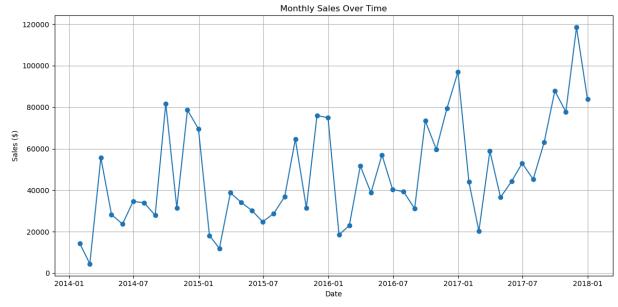
```
In [3]: import pandas as pd
        # Loading the cleaned Superstore dataset
        df = pd.read_csv("/Users/maazhussain/Desktop/Projects/EDA_Superstore/Superst
        # Convert Order Date to datetime and set as index
        df['Order Date'] = pd.to_datetime(df['Order Date'])
        df.set_index('Order Date', inplace=True)
        # Aggregate monthly sales
        monthly_sales = df['Sales'].resample('ME').sum()
        # View the data
        print(monthly_sales.head())
       Order Date
       2014-01-31
                    14236.895
       2014-02-28
                     4519.892
       2014-03-31
                     55691,009
       2014-04-30
                     28295.345
                     23648,287
       2014-05-31
       Freq: ME, Name: Sales, dtype: float64
In [5]: import matplotlib.pyplot as plt
        plt.figure(figsize=(12, 6))
        plt.plot(monthly_sales, marker='o', linestyle='-')
        plt.title('Monthly Sales Over Time')
        plt.xlabel('Date')
```





In [7]: pip install statsmodels

Requirement already satisfied: statsmodels in /opt/anaconda3/lib/python3.12/site-packages (0.14.2)

Requirement already satisfied: numpy>=1.22.3 in /opt/anaconda3/lib/python3.1 2/site-packages (from statsmodels) (1.26.4)

Requirement already satisfied: scipy!=1.9.2,>=1.8 in /opt/anaconda3/lib/pyth on3.12/site-packages (from statsmodels) (1.13.1)

Requirement already satisfied: pandas!=2.1.0,>=1.4 in /opt/anaconda3/lib/pyt hon3.12/site-packages (from statsmodels) (2.2.2)

Requirement already satisfied: patsy>=0.5.6 in /opt/anaconda3/lib/python3.1 2/site-packages (from statsmodels) (0.5.6)

Requirement already satisfied: packaging>=21.3 in /opt/anaconda3/lib/python 3.12/site-packages (from statsmodels) (24.1)

Requirement already satisfied: python-dateutil>=2.8.2 in /opt/anaconda3/lib/python3.12/site-packages (from pandas!=2.1.0,>=1.4->statsmodels) (2.9.0.post 0)

Requirement already satisfied: pytz>=2020.1 in /opt/anaconda3/lib/python3.1 2/site-packages (from pandas!=2.1.0,>=1.4->statsmodels) (2024.1)

Requirement already satisfied: tzdata>=2022.7 in /opt/anaconda3/lib/python3. 12/site-packages (from pandas!=2.1.0,>=1.4->statsmodels) (2023.3)

Requirement already satisfied: six in /opt/anaconda3/lib/python3.12/site-pac kages (from patsy>=0.5.6->statsmodels) (1.16.0)

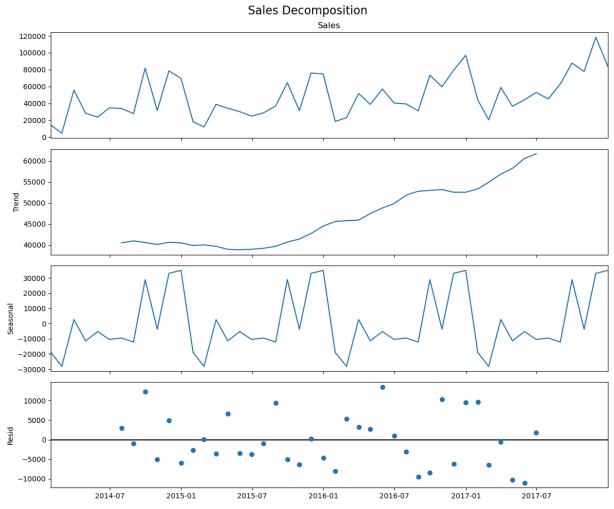
Note: you may need to restart the kernel to use updated packages.

```
In [9]: from statsmodels.tsa.seasonal import seasonal_decompose

# Decompose the time series
decomp = seasonal_decompose(monthly_sales, model='additive', period=12)

# Plot the decomposition
plt.rcParams.update({'figure.figsize': (12, 10)})
decomp.plot()
plt.suptitle('Sales Decomposition', fontsize=16)
plt.tight_layout()
plt.show()
```





