

```
In [83]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from statsmodels.tsa.arima_model import ARMA
%matplotlib inline
```

```
In [84]: df=pd.read_csv('/Users/kushimahar/Downloads/cloth_sales.csv')
df.head()
```

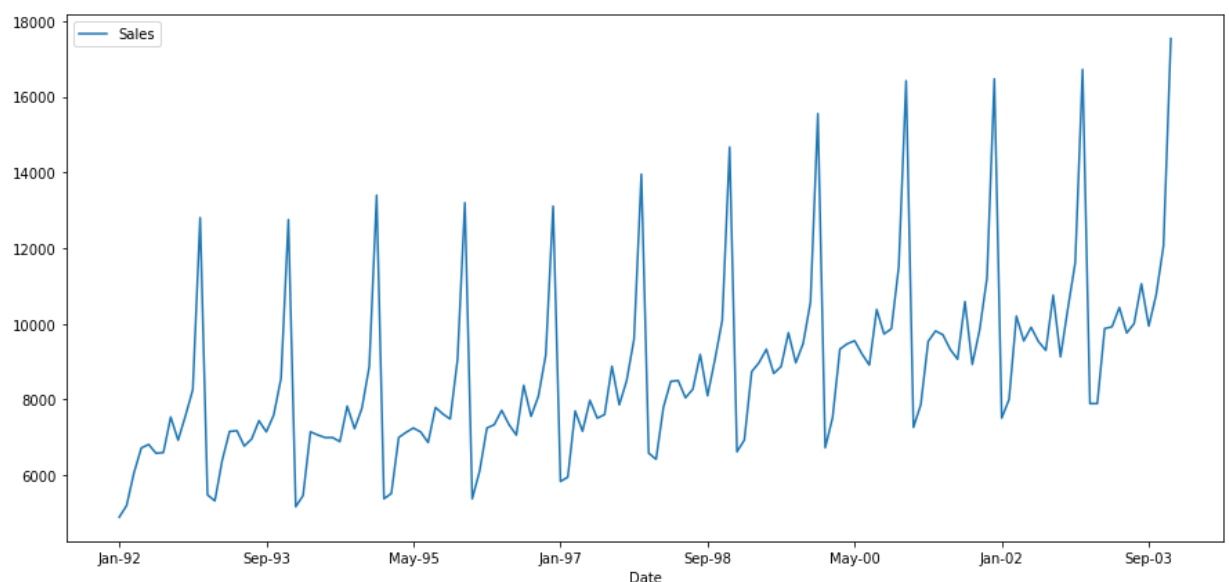
```
Out[84]:
```

	date	sales
0	Jan-92	4889
1	Feb-92	5198
2	Mar-92	6061
3	Apr-92	6720
4	May-92	6811

```
In [85]: df.columns=["Date","Sales"]
df.head()
df.describe()
df.set_index('Date',inplace=True)
```

```
In [86]: from pylab import rcParams
rcParams['figure.figsize'] = 15, 7
df.plot()
```

```
Out[86]: <AxesSubplot:xlabel='Date'>
```



```
In [87]: from statsmodels.tsa.stattools import adfuller
```

```
In [88]: test_result=adfuller(df['Sales'])
```

```
In [89]: #Ho: It is non-stationary
```

```
#H1: It is stationary
```

```
In [90]: def adfuller_test(sales):
          result=adfuller(sales)
          labels = ['ADF Test Statistic','p-value','#Lags Used','Number of Observations']
          for value,label in zip(result,labels):
              print(label+' : '+str(value) )

          if result[1] <= 0.05:
              print("strong evidence against the null hypothesis(Ho), reject the null hypothesis")
          else:
              print("weak evidence against null hypothesis,indicating it is non-stationary")

          adfuller_test(df['Sales'])
```

```
ADF Test Statistic : 0.6384826676448148
p-value : 0.9885267347007706
#Lags Used : 14
Number of Observations : 129
weak evidence against null hypothesis,indicating it is non-stationary
```

```
In [91]: df['Sales First Difference'] = df['Sales'] - df['Sales'].shift(1)
          df['Seasonal First Difference']=df['Sales']-df['Sales'].shift(12)
          df.head()
```

```
Out[91]:
```

	Sales	Sales First Difference	Seasonal First Difference
Jan-92	4889	NaN	NaN
Feb-92	5198	309.0	NaN
Mar-92	6061	863.0	NaN
Apr-92	6720	659.0	NaN
May-92	6811	91.0	NaN

