

```
In [51]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from statsmodels.tsa.arima_model import ARIMA
from sklearn.metrics import mean_squared_error
import seaborn as sns
import warnings
from joblib import Parallel, delayed
import seaborn as sns

# Set the style for the plots
sns.set(style="whitegrid")

warnings.filterwarnings('ignore')
```

```
In [5]: train_df = pd.read_csv("C:/Users/dhanu/OneDrive/train.csv")
test_df = pd.read_csv("C:/Users/dhanu/OneDrive/test.csv")

# Display the first few rows of the datasets
train_df.head(), test_df.head()
```

```

Out[5]: (
      ID      date      Item Id \
0  2022-04-12_B09KDTS4DC  2022-04-12  B09KDTS4DC
1  2022-04-12_B09MR2MLZH  2022-04-12  B09MR2MLZH
2  2022-04-12_B09KSYL73R  2022-04-12  B09KSYL73R
3  2022-04-12_B09KT5HMNY  2022-04-12  B09KT5HMNY
4  2022-04-12_B09KTF8ZDQ  2022-04-12  B09KTF8ZDQ

      Item Name      ad_spend      anarix_id
\
0  NapQueen Elizabeth 8" Gel Memory Foam Mattress...      NaN      NAPQUEEN
1  NapQueen 12 Inch Bamboo Charcoal Queen Size Me...      NaN      NAPQUEEN
2  NapQueen Elsa 8" Innerspring Mattress, Twin XL      NaN      NAPQUEEN
3  NapQueen Elsa 6" Innerspring Mattress, Twin      NaN      NAPQUEEN
4  NapQueen Elsa 6" Innerspring Mattress, Twin XL      NaN      NAPQUEEN

      units      unit_price
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 ,
      ID      date      Item Id \
0  2024-07-01_B09KDR64LT  2024-07-01  B09KDR64LT
1  2024-07-01_B09KDTS4DC  2024-07-01  B09KDTS4DC
2  2024-07-01_B09KDTHJ6V  2024-07-01  B09KDTHJ6V
3  2024-07-01_B09KDQ2BWY  2024-07-01  B09KDQ2BWY
4  2024-07-01_B09KDYY3SB  2024-07-01  B09KDYY3SB

      Item Name      ad_spend      anarix_id
\
0  NapQueen Elizabeth 10" Gel Memory Foam Mattres...      NaN      NAPQUEEN
1  NapQueen Elizabeth 8" Gel Memory Foam Mattress...      NaN      NAPQUEEN
2  NapQueen Elizabeth 12" Gel Memory Foam Mattres...      NaN      NAPQUEEN
3  NapQueen Elizabeth 12" Gel Memory Foam Mattres...      NaN      NAPQUEEN
4  NapQueen Elizabeth 10" Gel Memory Foam Mattres...      101.72      NAPQUEEN

      unit_price
0      0.0
1      0.0
2      0.0
3      0.0
4      1094.5 )

