# RETAIL ANALYSIS WITH WALMART DATA

```
In [11]:
#Importing the necessary libraries
import numpy as np
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.model selection import train test split
from matplotlib import style
import matplotlib.pyplot as plt
from datetime import datetime as dt
In [12]:
#reading the dataset
walmart dataset=pd.read csv('C:/Users/Administrator/Desktop/simplilearn/projects/Walmart Store sale
sv')
4
                                                                                                Þ
In [13]:
print(walmart dataset.describe().transpose())
filter=walmart_dataset['Weekly_Sales'] == walmart_dataset['Weekly_Sales'].max()
                             mean 55. 12.988182 12.954
                                             std
                                                        min
              count
                                                                     25% \
             6435.0 2.300000e+01
Store
                                                       1.000
                                                                  12.000
Weekly_Sales 6435.0 1.046965e+06 564366.622054 209986.250 553350.105
Holiday_Flag 6435.0 6.993007e-02
                                       0.255049
                                                     0.000
                                                                  0.000
Temperature 6435.0 6.066378e+01
                                       18.444933
                                                      -2.060
                                                                  47.460
             6435.0 3.358607e+00
Fuel Price
                                       0.459020
                                                      2.472
                                                                  2.933
                                                              131.735
             6435.0 1.715784e+02
                                      39.356712
                                                    126.064
                                                      3.879
Unemployment 6435.0 7.999151e+00
                                       1.875885
                                                                   6.891
                                     75%
                       50%
                 23.000000 3.400000e+01 4.500000e+01
Store
Weekly Sales 960746.040000 1.420159e+06 3.818686e+06
Holiday Flag
                 0.000000 0.000000e+00 1.000000e+00
                 62.670000 7.494000e+01 1.001400e+02 3.445000 3.735000e+00 4.468000e+00
Temperature
Fuel Price
                182.616521 2.127433e+02 2.272328e+02
CPT
                 7.874000 8.622000e+00 1.431300e+01
Unemployment
In [14]:
#store with maximum weekly sales
print(walmart dataset['Store'].where(filter).dropna())
      14.0
1905
Name: Store, dtype: float64
In [15]:
#store with maximum standard deviation
x=walmart dataset.groupby('Store').describe()
filter=x['Weekly Sales']['std']==x['Weekly Sales']['std'].max()
print(x.index.where(filter).dropna())
Float64Index([14.0], dtype='float64', name='Store')
In [16]:
##quartley growth
walmart dataset['Date']=[dt.strptime(x,'%d-%m-%Y') for x in list(walmart dataset['Date'])]
quarter 2=walmart dataset[walmart dataset['Date'].between('2012-01-01','2012-03-31')]
quarter 3=walmart dataset[walmart dataset['Date'].between('2012-04-01','2012-06-30')]
```

```
print((quarter 3.groupby('Store').describe()['Weekly Sales']
['std']-quarter 2.groupby('Store').describe()['Weekly Sales']
['std'])/quarter 2.groupby('Store').describe()['Weekly Sales']['std'])
Store
1
     -0.275527
     -0.428395
2
    -0.383404
4
    -0.344651
5
    -0.118503
     -0.175113
6
     1.009900
7
8
    -0.492255
9
     -0.371836
1.0
    -0.349078
11
     -0.264623
12
     -0.579519
    -0.290040
1.3
     0.497651
14
15
     0.111549
16
     1.306156
17
     -0.452162
    -0.344824
18
19
    0.175801
20
    -0.079545
    -0.510412
2.1
22
      0.082377
23
      0.187495
     0.104298
24
     0.563796
25
26
    -0.009503
27
     0.029331
28
    -0.279089
29
    -0.168859
30
     0.203751
31
    -0.456736
     -0.088438
32
33
      0.129064
34
     -0.403399
    -0.248750
3.5
     0.340981
36
37
      0.006670
      0.120106
38
39
      0.512016
40
      0.051050
41
     0.054635
42
     0.159690
    -0.169517
43
44
     -0.019341
45
     -0.285044
Name: std, dtype: float64
In [17]:
#holidays with higher sales
holiday sales=walmart dataset[['Store','Date','Weekly Sales','Holiday Flag']].where(walmart dataset
['Holiday Flag']==1)
non_holiday_sales=walmart_dataset[['Store','Date','Weekly_Sales','Holiday_Flag']].where(walmart_dat
aset['Holiday_Flag']==0)
holiday_sales=holiday_sales.dropna()
non_holiday_sales=non_holiday_sales.dropna()
impact=holiday_sales[holiday_sales['Weekly_Sales']>non_holiday_sales['Weekly_Sales'].mean()]
print(impact.groupby('Date').describe())
4
                                                                                                        •
            Store
                                      std min
                                                  25% 50%
                                                                 75%
            count
                        mean
                                                                        max
Date
2010-02-12 21.0 19.523810 12.035859 1.0 11.00 19.0 28.00
                                                                      41.0

      2010-09-10
      19.0
      18.736842
      11.737131
      1.0
      10.50
      19.0

      2010-11-26
      30.0
      21.533333
      12.159099
      1.0
      12.25
      21.5

                                                               26.50
                                                               30.25
                                                                      45.0
2010-12-31 16.0 17.000000 11.195237 1.0
                                                9.00 16.5 24.75 39.0
2011-02-11 20.0 18.750000 11.800424 1.0 10.75 18.5 27.25 41.0
```

```
2011-09-09 20.0 19.400000 11.802765 1.0 10.75 19.5 27.25 41.0
2011-11-25 31.0 21.387097 11.982424 1.0 12.50 21.0 29.50 45.0

      2011-12-30
      19.0
      18.789474
      12.122426
      1.0
      10.50
      19.0
      27.50

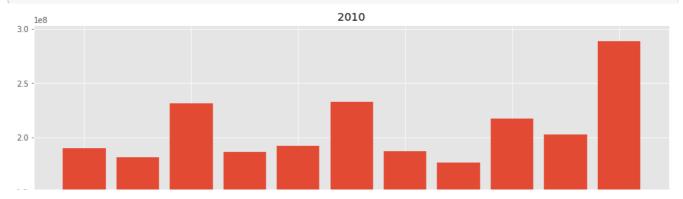
      2012-02-10
      22.0
      19.772727
      11.767315
      1.0
      11.25
      19.5
      27.75

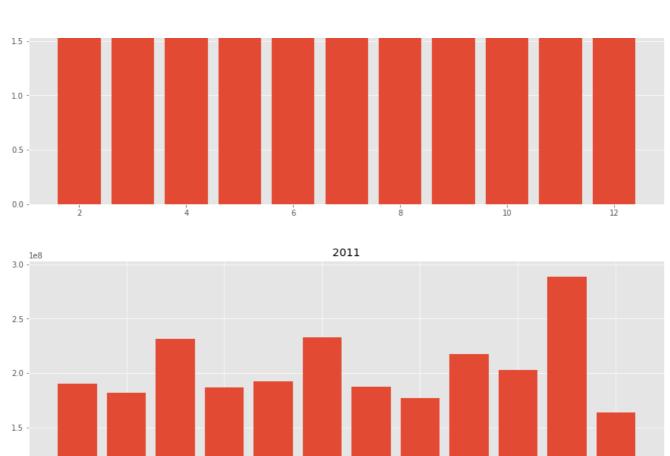
                                                                    41.0
                                                                    41.0
2012-09-07 22.0 20.272727 12.064043 1.0 11.50 19.5 27.75 41.0
           Weekly Sales
                                                        75%
                   count
                                  mean ...
                                                                    max
Date
                                         . . .
                                         ... 1.745363e+06 2188307.39
                    21.0 1.585199e+06
2010-02-12
                   19.0 1.529971e+06 ... 1.805636e+06 2191767.76
2010-09-10
2010-11-26
                   30.0 1.902568e+06 ... 2.237928e+06 2939946.38
2010-12-31
                   16.0 1.435490e+06 ... 1.683294e+06 1799737.79
                          1.573433e+06 ... 1.953430e+06 2211388.14
1.554863e+06 ... 1.779231e+06 2202742.90
2011-02-11
                    20.0
                    20.0
2011-09-09
                   31.0 1.891140e+06 ... 2.294322e+06 3004702.33
2011-11-25
2011-12-30
                   19.0 1.550523e+06 ... 1.894188e+06 2043245.00
                    22.0 1.619000e+06 ... 2.002583e+06 2462978.28
2012-02-10
2012-09-07
                    22.0 1.552976e+06 ... 1.807787e+06 2165796.31
           Holiday Flag
                   count mean std min 25% 50% 75% max
Date
2010-02-12
                    21.0 1.0 0.0 1.0 1.0 1.0 1.0 1.0
2010-09-10
                    19.0 1.0
                               0.0
                                    1.0
                                          1.0
                                               1.0
                                                    1.0
                                                          1.0
                    30.0 1.0
                                          1.0 1.0
2010-11-26
                               0.0
                                    1.0
                                                    1.0
                                                          1.0
                   16.0 1.0 0.0 1.0 1.0 1.0 1.0 1.0
2010-12-31
2011-02-11
                    20.0 1.0 0.0 1.0 1.0 1.0 1.0 1.0
2011-09-09
                    20.0 1.0 0.0 1.0 1.0 1.0 1.0 1.0
2011-11-25
                    31.0 1.0
                               0.0 1.0 1.0 1.0 1.0 1.0
2011-12-30
                    19.0
                          1.0
                               0.0
                                    1.0
                                          1.0 1.0
                                                    1.0
                    22.0 1.0 0.0 1.0 1.0 1.0 1.0 1.0
2012-02-10
                    22.0 1.0 0.0 1.0 1.0 1.0 1.0 1.0
2012-09-07
[10 rows x 24 columns]
In [25]:
#converting string date to date format
days = [(x - walmart dataset['Date'][0]) for x in list(walmart_dataset['Date'])]
walmart dataset['Week']=[np.timedelta64(x, 'D').astype(int)/7+1 for x in days]
print (walmart_dataset['Week'])
0
         1.0
1
          2.0
          3 0
2
3
          4.0
          5.0
4
          6.0
5
          7.0
6
7
         8.0
8
          9.0
9
         10.0
1.0
         11.0
11
        12.0
12
        13.0
13
         14.0
         15.0
14
1.5
         16.0
16
        17.0
17
        18.0
18
         19.0
19
         20.0
20
         21.0
         22.0
21
22
         23.0
23
         24.0
24
         25.0
25
         26.0
26
         27.0
27
         28.0
28
         29.0
         30.0
29
         . . .
6405
        114.0
```

