

```
In [18]: import pandas as pd
import matplotlib.pyplot as plt
from sklearn.metrics import mean_absolute_error, root_mean_squared_error, mean_a
from statsmodels.tsa.statespace.sarimax import SARIMAX
from statsmodels.tsa.stattools import grangercausalitytests
```

```
In [19]: df = pd.read_csv(r"C:\Users\HP\OneDrive\Desktop\NLP\Data\ML471_S4_Datafile_Conce

df['Datetime'] = pd.to_datetime(df['Datetime'])
df.set_index('Datetime', inplace=True)

y = df['Consumption']
exog = df['Festivals/Special_events']

train_size = int(len(y) * 0.8)
y_train, y_test = y[:train_size], y[train_size:]
exog_train, exog_test = exog[:train_size], exog[train_size:]

print("Granger Casualty Test Results:")

granger_data = df[['Consumption', 'Festivals/Special_events']]
grangercausalitytests(granger_data, maxlag=12, verbose=True)

model = SARIMAX(y_train, exog=exog_train, order=(1, 0, 2), seasonal_order=(0, 1,
model_fit = model.fit(disp=False)
print(model_fit.summary())

forecast = model_fit.predict(start=len(y_train), end=len(y_train)+len(y_test)-1,
mae = mean_absolute_error(y_test, forecast)
rmse = root_mean_squared_error(y_test, forecast)
mape = mean_absolute_percentage_error(y_test, forecast)

print(f"MAE: {mae:.4f}")
print(f"RMSE: {rmse:.4f}")
print(f"MAPE: {mape:.4f}")
```

Granger Causality Test Results:

Granger Causality
number of lags (no zero) 1
ssr based F test: F=6.3712 , p=0.0120 , df_denom=393, df_num=1
ssr based chi2 test: chi2=6.4199 , p=0.0113 , df=1
likelihood ratio test: chi2=6.3684 , p=0.0116 , df=1
parameter F test: F=6.3712 , p=0.0120 , df_denom=393, df_num=1

Granger Causality
number of lags (no zero) 2
ssr based F test: F=13.6644 , p=0.0000 , df_denom=390, df_num=2
ssr based chi2 test: chi2=27.6791 , p=0.0000 , df=2
likelihood ratio test: chi2=26.7524 , p=0.0000 , df=2
parameter F test: F=13.6644 , p=0.0000 , df_denom=390, df_num=2

Granger Causality
number of lags (no zero) 3
ssr based F test: F=3.1172 , p=0.0261 , df_denom=387, df_num=3
ssr based chi2 test: chi2=9.5208 , p=0.0231 , df=3
likelihood ratio test: chi2=9.4076 , p=0.0243 , df=3
parameter F test: F=3.1172 , p=0.0261 , df_denom=387, df_num=3

Granger Causality
number of lags (no zero) 4
ssr based F test: F=1.2370 , p=0.2946 , df_denom=384, df_num=4
ssr based chi2 test: chi2=5.0642 , p=0.2808 , df=4
likelihood ratio test: chi2=5.0318 , p=0.2840 , df=4
parameter F test: F=1.2370 , p=0.2946 , df_denom=384, df_num=4

Granger Causality
number of lags (no zero) 5
ssr based F test: F=1.0537 , p=0.3858 , df_denom=381, df_num=5
ssr based chi2 test: chi2=5.4205 , p=0.3667 , df=5
likelihood ratio test: chi2=5.3834 , p=0.3709 , df=5
parameter F test: F=1.0537 , p=0.3858 , df_denom=381, df_num=5

Granger Causality
number of lags (no zero) 6
ssr based F test: F=1.4520 , p=0.1937 , df_denom=378, df_num=6
ssr based chi2 test: chi2=9.0117 , p=0.1729 , df=6
likelihood ratio test: chi2=8.9094 , p=0.1787 , df=6
parameter F test: F=1.4520 , p=0.1937 , df_denom=378, df_num=6

Granger Causality
number of lags (no zero) 7
ssr based F test: F=1.1748 , p=0.3162 , df_denom=375, df_num=7
ssr based chi2 test: chi2=8.5526 , p=0.2864 , df=7
likelihood ratio test: chi2=8.4601 , p=0.2938 , df=7
parameter F test: F=1.1748 , p=0.3162 , df_denom=375, df_num=7