```

```
In [7]: #EDA-Exploratory Data Analysis

# Checking for missing values in the training dataset
missing_values_train = train_df.isnull().sum()

# Summary statistics of the training dataset
summary_statistics_train = train_df.describe(include='all')

# Checking the data types of each column in the training dataset
data_types_train = train_df.dtypes

print(missing_values_train)
print(summary_statistics_train)
print(data_types_train)
```

```
ID          0
date        0
Item Id     2
Item Name   1832
ad_spend    24187
anarix_id   0
units       17898
unit_price  0
dtype: int64
```

```
              ID      date      Item Id  \
count          101490      101490      101488
unique          101490          781          217
top    2022-04-12_B09KDTS4DC  2023-03-09  B09MR2FY5D
freq              1          189          781
mean            NaN          NaN          NaN
std             NaN          NaN          NaN
min             NaN          NaN          NaN
25%             NaN          NaN          NaN
50%             NaN          NaN          NaN
75%             NaN          NaN          NaN
max             NaN          NaN          NaN
```

```
              Item Name      ad_spend
\
count          99658  77303.000000
unique          199          NaN
top    NapQueen Margaret 8" Charcoal Memory Foam Matt...      NaN
freq              781          NaN
mean            NaN    110.771470
std             NaN    529.303777
min             NaN     0.000000
25%             NaN     0.000000
50%             NaN     4.230000
75%             NaN    44.310000
max             NaN   47934.990000
```

```
      anarix_id      units      unit_price
count    101490  83592.000000  101490.000000
unique         1          NaN          NaN
top      NAPQUEEN          NaN          NaN
freq    101490          NaN          NaN
mean         NaN    10.284381    106.750922
std          NaN    68.945915    425.704733
min          NaN   -173.000000   -8232.000000
25%          NaN     0.000000     0.000000
50%          NaN     1.000000     0.000000
75%          NaN     5.000000     0.000000
max          NaN   9004.000000   21557.390000
```

```
ID          object
date        object
Item Id     object
Item Name   object
ad_spend    float64
anarix_id   object
units       float64
unit_price  float64
dtype: object
```

```
In [9]: # Handle missing values in 'units' by filling with 0
train_df['units'].fillna(0, inplace=True)

# Replace negative values in 'units' with NaN and then fill with 0
train_df.loc[train_df['units'] < 0, 'units'] = None
train_df['units'].fillna(0, inplace=True)

# Replace negative values in 'unit_price' with absolute values
train_df['unit_price'] = train_df['unit_price'].abs()

# Convert 'date' column to datetime format and create new features
train_df['date'] = pd.to_datetime(train_df['date'])
train_df['year'] = train_df['date'].dt.year
train_df['month'] = train_df['date'].dt.month
train_df['day'] = train_df['date'].dt.day
train_df['day_of_week'] = train_df['date'].dt.dayofweek

# Convert 'date' column to datetime format and create new features in the
test_df['date'] = pd.to_datetime(test_df['date'])
test_df['year'] = test_df['date'].dt.year
test_df['month'] = test_df['date'].dt.month
test_df['day'] = test_df['date'].dt.day
test_df['day_of_week'] = test_df['date'].dt.dayofweek

print(test_df.head())
print(train_df.head())
```

	ID	date	Item Id	\
0	2024-07-01_B09KDR64LT	2024-07-01	B09KDR64LT	
1	2024-07-01_B09KDTS4DC	2024-07-01	B09KDTS4DC	
2	2024-07-01_B09KDTJH6V	2024-07-01	B09KDTJH6V	
3	2024-07-01_B09KDQ2BWY	2024-07-01	B09KDQ2BWY	
4	2024-07-01_B09KDYY3SB	2024-07-01	B09KDYY3SB	

	Item Name	ad_spend	anarix_id
0	NapQueen Elizabeth 10" Gel Memory Foam Mattres...	NaN	NAPQUEEN
1	NapQueen Elizabeth 8" Gel Memory Foam Mattres...	NaN	NAPQUEEN
2	NapQueen Elizabeth 12" Gel Memory Foam Mattres...	NaN	NAPQUEEN
3	NapQueen Elizabeth 12" Gel Memory Foam Mattres...	NaN	NAPQUEEN
4	NapQueen Elizabeth 10" Gel Memory Foam Mattres...	101.72	NAPQUEEN

	unit_price	year	month	day	day_of_week
0	0.0	2024	7	1	0
1	0.0	2024	7	1	0
2	0.0	2024	7	1	0
3	0.0	2024	7	1	0
4	1094.5	2024	7	1	0

	ID	date	Item Id	\
0	2022-04-12_B09KDTS4DC	2022-04-12	B09KDTS4DC	
1	2022-04-12_B09MR2MLZH	2022-04-12	B09MR2MLZH	
2	2022-04-12_B09KSYL73R	2022-04-12	B09KSYL73R	
3	2022-04-12_B09KT5HMNY	2022-04-12	B09KT5HMNY	
4	2022-04-12_B09KTF8ZDQ	2022-04-12	B09KTF8ZDQ	