Date

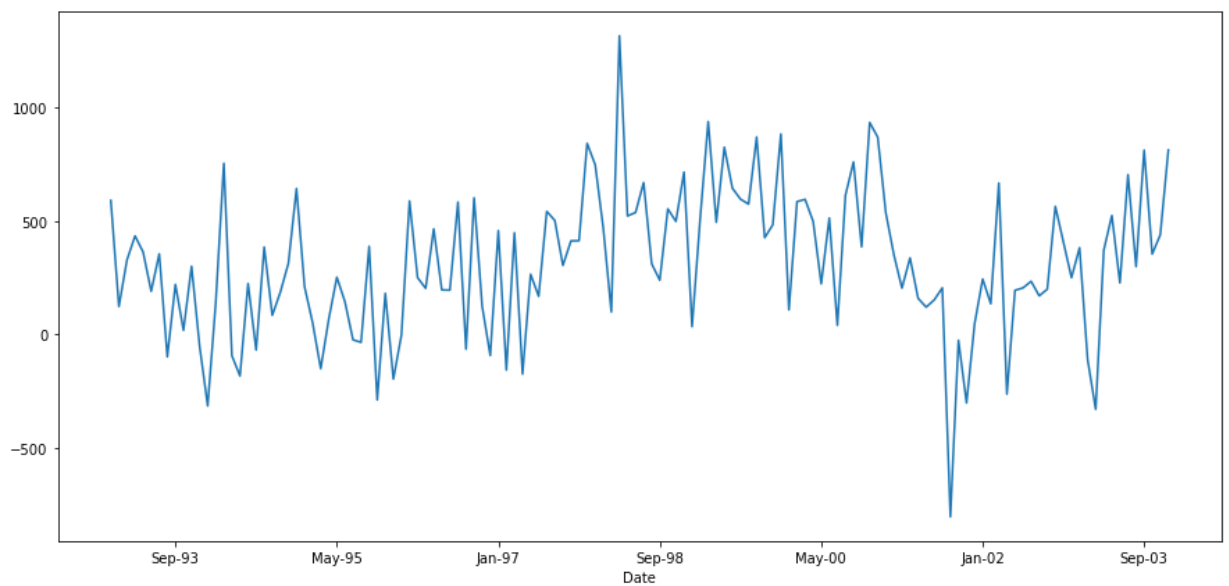
Date	Sales	Sales First Difference	Seasonal First Difference
Jan-92	4889	NaN	NaN
Feb-92	5198	309.0	NaN
Mar-92	6061	863.0	NaN
Apr-92	6720	659.0	NaN
May-92	6811	91.0	NaN

```
In [92]: # Again testing if data is stationary
          adfuller_test(df['Seasonal First Difference'].dropna())
```

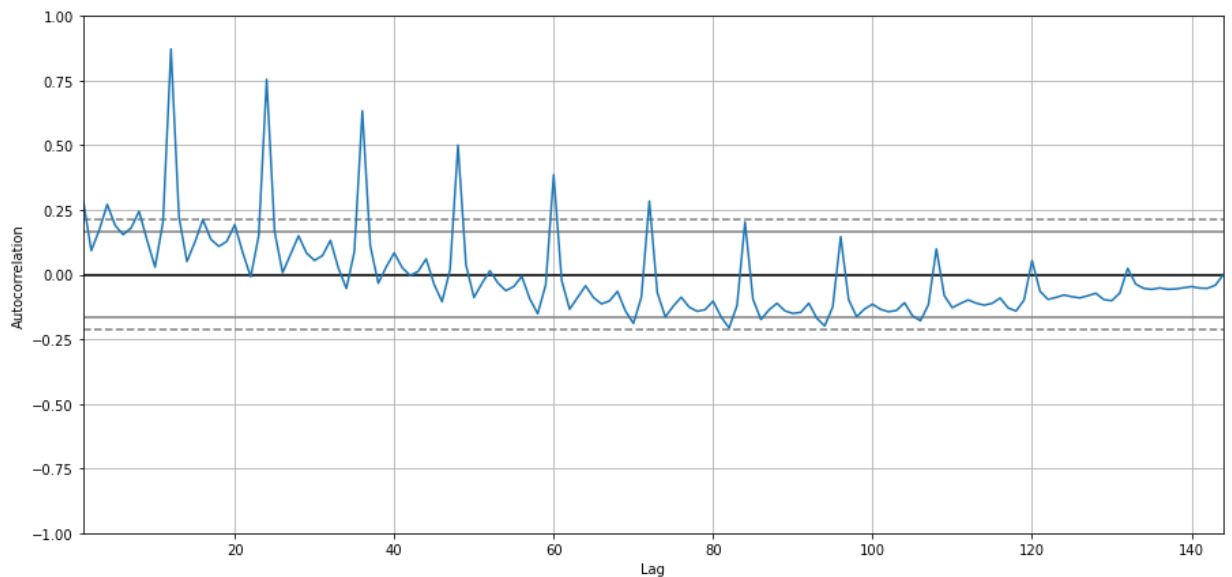
```
ADF Test Statistic : -3.442558535470895
p-value : 0.009590834550182063
#Lags Used : 2
Number of Observations : 129
strong evidence against the null hypothesis(Ho), reject the null hypothesis. Data is stationary
```

```
In [93]: df['Seasonal First Difference'].plot()
```