Granger Causality
number of lags (no zero) 8
ssr based F test: F=1.1357 , p=0.3382 , df_denom=372, df_num=8
ssr based chi2 test: chi2=9.5012 , p=0.3018 , df=8
likelihood ratio test: chi2=9.3870 , p=0.3107 , df=8
parameter F test: F=1.1357 , p=0.3382 , df_denom=372, df_num=8

Granger Causality
number of lags (no zero) 9

```
ssr based F test:      F=1.1146 , p=0.3512 , df_denom=369, df_num=9
ssr based chi2 test:  chi2=10.5481 , p=0.3080 , df=9
likelihood ratio test: chi2=10.4073 , p=0.3185 , df=9
parameter F test:      F=1.1146 , p=0.3512 , df_denom=369, df_num=9
```

Granger Causality

```
number of lags (no zero) 10
ssr based F test:      F=1.0533 , p=0.3980 , df_denom=366, df_num=10
ssr based chi2 test:  chi2=11.1372 , p=0.3469 , df=10
likelihood ratio test: chi2=10.9800 , p=0.3591 , df=10
parameter F test:      F=1.0533 , p=0.3980 , df_denom=366, df_num=10
```

Granger Causality

```
number of lags (no zero) 11
ssr based F test:      F=0.6228 , p=0.8094 , df_denom=363, df_num=11
ssr based chi2 test:  chi2=7.2852 , p=0.7755 , df=11
likelihood ratio test: chi2=7.2173 , p=0.7812 , df=11
parameter F test:      F=0.6228 , p=0.8094 , df_denom=363, df_num=11
```

Granger Causality

```
number of lags (no zero) 12
ssr based F test:      F=0.9273 , p=0.5195 , df_denom=360, df_num=12
ssr based chi2 test:  chi2=11.8998 , p=0.4538 , df=12
likelihood ratio test: chi2=11.7196 , p=0.4685 , df=12
parameter F test:      F=0.9273 , p=0.5195 , df_denom=360, df_num=12
```

```
c:\Users\HP\AppData\Local\Programs\Python\Python312\Lib\site-packages\statsmodels
\tsa\stattools.py:1556: FutureWarning: verbose is deprecated since functions shou
ld not print results
    warnings.warn(
c:\Users\HP\AppData\Local\Programs\Python\Python312\Lib\site-packages\statsmodels
\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provided,
so inferred frequency MS will be used.
    self._init_dates(dates, freq)
c:\Users\HP\AppData\Local\Programs\Python\Python312\Lib\site-packages\statsmodels
\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provided,
so inferred frequency MS will be used.
    self._init_dates(dates, freq)
```

SARIMAX Results

```
=====
=====
Dep. Variable: Consumption No. Observations: 846
Model: SARIMAX(1, 0, 2)x(0, 1, [1], 12) Log Likelihood -643.379
Date: Fri, 30 Jan 2026 AIC 1298.758
Time: 15:24:55 BIC 1320.777
Sample: 01-01-1988 HQIC 1307.580
                           - 05-01-2014
Covariance Type: opg
=====
=====
```

	coef	std err	z	P> z	[0.025
0.975]					
Festivals/Special_events	0.0871	0.060	1.454	0.146	-0.030
0.204					
ar.L1	0.9953	0.003	294.323	0.000	0.989
1.002					
ma.L1	-0.6029	0.069	-8.786	0.000	-0.737
-0.468					
ma.L2	-0.5299	0.057	-9.341	0.000	-0.641
-0.419					
ma.S.L12	-0.7262	0.047	-15.608	0.000	-0.817
-0.635					
sigma2	4.1056	0.315	13.036	0.000	3.488
4.723					

```
=====
==
```

Ljung-Box (L1) (Q):	0.52	Jarque-Bera (JB):	36.
37			
Prob(Q):	0.47	Prob(JB):	0.
00			
Heteroskedasticity (H):	2.82	Skew:	-0.
23			
Prob(H) (two-sided):	0.00	Kurtosis:	4.
67			

```
=====
==
```

Warnings:

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

MAE: 2.8428

RMSE: 3.6325

MAPE: 0.0205

```
In [22]: plt.figure(figsize=(12,6))
plt.plot(y_train.index, y_train, label='Train', color='blue')
plt.plot(y_test.index, y_test, label='Actual', color='yellow', linestyle='--')
plt.plot(y_test.index, forecast, label='Forecast', color='green', linestyle='--')
plt.legend()
plt.show()
```