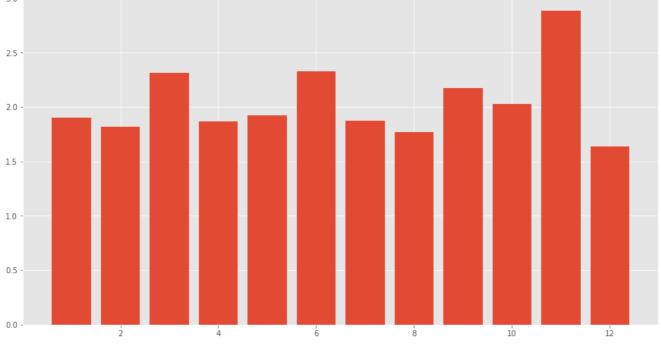
```
6406
       115.0
6407
       116.0
       117.0
6408
6409
       118.0
6410
        119.0
6411
        120.0
6412
        121.0
6413
       122.0
6414
       123.0
6415
       124.0
6416
        125.0
6417
        126.0
        127.0
6418
6419
       128.0
6420
       129.0
6421
        130.0
6422
        131.0
       132.0
6423
6424
       133.0
6425
       134.0
       135.0
6426
6427
        136.0
6428
        137.0
6429
       138.0
6430
       139.0
6431
       140.0
6432
        141.0
6433
        142.0
6434
        143.0
Name: Week, Length: 6435, dtype: float64
```

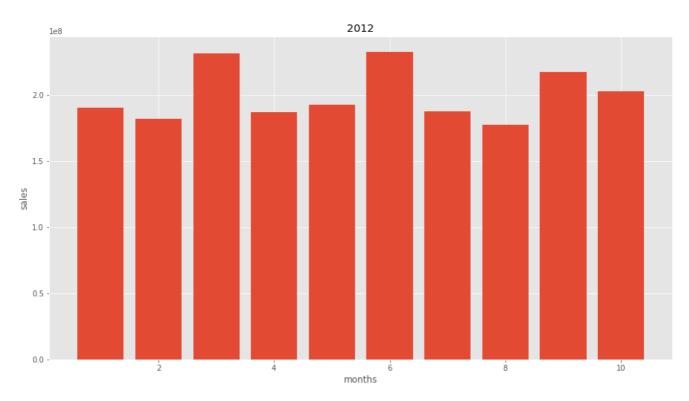
# In [334]:

```
#monthly insights of each year
style.use('ggplot')
plt.figure(figsize=[15,8])
month_plotting=pd.DataFrame()
month plotting['month'] = list(x.month for x in walmart dataset['Date'])
month_plotting['year']=list(x.year for x in walmart_dataset['Date'])
month_plotting['Weekly_Sales']=walmart_dataset['Weekly_Sales']
x=month plotting.groupby(['year','month']).sum()
x=x.reset_index(level=0,drop=True)
#print(x['Weekly_Sales'][23:33])
months=np.array(y for y in range(1,13))
plt.bar(x['Weekly_Sales'][:11].index,x['Weekly_Sales'][:11].values)
plt.title('2010')
plt.xlabel('months')
plt.ylabel('sales')
plt.show()
plt.figure(figsize=[15,8])
plt.bar(x['Weekly Sales'][11:23].index,x['Weekly Sales'][:12].values)
plt.title('2011')
plt.xlabel('months')
plt.ylabel('sales')
plt.show()
plt.figure(figsize=[15,8])
plt.bar(x['Weekly_Sales'][23:33].index,x['Weekly_Sales'][:10].values)
plt.title('2012')
plt.xlabel('months')
plt.ylabel('sales')
plt.show()
```



