

## Chapter 18 - Exercise 1: Sales of shampoo over a three year

Cho dữ liệu bán shampoo 3 năm trong tập tin sales-of-shampoo-over-a-three-year.csv.

- Thực hiện việc dự báo bán sản phẩm shampoo sử dụng thuật toán ARIMA
- Cho biết trong 3 tháng sau 3 năm trên thì giá trị bán sản phẩm như thế nào?

```
In [ ]: # from google.colab import drive
# drive.mount("/content/gdrive", force_remount=True)
```

Mounted at /content/gdrive

```
In [ ]: # %cd '/content/gdrive/My Drive/LDS6_MachineLearning/practice/Chapter18_ARIMA/'
/content/gdrive/My Drive/LDS6_MachineLearning/practice/Chapter18_ARIMA
```

```
In [ ]: import pandas as pd
```

```
In [ ]: data = pd.read_csv("sales-of-shampoo-over-a-three-year.csv", index_col=0)
data.head()
```

Out[4]:

Sales of shampoo over a three year period

| Month                    |       |
|--------------------------|-------|
| Friday, January 1, 2016  | 266.0 |
| Monday, February 1, 2016 | 145.9 |
| Tuesday, March 1, 2016   | 183.1 |
| Friday, April 1, 2016    | 119.3 |
| Sunday, May 1, 2016      | 180.3 |

```
In [ ]: data.index = pd.to_datetime(data.index)
```

```
In [ ]: data.index
```

```
Out[6]: DatetimeIndex(['2016-01-01', '2016-02-01', '2016-03-01', '2016-04-01',
                        '2016-05-01', '2016-06-01', '2016-07-01', '2016-08-01',
                        '2016-09-01', '2016-10-01', '2016-11-01', '2016-12-01',
                        '2017-01-01', '2017-02-01', '2017-03-01', '2017-04-01',
                        '2017-05-01', '2017-06-01', '2017-07-01', '2017-08-01',
                        '2017-09-01', '2017-10-01', '2017-11-01', '2017-12-01',
                        '2018-01-01', '2018-02-01', '2018-03-01', '2018-04-01',
                        '2018-05-01', '2018-06-01', '2018-07-01', '2018-08-01',
                        '2018-09-01', '2018-10-01', '2018-11-01', '2018-12-01'],
                        dtype='datetime64[ns]', name='Month', freq=None)
```

```
In [ ]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 36 entries, 2016-01-01 to 2018-12-01
Data columns (total 1 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Sales of shampoo over a three year period  36 non-null     float64
dtypes: float64(1)
memory usage: 576.0 bytes
```

```
In [ ]: data.columns = ['Sales_of_shampoo']
```

```
In [ ]: data.head()
```

```
Out[9]:
```