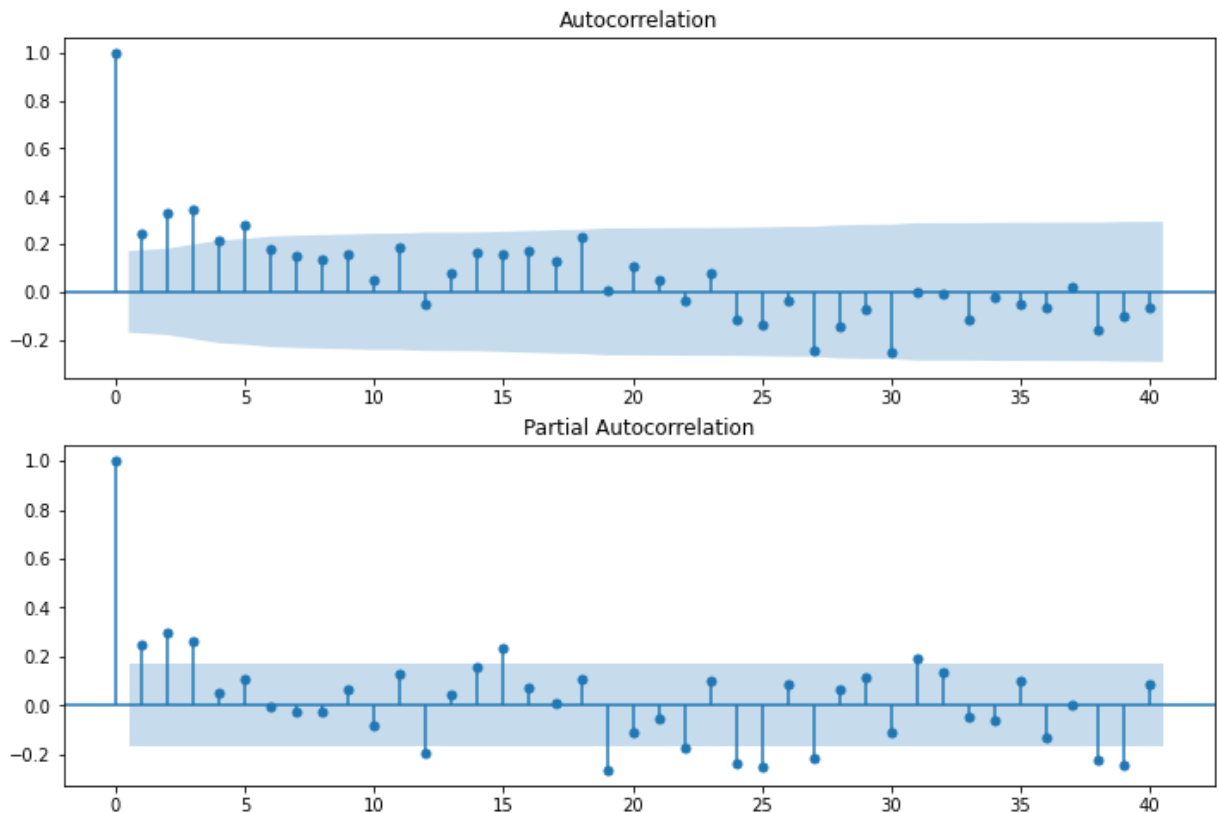
```
Out[93]: <AxesSubplot:xlabel='Date'>
```



```
In [94]: from pandas.plotting import autocorrelation_plot
autocorrelation_plot(df['Sales'])
plt.show()
```



```
In [95]: from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
import statsmodels.api as sm
fig = plt.figure(figsize=(12,8))
ax1 = fig.add_subplot(211)
fig = sm.graphics.tsa.plot_acf(df['Seasonal First Difference'].dropna(), lags=
ax2 = fig.add_subplot(212)
fig = sm.graphics.tsa.plot_pacf(df['Seasonal First Difference'].dropna(), lags=
```



```
In [96]: model = ARMA (df['Sales'], order = (3,0))
res = model.fit()
res.plot_predict()
model_fit=model.fit()
model_fit.summary()
```

```
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ar
ima_model.py:472: FutureWarning:
statsmodels.tsa.arima_model.ARMA and statsmodels.tsa.arima_model.ARIMA have
been deprecated in favor of statsmodels.tsa.arima.model.ARIMA (note the .
between arima and model) and
statsmodels.tsa.SARIMAX. These will be removed after the 0.12 release.
```

