Retail Analysis with Walmart Data

September 3, 2022

[1]:
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
from patsy import dmatrices
import sklearn
import seaborn as sns
```

[4]:
```
walmart_data = pd.read_csv("Walmart_Store_sales.csv")
walmart_data.head()
```

[4]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment
0	1	05-02-2010	1643690.90	0	42.31	2.572	211.096358	8.106
1	1	12-02-2010	1641957.44	1	38.51	2.548	211.242170	8.106
2	1	19-02-2010	1611968.17	0	39.93	2.514	211.289143	8.106
3	1	26-02-2010	1409727.59	0	46.63	2.561	211.319643	8.106
4	1	05-03-2010	1554806.68	0	46.50	2.625	211.350143	8.106

[3]:
```
walmart_data_groupby = walmart_data.groupby('Store')['Weekly_Sales'].sum()
print("Store Number {} has maximum Sales. Sum of Total Sales {}".
 ↪format(walmart_data_groupby.idxmax
(),walmart_data_groupby.max()))
```

```
Store Number 20 has maximum Sales. Sum of Total Sales 301397792.46000004
```

[5]:
```
walmart_data_std = walmart_data.groupby('Store').agg({'Weekly_Sales':'std'})
print("Store Number {} has maximum Standard Deviation. STD {}".
 ↪format(walmart_data_std['Weekly_Sales'].
 ↪idxmax(),walmart_data_std['Weekly_Sales'].max()))
```

```
Store Number 14 has maximum Standard Deviation. STD 317569.9494755081
```

[6]:
```
walmart_data_std = walmart_data.groupby('Store').agg({'Weekly_Sales':
 ↪['mean','std']})
walmart_data_std.head()
```

[6]:

	Weekly_Sales	
	mean	std
Store		
1	1.555264e+06	155980.767761
2	1.925751e+06	237683.694682
3	4.027044e+05	46319.631557
4	2.094713e+06	266201.442297
5	3.180118e+05	37737.965745

[7]:
```
walmart_data_Q32012 = walmart_data[(pd.to_datetime(walmart_data['Date']) >= pd.
 ↪to_datetime('07-01-2012')) & (pd.to_datetime(walmart_data['Date']) <= pd.
 ↪to_datetime('09-30-2012'))]
walmart_data_growth = walmart_data_Q32012.groupby(['Store'])['Weekly_Sales'].
 ↪sum()
print("Store Number {} has Good Quartely Growth in Q3'2012 {}".
 ↪format(walmart_data_growth.idxmax(),walmart_data_growth.max()))
```

```
Store Number 4 has Good Quartely Growth in Q3'2012 25652119.35
```

[8]:
```
stores_holiday_sales = walmart_data[walmart_data['Holiday_Flag'] == 1]
```

[9]:
```
stores_nonholiday_sales = walmart_data[walmart_data['Holiday_Flag'] == 0]
```

[10]:
```
stores_holiday_sales_superBowl = stores_holiday_sales[(pd.
 ↪to_datetime(stores_holiday_sales['Date']) == pd.to_datetime('12-02-2010'))␣
 ↪|(pd.to_datetime(stores_holiday_sales['Date']) == pd.
 ↪to_datetime('11-02-2011'))|(pd.to_datetime(stores_holiday_sales['Date']) ==␣
 ↪pd.to_datetime('10-02-2012'))|(pd.to_datetime(stores_holiday_sales['Date'])␣
 ↪== pd.to_datetime('08-02-2013'))]
```

[11]:
```
stores_holiday_sales_labourDay = stores_holiday_sales[(pd.
 ↪to_datetime(stores_holiday_sales['Date']) == pd.to_datetime('10-09-2010'))␣
 ↪|(pd.to_datetime(stores_holiday_sales['Date']) == pd.
 ↪to_datetime('09-09-2011'))|(pd.to_datetime(stores_holiday_sales['Date']) ==␣
 ↪pd.to_datetime('07-09-2012'))|(pd.to_datetime(stores_holiday_sales['Date'])␣
 ↪== pd.to_datetime('06-09-2013'))]
```

[12]:
```
stores_holiday_sales_thanksgiving = stores_holiday_sales[(pd.
 ↪to_datetime(stores_holiday_sales['Date']) == pd.to_datetime('26-11-2010'))␣
 ↪|(pd.to_datetime(stores_holiday_sales['Date']) == pd.
 ↪to_datetime('25-11-2011'))|(pd.to_datetime(stores_holiday_sales['Date']) ==
pd.to_datetime('23-11-2012'))|(pd.to_datetime(stores_holiday_sales['Date']) ==␣
 ↪pd.to_datetime('29-11-2013'))]
```

[13]:
```
stores_holiday_sales_Christmas = stores_holiday_sales[(pd.
 ↪to_datetime(stores_holiday_sales['Date']) == pd.to_datetime('31-12-2010'))␣
 ↪|(pd.to_datetime(stores_holiday_sales['Date']) == pd.
 ↪to_datetime('30-12-2011'))|(pd.to_datetime(stores_holiday_sales['Date']) ==␣
 ↪pd.to_datetime('28-12-2012'))|(pd.to_datetime(stores_holiday_sales['Date'])␣
 ↪== pd.to_datetime('27-12-2013'))]
stores_nonholiday_sales_mean = stores_nonholiday_sales.groupby(['Date']).
 ↪agg({'Weekly_Sales':'mean'}).reset_index()
stores_holiday_sales_sum = stores_holiday_sales.groupby(['Date']).
 ↪agg({'Weekly_Sales':'sum'}).reset_index()
```

[14]:
```
for row in stores_holiday_sales_sum.itertuples():
    for row1 in stores_nonholiday_sales_mean.itertuples():
        if row.Weekly_Sales > row1.Weekly_Sales:
            print("On this Date {} Holiday Sales is greater than Non Holiday␣
 ↪Sales and the Sales :- {}".format(row.Date,row.Weekly_Sales))
            break;
```

```
On this Date 07-09-2012 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 48330059.31
On this Date 09-09-2011 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 46763227.529999994
On this Date 10-02-2012 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 50009407.919999994
On this Date 10-09-2010 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 45634397.84
On this Date 11-02-2011 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 47336192.79000002
On this Date 12-02-2010 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 48336677.63000002
On this Date 25-11-2011 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 66593605.25999998
On this Date 26-11-2010 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 65821003.23999999
On this Date 30-12-2011 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 46042461.04000001
On this Date 31-12-2010 Holiday Sales is greater than Non Holiday Sales and the
Sales :- 40432519.0
```

[15]:
```
print("Super Bowl Day Sale",stores_holiday_sales_superBowl['Weekly_Sales'].
 ↪sum())
print("Labour Day Sale",stores_holiday_sales_labourDay['Weekly_Sales'].sum())
print("Thanksgiving Day Sale",stores_holiday_sales_thanksgiving['Weekly_Sales'].
 ↪sum())
print("Christmas Day Sale",stores_holiday_sales_Christmas['Weekly_Sales'].sum())
```

```
Super Bowl Day Sale 145682278.34
```

```
Labour Day Sale 140727684.68
Thanksgiving Day Sale 132414608.5
Christmas Day Sale 86474980.03999999
```

[16]:
```
x_features_object = walmart_data[walmart_data['Store'] ==1][['Store','Date']]
date_obj = walmart_data[walmart_data['Store'] ==1][['Date']]
date_obj.index +=1
x_features_object.Date = date_obj.index
x_features_object.head()
```

[16]:

	Store	Date
0	1	1
1	1	2
2	1	3
3	1	4
4	1	5

[17]:
```
y_target = walmart_data[walmart_data['Store'] ==1]['Weekly_Sales']
y_target.head()
```

[17]:
```
0    1643690.90
1    1641957.44
2    1611968.17
3    1409727.59
4    1554806.68
Name: Weekly_Sales, dtype: float64
```

[18]:
```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test =␣
 ↪train_test_split(x_features_object,y_target,random_state=1)
```

[19]:
```
from sklearn.linear_model import LinearRegression
linreg = LinearRegression()
linreg.fit(x_train,y_train)
feature_dataset = walmart_data[walmart_data['Store']␣
 ↪==1][['Store','CPI','Unemployment','Fuel_Price']]
feature_dataset.head()
```

[19]:

	Store	CPI	Unemployment	Fuel_Price
0	1	211.096358	8.106	2.572
1	1	211.242170	8.106	2.548
2	1	211.289143	8.106	2.514
3	1	211.319643	8.106	2.561
4	1	211.350143	8.106	2.625

[20]:
```
response_set_cpi = walmart_data[walmart_data['Store'] ==1]['CPI'].
 ↪astype('int64')
```

```
response_set_unemployment = walmart_data[walmart_data['Store']␣
↪==1]['Unemployment'].astype('int64')
```

[21]:
```
from sklearn.model_selection import train_test_split
x_train_cpi,x_test_cpi,y_train_cpi,y_test_cpi =␣
↪train_test_split(feature_dataset,response_set_cpi,random_state=1)
x_train_unemp, x_test_unemp, y_train_unemp, y_test_unemp =␣
↪train_test_split(feature_dataset,response_set_unemployment,random_state=1)
```

[22]:
```
from sklearn.linear_model import LogisticRegression
logreg = LogisticRegression(max_iter=10000)
logreg.fit(x_train_cpi,y_train_cpi)
y_pred = logreg.predict(x_test_cpi)
logreg.fit(x_train_unemp,y_train_unemp)
#y_pred_unemp = logreg.predict(x_test_unemp)
```

[22]:
```
LogisticRegression(max_iter=10000)
```

[23]:
```
y_pred_unemp = logreg.predict(x_test_unemp)
```

[24]:
```
from sklearn import metrics
print(metrics.accuracy_score(y_test_cpi,y_pred))
print(metrics.accuracy_score(y_test_unemp,y_pred_unemp))
```

```
0.7222222222222222
0.9444444444444444
```

[25]:
```
print('cpi actual :', y_test_cpi.values[0:30])
print('cpi Predicted :', y_pred[0:30])
print('actual Unemployment :', y_test_unemp.values[0:30])
print('Predicted Unemployment :', y_pred_unemp[0:30])
```

```
cpi actual : [215 221 211 211 221 211 210 211 215 217 221 212 216 218 211 210
211 217
 215 211 212 217 221 219 214 211 211 219 215 219]
cpi Predicted : [215 221 211 211 221 211 211 211 215 215 221 211 215 218 211 211
211 217
 215 211 211 217 221 220 215 211 211 221 215 220]
actual Unemployment : [7 7 7 8 7 7 7 7 7 7 6 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
7]
Predicted Unemployment : [7 7 7 7 6 7 7 7 7 7 6 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
7 7 7]
```

[26]:
```
walmart_data['Day'] = pd.to_datetime(walmart_data['Date']).dt.day_name()
walmart_data.head()
```

[26]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment	Day
0	1	05-02-2010	1643690.90	0	42.31	2.572	211.096358	8.106	Sunday
1	1	12-02-2010	1641957.44	1	38.51	2.548	211.242170	8.106	Thursday
2	1	19-02-2010	1611968.17	0	39.93	2.514	211.289143	8.106	Friday
3	1	26-02-2010	1409727.59	0	46.63	2.561	211.319643	8.106	Friday
4	1	05-03-2010	1554806.68	0	46.50	2.625	211.350143	8.106	Monday

[ ]: