

```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("/content/ELB-Sales-Data.csv", parse_dates=['Date'])

# Step 1: Preprocessing and basic exploration
df['Date'] = pd.to_datetime(df['Date'])
df.set_index('Date', inplace=True)

# Step 2: Handling missing values
# Check for missing values in the dataset
print(df.isnull().sum())

# Step 3: Descriptive statistics (mean, median, etc.)
print(df.describe())
```

```

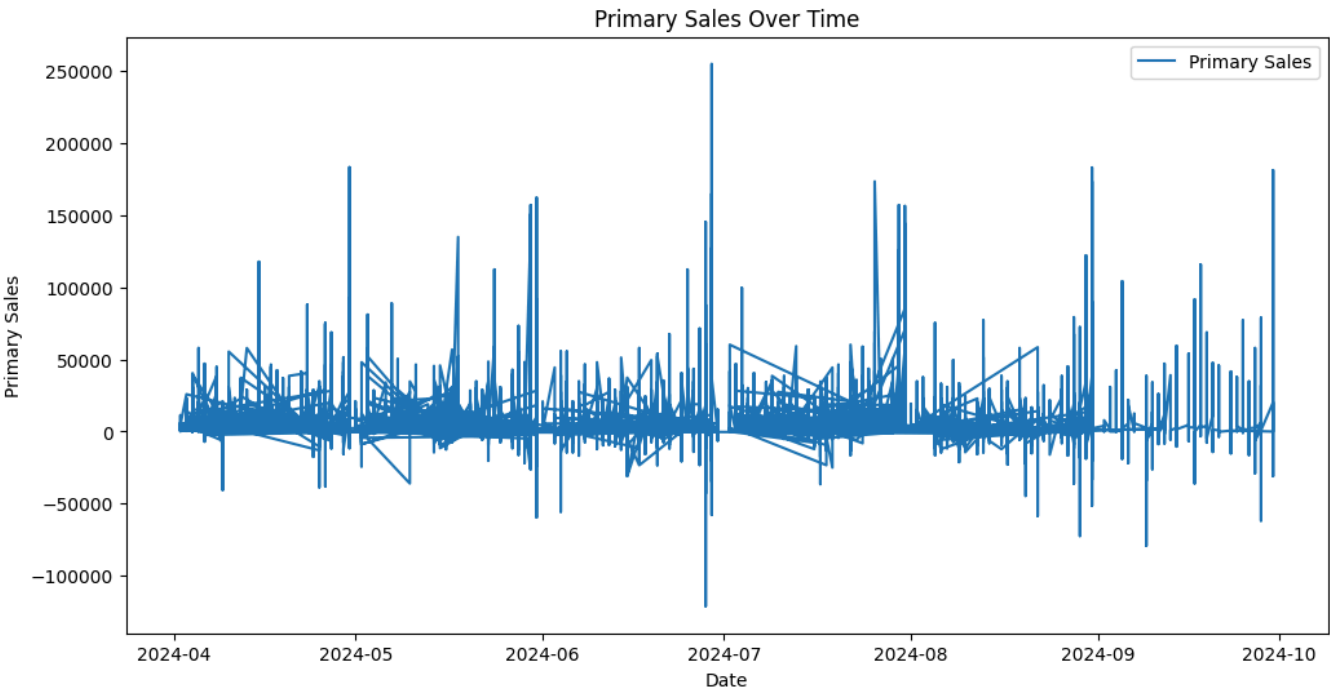
Sr      0
ID      0
Customer      0
Primary Sales      0
Cancelled      0
Is added to primary log      0
Against Expiry      0
Breakage      0
Claim Offer      0
Quantity      0
Net Rate      0
Free Item      0
HQ      180
Sales Team      180
Return for Reason      89300
Sales Return      0
Sales Invoice Name      2
Rate Difference      0
Item Name      0
Free Item Value      0
dtype: int64

Sr      Primary Sales      Cancelled      Is added to primary log \
count  107444.000000  107444.000000  107444.000000  107444.000000
mean    81398.343044    2514.400220     0.018726     0.998325
std     31018.861523    6077.772576     0.135556     0.040896
min     9354.000000   -120939.060000     0.000000     0.000000
25%     54538.750000     0.000000     0.000000     1.000000
50%     81399.500000    1157.200000     0.000000     1.000000
75%    108260.250000    3085.800000     0.000000     1.000000
max    135121.000000  254574.000000     1.000000     1.000000

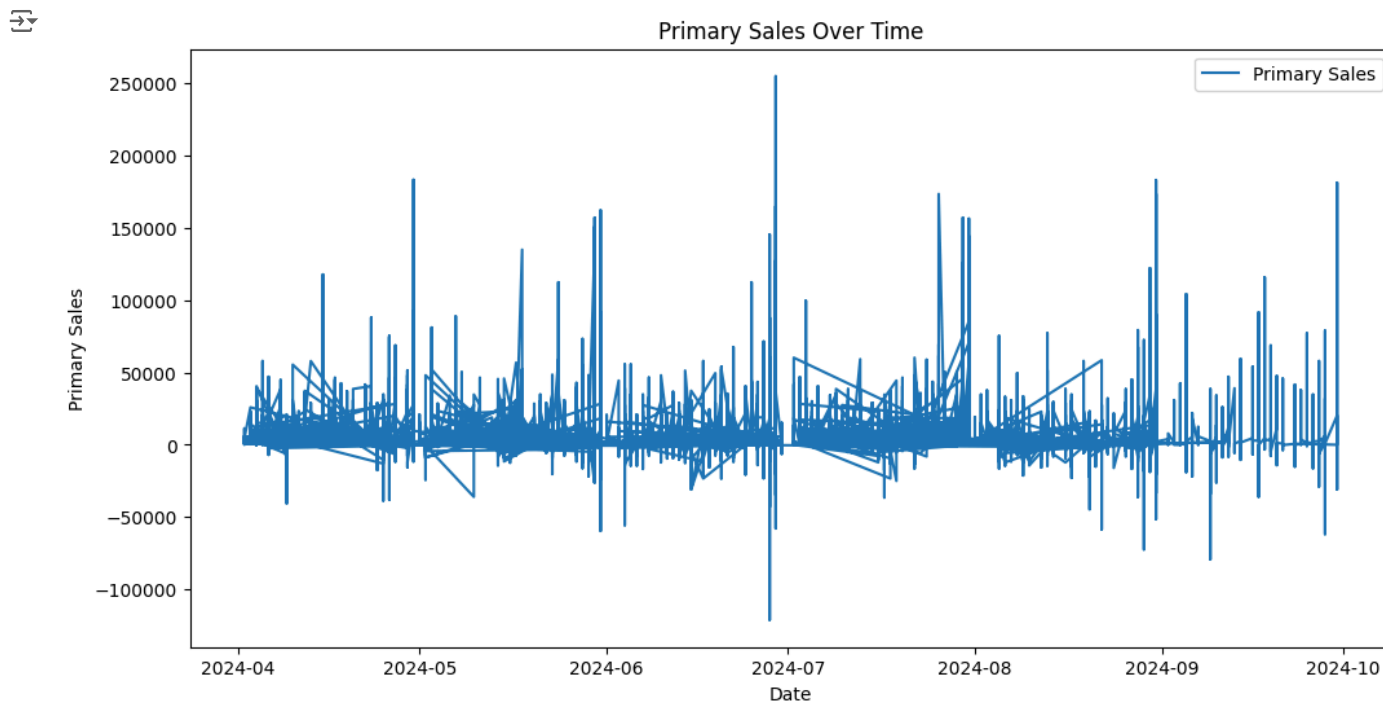
Against Expiry      Breakage      Claim Offer      Quantity \
count  107444.000000  107444.000000  107444.000000  107444.000000
mean   -66.504029    -9.788137    -34.735094    26.366619
std     535.978101    299.959327    315.690782    54.720325
