Rohit Bhise

RETAIL ANALYSIS WITH WALMART PROJECT

In [4]: #

#Import necessary libraries

import pandas as pd
import seaborn as sns
import numpy as np

import matplotlib.pyplot as plt
from matplotlib import dates
from datetime import datetime

In [5]:

#Load dataset

data=pd.read_csv("F:/Data science with Python/walmart_store_sales.csv")
data

Out[5]:		Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	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
	•••								
	6430	45	28- 09- 2012	713173.95	0	64.88	3.997	192.013558	8.684
	6431	45	05- 10- 2012	733455.07	0	64.89	3.985	192.170412	8.667
	6432	45	12- 10- 2012	734464.36	0	54.47	4.000	192.327265	8.667
	6433	45	19- 10- 2012	718125.53	0	56.47	3.969	192.330854	8.667

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	Unemployment
6434	45	26- 10- 2012	760281.43	0	58.85	3.882	192.308899	8.667

6435 rows × 8 columns

```
In [6]:
        #Convert date to datetime format and show dataset information
         data['Date'] = pd.to_datetime(data['Date'])
         data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 6435 entries, 0 to 6434
        Data columns (total 8 columns):
         #
             Column
                         Non-Null Count Dtype
            -----
                          -----
        ---
                         6435 non-null int64
         0
             Store
         1
             Date
                          6435 non-null datetime64[ns]
         2
            Weekly_Sales 6435 non-null float64
         3
            Holiday_Flag 6435 non-null int64
            Temperature 6435 non-null float64
         4
         5
             Fuel_Price
                          6435 non-null float64
                          6435 non-null float64
         6
             CPI
             Unemployment 6435 non-null float64
        dtypes: datetime64[ns](1), float64(5), int64(2)
        memory usage: 402.3 KB
In [7]:
        #Checking for missing values
         data.isnull().sum()
                       0
        Store
Out[7]:
        Date
                       0
        Weekly_Sales
                       0
        Holiday_Flag
                       0
        Temperature
        Fuel Price
        CPI
                       0
        Unemployment
        dtype: int64
In [8]:
        #Splitting Date and create new columns (Day, Month, and Year)
         data["Day"]= pd.DatetimeIndex(data['Date']).day
         data['Month'] = pd.DatetimeIndex(data['Date']).month
         data['Year'] = pd.DatetimeIndex(data['Date']).year
         data
```

Out[8]:		Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	Unemploymen
	0	1	2010- 05-02	1643690.90	0	42.31	2.572	211.096358	8.10
	1	1	2010- 12-02	1641957.44	1	38.51	2.548	211.242170	8.10
	2	1	2010- 02-19	1611968.17	0	39.93	2.514	211.289143	8.10

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	Unemploymen
3	1	2010- 02-26	1409727.59	0	46.63	2.561	211.319643	8.10
4	1	2010- 05-03	1554806.68	0	46.50	2.625	211.350143	8.10
•••								
6430	45	2012- 09-28	713173.95	0	64.88	3.997	192.013558	8.68
6431	45	2012- 05-10	733455.07	0	64.89	3.985	192.170412	8.66
6432	45	2012- 12-10	734464.36	0	54.47	4.000	192.327265	8.66
6433	45	2012- 10-19	718125.53	0	56.47	3.969	192.330854	8.66
6434	45	2012- 10-26	760281.43	0	58.85	3.882	192.308899	8.66

6435 rows × 11 columns



Q1: Which store has minimum and maximum sales?