	Item Name	ad_spend	anarix_id
0	NapQueen Elizabeth 8" Gel Memory Foam Mattress...	NaN	NAPQUEEN
1	NapQueen 12 Inch Bamboo Charcoal Queen Size Me...	NaN	NAPQUEEN
2	NapQueen Elsa 8" Innerspring Mattress, Twin XL	NaN	NAPQUEEN
3	NapQueen Elsa 6" Innerspring Mattress, Twin	NaN	NAPQUEEN
4	NapQueen Elsa 6" Innerspring Mattress, Twin XL	NaN	NAPQUEEN

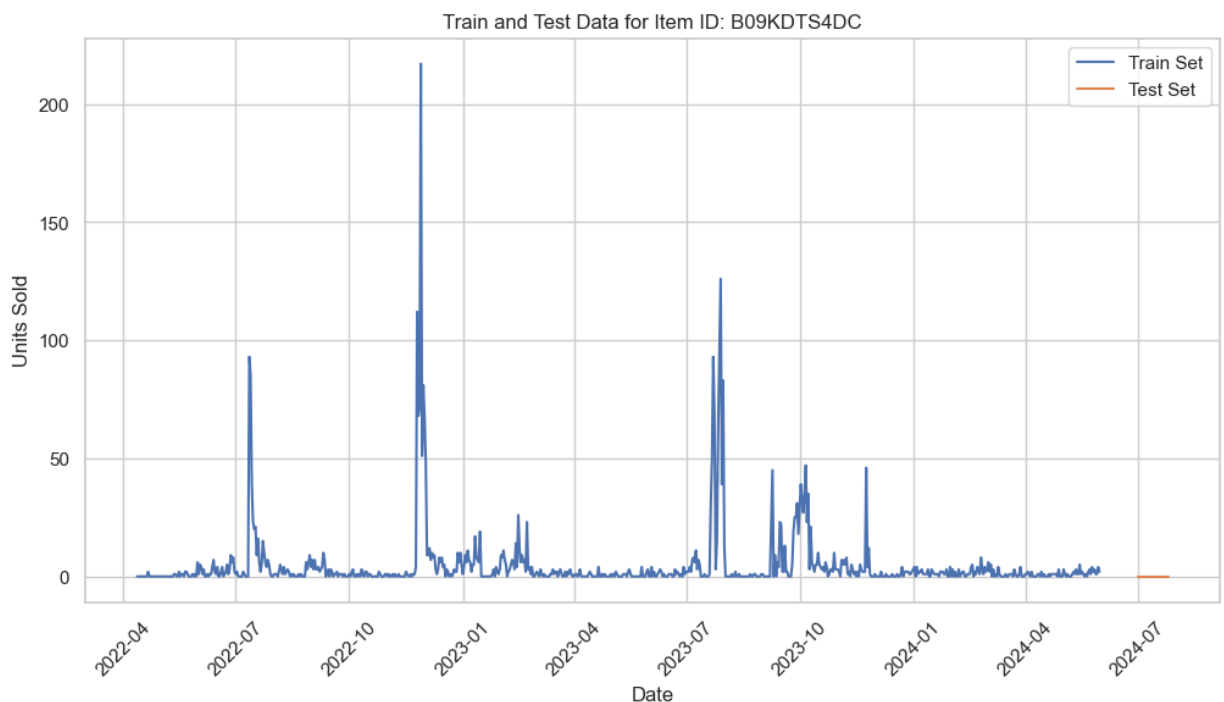
	units	unit_price	year	month	day	day_of_week
0	0.0	0.0	2022	4	12	1
1	0.0	0.0	2022	4	12	1
2	0.0	0.0	2022	4	12	1
3	0.0	0.0	2022	4	12	1
4	0.0	0.0	2022	4	12	1

```
In [11]: item_id = 'B09KDTS4DC' # Replace with the Item Id you want to visualize

# Filter the data for the selected Item Id
item_train_df = train_df[train_df['Item Id'] == item_id]
item_test_df = test_df[test_df['Item Id'] == item_id]
```

```
In [13]: # Ensure that the date columns are in datetime format
item_train_df['date'] = pd.to_datetime(item_train_df['date'])
item_test_df['date'] = pd.to_datetime(item_test_df['date'])

# Plotting train and test data separately
plt.figure(figsize=(12, 6))
plt.plot(item_train_df['date'], item_train_df['units'], label='Train Set')
plt.plot(item_test_df['date'], item_test_df['unit_price'], label='Test Set')
plt.title(f'Train and Test Data for Item ID: {item_id}')
plt.xlabel('Date')
plt.ylabel('Units Sold')
plt.legend()
plt.xticks(rotation=45)
plt.show()
```



```
In [14]: # Group data by 'Item Id' and 'date' for time series analysis
item_sales = train_df.groupby(['date', 'Item Id']).agg({'units': 'sum'})

# Pivot table to create a time series for each item
time_series_data = item_sales.pivot(index='date', columns='Item Id', values='units')

# Initialize dictionary to store ARIMA predictions
arima_predictions = {}
```

```