min   -44550.000000   -61088.000000   -24940.285710   -490.000000
25%      0.000000      0.000000      0.000000      3.000000
50%      0.000000      0.000000      0.000000     10.000000
75%      0.000000      0.000000      0.000000     30.000000
max     21901.490000     8100.000000     7675.714286    2000.000000

Net Rate      Free Item      Sales Return      Rate Difference \
count  107444.000000  107444.000000  107444.000000  107444.000000
mean     88.986826      0.054252    -60.959267    -13.028095
std      76.230676      0.226514     983.102573     200.668543
min    -125.360000      0.000000   -118743.300000   -17607.700000
25%      51.430000      0.000000      0.000000      0.000000
50%      86.790000      0.000000      0.000000      0.000000
75%     122.030000      0.000000      0.000000      0.000000
max     1684.730000      1.000000     5785.800000     1225.000000

Free Item Value
count  107444.000000
mean   -47.930296
std     428.920348
min   -38507.142860
25%      0.000000
50%      0.000000
75%      0.000000
max     6171.428571
```



```
# Step 4: Basic analysis of primary sales over time
plt.figure(figsize=(12,6))
plt.plot(df['Primary Sales'], label='Primary Sales')
plt.title('Primary Sales Over Time')
plt.xlabel('Date')
plt.ylabel('Primary Sales')
plt.legend()
plt.show()
```



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# Step 5: Check for correlations between columns, if needed
# Step 5: Check for correlations between numeric columns
# Select only numeric columns
numeric_df = df.select_dtypes(include=['number'])

# Compute the correlation matrix
correlation_matrix = numeric_df.corr()

# Print the correlation matrix
print(correlation_matrix)
```

```
Sr          1.000000      0.022398      0.016671