In [15]: !pip install prophet

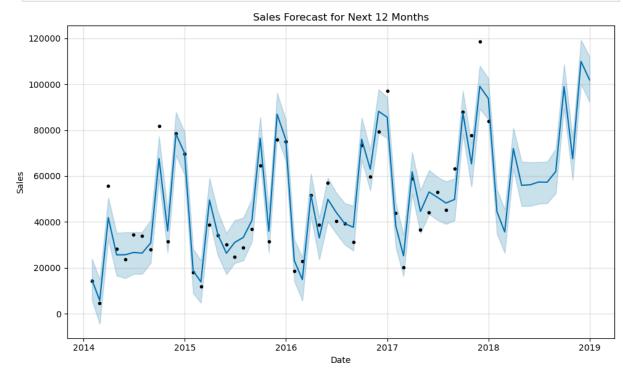
```
Collecting prophet
  Downloading prophet-1.1.6-py3-none-macosx 11 0 arm64.whl.metadata (3.5 kB)
Collecting cmdstanpy>=1.0.4 (from prophet)
  Downloading cmdstanpy-1.2.5-py3-none-any.whl.metadata (4.0 kB)
Requirement already satisfied: numpy>=1.15.4 in /opt/anaconda3/lib/python3.1
2/site-packages (from prophet) (1.26.4)
Requirement already satisfied: matplotlib>=2.0.0 in /opt/anaconda3/lib/pytho
n3.12/site-packages (from prophet) (3.9.2)
Requirement already satisfied: pandas>=1.0.4 in /opt/anaconda3/lib/python3.1
2/site-packages (from prophet) (2.2.2)
Collecting holidays<1,>=0.25 (from prophet)
  Downloading holidays-0.70-py3-none-any.whl.metadata (34 kB)
Requirement already satisfied: tqdm>=4.36.1 in /opt/anaconda3/lib/python3.1
2/site-packages (from prophet) (4.66.5)
Collecting importlib-resources (from prophet)
  Downloading importlib resources-6.5.2-py3-none-any.whl.metadata (3.9 kB)
Collecting stanio<2.0.0,>=0.4.0 (from cmdstanpy>=1.0.4->prophet)
  Downloading stanio-0.5.1-py3-none-any.whl.metadata (1.6 kB)
Requirement already satisfied: python-dateutil in /opt/anaconda3/lib/python
3.12/site-packages (from holidays<1,>=0.25->prophet) (2.9.0.post0)
Requirement already satisfied: contourpy>=1.0.1 in /opt/anaconda3/lib/python
3.12/site-packages (from matplotlib>=2.0.0->prophet) (1.2.0)
Requirement already satisfied: cycler>=0.10 in /opt/anaconda3/lib/python3.1
2/site-packages (from matplotlib>=2.0.0->prophet) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in /opt/anaconda3/lib/pytho
n3.12/site-packages (from matplotlib>=2.0.0->prophet) (4.51.0)
Requirement already satisfied: kiwisolver>=1.3.1 in /opt/anaconda3/lib/pytho
n3.12/site-packages (from matplotlib>=2.0.0->prophet) (1.4.4)
Requirement already satisfied: packaging>=20.0 in /opt/anaconda3/lib/python
3.12/site-packages (from matplotlib>=2.0.0->prophet) (24.1)
Requirement already satisfied: pillow>=8 in /opt/anaconda3/lib/python3.12/si
te-packages (from matplotlib>=2.0.0->prophet) (10.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in /opt/anaconda3/lib/python
3.12/site-packages (from matplotlib>=2.0.0->prophet) (3.1.2)
Requirement already satisfied: pytz>=2020.1 in /opt/anaconda3/lib/python3.1
2/site-packages (from pandas>=1.0.4->prophet) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in /opt/anaconda3/lib/python3.
12/site-packages (from pandas>=1.0.4->prophet) (2023.3)
Requirement already satisfied: six>=1.5 in /opt/anaconda3/lib/python3.12/sit
e-packages (from python-dateutil->holidays<1,>=0.25->prophet) (1.16.0)
Downloading prophet-1.1.6-py3-none-macosx_11_0_arm64.whl (8.2 MB)
                                        ---- 8.2/8.2 MB 12.0 MB/s eta 0:00:00
a 0:00:01
Downloading cmdstanpy-1.2.5-py3-none-any.whl (94 kB)
Downloading holidays-0.70-py3-none-any.whl (903 kB)
                                        --- 903.1/903.1 kB 11.6 MB/s eta 0:0
0:00
Downloading importlib resources-6.5.2-py3-none-any.whl (37 kB)
Downloading stanio-0.5.1-py3-none-any.whl (8.1 kB)
Installing collected packages: stanio, importlib-resources, holidays, cmdsta
npy, prophet
Successfully installed cmdstanpy-1.2.5 holidays-0.70 importlib-resources-6.
5.2 prophet-1.1.6 stanio-0.5.1
```

```
In [25]: # Reset the index and rename columns for Prophet
         prophet_df = monthly_sales.reset_index()
         prophet_df.columns = ['ds', 'y'] # 'ds' = datetime, 'y' = sales
         # Check the result
         print(prophet_df.head())
                  ds
        0 2014-01-31 14236.895
        1 2014-02-28 4519.892
        2 2014-03-31 55691.009
        3 2014-04-30 28295.345
        4 2014-05-31 23648.287
In [27]: # Initialize the Prophet model
         model = Prophet()
         # Fit the model to your monthly sales data
         model.fit(prophet_df)
        12:24:07 - cmdstanpy - INFO - Chain [1] start processing
        12:24:08 - cmdstanpy - INFO - Chain [1] done processing
Out[27]: cprophet.forecaster.Prophet at 0x160f528d0>
In [33]: # Create a DataFrame for 12 months into the future
         future = model.make_future_dataframe(periods=12, freq='ME')
         # Forecast future sales
         forecast = model.predict(future)
         # Preview the forecast
         forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail(12)
```