In [17]: # Convert 'date' to datetime if it's not already
train_df['date'] = pd.to_datetime(train_df['date'])
test_df['date'] = pd.to_datetime(test_df['date'])

# Extract date features
train_df['year'] = train_df['date'].dt.year
train_df['month'] = train_df['date'].dt.month
train_df['day'] = train_df['date'].dt.day
train_df['day_of_week'] = train_df['date'].dt.dayofweek

test_df['year'] = test_df['date'].dt.year
test_df['month'] = test_df['date'].dt.month
test_df['day'] = test_df['date'].dt.day
test_df['day_of_week'] = test_df['date'].dt.dayofweek
```

```
In [19]: # Create lag features for the 'units' column
lag_days = [1, 7, 14] # Example lags: 1 day, 1 week, 2 weeks

for lag in lag_days:
    train_df[f'units_lag_{lag}'] = train_df.groupby('Item Id')['units'].shift(lag)
    test_df[f'units_lag_{lag}'] = test_df.groupby('Item Id')['unit_price'].shift(lag)
```

```
In [21]: # Create rolling window features for the 'units' column
window_sizes = [3, 7, 14] # Example windows: 3 days, 1 week, 2 weeks

for window in window_sizes:
    train_df[f'units_roll_mean_{window}'] = train_df.groupby('Item Id')['units'].rolling(window=window).mean()
    test_df[f'units_roll_mean_{window}'] = test_df.groupby('Item Id')['unit_price'].rolling(window=window).mean()
```

```
In [23]: # Interaction between ad spend and units
train_df['ad_spend_units'] = train_df['ad_spend'] * train_df['units']
test_df['ad_spend_units'] = test_df['ad_spend'] * test_df['unit_price']
```

```
In [25]: # Fill NaN values after creating lag and rolling features
train_df.fillna(0, inplace=True)
test_df.fillna(0, inplace=True)
```

```
In [27]: # Display the first few rows to check the new features
print(train_df.head())
print(test_df.head())
```