Primary Sales 0.022398      1.000000     -0.127082
Cancelled     0.016671     -0.127082      1.000000
Is added to primary log 0.003917      0.007564      0.005659
Against Expiry 0.003879      0.140877      0.054284
Breakage     -0.014025      0.025809      0.012673
Claim Offer   0.071049      0.045520      0.030296
Quantity      0.017140      0.823273      0.012064
Net Rate     -0.009896      0.252527      0.023944
Free Item     0.018841     -0.099085      0.019049
Sales Return  0.026632      0.133903      0.012378
Rate Difference 0.001803      0.056880      0.011003
Free Item Value -0.000917      0.046230      0.040957

Is added to primary log Against Expiry Breakage \
Sr          0.003917      0.003879     -0.014025
Primary Sales 0.007564      0.140877      0.025809
Cancelled     0.005659      0.054284      0.012673
Is added to primary log 1.000000      0.005041     -0.000897
Against Expiry 0.005041      1.000000     -0.004049
Breakage     -0.000897     -0.004049      1.000000
Claim Offer   -0.004507     -0.013652     -0.003590
Quantity      0.009441      0.113441      0.041105
Net Rate      0.005188     -0.099676     -0.008352
```

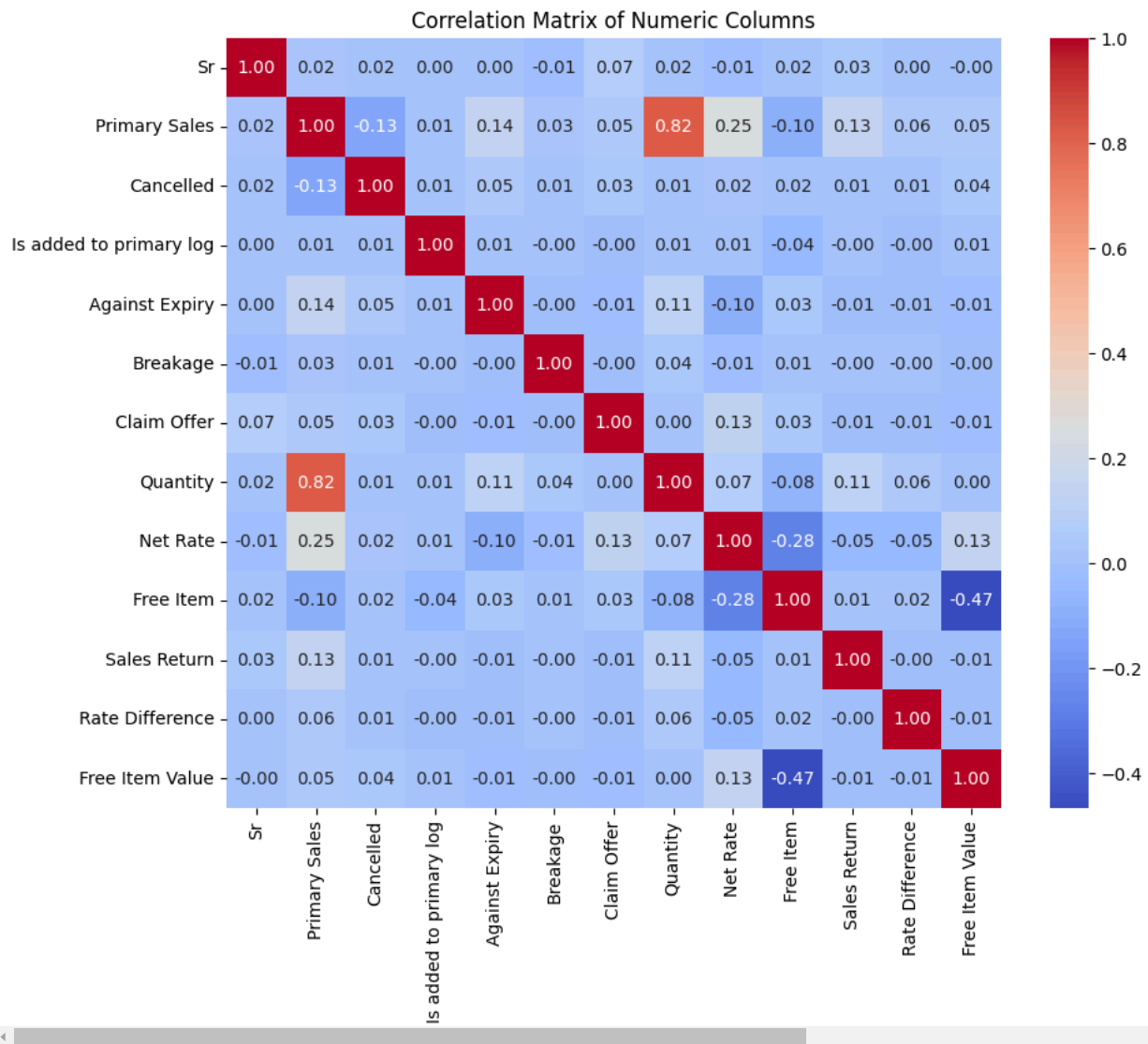
Sales Return	-0.001908	-0.007694	-0.002023
Rate Difference	-0.002660	-0.008056	-0.002119
Free Item Value	0.009617	-0.013866	-0.003646

	Claim Offer	Quantity	Net Rate	Free Item \
Sr	0.071049	0.017140	-0.009896	0.018841
Primary Sales	0.045520	0.823273	0.252527	-0.099085
Cancelled	0.030296	0.012064	0.023944	0.019049
Is added to primary log	-0.004507	0.009441	0.005188	-0.044443
Against Expiry	-0.013652	0.113441	-0.099676	0.029718
Breakage	-0.003590	0.041105	-0.008352	0.007816
Claim Offer	1.000000	0.002637	0.127351	0.026353