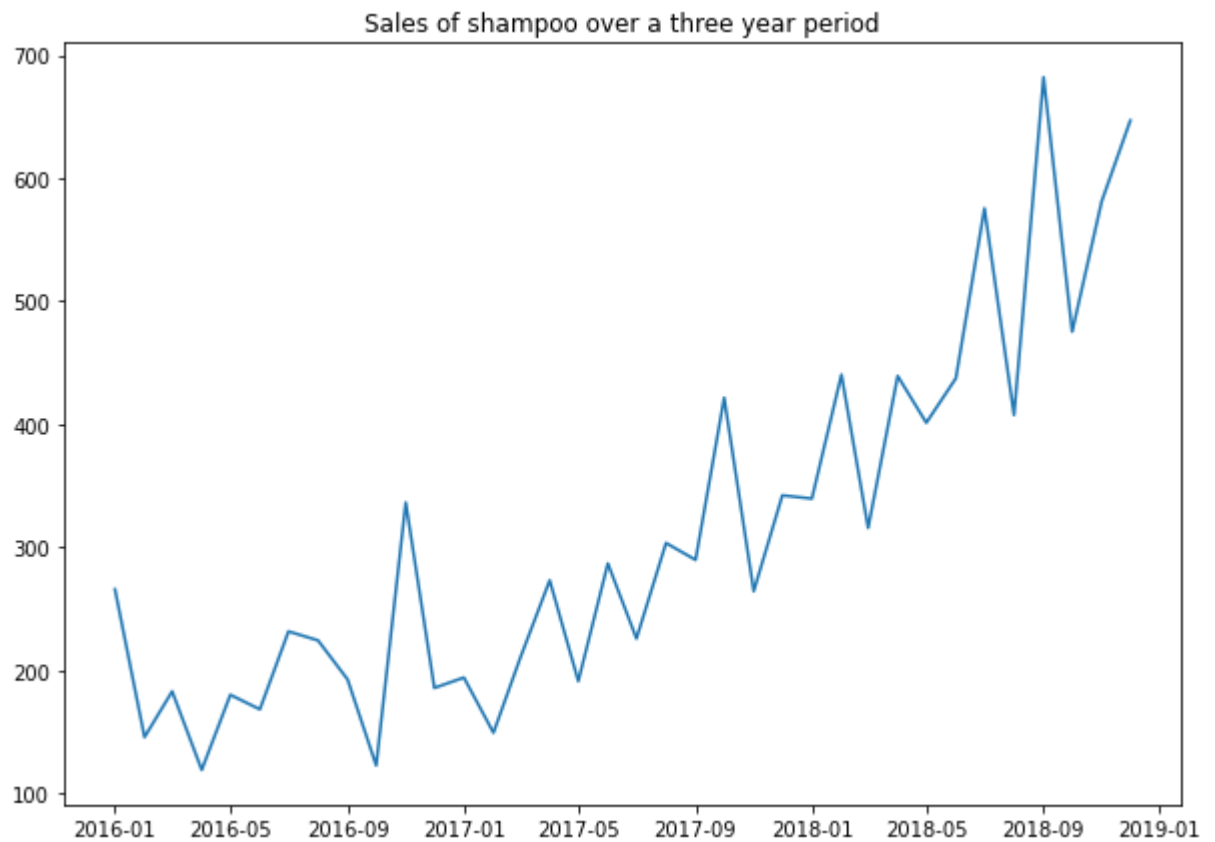
| Sales_of_shampoo |       |
|------------------|-------|
| Month            |       |
| 2016-01-01       | 266.0 |
| 2016-02-01       | 145.9 |
| 2016-03-01       | 183.1 |
| 2016-04-01       | 119.3 |
| 2016-05-01       | 180.3 |

| Month      |       |
|------------|-------|
| 2016-01-01 | 266.0 |
| 2016-02-01 | 145.9 |
| 2016-03-01 | 183.1 |
| 2016-04-01 | 119.3 |
| 2016-05-01 | 180.3 |

```
In [ ]: from datetime import datetime
```

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

```
In [ ]: plt.figure(figsize=(10,7))  
plt.plot(data)  
plt.title("Sales of shampoo over a three year period")  
plt.show()
```



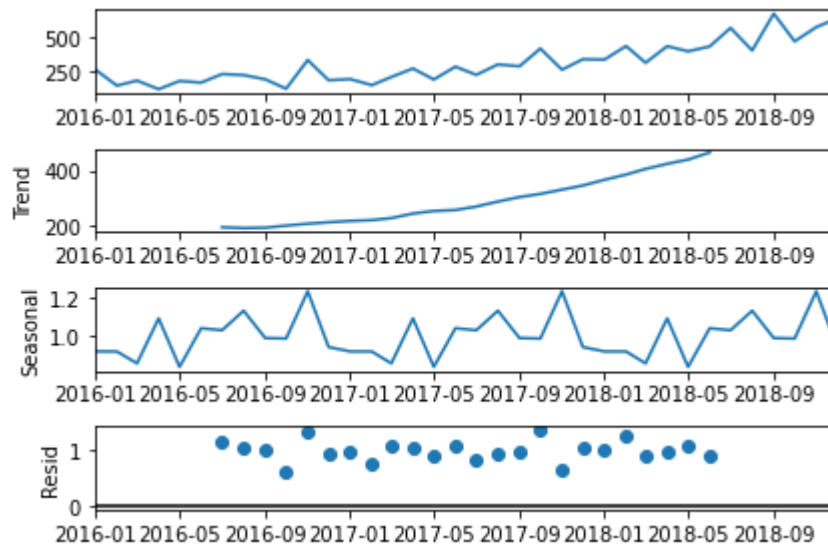
```
In [ ]: type(data)
```

```
Out[13]: pandas.core.frame.DataFrame
```

```
In [ ]: from statsmodels.tsa.seasonal import seasonal_decompose  
result = seasonal_decompose(x = data, model='multiplicative')  
result
```

```
Out[14]: <statsmodels.tsa.seasonal.DecomposeResult at 0x7f17f5b72eb8>
```

```
In [ ]: result.plot()  
plt.show()
```



```
In [ ]: ! pip install pmdarima
```

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

```
In [ ]: stepwise_model = auto_arima(data, start_p=2, start_q= 2,
                                   max_p=5, max_q=5, m=12,
                                   start_P=1, seasonal=True,
                                   d=1, D=1, trace=True,
                                   error_action='ignore',
                                   suppress_warnings=True,
                                   stepwise=True)
```

Performing stepwise search to minimize aic

```
ARIMA(2,1,2)(1,1,1)[12]      : AIC=inf, Time=1.32 sec
ARIMA(0,1,0)(0,1,0)[12]      : AIC=305.954, Time=0.02 sec
ARIMA(1,1,0)(1,1,0)[12]      : AIC=287.610, Time=0.16 sec
ARIMA(0,1,1)(0,1,1)[12]      : AIC=289.951, Time=0.28 sec
ARIMA(1,1,0)(0,1,0)[12]      : AIC=287.696, Time=0.04 sec
ARIMA(1,1,0)(2,1,0)[12]      : AIC=inf, Time=1.40 sec
ARIMA(1,1,0)(1,1,1)[12]      : AIC=inf, Time=0.67 sec
ARIMA(1,1,0)(0,1,1)[12]      : AIC=287.722, Time=0.25 sec
ARIMA(1,1,0)(2,1,1)[12]      : AIC=291.042, Time=1.73 sec
ARIMA(0,1,0)(1,1,0)[12]      : AIC=303.591, Time=0.11 sec
ARIMA(2,1,0)(1,1,0)[12]      : AIC=286.460, Time=0.27 sec
ARIMA(2,1,0)(0,1,0)[12]      : AIC=288.501, Time=0.07 sec
ARIMA(2,1,0)(2,1,0)[12]      : AIC=inf, Time=1.84 sec
ARIMA(2,1,0)(1,1,1)[12]      : AIC=inf, Time=0.89 sec
ARIMA(2,1,0)(0,1,1)[12]      : AIC=287.268, Time=0.36 sec
ARIMA(2,1,0)(2,1,1)[12]      : AIC=288.971, Time=2.39 sec
ARIMA(3,1,0)(1,1,0)[12]      : AIC=287.153, Time=0.33 sec
ARIMA(2,1,1)(1,1,0)[12]      : AIC=287.237, Time=0.39 sec
ARIMA(1,1,1)(1,1,0)[12]      : AIC=285.464, Time=0.24 sec
ARIMA(1,1,1)(0,1,0)[12]      : AIC=286.063, Time=0.08 sec
ARIMA(1,1,1)(2,1,0)[12]      : AIC=286.473, Time=1.61 sec
ARIMA(1,1,1)(1,1,1)[12]      : AIC=inf, Time=0.88 sec
ARIMA(1,1,1)(0,1,1)[12]      : AIC=285.779, Time=0.35 sec
ARIMA(1,1,1)(2,1,1)[12]      : AIC=288.479, Time=1.90 sec
ARIMA(0,1,1)(1,1,0)[12]      : AIC=289.967, Time=0.19 sec
ARIMA(1,1,2)(1,1,0)[12]      : AIC=inf, Time=0.48 sec
ARIMA(0,1,2)(1,1,0)[12]      : AIC=inf, Time=0.48 sec
ARIMA(2,1,2)(1,1,0)[12]      : AIC=inf, Time=0.72 sec
ARIMA(1,1,1)(1,1,0)[12] intercept : AIC=inf, Time=0.40 sec
```

Best model: ARIMA(1,1,1)(1,1,0)[12]

Total fit time: 19.888 seconds

```
In [ ]: print(stepwise_model.aic())
```

285.4635906667674

```
In [ ]: train = data.loc['2016-01-01':'2018-02-01']
        test = data.loc['2018-02-01':]
```

```
In [ ]: test
```

```
Out[21]: Sales_of_shampoo
```

| Month      |       |
|------------|-------|
| 2018-02-01 | 440.4 |
| 2018-03-01 | 315.9 |
| 2018-04-01 | 439.3 |
| 2018-05-01 | 401.3 |
| 2018-06-01 | 437.4 |
| 2018-07-01 | 575.5 |
| 2018-08-01 | 407.6 |
| 2018-09-01 | 682.0 |
| 2018-10-01 | 475.3 |
| 2018-11-01 | 581.3 |
| 2018-12-01 | 646.9 |

```
In [ ]: len(test)
```

```
Out[22]: 11
```

```
In [ ]: len(train)
```

```
Out[23]: 26
```

```
In [ ]: stepwise_model.fit(train)
```

```
Out[24]: ARIMA(maxiter=50, method='lbfgs', order=(1, 1, 1), out_of_sample_size=0,
              scoring='mse', scoring_args={}, seasonal_order=(1, 1, 0, 12),
              start_params=None, suppress_warnings=True, trend=None,
              with_intercept=False)
```

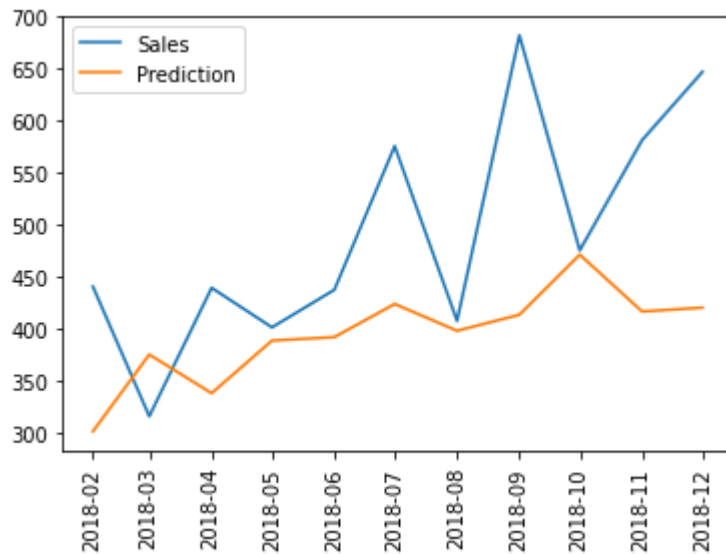
```
In [ ]: future_forecast = stepwise_model.predict(n_periods=len(test))
```

```
In [ ]: future_forecast
```