statsmodels.tsa.arima.model.ARIMA makes use of the statespace framework and is both well tested and maintained.

To silence this warning and continue using ARMA and ARIMA until they are removed, use:

```
import warnings
warnings.filterwarnings('ignore', 'statsmodels.tsa.arima_model.ARMA',
                        FutureWarning)
warnings.filterwarnings('ignore', 'statsmodels.tsa.arima_model.ARIMA',
                        FutureWarning)
```

```
warnings.warn(ARIMA_DEPRECATION_WARN, FutureWarning)
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ba
se/tsa_model.py:578: ValueWarning: An unsupported index was provided and will
be ignored when e.g. forecasting.
warnings.warn('An unsupported index was provided and will be'
```

Out[96]:

```
ARMA Model Results
Dep. Variable:          Sales    No. Observations:          144
Model:          ARMA(3, 0)      Log Likelihood    -1315.322
Method:          css-mle      S.D. of innovations    2240.561
Date:    Fri, 29 Apr 2022      AIC    2640.644
```

Time:

10:20:52

BIC

2655.493

Sample:

0

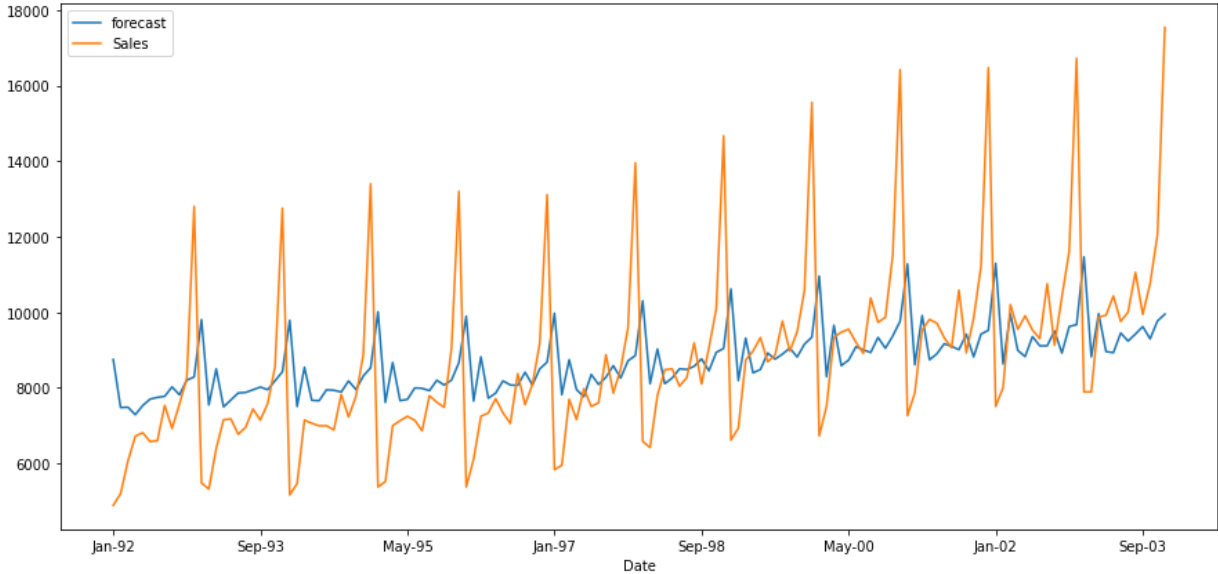
HQIC

2646.678

	coef	std err	z	P> z	[0.025	0.975]
const	8747.5907	349.534	25.026	0.000	8062.516	9432.665
ar.L1.Sales	0.3112	0.086	3.632	0.000	0.143	0.479
ar.L2.Sales	-0.0233	0.089	-0.261	0.794	-0.199	0.152
ar.L3.Sales	0.1831	0.087	2.112	0.035	0.013	0.353

