Tecnológico de Monterrey

Instituto Tecnológico y de Estudios Superiores de Monterrey

A8-Series de tiempo no estacionarias. Tendencia

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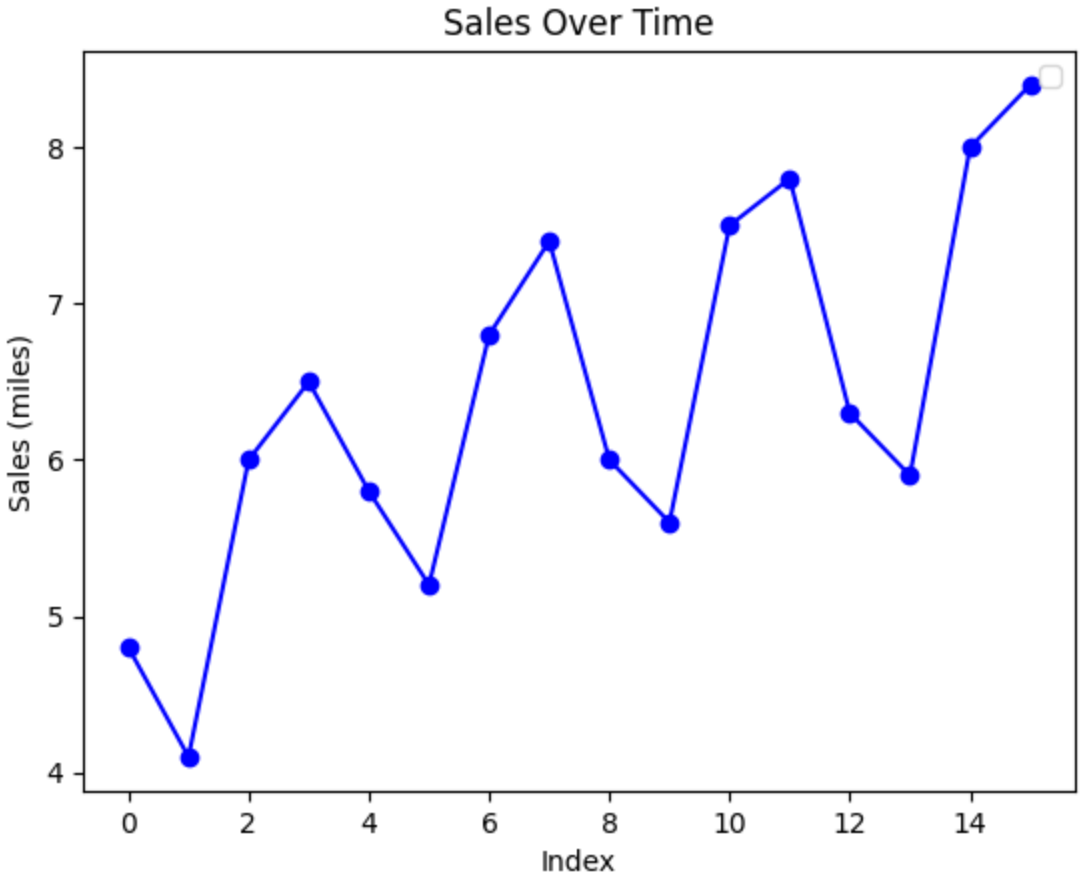
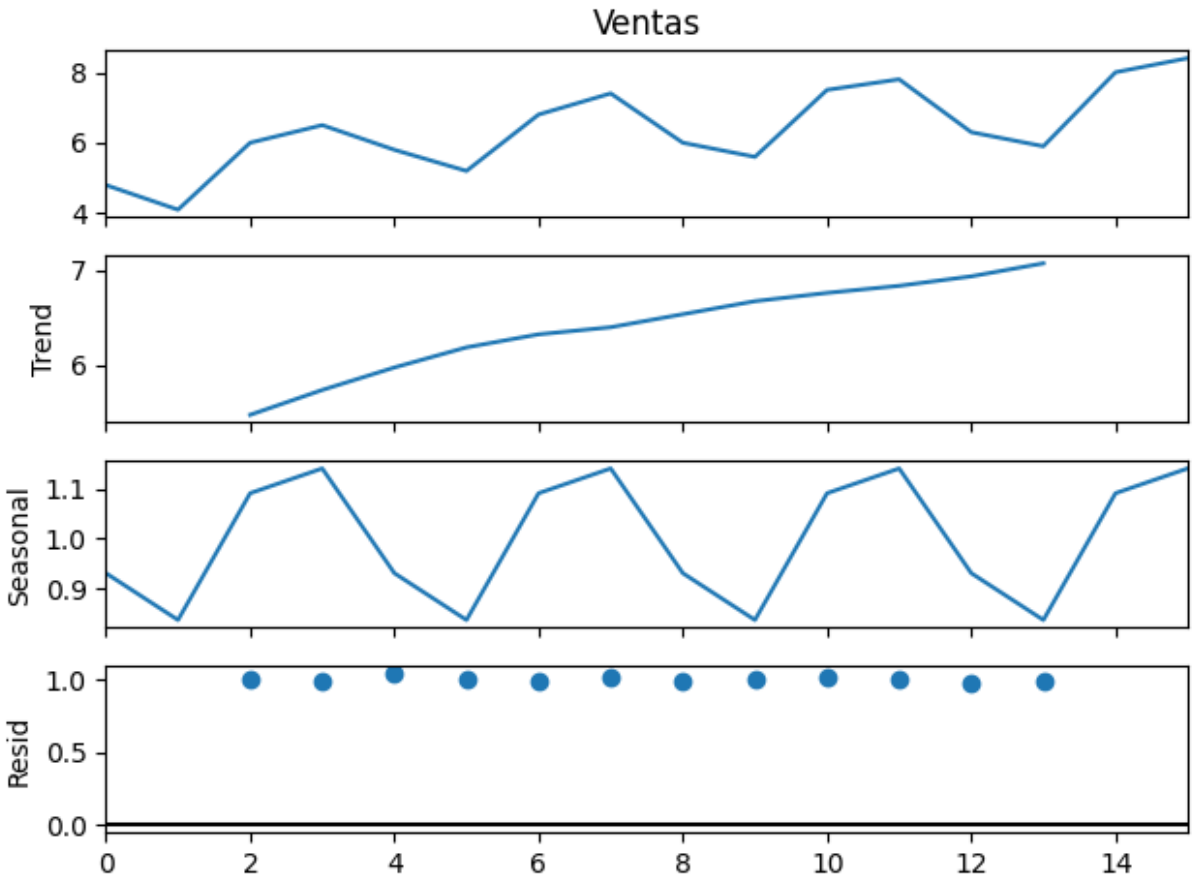
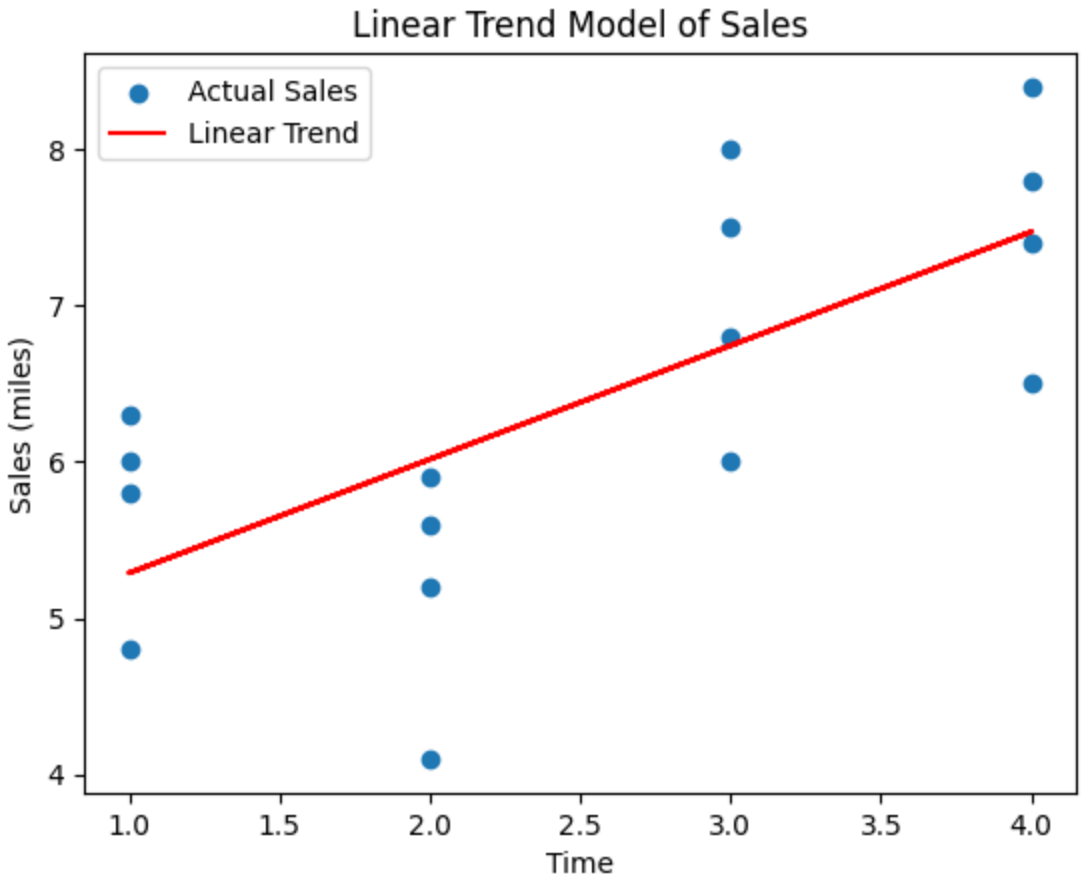
In [ ]:
```
import pandas as pd
import matplotlib.pyplot as plt
import statsmodels.api as sm

data = {
    'Año': [1, 1, 1, 1, 2, 2, 2, 2, 3, 3, 3, 3, 4, 4, 4, 4],
    'Trimestre': [1, 2, 3, 4] * 4,
    'Ventas': [4.8, 4.1, 6.0, 6.5, 5.8, 5.2, 6.8, 7.4, 6.0, 5.6, 7.5, 7.8, 6.3, 5.9
}

df = pd.DataFrame(data)

plt.plot(df.index, df['Ventas'], marker='o', linestyle='-', color='b')
plt.title('Sales Over Time')
plt.xlabel('Index')
plt.ylabel('Sales (miles)')
plt.legend()
plt.show()

result = sm.tsa.seasonal_decompose(df['Ventas'], model='multiplicative', period=4)
result.plot()
plt.show()
df['Time'] = df.groupby('Año').cumcount() + 1

X = sm.add_constant(df['Time'])
model = sm.OLS(df['Ventas'], X).fit()

plt.scatter(df['Time'], df['Ventas'], label='Actual Sales')
plt.plot(df['Time'], model.predict(X), label='Linear Trend', color='red')
plt.title('Linear Trend Model of Sales')
plt.xlabel('Time')
plt.ylabel('Sales (miles)')
plt.legend()
plt.show()

print(model.summary())

next_year_time = df['Time'].max() + 1
next_year_sales = model.predict([1, next_year_time])[0]
print(f'Predicted Sales for the Next Year: {next_year_sales}')

plt.plot(df['Time'], df['Ventas'], label='Actual Sales')
plt.scatter(next_year_time, next_year_sales, color='red', label='Predicted Sales')
plt.title('Actual Sales and Predictions Over Time')
plt.xlabel('Time')
plt.ylabel('Sales (miles)')
plt.legend()
plt.show()
```

```
WARNING:matplotlib.legend:No artists with labels found to put in legend.  Note that
artists whose label start with an underscore are ignored when legend() is called wit
h no argument.
```


Sales Over Time
Sales (miles)
Index
4
5
6
7
8
0
2
4
6
8
10
12
14


Ventas
Trend
Seasonal
Resid
4
6
8
6
7
0.9
1.0
1.1
0.0
0.5
1.0
0
2
4
6
8
10
12
14




```
/usr/local/lib/python3.10/dist-packages/scipy/stats/_stats_py.py:1806: UserWarning:
kurtosistest only valid for n>=20 ... continuing anyway, n=16
  warnings.warn("kurtosistest only valid for n>=20 ... continuing "
```

OLS Regression Results

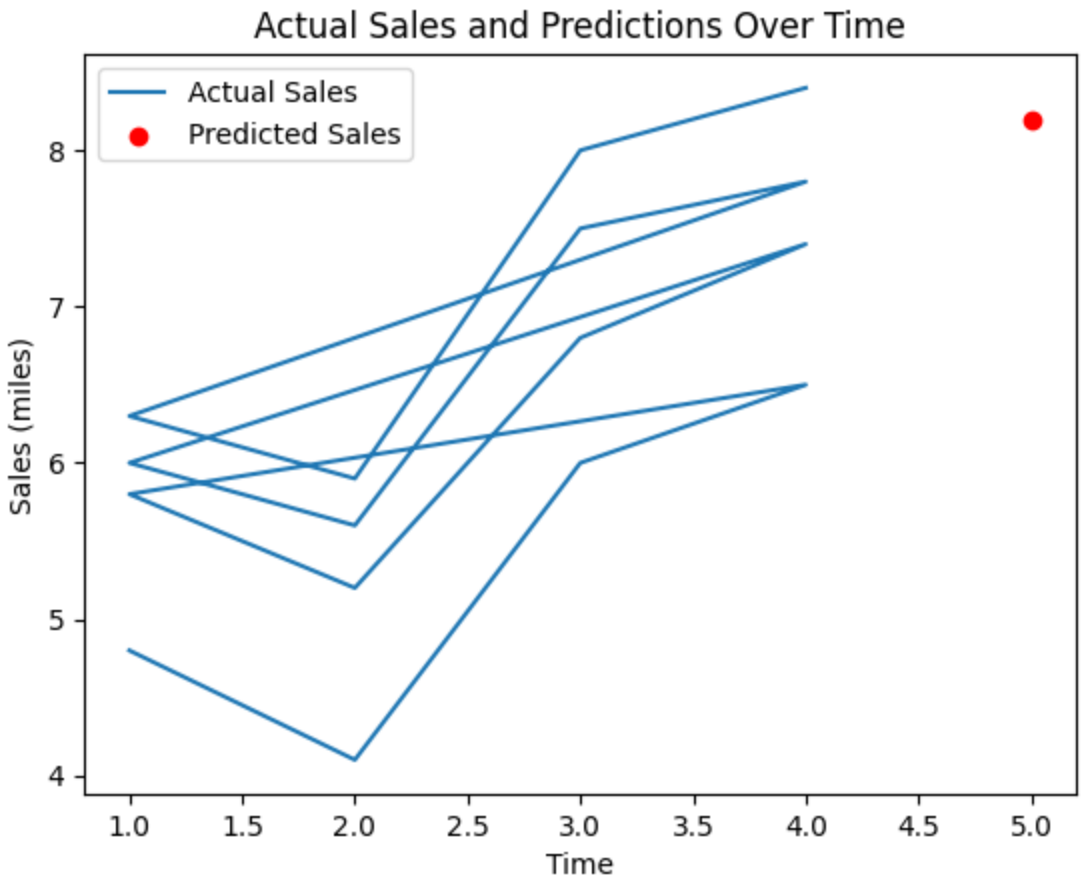
Dep. Variable:	Ventas	R-squared:	0.486
Model:	OLS	Adj. R-squared:	0.450
Method:	Least Squares	F-statistic:	13.26
Date:	Fri, 17 Nov 2023	Prob (F-statistic):	0.00267
Time:	06:20:22	Log-Likelihood:	-19.835
No. Observations:	16	AIC:	43.67
Df Residuals:	14	BIC:	45.21
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
const	4.5625	0.547	8.338	0.000	3.389	5.736
Time	0.7275	0.200	3.641	0.003	0.299	1.156

Omnibus:	0.927	Durbin-Watson:	1.375
Prob(Omnibus):	0.629	Jarque-Bera (JB):	0.697
Skew:	-0.467	Prob(JB):	0.706
Kurtosis:	2.583	Cond. No.	7.47

Notes:
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

Predicted Sales for the Next Year: 8.199999999999996



In [ ]:
```
import pandas as pd

data = {
    'Trimestre': [1, 2, 3, 4],
    '1': [1690, 940, 2625, 2500],
    '2': [1800, 900, 2900, 2360],
    '3': [1850, 1100, 2930, 2615]
}

df = pd.DataFrame(data)

df['Promedio_Movil'] = df[['1', '2', '3']].mean(axis=1)

df['Promedio_Movil_Centrado'] = df['Promedio_Movil'].rolling(window=2, center=True)

print(df[['Trimestre', '1', '2', '3', 'Promedio_Movil', 'Promedio_Movil_Centrado']]

df['Promedio_Trimestre'] = df[['1', '2', '3']].mean(axis=1)

df['Indice_Estacional'] = df['1'] / df['Promedio_Trimestre']
df['Indice_Estacional_Acumulado'] = df['Indice_Estacional'].cumsum()

max_estacional_trimestre = df.loc[df['Indice_Estacional'].idxmax(), 'Trimestre']

print(f'El mayor índice estacional se obtiene en el trimestre {max_estacional_trime
```

```
   Trimestre     1     2     3  Promedio_Movil  Promedio_Movil_Centrado
0          1  1690  1800  1850     1780.000000                      NaN
1          2   940   900  1100      980.000000              1380.000000
2          3  2625  2900  2930     2818.333333              1899.166667
3          4  2500  2360  2615     2491.666667              2655.000000
El mayor índice estacional se obtiene en el trimestre 4.
```