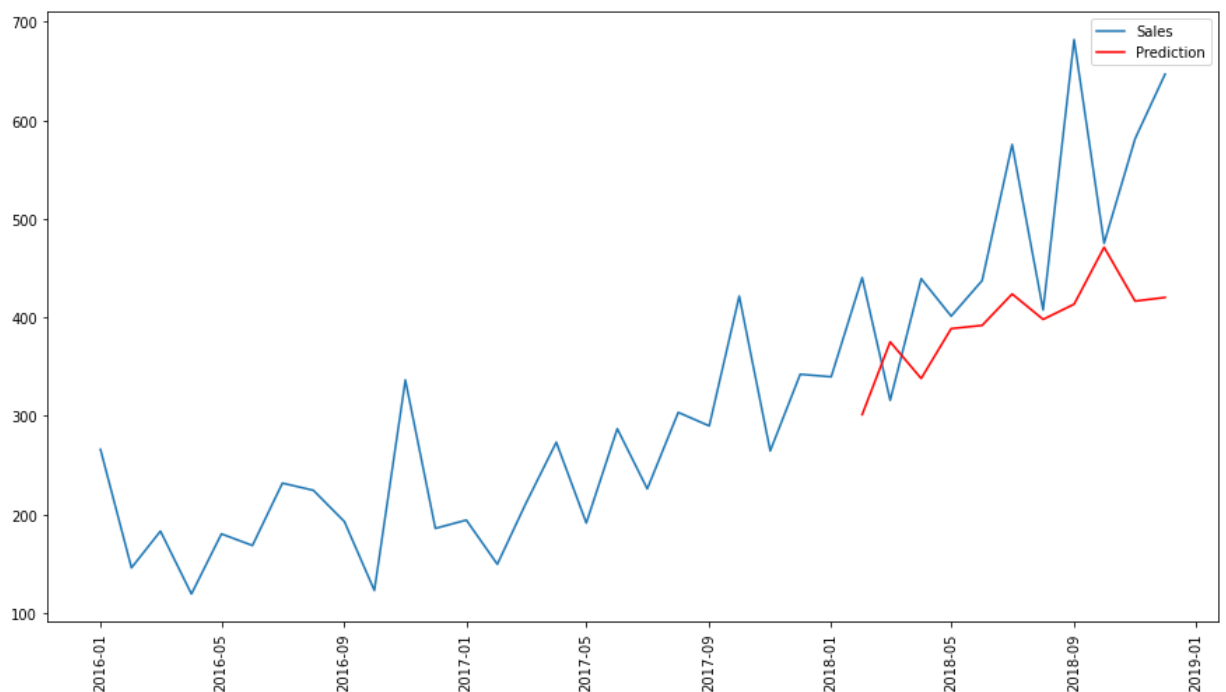
```
Out[26]: array([301.37822619, 375.19264882, 338.01219278, 388.5995918 ,
                391.89687524, 423.77554979, 398.01232103, 413.41907471,
                471.17193988, 416.54027228, 420.25738927])
```

```
In [ ]: future_forecast = pd.DataFrame(future_forecast,
                                       index = test.index,
                                       columns=['Prediction'])
```

```
In [ ]: plt.plot(test, label='Sales')
plt.plot(future_forecast, label='Prediction')
plt.xticks(rotation='vertical')
plt.legend()
plt.show()
```



```
In [ ]: plt.figure(figsize=(15,8))
plt.plot(data, label='Sales')
plt.plot(future_forecast, label='Prediction', color='red')
plt.xticks(rotation='vertical')
plt.legend()
plt.show()
```



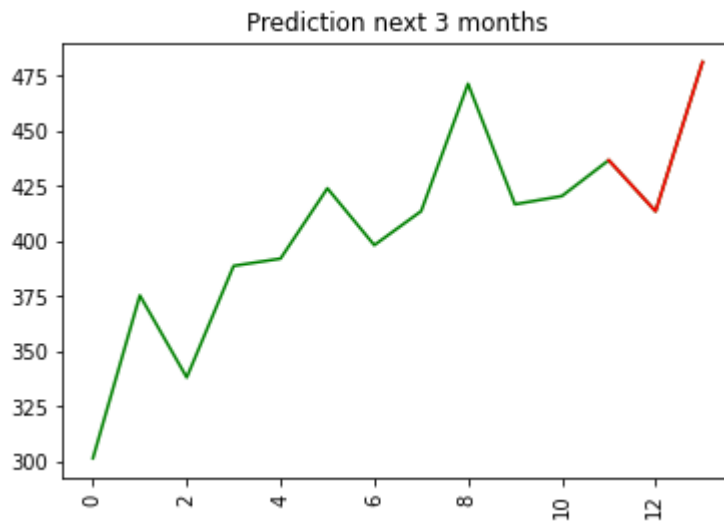
```
In [ ]: # Dự đoán 3 tháng sau
```

```
In [ ]: future_forecast = stepwise_model.predict(n_periods=len(test)+3)
future_forecast
```

```
Out[31]: array([301.37822619, 375.19264882, 338.01219278, 388.5995918 ,
               391.89687524, 423.77554979, 398.01232103, 413.41907471,
               471.17193988, 416.54027228, 420.25738927, 436.52817491,
               413.36922593, 481.02851979])
```

```
In [ ]: import numpy as np
```

```
In [ ]: plt.plot(np.arange(14), future_forecast, color='green')
plt.plot(np.array([11, 12, 13]), future_forecast[len(test):], color='red')
plt.xticks(rotation='vertical')
plt.title("Prediction next 3 months")
plt.show()
```



```
In [ ]:
```