

TIME SERIES FORECASTING

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PGPDSBA – JULY 2023

SOLUTION

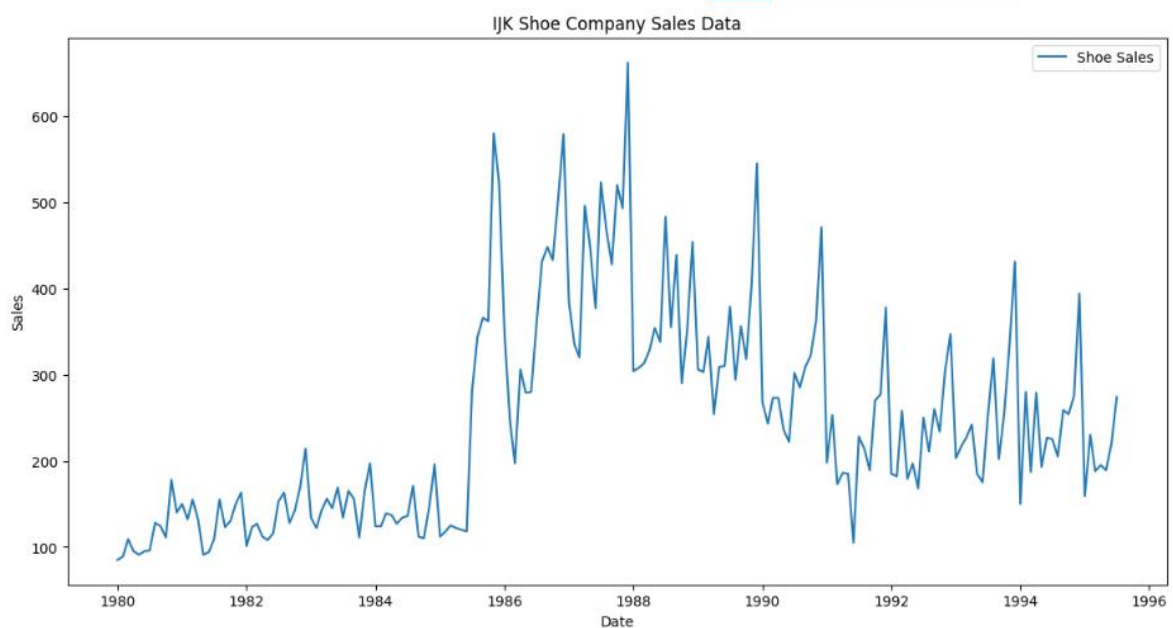
1. Define the problem and perform Exploratory Data Analysis

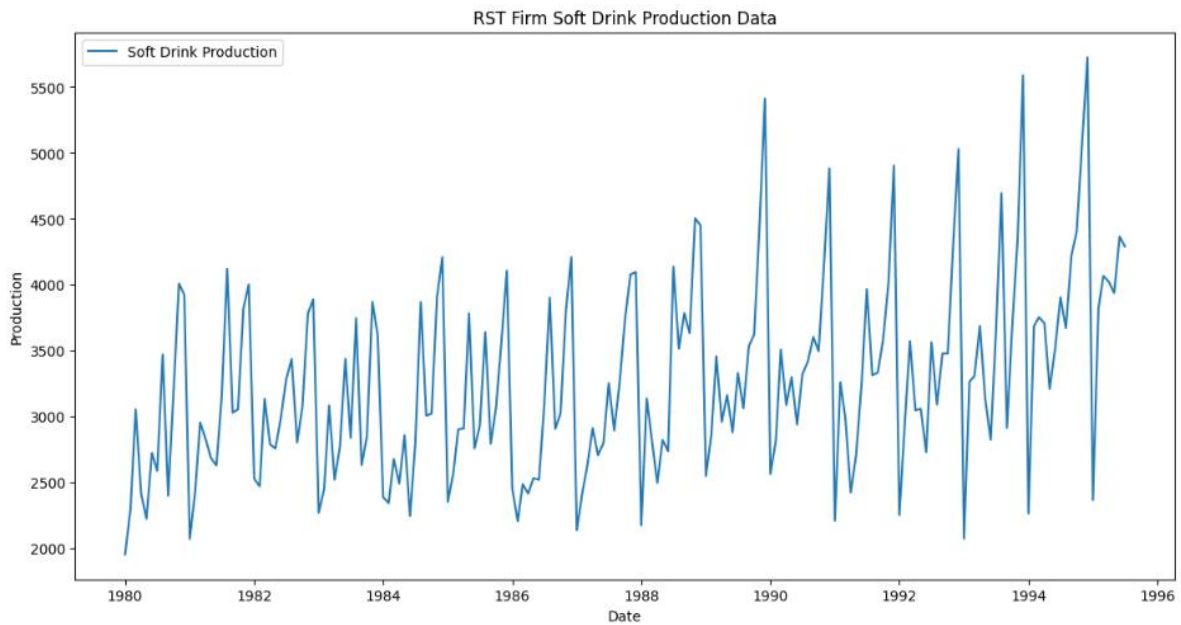
- Read the data as an appropriate time series data - Plot the data - Perform EDA - Perform Decomposition

- Top 5 rows of Shoe Sales data and Soft Drinks data

```
(
  Shoe_Sales
  YearMonth
  1980-01-01      85
  1980-02-01      89
  1980-03-01     109
  1980-04-01      95
  1980-05-01      91,
  SoftDrinkProduction
  YearMonth
  1980-01-01     1954
  1980-02-01     2302
  1980-03-01     3054
  1980-04-01     2414
  1980-05-01     2226)
```