	ID	date	Item Id \		
0	2022-04-12_B09KDTS4DC	2022-04-12	B09KDTS4DC		
1	2022-04-12_B09MR2MLZH	2022-04-12	B09MR2MLZH		
2	2022-04-12_B09KSYL73R	2022-04-12	B09KSYL73R		
3	2022-04-12_B09KT5HMNY	2022-04-12	B09KT5HMNY		
4	2022-04-12_B09KTF8ZDQ	2022-04-12	B09KTF8ZDQ		

	Item Name	ad_spend	anarix_id
0	NapQueen Elizabeth 8" Gel Memory Foam Mattress...	0.0	NAPQUEEN
1	NapQueen 12 Inch Bamboo Charcoal Queen Size Me...	0.0	NAPQUEEN
2	NapQueen Elsa 8" Innerspring Mattress, Twin XL	0.0	NAPQUEEN
3	NapQueen Elsa 6" Innerspring Mattress, Twin	0.0	NAPQUEEN



```

4      NapQueen Elsa 6" Innerspring Mattress, Twin XL      0.0  NAPQUEEN

      units  unit_price  year  month  day  day_of_week  units_lag_1  units_l
ag_7  \
0      0.0          0.0  2022      4   12              1          0.0
0.0
1      0.0          0.0  2022      4   12              1          0.0
0.0
2      0.0          0.0  2022      4   12              1          0.0
0.0
3      0.0          0.0  2022      4   12              1          0.0
0.0
4      0.0          0.0  2022      4   12              1          0.0
0.0

      units_lag_14  units_roll_mean_3  units_roll_mean_7  units_roll_mean_14
\
0              0.0              0.0              0.0              0.0
1              0.0              0.0              0.0              0.0
2              0.0              0.0              0.0              0.0
3              0.0              0.0              0.0              0.0
4              0.0              0.0              0.0              0.0

      ad_spend_units
0              0.0
1              0.0
2              0.0
3              0.0
4              0.0

      ID      date      Item Id  \
0  2024-07-01_B09KDR64LT  2024-07-01  B09KDR64LT
1  2024-07-01_B09KDTS4DC  2024-07-01  B09KDTS4DC
2  2024-07-01_B09KDTHJ6V  2024-07-01  B09KDTHJ6V
3  2024-07-01_B09KDQ2BWY  2024-07-01  B09KDQ2BWY
4  2024-07-01_B09KDYY3SB  2024-07-01  B09KDYY3SB

      Item Name  ad_spend  anarix_id
\
0  NapQueen Elizabeth 10" Gel Memory Foam Mattres...      0.00  NAPQUEEN
1  NapQueen Elizabeth 8" Gel Memory Foam Mattress...      0.00  NAPQUEEN
2  NapQueen Elizabeth 12" Gel Memory Foam Mattres...      0.00  NAPQUEEN
3  NapQueen Elizabeth 12" Gel Memory Foam Mattres...      0.00  NAPQUEEN
4  NapQueen Elizabeth 10" Gel Memory Foam Mattres...     101.72  NAPQUEEN

      unit_price  year  month  day  day_of_week  units_lag_1  units_lag_7  \
0              0.0  2024      7   1              0          0.0      0.0
1              0.0  2024      7   1              0          0.0      0.0
2              0.0  2024      7   1              0          0.0      0.0
3              0.0  2024      7   1              0          0.0      0.0
4             1094.5  2024      7   1              0          0.0      0.0

      units_lag_14  units_roll_mean_3  units_roll_mean_7  units_roll_mean_14
\
0              0.0              0.0              0.0              0.0
1              0.0              0.0              0.0              0.0
2              0.0              0.0              0.0              0.0
3              0.0              0.0              0.0              0.0