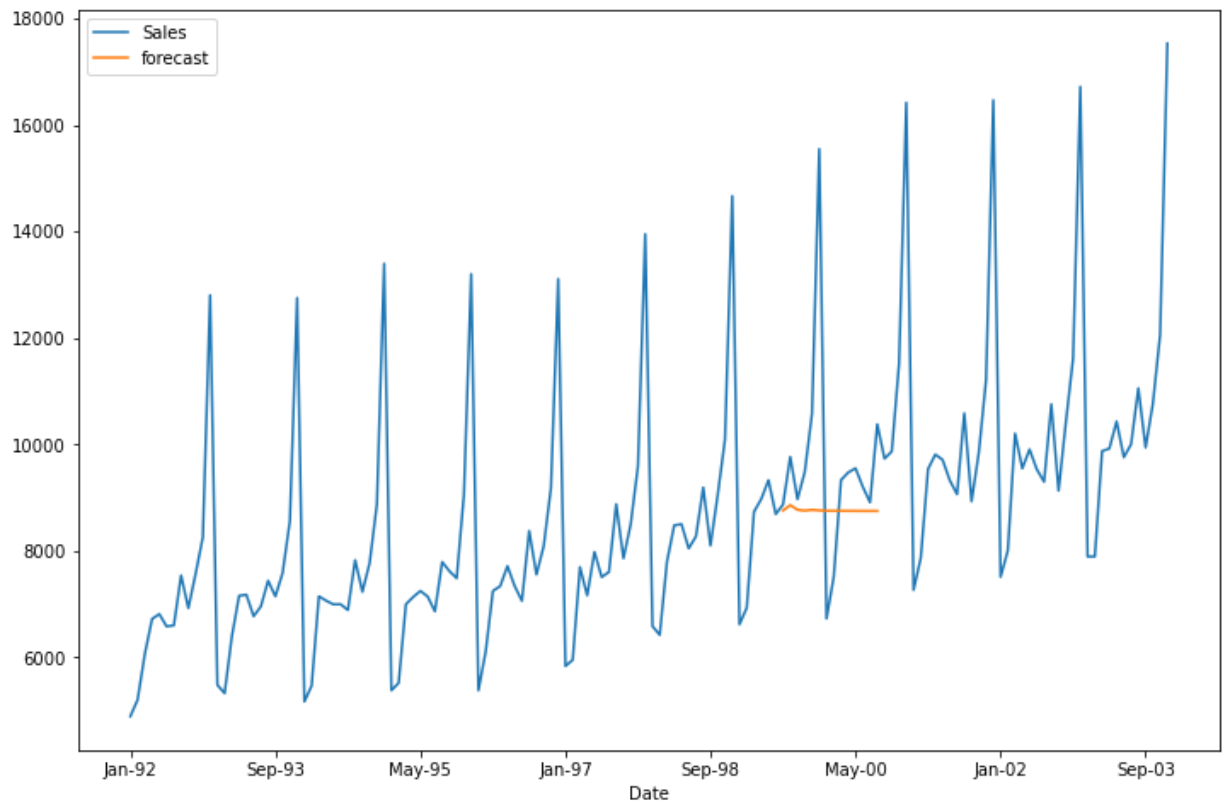
Roots

	Real	Imaginary	Modulus	Frequency
AR.1	1.4778	-0.0000j	1.4778	-0.0000
AR.2	-0.6752	-1.7998j	1.9223	-0.3071
AR.3	-0.6752	+1.7998j	1.9223	0.3071



```
In [97]: df['forecast']=model_fit.predict(start=90,end=103,dynamic=True)
df[['Sales','forecast']].plot(figsize=(12,8))
```

Out[97]: <AxesSubplot:xlabel='Date'>



```
In [98]: import statsmodels.api as sm
model=sm.tsa.statespace.SARIMAX(df['Sales'],order=(1, 1, 1),seasonal_order=(1
results=model.fit()
df['forecast']=results.predict(start=90,end=103,dynamic=True)
df[['Sales', 'forecast']].plot(figsize=(12,8))
```

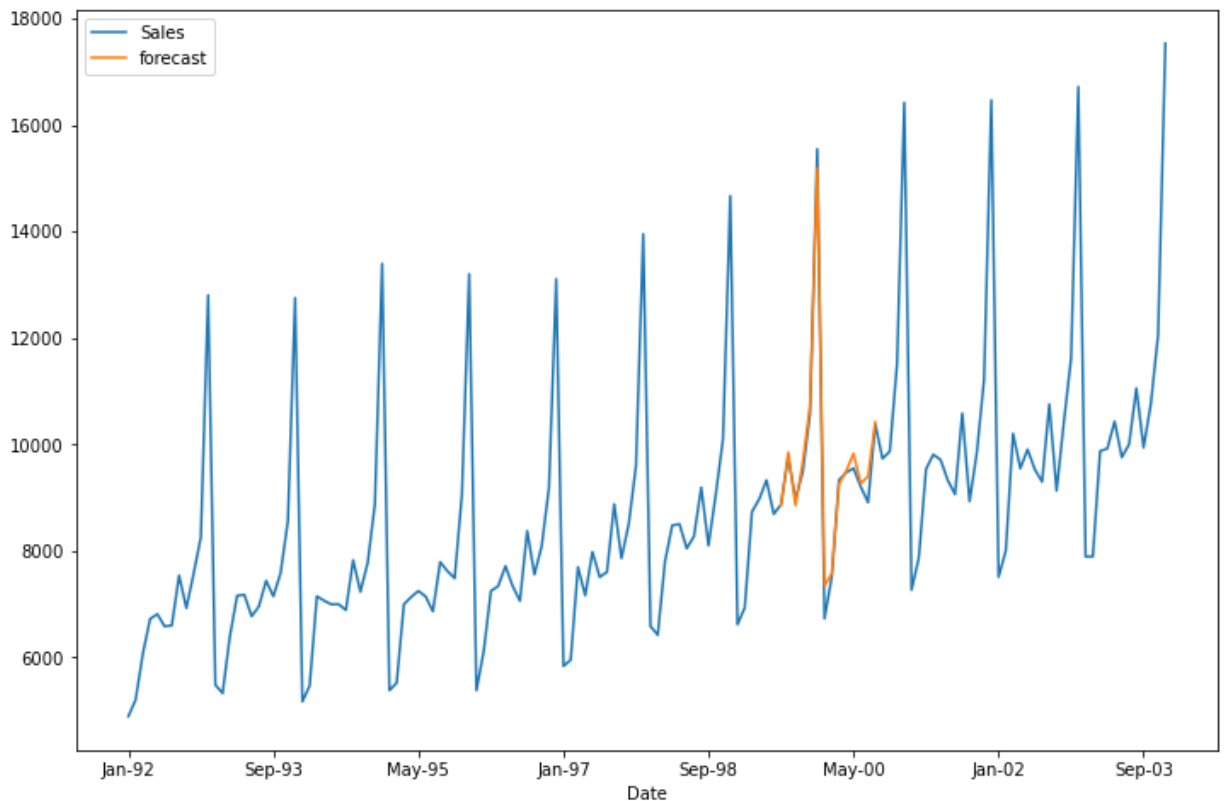
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/base/tsa_model.py:578: ValueWarning: An unsupported index was provided and will be ignored when e.g. forecasting.

warnings.warn('An unsupported index was provided and will be'

/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/base/tsa_model.py:578: ValueWarning: An unsupported index was provided and will be ignored when e.g. forecasting.

warnings.warn('An unsupported index was provided and will be'

```
Out[98]: <AxesSubplot:xlabel='Date'>
```



In [99]:

```
# For non-seasonal data
#p=1, d=1, q=0 or 1

from statsmodels.tsa.arima_model import ARIMA
model=ARIMA(df['Sales'],order=(1,1,1))
model_fit=model.fit()
model_fit.summary()
```

```
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ar
ima_model.py:472: FutureWarning:
statsmodels.tsa.arima_model.ARMA and statsmodels.tsa.arima_model.ARIMA have
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```

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warnings.filterwarnings('ignore', 'statsmodels.tsa.arima_model.ARIMA',
                        FutureWarning)

warnings.warn(ARIMA_DEPRECATION_WARN, FutureWarning)
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ba
se/tsa_model.py:578: ValueWarning: An unsupported index was provided and will
be ignored when e.g. forecasting.
warnings.warn('An unsupported index was provided and will be'
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ba
se/tsa_model.py:578: ValueWarning: An unsupported index was provided and will
be ignored when e.g. forecasting.
warnings.warn('An unsupported index was provided and will be'
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ar
ima_model.py:472: FutureWarning:
statsmodels.tsa.arima_model.ARMA and statsmodels.tsa.arima_model.ARIMA have
```

been deprecated in favor of `statsmodels.tsa.arima.model.ARIMA` (note the . between `arima` and `model`) and `statsmodels.tsa.SARIMAX`. These will be removed after the 0.12 release.