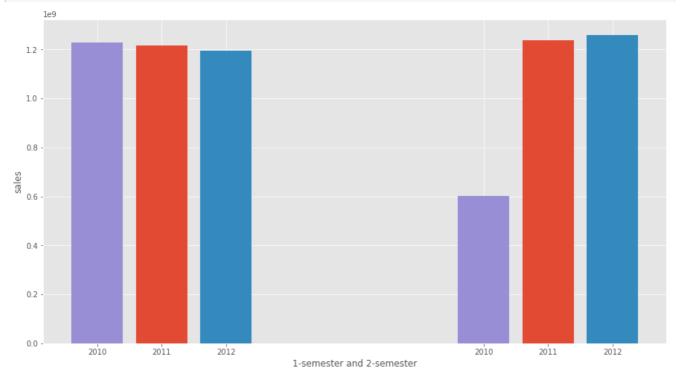






In [488]:

```
semester['month'] = np.array([6,12,6,12,6,12])
semester['sales'] = pd.DataFrame(x.iloc[i:i+6].sum().values for i in range(0,33,6))
semester.set_index('month',inplace=True)
plt.figure(figsize=[15,8])
plt.bar(semester[:2].index.values+1, (semester[:2].values.reshape(2)))
plt.bar(semester[2:4].index.values+2, (semester[2:4].values.reshape(2)))
plt.bar(semester[4:6].index.values, (semester[4:6].values.reshape(2)))
plt.xticks([6,7,8,12,13,14],[2010,2011,2012,2010,2011,2012])
plt.xlabel("1-semester and 2-semester")
plt.ylabel("sales")
plt.show()
```



### In [495]:

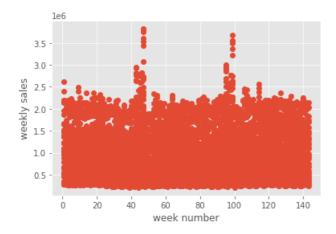
```
#assigning features and label
X_feature=walmart_dataset.drop(['Weekly_Sales','Date','Fuel_Price'],axis=1)
y_label=walmart_dataset['Weekly_Sales']
```

### In [496]:

```
plt.plot(X_feature['Week'],y_label,'o')
plt.xlabel("week number")
plt.ylabel("weekly sales")
```

### Out[496]:

Text(0, 0.5, 'weekly sales')



# In [497]:

```
#using the linear regression algorithm
linreg=LinearRegression()
```

### In [498]:

```
#training and testing the data
X_train,X_test,y_train,y_test=train_test_split(X_feature,y_label)
linreg.fit(X_train,y_train)
```

### Out[498]:

LinearRegression(copy X=True, fit intercept=True, n jobs=None, normalize=False)

### In [499]:

```
#the predicted values
y_predict=linreg.predict(X_test)
print(y_predict)
```

[1206828.25653116 811432.45265622 1183192.92186963 ... 916171.40195039 1167210.16087863 947836.2525009 ]

# In [500]:

```
#root mean square value of the algorithm
print(np.sqrt(((y_predict-y_test) ** 2).mean()))
```

513613.17274759343