```

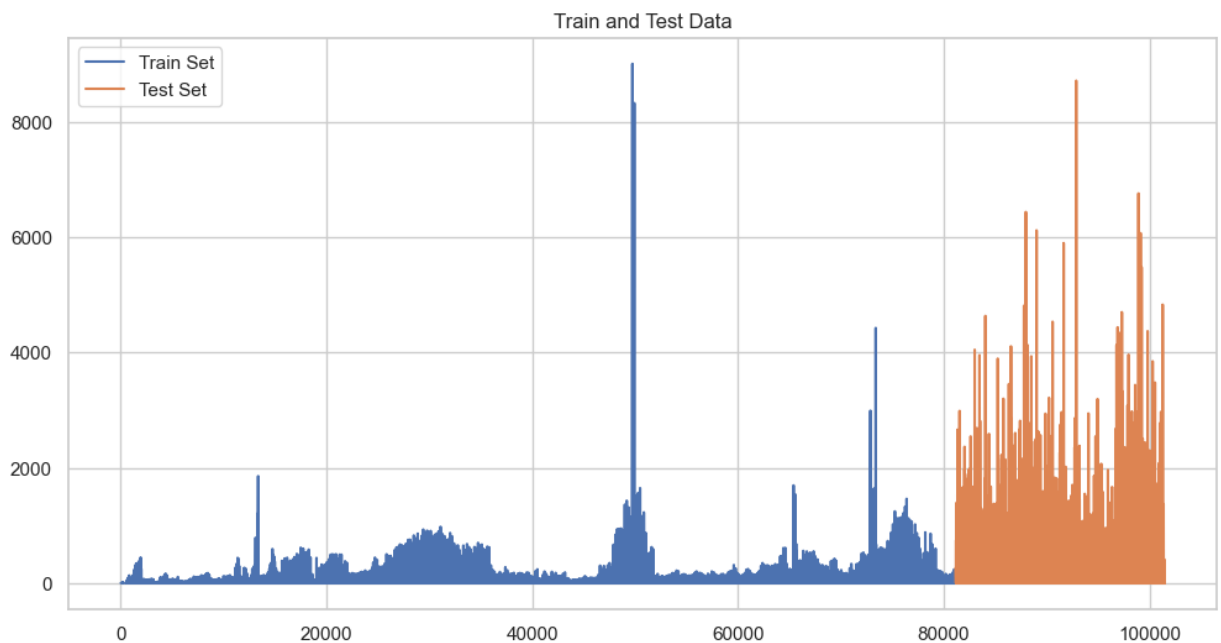
4	0.0	0.0	0.0	0.0
---	-----	-----	-----	-----