- Plotted the Data





- Performed EDA

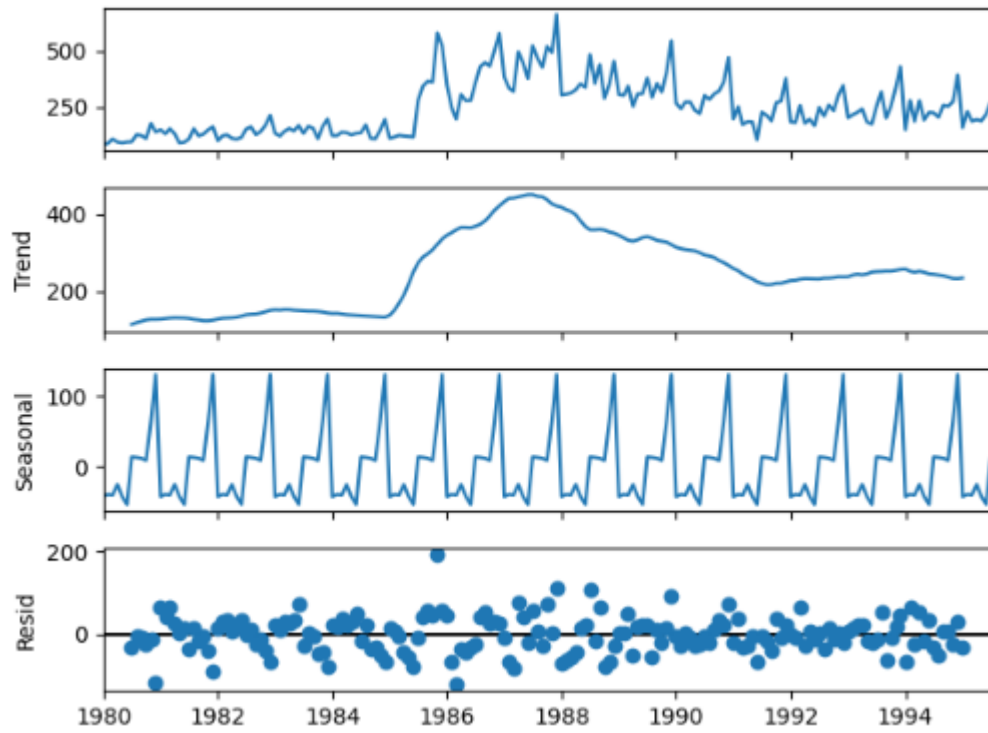
```

(Shoe_Sales    0
 dtype: int64,
 SoftDrinkProduction    0
 dtype: int64,
  Shoe_Sales
 count    187.000000
 mean     245.636364
 std      121.390804
 min       85.000000
 25%      143.500000
 50%      220.000000
 75%      315.500000
 max       662.000000,
  SoftDrinkProduction
 count          187.000000
 mean          3262.609626
 std           728.357367
 min          1954.000000
 25%          2748.000000
 50%          3134.000000
 75%          3741.000000
 max          5725.000000)

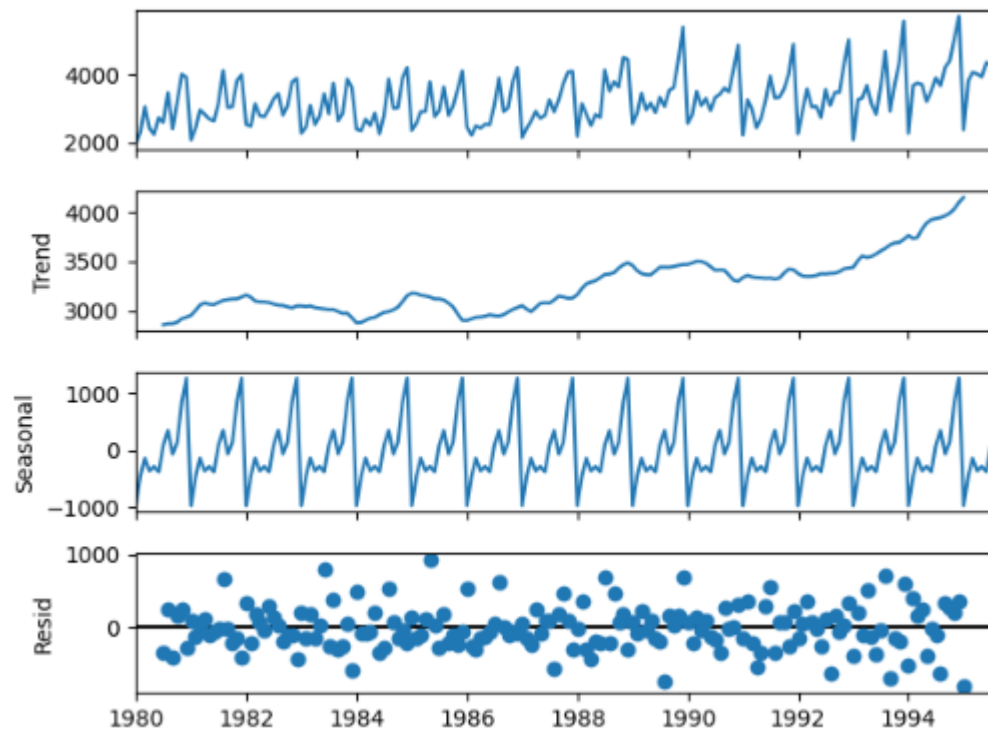
```

- Performed Decomposition

Shoe Sales data



Soft Drinks Data



2. Data Pre-processing

- Missing value treatment - Visualize the processed data - Train-test split

- Missing Values

```
Shoe_Sales    0  
dtype: int64
```

```
SoftDrinkProduction    0  
dtype: int64
```

- Train-test split

```
Shoe sales training set length: 149  
Shoe sales testing set length: 38  
Soft drink production training set length: 149  
Soft drink production testing set length: 38
```

3. Model Building - Original Data

- Build forecasting models - Linear regression - Simple Average - Moving Average - Exponential Models (Single, Double, Triple) - Check the performance of the models built

```
Mean Squared Error for Linear Regression (Shoe Sales): 29324.28073336573  
Mean Squared Error for Linear Regression (Soft Drink Production): 699768.9626079414
```

```
Mean Squared Error for Simple Average (Shoe Sales): 3809.0410922694327  
Mean Squared Error for Simple Average (Soft Drink Production): 1051067.673177358
```

```
Mean Squared Error for Moving Average (Shoe Sales): 4738.225146198831  
Mean Squared Error for Moving Average (Soft Drink Production): 945913.4970760236
```

```
Mean Squared Error for Single Exponential Smoothing (Shoe Sales): 5518.0431338778335  
Mean Squared Error for Double Exponential Smoothing (Shoe Sales): 4807.498605606696  
Mean Squared Error for Triple Exponential Smoothing (Shoe Sales): 2203.8379347829373
```

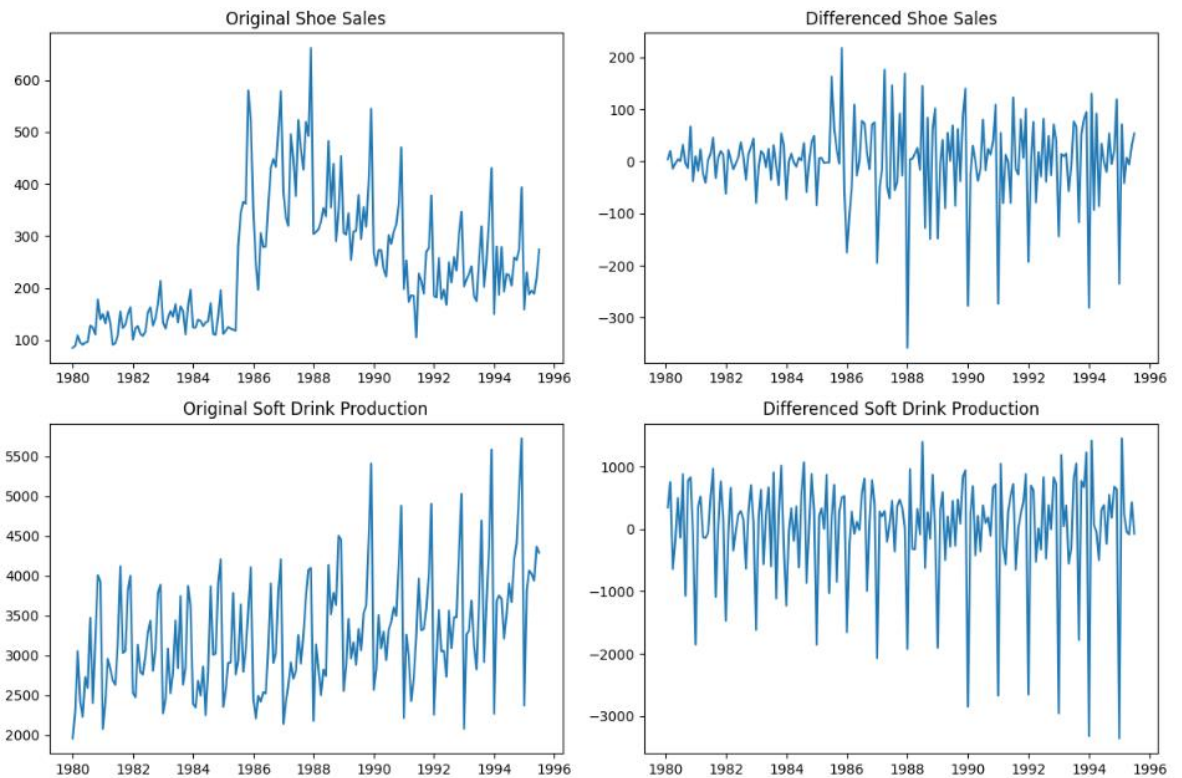
```
Mean Squared Error for Single Exponential Smoothing (Soft Drink Production): 843611.0340585107  
Mean Squared Error for Double Exponential Smoothing (Soft Drink Production): 1116604.2516877097  
Mean Squared Error for Triple Exponential Smoothing (Soft Drink Production): 282571.79270644934
```

4. Check for Stationarity

- Check for stationarity - Make the data stationary (if needed)

```
➡ Augmented Dickey-Fuller Test: Shoe Sales
ADF Test Statistic      -1.717397
p-value                 0.422172
#Lags Used              13.000000
Number of Observations Used 173.000000
Critical Value (1%)     -3.468726
Critical Value (5%)     -2.878396
Critical Value (10%)    -2.575756
=> The series is non-stationary.
Augmented Dickey-Fuller Test: Soft Drink Production
ADF Test Statistic      1.098734
p-value                 0.995206
#Lags Used              12.000000
Number of Observations Used 174.000000
Critical Value (1%)     -3.468502
Critical Value (5%)     -2.878298
Critical Value (10%)    -2.575704
=> The series is non-stationary.
```

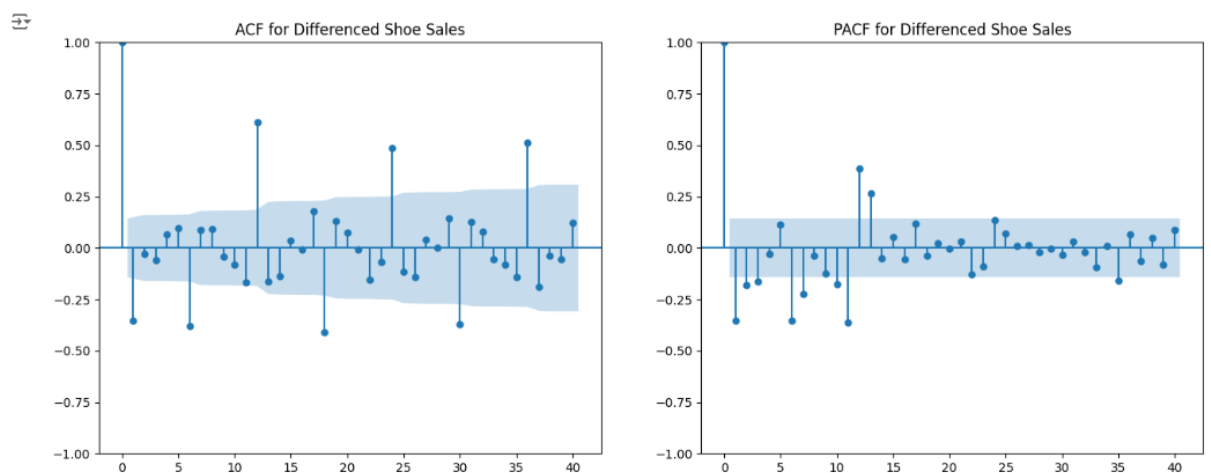
```
➡ Augmented Dickey-Fuller Test: Differenced Shoe Sales
ADF Test Statistic      -3.479160
p-value                 0.008539
#Lags Used              12.000000
Number of Observations Used 173.000000
Critical Value (1%)     -3.468726
Critical Value (5%)     -2.878396
Critical Value (10%)    -2.575756
=> The series is stationary.
Augmented Dickey-Fuller Test: Differenced Soft Drink Production
ADF Test Statistic      -9.313527e+00
p-value                 1.033701e-15
#Lags Used              1.100000e+01
Number of Observations Used 1.740000e+02
Critical Value (1%)     -3.468502e+00
Critical Value (5%)     -2.878298e+00
Critical Value (10%)    -2.575704e+00
=> The series is stationary.
```

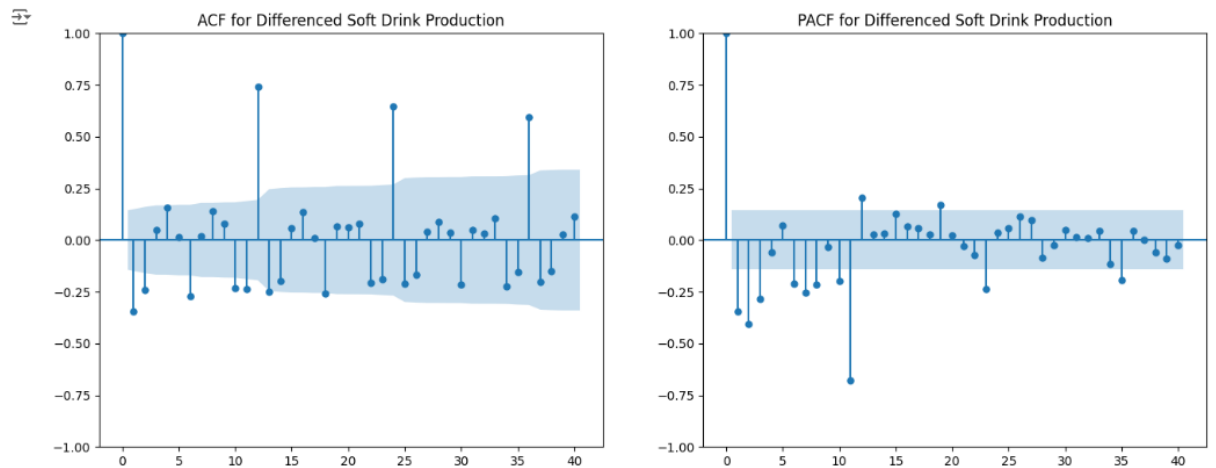


5. Model Building - Stationary Data

- Generate ACF & PACF Plot and find the AR, MA values. - Build different ARIMA models - Auto ARIMA - Manual ARIMA - Build different SARIMA models - Auto SARIMA - Manual SARIMA - Check the performance of the models built

```
➡ Mean Squared Error for Linear Regression (Differenced Shoe Sales): 7868.716214140065
   Mean Squared Error for Linear Regression (Differenced Soft Drink Production): 1321667.8618776456
```





Performing stepwise search to minimize aic

```
ARIMA(2,0,2)(0,0,0)[0]      : AIC=2128.408, Time=0.32 sec
ARIMA(0,0,0)(0,0,0)[0]      : AIC=2164.586, Time=0.03 sec
ARIMA(1,0,0)(0,0,0)[0]      : AIC=2141.667, Time=0.07 sec
ARIMA(0,0,1)(0,0,0)[0]      : AIC=2130.896, Time=0.08 sec
ARIMA(1,0,2)(0,0,0)[0]      : AIC=2126.548, Time=0.35 sec
ARIMA(0,0,2)(0,0,0)[0]      : AIC=2128.488, Time=0.21 sec
ARIMA(1,0,1)(0,0,0)[0]      : AIC=2125.556, Time=0.10 sec
ARIMA(2,0,1)(0,0,0)[0]      : AIC=2126.429, Time=0.14 sec
ARIMA(2,0,0)(0,0,0)[0]      : AIC=2137.572, Time=0.08 sec
ARIMA(1,0,1)(0,0,0)[0] intercept : AIC=2127.278, Time=0.19 sec
```

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

Total fit time: 1.596 seconds

SARIMAX Results

```
=====
Dep. Variable:          y      No. Observations:          186
Model:                 SARIMAX(1, 0, 1)  Log Likelihood      -1059.778
Date:                 Fri, 26 Jul 2024    AIC                  2125.556
Time:                 06:12:33           BIC                  2135.233
Sample:              02-01-1980         HQIC                  2129.478
                  - 07-01-1995
Covariance Type:      opg
=====
              coef    std err          z      P>|z|      [0.025     0.975]
-----
ar.L1         0.3432     0.087       3.948     0.000     0.173     0.514
ma.L1        -0.8240     0.053    -15.621     0.000    -0.927    -0.721
sigma2       5189.4172   401.925    12.911     0.000   4401.659   5977.175
=====
Ljung-Box (L1) (Q):           0.15   Jarque-Bera (JB):          30.22
Prob(Q):                     0.70   Prob(JB):              0.00
Heteroskedasticity (H):       7.56   Skew:                  0.06
Prob(H) (two-sided):          0.00   Kurtosis:              4.97
=====
```


Warnings:

[1] Covariance matrix calculated using the outer product of gradients (complex-step).
Performing stepwise search to minimize aic

```
ARIMA(2,0,2)(0,0,0)[0]      : AIC=2958.950, Time=0.37 sec
ARIMA(0,0,0)(0,0,0)[0]      : AIC=3045.878, Time=0.03 sec
ARIMA(1,0,0)(0,0,0)[0]      : AIC=3024.344, Time=0.04 sec
ARIMA(0,0,1)(0,0,0)[0]      : AIC=2961.179, Time=0.10 sec
ARIMA(1,0,2)(0,0,0)[0]      : AIC=2957.933, Time=0.21 sec
ARIMA(0,0,2)(0,0,0)[0]      : AIC=2957.763, Time=0.16 sec
ARIMA(0,0,3)(0,0,0)[0]      : AIC=2957.298, Time=0.50 sec
ARIMA(1,0,3)(0,0,0)[0]      : AIC=2950.532, Time=1.00 sec
ARIMA(2,0,3)(0,0,0)[0]      : AIC=inf, Time=1.69 sec
ARIMA(1,0,4)(0,0,0)[0]      : AIC=2956.036, Time=1.36 sec
ARIMA(0,0,4)(0,0,0)[0]      : AIC=2959.290, Time=0.67 sec
ARIMA(2,0,4)(0,0,0)[0]      : AIC=inf, Time=2.08 sec
ARIMA(1,0,3)(0,0,0)[0] intercept : AIC=inf, Time=2.99 sec
```

Best model: ARIMA(1,0,3)(0,0,0)[0]

Total fit time: 11.230 seconds

SARIMAX Results

```
=====
Dep. Variable:          y      No. Observations:          186
Model:                 SARIMAX(1, 0, 3)      Log Likelihood      -1470.266
Date:                 Fri, 26 Jul 2024      AIC                  2950.532
Time:                 06:12:44      BIC                  2966.660
Sample:              02-01-1980      HQIC                 2957.068
                  - 07-01-1995
Covariance Type:      opg
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
ar.L1          0.7297      0.079        9.186      0.000        0.574        0.885
ma.L1         -1.5614      0.115       -13.620      0.000       -1.786       -1.337
ma.L2          0.2816      0.204         1.378      0.168       -0.119         0.682
ma.L3          0.3286      0.102         3.222      0.001         0.129         0.528
sigma2        4.218e+05  4.33e+04         9.751      0.000      3.37e+05      5.07e+05
=====
Ljung-Box (L1) (Q):                0.00      Jarque-Bera (JB):                4.01
Prob(Q):                          1.00      Prob(JB):                0.13
Heteroskedasticity (H):            1.87      Skew:                    0.36
Prob(H) (two-sided):              0.02      Kurtosis:                3.10
=====
```

SARIMAX Results

```
=====
Dep. Variable:      Shoe_Sales      No. Observations:          186
Model:              ARIMA(1, 1, 1)      Log Likelihood      -1066.470
Date:              Fri, 26 Jul 2024      AIC                  2138.941
Time:              06:12:49      BIC                  2148.602
Sample:            02-01-1980      HQIC                 2142.856
                  - 07-01-1995
Covariance Type:    opg
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
ar.L1          -0.3502      0.055       -6.354      0.000       -0.458       -0.242
ma.L1          -0.9989      0.417       -2.397      0.017       -1.816       -0.182
sigma2        5769.1421  2437.014         2.367      0.018      992.683      1.05e+04
=====
Ljung-Box (L1) (Q):                0.89      Jarque-Bera (JB):                61.10
Prob(Q):                          0.35      Prob(JB):                0.00
Heteroskedasticity (H):            6.55      Skew:                    -0.78
Prob(H) (two-sided):              0.00      Kurtosis:                5.35
=====
```

```

=====
SARIMAX Results
=====
Dep. Variable:    SoftDrinkProduction    No. Observations:    186
Model:            ARIMA(1, 1, 1)         Log Likelihood       -1505.106
Date:             Fri, 26 Jul 2024       AIC                  3016.212
Time:             06:12:49              BIC                  3025.873
Sample:           02-01-1980            HQIC                 3020.127
                  - 07-01-1995
Covariance Type:    opg
=====
              coef    std err          z      P>|z|      [0.025    0.975]
-----
ar.L1         -0.3429     0.102     -3.363     0.001     -0.543     -0.143
ma.L1         -1.0000     0.084    -11.951     0.000     -1.164     -0.836
sigma2        6.641e+05    1.26e-07    5.27e+12     0.000    6.64e+05    6.64e+05
=====
Ljung-Box (L1) (Q):                3.80   Jarque-Bera (JB):                113.71
Prob(Q):                           0.05   Prob(JB):                      0.00
Heteroskedasticity (H):              1.85   Skew:                          -1.41
Prob(H) (two-sided):                0.02   Kurtosis:                      5.62
=====

```

```

ARIMA(2,0,2)(1,0,0)[12] intercept : AIC=inf, Time=3.23 sec
ARIMA(2,0,2)(2,0,1)[12] intercept : AIC=2021.372, Time=7.12 sec
ARIMA(2,0,2)(1,0,2)[12] intercept : AIC=2021.239, Time=3.26 sec
ARIMA(2,0,2)(0,0,0)[12] intercept : AIC=2130.113, Time=0.44 sec
ARIMA(2,0,2)(0,0,2)[12] intercept : AIC=inf, Time=2.70 sec
ARIMA(2,0,2)(2,0,0)[12] intercept : AIC=inf, Time=4.57 sec
ARIMA(2,0,2)(2,0,2)[12] intercept : AIC=inf, Time=3.70 sec
ARIMA(1,0,2)(1,0,1)[12] intercept : AIC=inf, Time=1.21 sec
ARIMA(2,0,1)(1,0,1)[12] intercept : AIC=inf, Time=1.42 sec
ARIMA(3,0,2)(1,0,1)[12] intercept : AIC=2017.482, Time=1.57 sec
ARIMA(3,0,2)(0,0,1)[12] intercept : AIC=inf, Time=1.67 sec
ARIMA(3,0,2)(1,0,0)[12] intercept : AIC=inf, Time=2.73 sec
ARIMA(3,0,2)(2,0,1)[12] intercept : AIC=inf, Time=5.22 sec
ARIMA(3,0,2)(1,0,2)[12] intercept : AIC=inf, Time=3.62 sec
ARIMA(3,0,2)(0,0,0)[12] intercept : AIC=inf, Time=0.79 sec
ARIMA(3,0,2)(0,0,2)[12] intercept : AIC=inf, Time=3.88 sec
ARIMA(3,0,2)(2,0,0)[12] intercept : AIC=inf, Time=4.54 sec
ARIMA(3,0,2)(2,0,2)[12] intercept : AIC=inf, Time=4.19 sec
ARIMA(3,0,1)(1,0,1)[12] intercept : AIC=2021.099, Time=1.50 sec
ARIMA(4,0,2)(1,0,1)[12] intercept : AIC=inf, Time=1.75 sec
ARIMA(3,0,3)(1,0,1)[12] intercept : AIC=inf, Time=3.95 sec
ARIMA(2,0,3)(1,0,1)[12] intercept : AIC=inf, Time=2.08 sec
ARIMA(4,0,1)(1,0,1)[12] intercept : AIC=2019.076, Time=1.62 sec
ARIMA(4,0,3)(1,0,1)[12] intercept : AIC=inf, Time=2.08 sec
ARIMA(3,0,2)(1,0,1)[12] intercept : AIC=inf, Time=1.64 sec

```

Best model: ARIMA(3,0,2)(1,0,1)[12] intercept
Total fit time: 79.830 seconds

SARIMAX Results

```

=====
Dep. Variable: y No. Observations: 186
Model: SARIMAX(3, 0, 2)x(1, 0, [1], 12) Log Likelihood -999.741
Date: Fri, 26 Jul 2024 AIC 2017.482
Time: 06:14:15 BIC 2046.513
Sample: 02-01-1980 HQIC 2029.246
- 07-01-1995

Covariance Type: opg
=====

```

	coef	std err	z	P> z	[0.025	0.975]
intercept	2.8883	6.419	0.450	0.653	-9.694	15.470
ar.L1	-2.3461	0.077	-30.392	0.000	-2.497	-2.195
ar.L2	-1.7350	0.140	-12.402	0.000	-2.009	-1.461
ar.L3	-0.3762	0.066	-5.669	0.000	-0.506	-0.246
ma.L1	1.9449	0.060	32.268	0.000	1.827	2.063
ma.L2	0.9600	0.060	16.108	0.000	0.843	1.077
ar.S.L12	0.9155	0.044	21.039	0.000	0.830	1.001
ma.S.L12	-0.5137	0.093	-5.536	0.000	-0.696	-0.332
sigma2	2484.4137	212.301	11.702	0.000	2068.311	2900.517

```

=====
Ljung-Box (L1) (Q): 0.30 Jarque-Bera (JB): 27.96
Prob(Q): 0.58 Prob(JB): 0.00
Heteroskedasticity (H): 3.88 Skew: -0.11
Prob(H) (two-sided): 0.00 Kurtosis: 4.89
=====

```

```

ARIMA(3,0,1)(2,0,2)[12] : AIC=inf, Time=nan sec
ARIMA(2,0,1)(1,0,1)[12] : AIC=2742.132, Time=1.63 sec
ARIMA(3,0,0)(1,0,1)[12] : AIC=2746.930, Time=1.26 sec
ARIMA(4,0,1)(1,0,1)[12] : AIC=inf, Time=0.91 sec
ARIMA(3,0,2)(1,0,1)[12] : AIC=inf, Time=2.20 sec
ARIMA(2,0,0)(1,0,1)[12] : AIC=2766.968, Time=0.89 sec
ARIMA(2,0,2)(1,0,1)[12] : AIC=inf, Time=2.58 sec
ARIMA(4,0,2)(1,0,1)[12] : AIC=inf, Time=3.81 sec
ARIMA(3,0,1)(1,0,1)[12] intercept : AIC=2801.119, Time=1.61 sec

```

Best model: ARIMA(3,0,1)(1,0,1)[12]
Total fit time: 149.251 seconds

SARIMAX Results

```

=====
Dep. Variable: y No. Observations: 186
Model: SARIMAX(3, 0, 1)x(1, 0, 1, 12) Log Likelihood -1359.788
Date: Fri, 26 Jul 2024 AIC 2733.576
Time: 06:16:45 BIC 2756.156
Sample: 02-01-1980 HQIC 2742.727
- 07-01-1995
Covariance Type: opg
=====

```

	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-0.0691	0.097	-0.714	0.475	-0.259	0.121
ar.L2	-0.1122	0.097	-1.158	0.247	-0.302	0.078
ar.L3	-0.0827	0.092	-0.900	0.368	-0.263	0.097
ma.L1	-0.8000	0.067	-11.949	0.000	-0.931	-0.669
ar.S.L12	0.9827	0.012	82.226	0.000	0.959	1.006
ma.S.L12	-0.6362	0.071	-8.907	0.000	-0.776	-0.496
sigma2	1.149e+05	1.01e+04	11.339	0.000	9.5e+04	1.35e+05

```

=====
Ljung-Box (L1) (Q): 0.00 Jarque-Bera (JB): 16.10
Prob(Q): 0.95 Prob(JB): 0.00
Heteroskedasticity (H): 1.79 Skew: 0.49
Prob(H) (two-sided): 0.02 Kurtosis: 4.05
=====

```

SARIMAX Results

```

=====
Dep. Variable: Shoe_Sales No. Observations: 186
Model: SARIMAX(1, 1, 1)x(1, 1, 1, 12) Log Likelihood -937.683
Date: Fri, 26 Jul 2024 AIC 1885.367
Time: 06:18:01 BIC 1901.133
Sample: 02-01-1980 HQIC 1891.763
- 07-01-1995
Covariance Type: opg
=====

```

	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-0.4703	0.051	-9.268	0.000	-0.570	-0.371
ma.L1	-0.9992	1.101	-0.907	0.364	-3.158	1.160
ar.S.L12	0.2000	0.110	1.814	0.070	-0.016	0.416
ma.S.L12	-0.7897	0.084	-9.358	0.000	-0.955	-0.624
sigma2	2722.9141	2968.725	0.917	0.359	-3095.680	8541.508

```

=====
Ljung-Box (L1) (Q): 0.14 Jarque-Bera (JB): 16.72
Prob(Q): 0.71 Prob(JB): 0.00
Heteroskedasticity (H): 1.29 Skew: -0.34
Prob(H) (two-sided): 0.33 Kurtosis: 4.36
=====

```

```

=====
SARIMAX Results
=====
Dep. Variable:          SoftDrinkProduction    No. Observations:          186
Model:                 SARIMAX(1, 1, 1)x(1, 1, 1, 12)    Log Likelihood             -1288.552
Date:                  Fri, 26 Jul 2024    AIC                       2587.103
Time:                  06:18:02    BIC                       2602.870
Sample:                02-01-1980    HQIC                      2593.500
                    - 07-01-1995

Covariance Type:      opg
=====
              coef    std err          z      P>|z|      [0.025    0.975]
-----
ar.L1          -0.4696     0.052     -9.076     0.000     -0.571     -0.368
ma.L1          -1.0000     0.097    -10.269     0.000     -1.191     -0.809
ar.S.L12       -0.0866     0.120     -0.721     0.471     -0.322     0.149
ma.S.L12       -0.6053     0.108     -5.585     0.000     -0.818     -0.393
sigma2         1.589e+05    6.13e-07    2.59e+11     0.000    1.59e+05    1.59e+05
=====
Ljung-Box (L1) (Q):                4.50    Jarque-Bera (JB):                7.78
Prob(Q):                          0.03    Prob(JB):                  0.02
Heteroskedasticity (H):            1.62    Skew:                      0.18
Prob(H) (two-sided):              0.07    Kurtosis:                  3.97
=====

```

6. Compare the performance of the models

- Compare the performance of all the models built - Choose the best model with proper rationale - Rebuild the best model using the entire data - Make a forecast for the next 12 months

```

➡ Auto ARIMA MSE (Shoe Sales): 7878.246561908018
Auto ARIMA MAE (Shoe Sales): 64.7812159022766
Manual ARIMA MSE (Shoe Sales): 7752.277289898592
Manual ARIMA MAE (Shoe Sales): 64.4821713276115
Auto SARIMA MSE (Shoe Sales): 2035.743450900802
Auto SARIMA MAE (Shoe Sales): 33.07547505620904
Manual SARIMA MSE (Shoe Sales): 8910.586056004186
Manual SARIMA MAE (Shoe Sales): 75.12293109645015
Auto ARIMA MSE (Soft Drink Production): 1336586.0037813315
Auto ARIMA MAE (Soft Drink Production): 779.3194054047856
Manual ARIMA MSE (Soft Drink Production): 1294846.7625934277
Manual ARIMA MAE (Soft Drink Production): 773.6451549565966
Auto SARIMA MSE (Soft Drink Production): 205443.38448665684
Auto SARIMA MAE (Soft Drink Production): 322.17460857838745
Manual SARIMA MSE (Soft Drink Production): 1949455.7012737866
Manual SARIMA MAE (Soft Drink Production): 1113.8720799747234

```

```

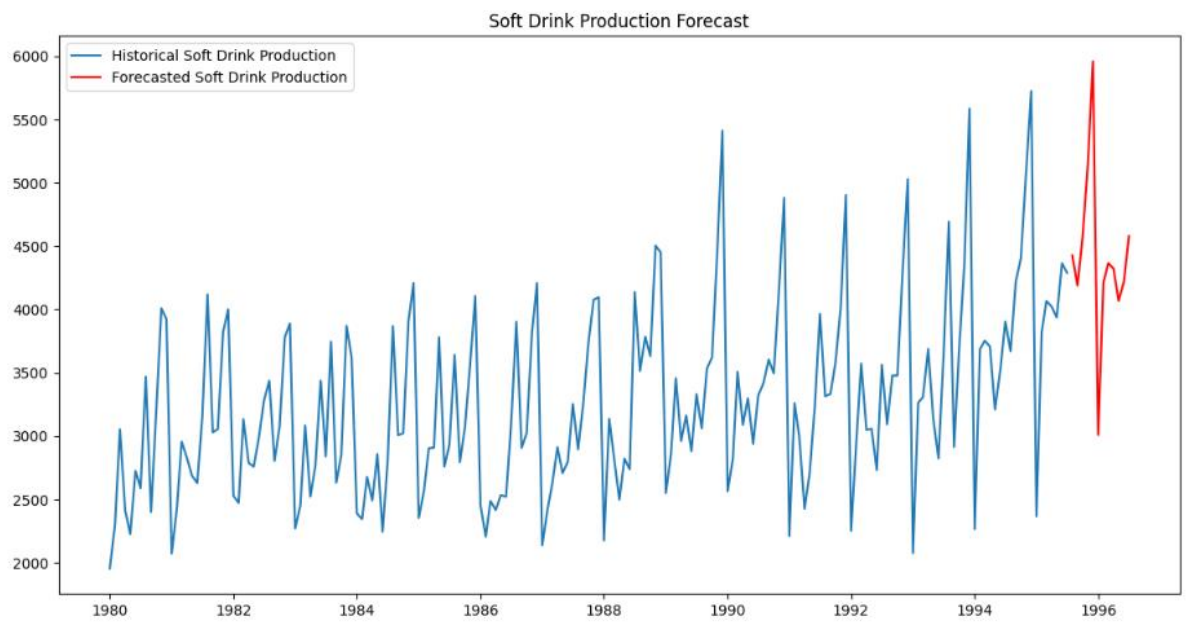
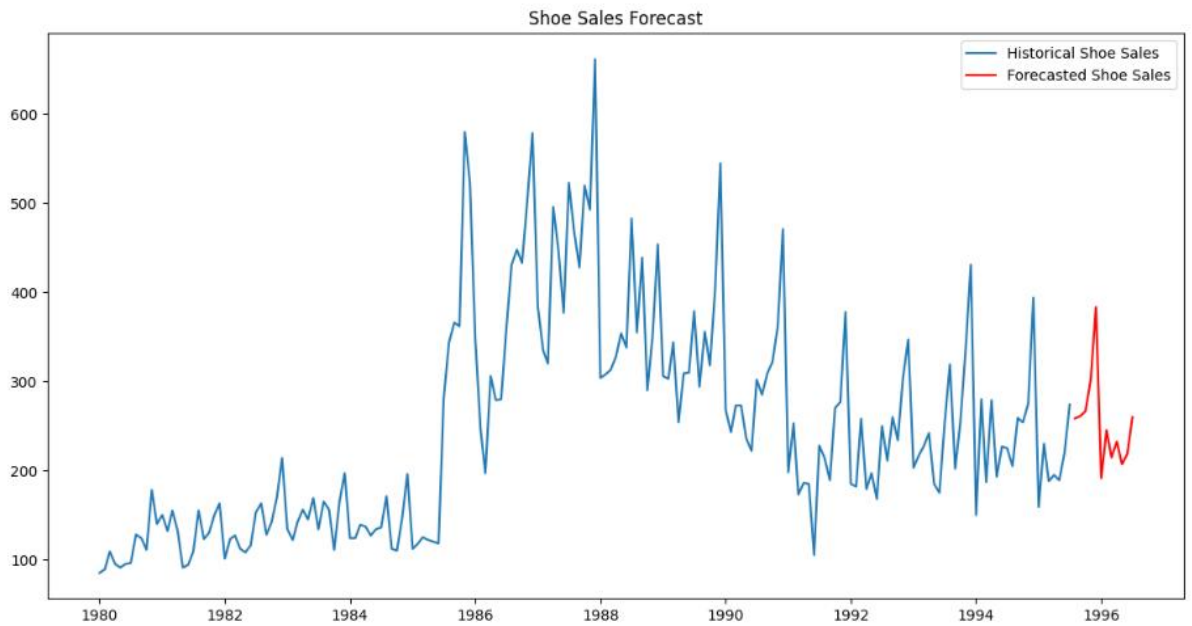
➡

```

	Model	MSE	MAE
0	Auto ARIMA (Shoe Sales)	7.878247e+03	64.781216
1	Manual ARIMA (Shoe Sales)	7.752277e+03	64.482171
2	Auto SARIMA (Shoe Sales)	2.035743e+03	33.075475
3	Manual SARIMA (Shoe Sales)	8.910586e+03	75.122931
4	Auto ARIMA (Soft Drink Production)	1.336586e+06	779.319405
5	Manual ARIMA (Soft Drink Production)	1.294847e+06	773.645155
6	Auto SARIMA (Soft Drink Production)	2.054434e+05	322.174609
7	Manual SARIMA (Soft Drink Production)	1.949456e+06	1113.872080

```
➡ Best Model for Shoe Sales:
Model    Auto SARIMA (Shoe Sales)
MSE              2035.743451
MAE              33.075475
Name: 2, dtype: object

Best Model for Soft Drink Production:
Model    Auto SARIMA (Soft Drink Production)
MSE      205443.384487
MAE      322.174609
Name: 6, dtype: object
```



7. Actionable Insights & Recommendations

- Conclude with the key takeaways (actionable insights and recommendations) for the business

Based on the analysis and forecasting of the Shoe Sales and Soft Drink Production data, here are some key takeaways and recommendations for the business:

Key Takeaways

1. Seasonal Patterns:

- Both the Shoe Sales and Soft Drink Production data exhibit clear seasonal patterns. This indicates that there are specific times of the year when sales and production peak, and other times when they decline.

2. Model Performance:

- The Auto SARIMA models were found to be the best fit for both datasets. For Shoe Sales, the Auto SARIMA model achieved an MSE of 2035.74 and an MAE of 33.08. For Soft Drink Production, the Auto SARIMA model achieved an MSE of 205443.38 and an MAE of 322.17.

3. Forecast Accuracy:

- The forecasted values for the next 12 months provide a reliable estimate of future sales and production. This can be crucial for inventory management, resource allocation, and strategic planning.

Recommendations

1. Inventory Management:

- Use the forecasted values to optimize inventory levels. Ensure that sufficient stock is available during peak sales periods to avoid stockouts and reduce inventory during off-peak periods to minimize holding costs.

2. Marketing Campaigns:

- Align marketing campaigns with the seasonal peaks identified in the data. For example, if Shoe Sales typically peak in certain months, plan marketing efforts, promotions, and discounts to coincide with these periods to maximize impact.

3. Resource Allocation:

- Allocate resources, such as labour and production capacity, in line with the forecasted production requirements. This will help in maintaining operational efficiency and meeting demand without overproducing.

4. Supply Chain Coordination:

- Coordinate with suppliers based on the forecasted demand to ensure timely delivery of raw materials and avoid any disruptions in the supply chain.

5. Monitor Model Performance:

- Regularly update and monitor the performance of the forecasting models. As new data becomes available, retrain the models to ensure they remain accurate and reflective of current trends.

6. Explore External Factors:

- Investigate external factors that might influence sales and production, such as economic conditions, consumer trends, and competitor actions. Incorporate these insights into future forecasting and planning.

Conclusion

The analysis and forecasting provide a data-driven foundation for decision-making in inventory management, marketing, resource allocation, and supply chain coordination. By leveraging these insights, the business can enhance operational efficiency, meet customer demand effectively, and ultimately drive growth and profitability.