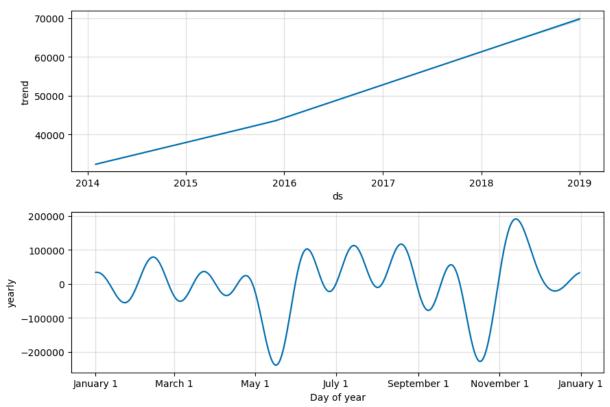
Out[33]:		ds	yhat	yhat_lower	yhat_upper
	48	2018-01-31	44491.106008	34764.866586	54134.191653
	49	2018-02-28	35661.339165	26508.300994	45447.269821

```
2018-03-31
                  71959.875166
                                62966.505930
                                                80937.018669
50
    2018-04-30
                 55942.100032
                                 46897.436801
                                                66261.231412
51
                                               65934.134932
52
    2018-05-31
                 56193.090655
                                46921.292093
   2018-06-30
                 57388.058290
                                 47887.940557
                                               66087.244889
53
54
    2018-07-31
                  57316.111004
                                 48139.221771
                                                66184.136886
55
    2018-08-31
                 61936.252857
                                52456.545850
                                                71826.574968
56
   2018-09-30
                 98895.638706
                                89322.033833
                                               108617.023879
57
    2018-10-31
                 67553.859076
                                58398.303369
                                                 76451.611001
58
    2018-11-30
                109972.884383
                                100051.399215
                                               119281.785612
59
    2018-12-31
                101864.663898
                                92248.297578
                                               112040.778421
```

```
In [35]:
         # Plot the forecast
         model.plot(forecast)
         plt.title('Sales Forecast for Next 12 Months')
         plt.xlabel('Date')
         plt.ylabel('Sales')
         plt.grid(True)
         plt.tight_layout()
         plt.show()
```



In [37]: # View trend and seasonality
 model.plot_components(forecast)
 plt.tight_layout()
 plt.show()



Sales Forecasting Using Time Series Analysis − Superstore Sales

✓ Objective

The goal of this project is to analyze historical monthly sales data and develop a predictive model that forecasts future sales for the next 12 months. This analysis supports informed business planning, inventory management, and financial decision-making.

👜 Tools & Libraries Used

- Python
- Pandas for data wrangling
- Matplotlib for plotting
- Statsmodels for time series decomposition
- Prophet (by Facebook) for building the forecasting model

Key Steps

1. Data Aggregation

- Monthly sales were calculated by resampling the transaction-level dataset using the
 Order Date field.
- Result: A continuous monthly time series of total sales from January 2014 to December 2017.

2. Time Series Visualization

- A line plot was used to visualize overall sales trends and fluctuations over time.
- Clear seasonal spikes and dips were visible, particularly during year-end months.

3. Decomposition

- Applied additive seasonal decomposition (STL) to break down sales into:
 - Trend: Long-term growth in sales
 - Seasonality: Repeating monthly patterns
 - Residual: Irregularities and noise

4. Forecasting with Prophet

- Converted the series into Prophet's required format (ds for date, y for sales).
- Trained the model on historical data and generated forecasts for the next 12 months.
- · Forecast included:
 - yhat : Predicted sales
 - yhat_lower and yhat_upper: 95% confidence interval bounds

Key Results

Month	Forecasted Sales (\$)	Confidence Interval (Lower - Upper)
Jan 2018	44,491	34,765 – 54,134
Mar 2018	71,960	62,967 – 80,937
Jun 2018	57,388	47,888 – 66,087
Sep 2018	98,896	89,322 – 108,617
Dec 2018	101,865	92,248 – 112,041

(Full forecast available in the notebook output)

Business Implications

- **Seasonal Insight:** December and September historically drive high sales optimal for promotions.
- **Strategic Planning:** Forecast supports quarterly sales targets and stock optimization.
- Revenue Management: Helps project cash flow and resource allocation.

In [43]: forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].to_csv("forecast_output