Quantity	0.002637	1.000000	0.069684	-0.079472
Net Rate	0.127351	0.069684	1.000000	-0.277883
Free Item	0.026353	-0.079472	-0.277883	1.000000
Sales Return	-0.006823	0.110531	-0.048036	0.014851
Rate Difference	-0.007144	0.058586	-0.049906	0.015550
Free Item Value	-0.012295	0.001484	0.129866	-0.466571

	Sales Return	Rate Difference	Free Item Value
Sr	0.026632	0.001803	-0.000917
Primary Sales	0.133903	0.056880	0.046230
Cancelled	0.012378	0.011003	0.040957
Is added to primary log	-0.001908	-0.002660	0.009617
Against Expiry	-0.007694	-0.008056	-0.013866
Breakage	-0.002023	-0.002119	-0.003646
Claim Offer	-0.006823	-0.007144	-0.012295
Quantity	0.110531	0.058586	0.001484
Net Rate	-0.048036	-0.049906	0.129866
Free Item	0.014851	0.015550	-0.466571
Sales Return	1.000000	-0.004026	-0.006929
Rate Difference	-0.004026	1.000000	-0.007255
Free Item Value	-0.006929	-0.007255	1.000000

```
# Optionally, visualize the correlation matrix as a heatmap
import seaborn as sns
```

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plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Matrix of Numeric Columns')
plt.show()
```



```
# Step 6: Investigating seasonality or trends (optional)
# For example, resample by month to analyze monthly trends
monthly_sales = df['Primary Sales'].resample('M').sum()
monthly_sales.plot(figsize=(12,6), title='Monthly Primary Sales')
plt.ylabel('Total Primary Sales')
plt.show()
```

```

↳ <ipython-input-30-70c8bf812830>:3: FutureWarning: 'M' is deprecated and will be removed in a future version, please use 'ME' instead
monthly_sales = df['Primary Sales'].resample('M').sum()

from statsmodels.tsa.holtwinters import ExponentialSmoothing

# Step 7: Forecasting with Exponential Smoothing (Holt-Winters method)
# Ensure the dataset has no missing values in 'Primary Sales'
df.dropna(subset=['Primary Sales'], inplace=True)

# Fit the Holt-Winters model for trend and seasonality
model = ExponentialSmoothing(df['Primary Sales'], trend='add', seasonal='add', seasonal_periods=12)
model_fit = model.fit()

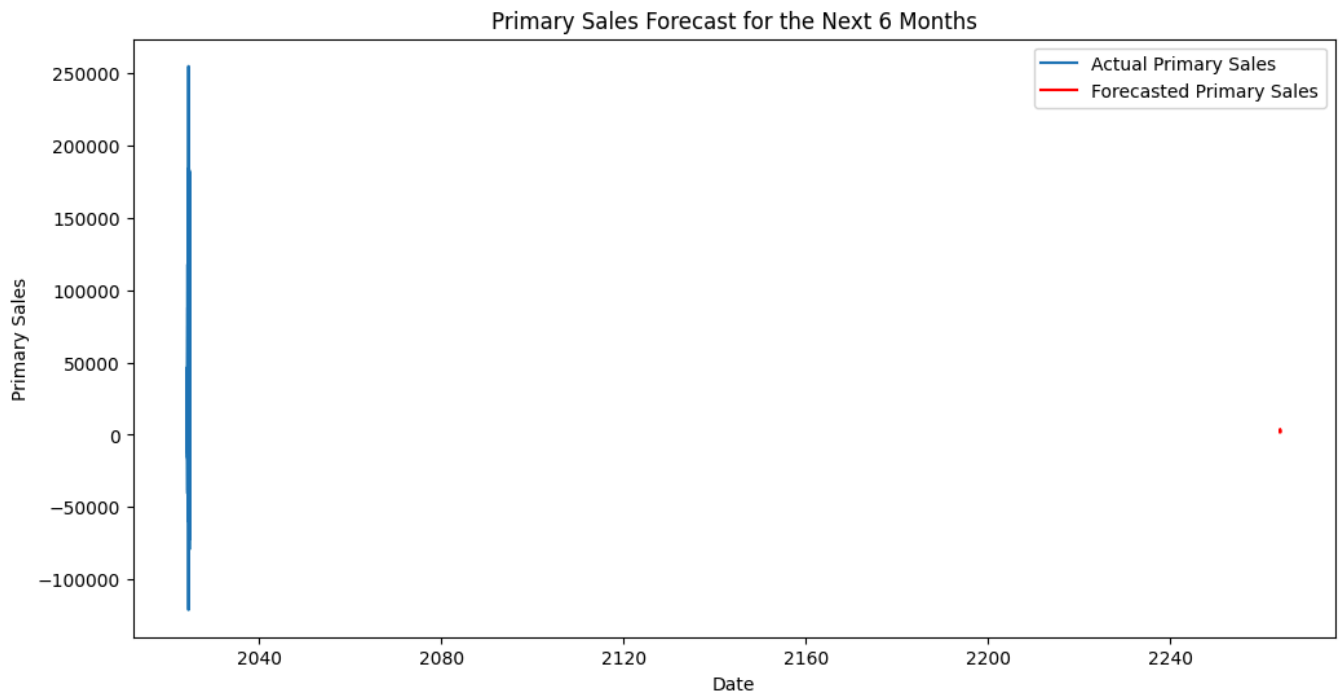
# Step 8: Forecast the next 6 months (adjust as needed)
forecast = model_fit.forecast(steps=6)

# Step 9: Plot the forecast
plt.figure(figsize=(12,6))
plt.plot(df['Primary Sales'], label='Actual Primary Sales')
plt.plot(forecast.index, forecast, label='Forecasted Primary Sales', color='red')
plt.title('Primary Sales Forecast for the Next 6 Months')
plt.xlabel('Date')
plt.ylabel('Primary Sales')
plt.legend()
plt.show()

# Output the forecasted sales for the next 6 months
print("Forecasted Primary Sales for the Next 6 Months:")
print(forecast)

↳ /usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/holtwinters/model.py:84: RuntimeWarning: overflow encountered in matmul
return err.T @ err
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/holtwinters/model.py:918: ConvergenceWarning: Optimization failed to conver
warnings.warn(
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: ValueWarning: No supported index is available. Predic
return get_prediction_index(
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: FutureWarning: No supported index is available. In t
return get_prediction_index(

```



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Forecasted Primary Sales for the Next 6 Months:
107444    1889.948875
107445    2425.420405
107446    2236.112058
107447    3904.247024
107448    2518.473797
107449    2112.713454
dtype: float64

```