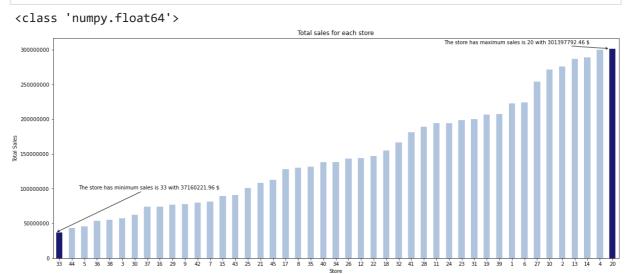
```
In [9]:
         plt.figure(figsize=(20,8))
         #Sum Weekly Sales for each store, then sortded by total sales
         total_sales_for_each_store = data.groupby('Store')['Weekly_Sales'].sum().sort_values
         total sales for each store array = np.array(total sales for each store) # convert to
         #Assigning a specific color for the stores have the lowest and highest sales
         clrs = ['lightsteelblue' if ((x < max(total_sales_for_each_store_array)) and</pre>
                                       (x > min(total sales for each store array)))
                 else 'midnightblue' for x in total sales for each store array]
         ax = total sales for each store.plot(kind='bar',color=clrs);
         #Store have minimum sales
         p = ax.patches[0]
         print(type(p.get_height()))
         ax.annotate("The store has minimum sales is 33 with {0:.2f} $".format((p.get height(
                     xy=(p.get_x(), p.get_height()), xycoords='data',
                     xytext=(0.17, 0.32), textcoords='axes fraction',
                     arrowprops=dict(arrowstyle="->", connectionstyle="arc3"),
                     horizontalalignment='center', verticalalignment='center')
```

```
#Store have maximum sales

p = ax.patches[44]
ax.annotate("The store has maximum sales is 20 with {0:.2f} $".format((p.get_height(), xy=(p.get_x(), p.get_height()), xycoords='data', xytext=(0.82, 0.98), textcoords='axes fraction', arrowprops=dict(arrowstyle="->", connectionstyle="arc3"), horizontalalignment='center', verticalalignment='center')

#Plot properties

plt.xticks(rotation=0)
plt.ticklabel_format(useOffset=False, style='plain', axis='y')
plt.title('Total sales for each store')
plt.xlabel('Store')
plt.ylabel('Total Sales');
```

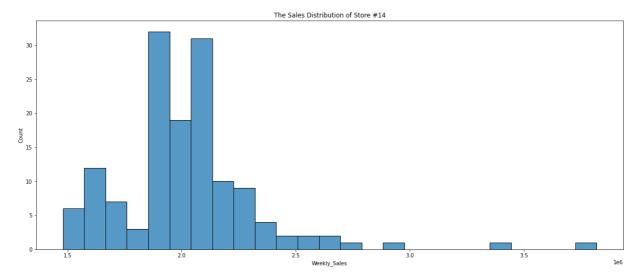


Q2: Which store has maximum standard deviation i.e., the sales vary a lot. Also, find out the coefficient of mean to standard deviation?

The store has maximum standard deviation is 14 with 317570 \$

```
In [21]: #Distribution of store has maximum standard deviation

plt.figure(figsize=(20,8))
    sns.histplot(data[data['Store'] == data_std.head(1).index[0]]['Weekly_Sales'])
    plt.title('The Sales Distribution of Store #'+ str(data_std.head(1).index[0]));
```



In [22]:

#Coefficient of mean to standard deviation

coef_mean_std = pd.DataFrame(data.groupby('Store')['Weekly_Sales'].std() / data.grou
coef_mean_std = coef_mean_std.rename(columns={'Weekly_Sales':'Coefficient of mean to
coef_mean_std

Out [22]: Coefficient of mean to standard deviation

-	
Store	
1	0.100292
2	0.123424
3	0.115021
4	0.127083
5	0.118668
6	0.135823
7	0.197305
8	0.116953
9	0.126895
10	0.159133
11	0.122262
12	0.137925
13	0.132514
14	0.157137
15	0.193384
16	0.165181
17	0.125521
18	0.162845
19	0.132680
20	0.130903
21	0.170292

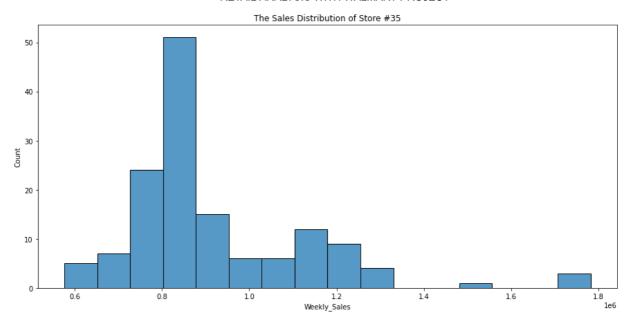
Coefficient of mean to standard deviation

Store	
22	0.156783
23	0.179721
24	0.123637
25	0.159860
26	0.110111
27	0.135155
28	0.137330
29	0.183742
30	0.052008
31	0.090161
32	0.118310
33	0.092868
34	0.108225
35	0.229681
36	0.162579
37	0.042084
38	0.110875
39	0.149908
40	0.123430
41	0.148177
42	0.090335
43	0.064104
44	0.081793
45	0.165613

In [23]:

#Distribution of store has maximum coefficient of mean to standard deviation

```
coef_mean_std_max = coef_mean_std.sort_values(by='Coefficient of mean to standard de
plt.figure(figsize=(15,7))
sns.histplot(data[data['Store'] == coef_mean_std_max.tail(1).index[0]]['Weekly_Sales
plt.title('The Sales Distribution of Store #'+str(coef_mean_std_max.tail(1).index[0])
```

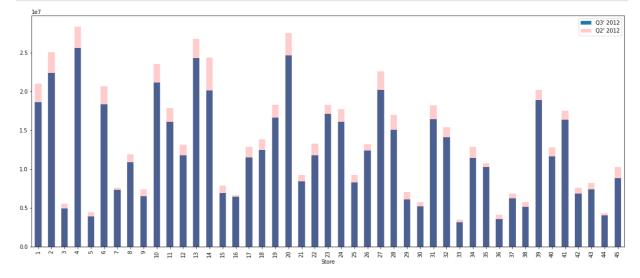


Q3 Which store/s has good quarterly growth rate in Q3'2012

```
plt.figure(figsize=(20,8))
# Sales for third quarterly in 2012
Q3 = data[(data['Date'] > '2012-07-01') & (data['Date'] < '2012-09-30')].groupby('St

# Sales for second quarterly in 2012
Q2 = data[(data['Date'] > '2012-04-01') & (data['Date'] < '2012-06-30')].groupby('St

# Plotting the difference between sales for second and third quarterly
Q2.plot(ax=Q3.plot(kind='bar',legend=True),kind='bar',color='r',alpha=0.2,legend=True)
plt.legend(["Q3' 2012", "Q2' 2012"]);</pre>
```



```
In [29]: #Store/s has good quarterly growth rate in Q3'2012 - .sort_values(by='Weekly_Sales')
print('Store have good quarterly growth rate in Q3'2012 is Store '+str(Q3.idxmax())+
```

Store have good quarterly growth rate in Q3'2012 is Store 4 With 25652119.35 \$

Q4: 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?

Holiday Events:

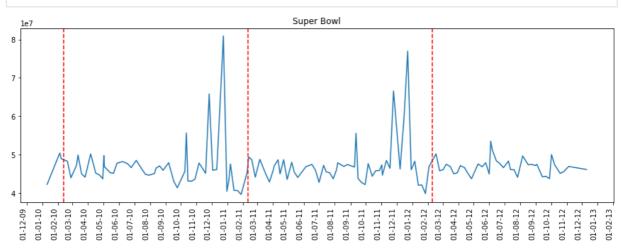
Super Bowl: 12-Feb-10, 11-Feb-11, 10-Feb-12, 8-Feb-13

Labour Day: 10-Sep-10, 9-Sep-11, 7-Sep-12, 6-Sep-13

Thanksgiving: 26-Nov-10, 25-Nov-11, 23-Nov-12, 29-Nov-13

Christmas: 31-Dec-10, 30-Dec-11, 28-Dec-12, 27-Dec-13

```
In [30]:
          def plot_line(df,holiday_dates,holiday_label):
              fig, ax = plt.subplots(figsize = (15,5))
              ax.plot(df['Date'],df['Weekly_Sales'],label=holiday_label)
              for day in holiday_dates:
                  day = datetime.strptime(day, '%d-%m-%Y')
                  plt.axvline(x=day, linestyle='--', c='r')
              plt.title(holiday label)
              x_dates = df['Date'].dt.strftime('%Y-%m-%d').sort_values().unique()
              xfmt = dates.DateFormatter('%d-%m-%y')
              ax.xaxis.set_major_formatter(xfmt)
              ax.xaxis.set_major_locator(dates.DayLocator(1))
              plt.gcf().autofmt_xdate(rotation=90)
              plt.show()
          total_sales = data.groupby('Date')['Weekly_Sales'].sum().reset index()
          Super_Bowl =['12-2-2010', '11-2-2011', '10-2-2012']
          Labour_Day = ['10-9-2010', '9-9-2011', '7-9-2012']
          Thanksgiving = ['26-11-2010', '25-11-2011', '23-11-2012']
          Christmas = ['31-12-2010', '30-12-2011', '28-12-2012']
          plot_line(total_sales,Super_Bowl,'Super Bowl')
          plot line(total sales, Labour Day, 'Labour Day')
          plot line(total sales, Thanksgiving, 'Thanksgiving')
          plot line(total sales, Christmas, 'Christmas')
```





The sales increased during thanksgiving. And the sales decreased during christmas.

In [31]: data.loc[data.Date.isin(Super_Bowl)]

Out[31]:		Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	Unemploymen
	1	1	2010- 12-02	1641957.44	1	38.51	2.548	211.242170	8.10
				1649614.93		36.39	3.022	212.936705	7.74
	105	1	2012- 10-02	1802477.43	1	48.02	3.409	220.265178	7.34
	144	2	2010- 12-02	2137809.50	1	38.49	2.548	210.897994	8.32

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	Unemploymen
196	2	2011- 11-02	2168041.61	1	33.19	3.022	212.592862	8.02
•••								
6202	44	2011- 11-02	307486.73	1	30.83	3.034	127.859129	7.22
6254	44	2012- 10-02	325377.97	1	33.73	3.116	130.384903	5.77
6293	45	2010- 12-02	656988.64	1	27.73	2.773	181.982317	8.99
6345	45	2011- 11-02	766456.00	1	30.30	3.239	183.701613	8.54
6397	45	2012- 10-02	803657.12	1	37.00	3.640	189.707605	8.42

135 rows × 11 columns



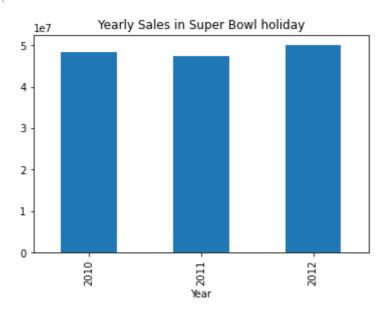
In [32]:

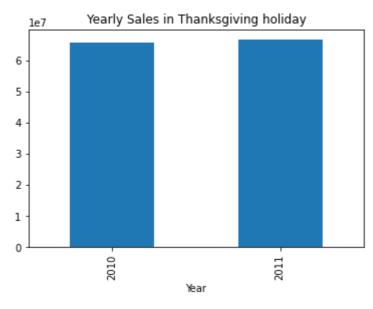
#Yearly Sales in holidays

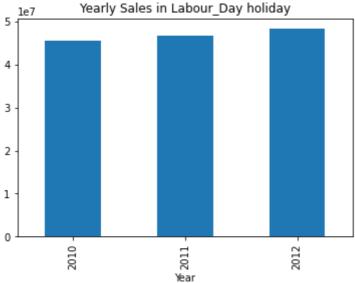
Super_Bowl_df = pd.DataFrame(data.loc[data.Date.isin(Super_Bowl)].groupby('Year')['W
Thanksgiving_df = pd.DataFrame(data.loc[data.Date.isin(Thanksgiving)].groupby('Year')
Labour_Day_df = pd.DataFrame(data.loc[data.Date.isin(Labour_Day)].groupby('Year')['W
Christmas_df = pd.DataFrame(data.loc[data.Date.isin(Christmas)].groupby('Year')['Wee

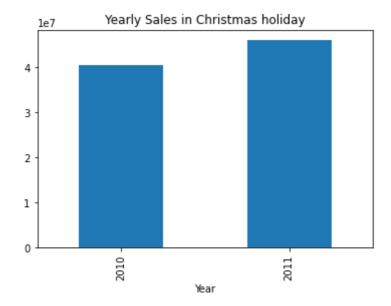
Super_Bowl_df.plot(kind='bar',legend=False,title='Yearly Sales in Super Bowl holiday Thanksgiving_df.plot(kind='bar',legend=False,title='Yearly Sales in Thanksgiving hol Labour_Day_df.plot(kind='bar',legend=False,title='Yearly Sales in Labour_Day holiday Christmas_df.plot(kind='bar',legend=False,title='Yearly Sales in Christmas holiday')

Out[32]: <AxesSubplot:title={'center':'Yearly Sales in Christmas holiday'}, xlabel='Year'>









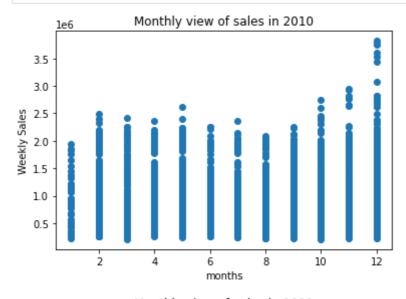
Q5: Provide a monthly and semester view of sales in units and give insights

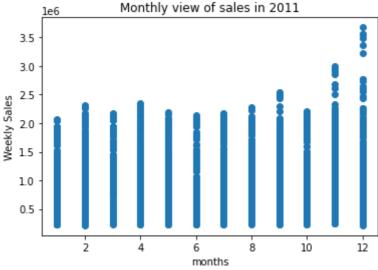
```
In [33]: #Monthly view of sales for each years
    plt.scatter(data[data.Year==2010]["Month"],data[data.Year==2010]["Weekly_Sales"])
```

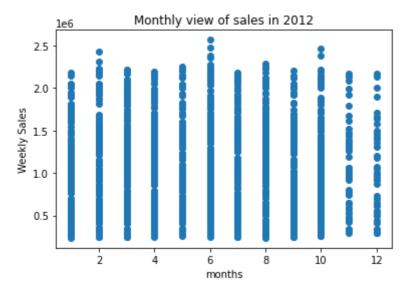
```
plt.xlabel("months")
plt.ylabel("Weekly Sales")
plt.title("Monthly view of sales in 2010")
plt.show()

plt.scatter(data[data.Year==2011]["Month"],data[data.Year==2011]["Weekly_Sales"])
plt.xlabel("months")
plt.ylabel("Weekly Sales")
plt.title("Monthly view of sales in 2011")
plt.show()

plt.scatter(data[data.Year==2012]["Month"],data[data.Year==2012]["Weekly_Sales"])
plt.xlabel("months")
plt.ylabel("Weekly Sales")
plt.title("Monthly view of sales in 2012")
plt.show()
```



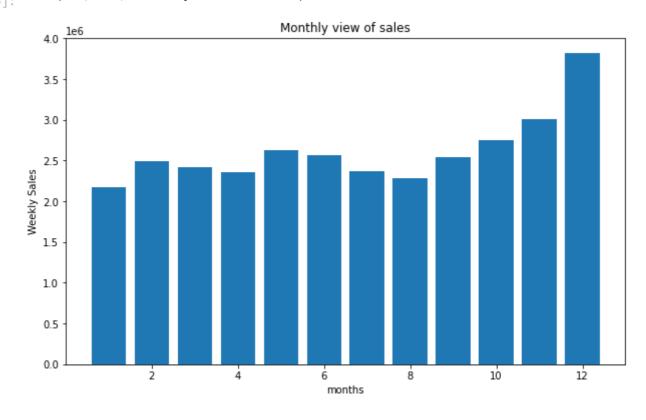




```
In [35]: #Monthly view of sales for all years

plt.figure(figsize=(10,6))
plt.bar(data["Month"],data["Weekly_Sales"])
plt.xlabel("months")
plt.ylabel("Weekly Sales")
plt.title("Monthly view of sales")
```

Out[35]: Text(0.5, 1.0, 'Monthly view of sales')



Prediction Model

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

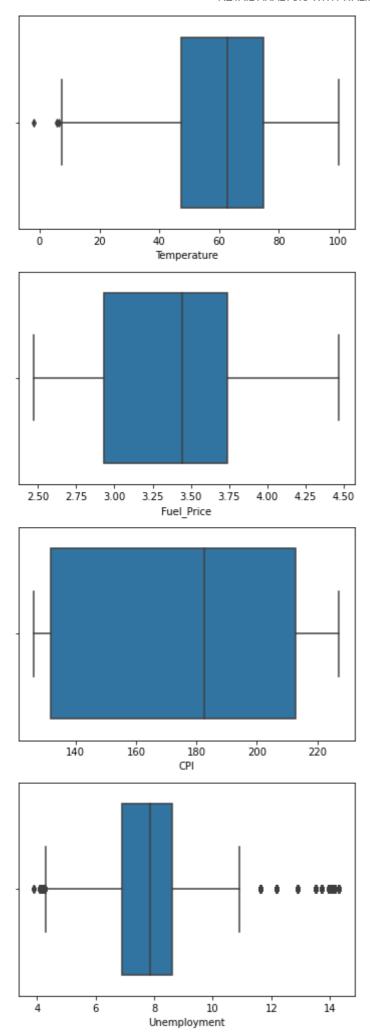
warnings.warn(

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(



```
In [37]: #Dropping the outliers
    data_new = data[(data['Unemployment']<10) & (data['Unemployment']>4.5) & (data['Temp data_new
```

Out[37]:		Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	СРІ	Unemploymen
	0	1	2010- 05-02	1643690.90	0	42.31	2.572	211.096358	8.10
	1	1	2010- 12-02	1641957.44	1	38.51	2.548	211.242170	8.10
	2	1	2010- 02-19	1611968.17	0	39.93	2.514	211.289143	8.10
	3	1	2010- 02-26	1409727.59	0	46.63	2.561	211.319643	8.10
	4	1	2010- 05-03	1554806.68	0	46.50	2.625	211.350143	8.10
	•••								
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	6431	45	2012- 05-10	733455.07	0	64.89	3.985	192.170412	8.66
	6432	45	2012- 12-10	734464.36	0	54.47	4.000	192.327265	8.66
	6433	45	2012- 10-19	718125.53	0	56.47	3.969	192.330854	8.66
	6434	45	2012- 10-26	760281.43	0	58.85	3.882	192.308899	8.66

5658 rows × 11 columns

```
In [38]: # Checking outliers
fig, axs = plt.subplots(4,figsize=(6,18))
X = data_new[['Temperature','Fuel_Price','CPI','Unemployment']]
for i,column in enumerate(X):
    sns.boxplot(data_new[column], ax=axs[i])
```

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(

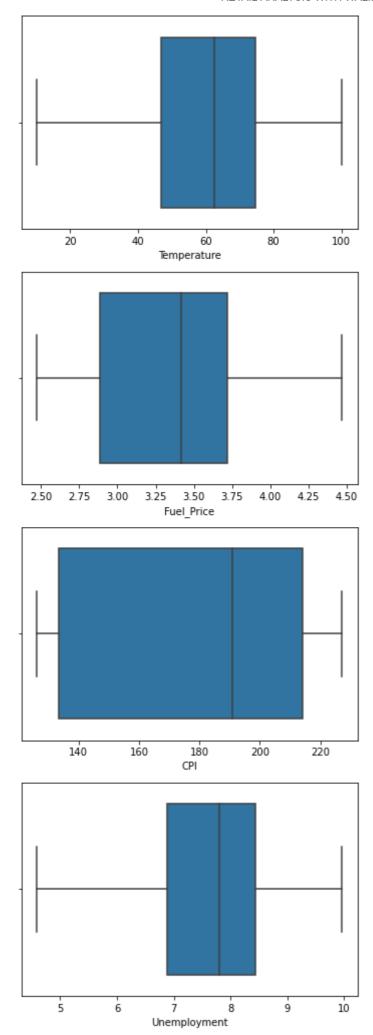
C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variable as a keyword arg: x. From version 0.12, the only valid pos itional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa

ss the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit key word will result in an error or misinterpretation.

warnings.warn(



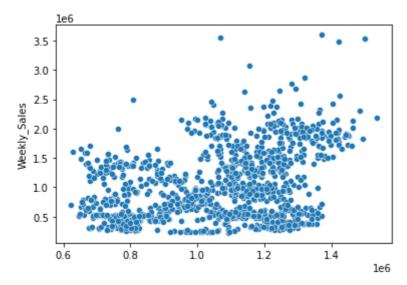
Building a Model

```
In [39]:
          #Import sklearn
          from sklearn.ensemble import RandomForestRegressor
          from sklearn.model_selection import train_test_split
          from sklearn import metrics
          from sklearn.linear_model import LinearRegression
In [41]:
         #Select features and target
          X = data_new[['Store','Fuel_Price','CPI','Unemployment','Day','Month','Year']]
          y = data_new['Weekly_Sales']
          # Split data to train and test (0.80:0.20)
          X train, X test, y train, y test = train test split(X,y,test size=0.2)
In [42]:
          # Linear Regression model
          print('Linear Regression:')
          print()
          reg = LinearRegression()
          reg.fit(X_train, y_train)
          y_pred = reg.predict(X_test)
          print('Accuracy:',reg.score(X_train, y_train)*100)
          print('Mean Absolute Error:', metrics.mean_absolute_error(y_test, y pred))
          print('Mean Squared Error:', metrics.mean squared error(y test, y pred))
          print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test, y_pred)
          sns.scatterplot(y_pred, y_test);
         Linear Regression:
         Accuracy: 13.001420675209596
```

Mean Absolute Error: 448176.39687227097 Mean Squared Error: 297943223046.5909 Root Mean Squared Error: 545841.7564153469

C:\Users\HP\anaconda3\lib\site-packages\seaborn\ decorators.py:36: FutureWarning: Pa ss the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



```
In [43]: # Random Forest Regressor
    print('Random Forest Regressor:')
    print()
    rfr = RandomForestRegressor(n_estimators = 400,max_depth=15,n_jobs=5)
    rfr.fit(X_train,y_train)
    y_pred=rfr.predict(X_test)
    print('Accuracy:',rfr.score(X_test, y_test)*100)

    print('Mean Absolute Error:', metrics.mean_absolute_error(y_test, y_pred))
    print('Mean Squared Error:', metrics.mean_squared_error(y_test, y_pred))
    print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test, y_pred))
    sns.scatterplot(y_pred, y_test);
```

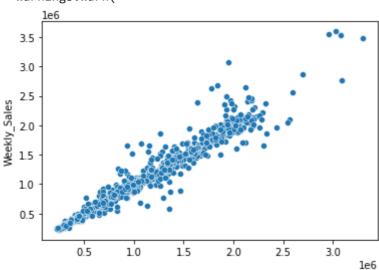
Random Forest Regressor:

Accuracy: 95.16473908382495

Mean Absolute Error: 67890.13043525523 Mean Squared Error: 16416976664.385538 Root Mean Squared Error: 128128.75034271402

C:\Users\HP\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pa ss the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.





Result:

Accuracy: 95.16473908382495

Mean Absolute Error: 67890.13043525523

Mean Squared Error:16416976664.385538

Root Mean Squared Error: 128128.75034271402