	ad_spend_units
0	0.00
1	0.00
2	0.00
3	0.00
4	111332.54

```
In [29]: # Assuming `data` is already loaded and combined as shown in the previous
train_size = int(len(train_df) * 0.8)
train, test = train_df.iloc[:train_size], train_df.iloc[train_size:]

# Plot the train and test sets
plt.figure(figsize=(12, 6))
plt.plot(train.index, train['units'], label='Train Set')
plt.plot(test.index, test['unit_price'], label='Test Set')
plt.legend()
plt.title('Train and Test Data')
plt.show()
```



```
In [37]: def fit_arima_model(item_id, item_series):
# Ensure that the series has enough data points
if len(item_series) < 30: # Adjust the threshold as necessary
    print(f'Item ID: {item_id} has too few data points, skipping...')
    return item_id, None, None

# Split the data into training and testing sets (80-20 split)
train_size = int(len(item_series) * 0.8)
train_data, test_data = item_series[:train_size], item_series[train_s

# Automatically select the best ARIMA model using auto_arima
try:
    model = auto_arima(train_data, seasonal=False, stepwise=True, sup
                        error_action='ignore', trace=False)
    model_fit = model.fit(train_data)

    # Forecast the future values
    forecast = model_fit.predict(n_periods=len(test_data))

    # Calculate MSE for the forecast
    mse = mean_squared_error(test_data, forecast)
    print(f'Item ID: {item_id}, Best ARIMA Order: {model.order}, MSE:

    return item_id, forecast, mse

except Exception as e:
    print(f'Item ID: {item_id} encountered an error: {e}')
    return item_id, None, None
```

```

In [41]: from statsmodels.tsa.arima.model import ARIMA
from sklearn.metrics import mean_squared_error
import numpy as np

# Select the target column
target_column = 'unit_price'
train_data = train_df.set_index('date')[target_column]
test_data = test_df.set_index('date')[target_column]

# Split the data (assuming 'date' is the index)
train_size = int(len(train_data) * 0.8)
train_set, test_set = train_data[:train_size], train_data[train_size:]

# Define the ARIMA model parameters (p, d, q) - replace with your best pa
p, d, q = 1, 1, 1 # Example values, adjust according to your model selec
arima_model = ARIMA(train_set, order=(p, d, q))

# Fit the ARIMA model
arima_model_fit = arima_model.fit()

# Make predictions on the test set
forecast_steps = len(test_set)
forecast = arima_model_fit.forecast(steps=forecast_steps)

# Create a DataFrame for the forecast
forecast_index = test_set.index
forecast_df = pd.DataFrame(data=forecast, index=forecast_index, columns=[

# Plot actual vs predicted values
plt.figure(figsize=(12, 6))
plt.plot(train_set.index, train_set, label='Train Set')
plt.plot(test_set.index, test_set, label='Test Set')
plt.plot(forecast_df.index, forecast_df['Forecast'], label='Forecast', li
plt.legend()
plt.title('ARIMA Model Forecast')
plt.xlabel('Date')
plt.ylabel('Units Sold')
plt.xticks(rotation=45)
plt.show()

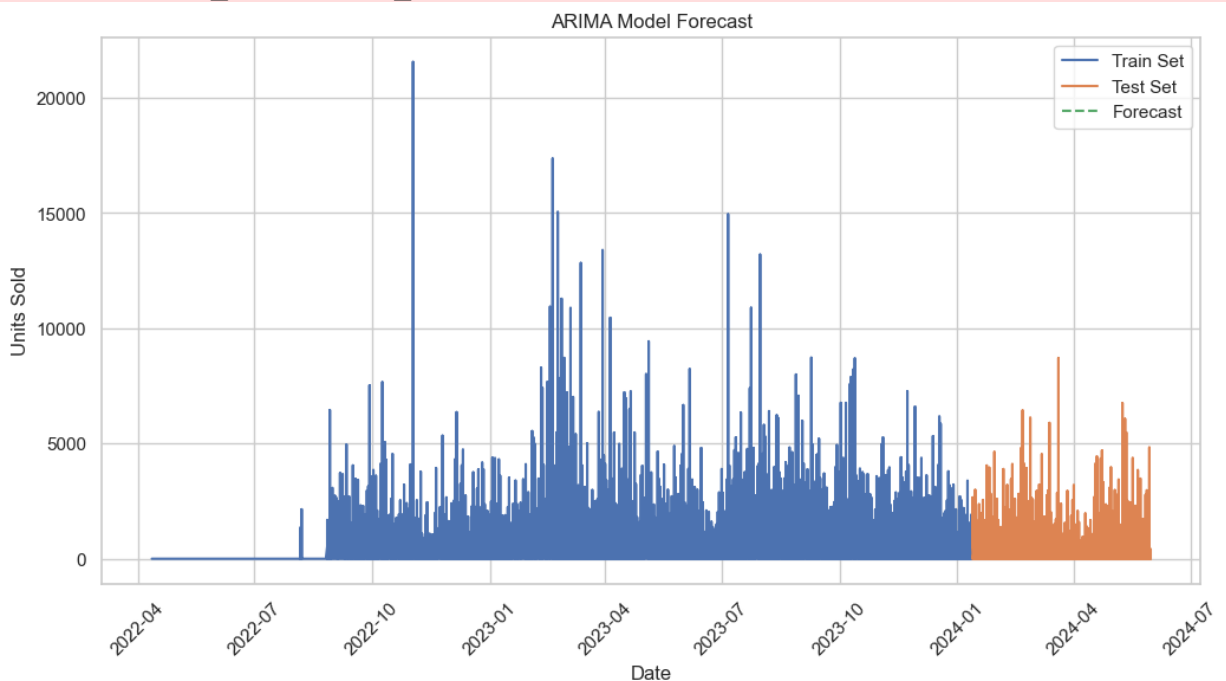
# Calculate and print the Mean Squared Error (MSE) of the forecast
mse = mean_squared_error(test_set, forecast)
print(f'Mean Squared Error: {mse:.2f}')

```

```

C:\Users\dhanu\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: A date index has been provided, but it has no associated frequency information and so will be ignored when e.g. forecasting.
    self._init_dates(dates, freq)
C:\Users\dhanu\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: A date index has been provided, but it has no associated frequency information and so will be ignored when e.g. forecasting.
    self._init_dates(dates, freq)
C:\Users\dhanu\anaconda3\Lib\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: A date index has been provided, but it has no associated frequency information and so will be ignored when e.g. forecasting.
    self._init_dates(dates, freq)
C:\Users\dhanu\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(

```



Mean Squared Error: 111288.14

In [ ]: