

Retail Analysis with Walmart Data Screenshots

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- Which store has maximum sales
- Which store has maximum standard deviation i.e., the sales vary a lot. Also, find out the coefficient of mean to standard deviation
- Which store/s has good quarterly growth rate in Q3'2012

```
[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

[11]: walmart = pd.read_csv("Walmart_Store_sales.csv")
walmart.head()

[11]:
```

	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

```
[12]: walmart_group = walmart.groupby('Store')['Weekly_Sales'].sum()
print("Store Number {} has maximum Sales. Sum of Total Sales {}".format(walmart_group.idxmax(),walmart_group.max()))
Store Number 20 has maximum Sales. Sum of Total Sales 301397792.46000004

[13]: walmart_std = walmart.groupby('Store').agg({'Weekly_Sales':'std'})
print("Store Number {} has maximum Standard Deviation. STD {}".format(walmart_std['Weekly_Sales'].idxmax(),walmart_std['Weekly_Sales'].max()))
Store Number 14 has maximum Standard Deviation. STD 317569.9494755081

[14]: walmart2012 = walmart[(pd.to_datetime(walmart['Date']) >= pd.to_datetime('07-01-2012')) & (pd.to_datetime(walmart['Date']) <= pd.to_datetime('09-30-2012'))]
growth = walmart2012.groupby(['Store'])['Weekly_Sales'].sum()
print("Store Number {} has Good Quartely Growth in Q3'2012 {}".format(growth.idxmax(),growth.max()))
Store Number 4 has Good Quartely Growth in Q3'2012 25652119.35
```

- Some holidays have a negative impact on sales. Find out holidays which have higher sales than the mean sales in non-holiday season for all stores together
- Provide a monthly and semester view of sales in units and give insights

```
[37]: holiday = walmart[walmart['Holiday_Flag'] == 1]
nonholiday = walmart[walmart['Holiday_Flag'] == 0]

superBowl = holiday[(pd.to_datetime(holiday['Date']) == pd.to_datetime('12-02-2010')) | (pd.to_datetime(holiday['Date']) == pd.to_datetime('11-12-2010'))]
labourDay = holiday[(pd.to_datetime(holiday['Date']) == pd.to_datetime('10-09-2010')) | (pd.to_datetime(holiday['Date']) == pd.to_datetime('09-05-2010'))]
thanksgiving = holiday[(pd.to_datetime(holiday['Date']) == pd.to_datetime('11-26-2010')) | (pd.to_datetime(holiday['Date']) == pd.to_datetime('11-23-2010'))]
christmas = holiday[(pd.to_datetime(holiday['Date']) == pd.to_datetime('12-31-2010')) | (pd.to_datetime(holiday['Date']) == pd.to_datetime('12-25-2010'))]

nonholiday_mean = nonholiday.groupby(['Date']).agg({'Weekly_Sales':'mean'}).reset_index()
holiday_sum = holiday.groupby(['Date']).agg({'Weekly_Sales':'sum'}).reset_index()

print("Super Bowl Day Sale",superBowl['Weekly_Sales'].sum())
print("Labour Day Sale",labourDay['Weekly_Sales'].sum())
print("Thanksgiving Day Sale",thanksgiving['Weekly_Sales'].sum())
print("Christmas Day Sale",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
```

For Store 1 – Build prediction models to forecast demand

- Linear Regression – Utilize variables like date and restructure dates as 1 for 5 Feb 2010 (starting from the earliest date in order). Hypothesize if CPI, unemployment, and fuel price have any impact on sales.

- Change dates into days by creating new variable.

```
[15]: x = walmart[walmart['Store'] == 1][['Store', 'Date']]
      date = walmart[walmart['Store'] == 1][['Date']]
      date.index += 1
      x.Date = date.index
      x.head()
```

```
[15]:   Store  Date
      0    1    1
      1    1    2
      2    1    3
      3    1    4
      4    1    5
```

```
[ ]:
```

```
[16]: y = walmart[walmart['Store'] == 1]['Weekly_Sales']
      y.head()
```

```
[16]: 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, y, random_state=1)
```

```
[19]: from sklearn.linear_model import LinearRegression
      reg = LinearRegression()
      reg.fit(x_train, y_train)
      linear = walmart[walmart['Store'] == 1][['Store', 'CPI', 'Unemployment', 'Fuel_Price']]
      linear.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
```

```
[22]: from sklearn.model_selection import train_test_split
      x_train_cpi, x_test_cpi, y_train_cpi, y_test_cpi = train_test_split(linear, cpi, random_state=1)
      x_train_unemp, x_test_unemp, y_train_unemp, y_test_unemp = train_test_split(linear, unemployment, random_state=1)
      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)
```

```
[23]: 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
```

```
[24]: 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 6 7 7 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 7]
```

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