].set_ylabel('Seasonal')

decomposition_result.resid.plot(ax=axes[3], title='Residual')
axes[3].set_ylabel('Residual')

plt.tight_layout()
plt.show()
```



```
In [90]: df_INDPRO= pd.read_csv("INDPRO.csv")
```



In [91]: `df_INDPRO.head()`

Out[91]:

	DATE	INDPRO
0	1919-01-01	4.8665
1	1919-02-01	4.6514
2	1919-03-01	4.5170
3	1919-04-01	4.5976
4	1919-05-01	4.6245

In [92]: `df_INDPRO.shape`

Out[92]: (1261, 2)

In [93]: `df_INDPRO.describe`

Out[93]: <bound method NDFrame.describe of

	DATE	INDPRO
0	1919-01-01	4.8665
1	1919-02-01	4.6514
2	1919-03-01	4.5170
3	1919-04-01	4.5976
4	1919-05-01	4.6245
...
1256	2023-09-01	103.2096
1257	2023-10-01	102.3722
1258	2023-11-01	102.6710
1259	2023-12-01	102.6715
1260	2024-01-01	102.5739

[1261 rows x 2 columns]>

In [94]: `df_INDPRO.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1261 entries, 0 to 1260
Data columns (total 2 columns):
#   Column  Non-Null Count  Dtype
---  -
0   DATE    1261 non-null     object
1   INDPRO  1261 non-null     float64
dtypes: float64(1), object(1)
memory usage: 19.8+ KB
```

In [95]: `df_INDPRO['DATE'] = pd.to_datetime(df_INDPRO['DATE'])`
`df_INDPRO.corr()`

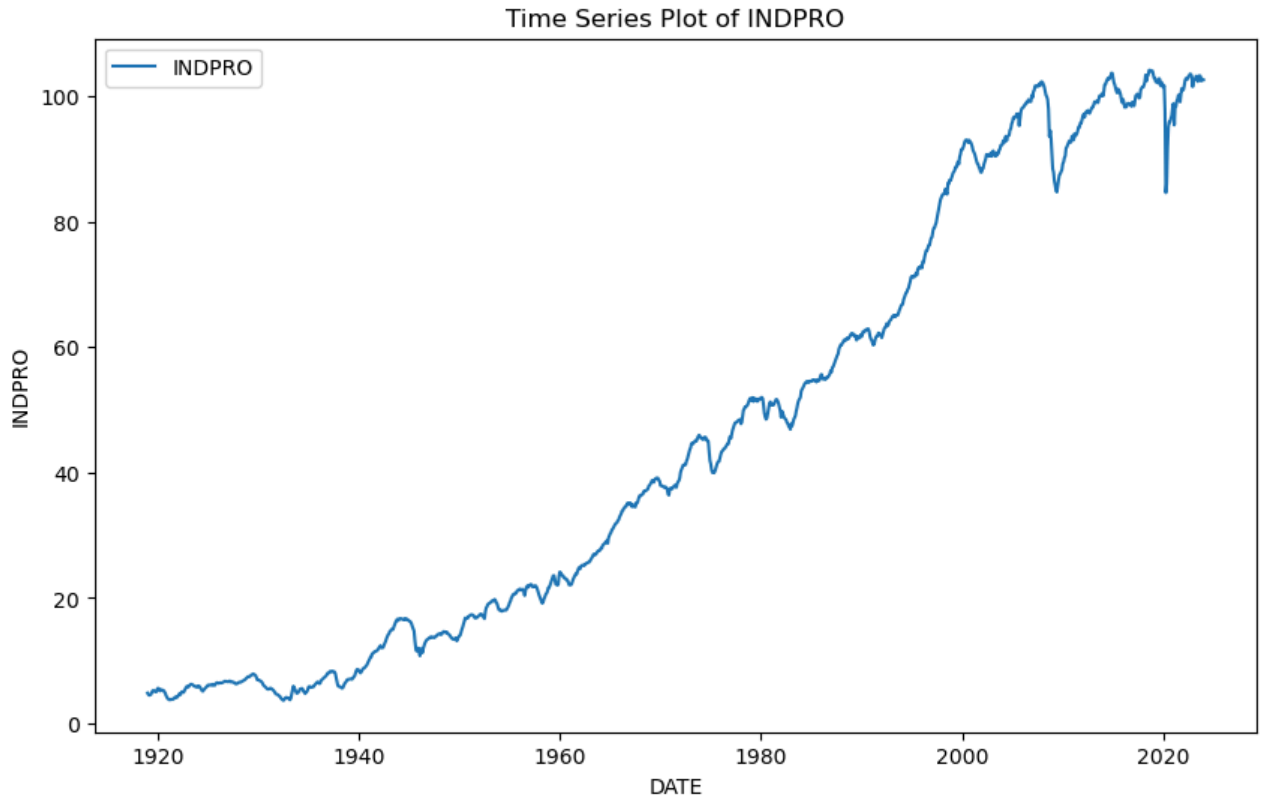
Out[95]:

	DATE	INDPRO
DATE	1.000000	0.974652
INDPRO	0.974652	1.000000



```
In [96]: plt.figure(figsize=(10, 6))
plt.plot(df_INDPRO['DATE'], df_INDPRO['INDPRO'], label='INDPRO')

plt.xlabel('DATE')
plt.ylabel('INDPRO')
plt.title('Time Series Plot of INDPRO')
plt.legend()
plt.show()
```



```
In [97]: df_month = df.resample('M', on='DATE').mean().reset_index()

df_month['DATE'] = df_month['DATE'] + pd.offsets.MonthEnd(0)
df_INDPRO['DATE'] = df_INDPRO['DATE'] + pd.offsets.MonthEnd(0)

merged_df = pd.merge(df_month, df_INDPRO, on='DATE', how='inner')
print(merged_df.head(10))
```

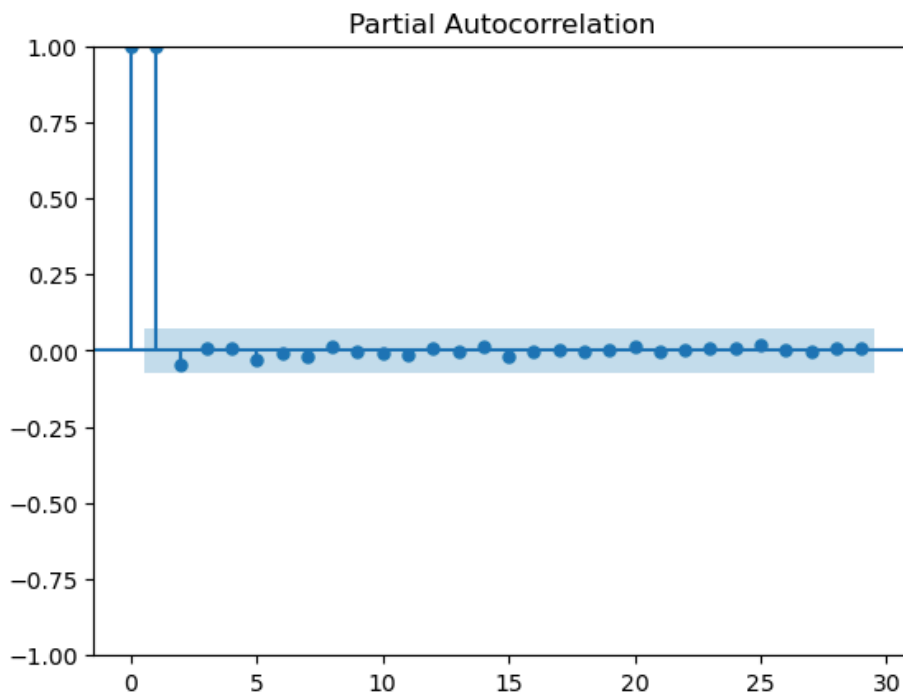
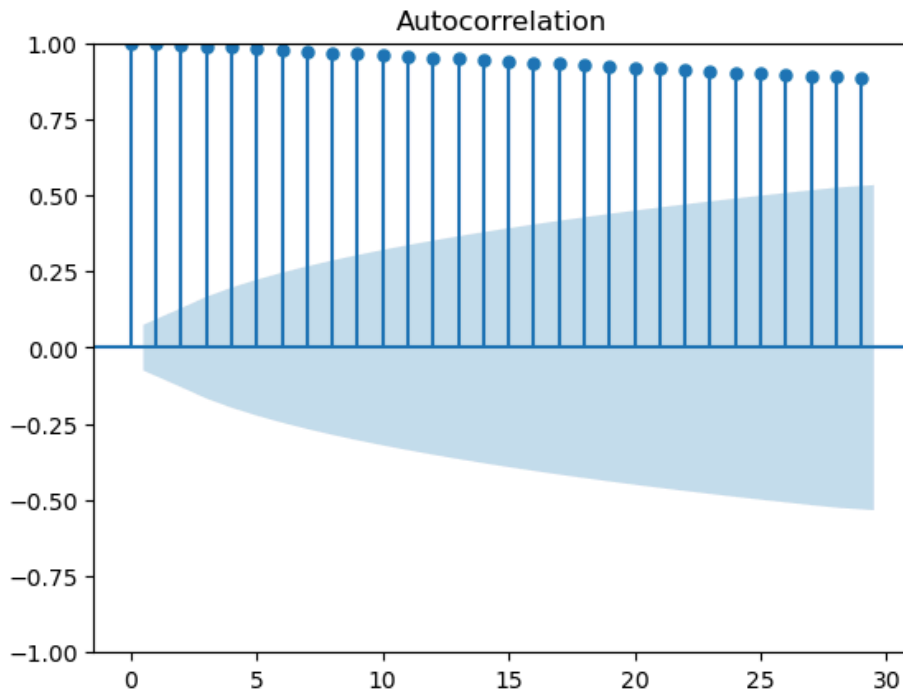
	DATE	ICNSA	ICNSA_imputed	INDPRO
0	1967-01-31	302250.0	112357.076327	35.1716
1	1967-02-28	261250.0	201711.651656	34.7727
2	1967-03-31	251500.0	219033.014210	34.5767
3	1967-04-30	239400.0	216012.687755	34.9028
4	1967-05-31	189000.0	172660.052122	34.5983
5	1967-06-30	179250.0	172408.894090	34.5940
6	1967-07-31	272400.0	258204.059244	34.5152
7	1967-08-31	194250.0	178532.132919	35.1765
8	1967-09-30	158600.0	164476.668589	35.1194
9	1967-10-31	177250.0	165251.438385	35.4054



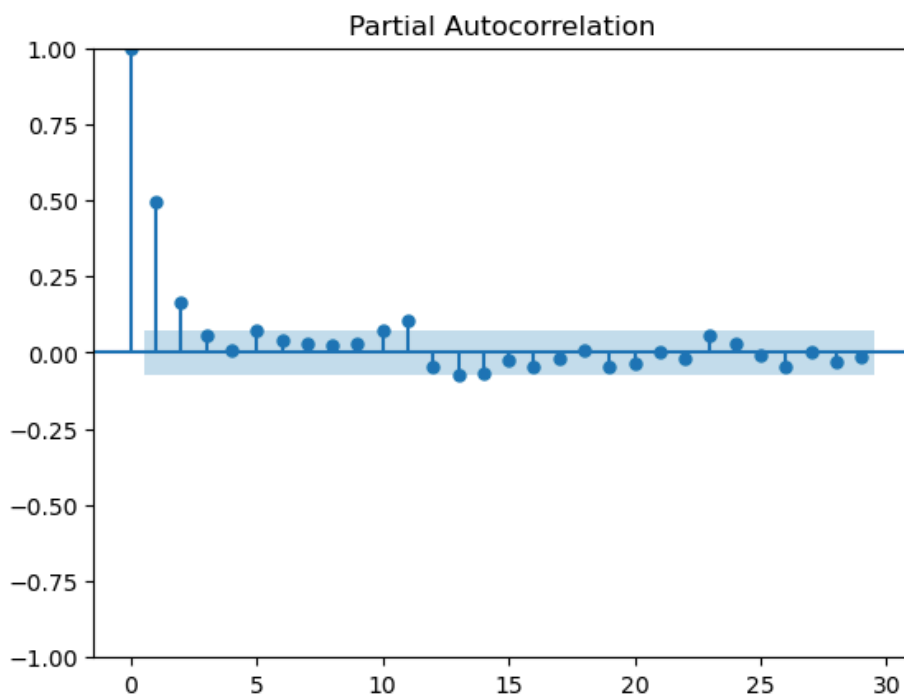
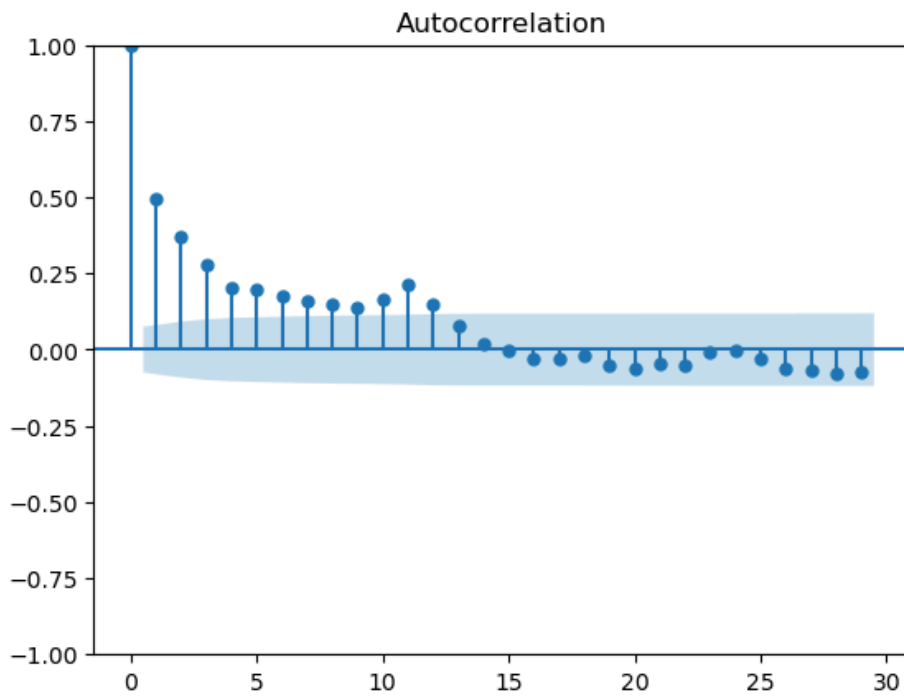

```
In [98]: from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
```

```
plot_acf(merged_df['INDPRO'])  
plt.show()
```

```
plot_pacf(merged_df['INDPRO'])  
plt.show()
```



```
In [99]: plot_acf(merged_df['ICNSA_imputed'])  
plt.show()  
  
plot_pacf(merged_df['ICNSA_imputed'])  
plt.show()
```



```
In [100]: from sklearn.model_selection import train_test_split

X = merged_df[['INDPRO']]
y = merged_df['ICNSA_imputed']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
In [101]: from pmdarima import auto_arima

arima_model = auto_arima(y_train, exogenous=X_train, seasonal=True, suppress_warnings=True)
print(arima_model.summary())
```

```

=====
SARIMAX Results
=====
Dep. Variable:          y      No. Observations:          548
Model:                SARIMAX  Log Likelihood             -7626.683
Date:                 Thu, 07 Mar 2024  AIC                   15257.365
Time:                 16:16:00    BIC                   15265.978
Sample:              0      HQIC                   15260.732
                    - 548
Covariance Type:      opg
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
intercept  3.695e+05  2.22e+04   16.661    0.000    3.26e+05  4.13e+05
sigma2     7.177e+10  7.38e+08   97.246    0.000    7.03e+10  7.32e+10
=====
Ljung-Box (L1) (Q):          0.73  Jarque-Bera (JB):          1529023.78
Prob(Q):                    0.39  Prob(JB):                   0.00
Heteroskedasticity (H):     5.61  Skew:                       13.79
Prob(H) (two-sided):        0.00  Kurtosis:                   260.30
=====

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



```
In [102]: forecast, conf_int = arima_model.predict(n_periods=len(y_test), exogenous=X_test, return_conf_int=True)
forecast_df = pd.DataFrame({
    'Date': y_test.index,
    'Actual': y_test.values,
    'Forecast': forecast,
    'Lower_CI': conf_int[:, 0],
    'Upper_CI': conf_int[:, 1]
})

print(forecast_df)
```

	Date	Actual	Forecast	Lower_CI	Upper_CI
548	294	441439.203202	369548.570945	-155524.032143	894621.174033
549	545	374631.644097	369548.570945	-155524.032143	894621.174033
550	213	361496.677692	369548.570945	-155524.032143	894621.174033
551	328	295644.387567	369548.570945	-155524.032143	894621.174033
552	164	385318.933394	369548.570945	-155524.032143	894621.174033
..
680	275	466455.617971	369548.570945	-155524.032143	894621.174033
681	110	359844.177967	369548.570945	-155524.032143	894621.174033
682	82	251707.670403	369548.570945	-155524.032143	894621.174033
683	51	265251.730035	369548.570945	-155524.032143	894621.174033
684	212	309999.419005	369548.570945	-155524.032143	894621.174033

[137 rows x 5 columns]

C:\ProgramData\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:836: ValueWarning: No supported index is available. Prediction results will be given with an integer index beginning at `start`.

```
return get_prediction_index()
```

C:\ProgramData\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:836: FutureWarning: No supported index is available. In the next version, calling this method in a model without a supported index will result in an exception.

```
return get_prediction_index()
```

```
In [103]: last_forecast_row = forecast_df.iloc[-1]

forecast_value = last_forecast_row['Forecast']
lower_ci = last_forecast_row['Lower_CI']
upper_ci = last_forecast_row['Upper_CI']

print(f"So I forecast the new data release value will be {forecast_value:.1f} with confidence interval")
```

So I forecast the new data release value will be 369548.6 with confidence interval (-155524.0, 894621.2)