`statsmodels.tsa.arima.model.ARIMA` makes use of the statespace framework and is both well tested and maintained.

To silence this warning and continue using ARMA and ARIMA until they are removed, use:

```
import warnings
warnings.filterwarnings('ignore', 'statsmodels.tsa.arima_model.ARMA',
                        FutureWarning)
warnings.filterwarnings('ignore', 'statsmodels.tsa.arima_model.ARIMA',
                        FutureWarning)
```

```
warnings.warn(ARIMA_DEPRECATION_WARN, FutureWarning)
```

Out[99]:

ARIMA Model Results

Dep. Variable:	D.Sales	No. Observations:	143
Model:	ARIMA(1, 1, 1)	Log Likelihood	-1293.257
Method:	css-mle	S.D. of innovations	2013.668
Date:	Fri, 29 Apr 2022	AIC	2594.514
Time:	10:20:55	BIC	2606.366
Sample:	1	HQIC	2599.330

	coef	std err	z	P> z	[0.025	0.975]
const	31.1829	4.198	7.428	0.000	22.955	39.411
ar.L1.D.Sales	0.0378	0.087	0.433	0.665	-0.133	0.209
ma.L1.D.Sales	-1.0000	0.019	-53.202	0.000	-1.037	-0.963

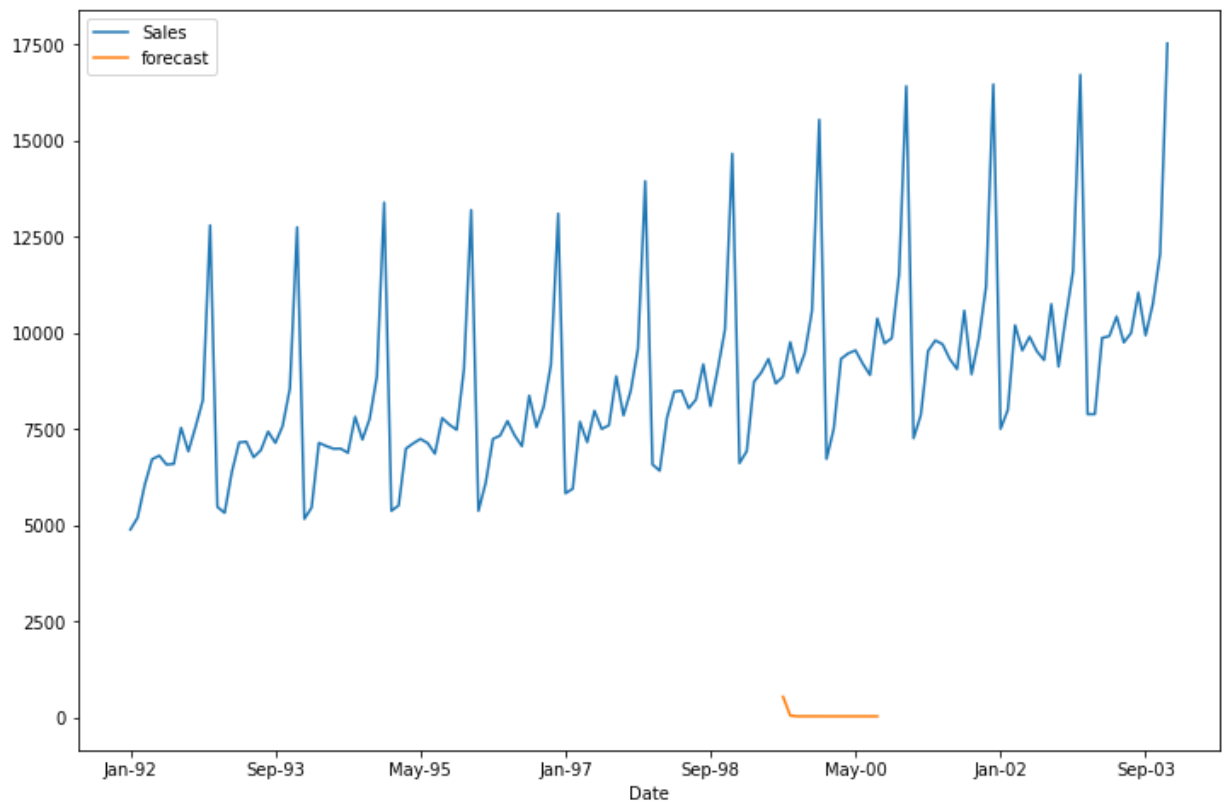
Roots

	Real	Imaginary	Modulus	Frequency
AR.1	26.4512	+0.0000j	26.4512	0.0000
MA.1	1.0000	+0.0000j	1.0000	0.0000

In [100...

```
df['forecast']=model_fit.predict(start=90,end=103,dynamic=True)
df[['Sales','forecast']].plot(figsize=(12,8))
```

Out[100... <AxesSubplot:xlabel='Date'>



```
In [101... import statsmodels.api as sm
model=sm.tsa.statespace.SARIMAX(df['Sales'],order=(1, 1, 1),seasonal_order=(1
results=model.fit()
df['forecast']=results.predict(start=90,end=103,dynamic=True)
df[['Sales', 'forecast']].plot(figsize=(12,8))
```

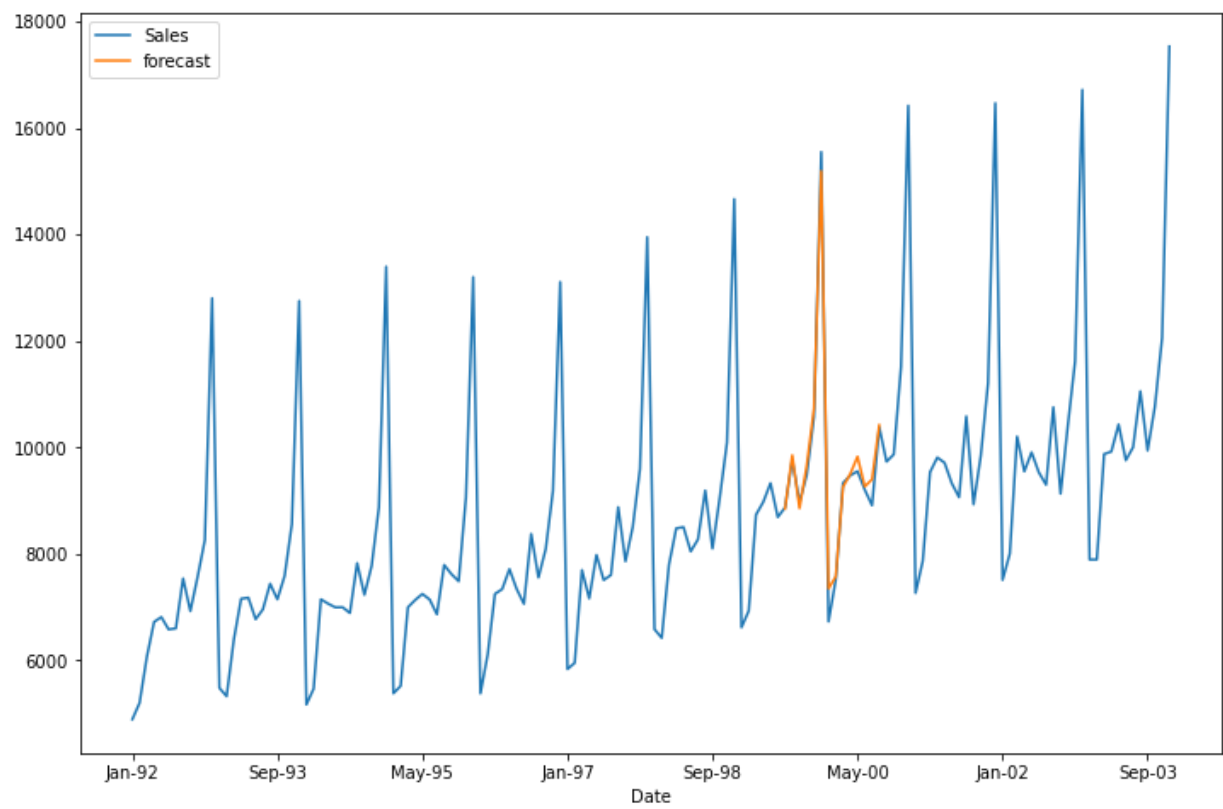
/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ba
se/tsa_model.py:578: ValueWarning: An unsupported index was provided and will
be ignored when e.g. forecasting.

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/Users/kushimahar/opt/anaconda3/lib/python3.8/site-packages/statsmodels/tsa/ba
se/tsa_model.py:578: ValueWarning: An unsupported index was provided and will
be ignored when e.g. forecasting.

warnings.warn('An unsupported index was provided and will be'

```
Out[101... <AxesSubplot:xlabel='Date'>
```



In []:

In []:

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In []: