

In [82]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
import datetime as dt
```

In []:

In [53]:

```
dfk=pd.read_csv("E:\walmart-sales-dataset-of-45stores.csv")
print(dfk.to_string())
```

	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
5	1	12-03-2010	1439541.59	0	57.79	2.667	211.380643	8.106
6	1	19-03-2010	1472515.79	0	54.58	2.720	211.215635	8.106
7	1	26-03-2010	1404429.92	0	51.45	2.732	211.018042	8.106
8	1	02-04-2010	1594968.28	0	62.27	2.719	210.820450	7.808
9	1	09-04-2010	1545418.53	0	65.86	2.770	210.622857	7.808
10	1	16-04-2010	1466058.28	0	66.32	2.808	210.488700	7.808
11	1	23-04-2010	1391256.12	0	64.84	2.795	210.439123	7.808
12	1	30-04-2010	1425100.71	0	67.41	2.780	210.389546	7.808
13	1	07-05-2010	1603955.12	0	72.55	2.835	210.339968	7.808
14	1	14-05-2010	1494251.50	0	74.78	2.854	210.337426	7.808
15	1	21-05-2010	1399662.07	0	76.44	2.826	210.617093	7.808
16	1	28-05-2010	1432069.95	0	80.44	2.759	210.896761	7.808
17	1	04-06-2010	1615524.71	0	80.69	2.705	211.176428	7.808

In [54]:

```
for x in dfk.index:
    if dfk.loc[x,"Holiday_Flag"] > 0:
        dfk.loc[x,"Holiday_Flag"] = "Holiday_week"
    elif dfk.loc[x,"Holiday_Flag"] < 1:
        dfk.loc[x,"Holiday_Flag"] = "Non_holiday_week"
print(dfk.to_string())
```

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment
0	1	05-02-2010	1643690.90	Non_holiday_week	42.31	2.572	211.096358	8.106
1	1	12-02-2010	1641957.44	Holiday_week	38.51	2.548	211.242170	8.106
2	1	19-02-2010	1611968.17	Non_holiday_week	39.93	2.514	211.289143	8.106
3	1	26-02-2010	1409727.59	Non_holiday_week	46.63	2.561	211.319643	8.106
4	1	05-03-2010	1554806.68	Non_holiday_week	46.50	2.625	211.350143	8.106
5	1	12-03-2010	1439541.59	Non_holiday_week	57.79	2.667	211.380643	8.106
6	1	19-03-2010	1472515.79	Non_holiday_week	54.58	2.720	211.215635	8.106
7	1	26-03-2010	1404429.92	Non_holiday_week	51.45	2.732	211.018042	8.106
8	1	02-04-2010	1594968.28	Non_holiday_week	62.27	2.719	210.820450	7.808
9	1	09-04-2010	1545418.53	Non_holiday_week	65.86	2.770	210.622857	7.808
10	1	16-04-2010	1466058.28	Non_holiday_week	66.32	2.808	210.488700	7.808
11	1	23-04-2010	1391256.12	Non_holiday_week	64.84	2.795	210.439123	7.808
12	1	30-04-2010	1425100.71	Non_holiday_week	67.41	2.780	210.389546	7.808
13	1	07-05-2010	1603955.12	Non_holiday_week	72.55	2.835	210.339968	7.808
14	1	14-05-2010	1494251.50	Non_holiday_week	74.78	2.854	210.337426	7.808
15	1	21-05-2010	1399662.07	Non_holiday_week	76.44	2.826	210.617093	7.808
16	1	28-05-2010	1432069.95	Non_holiday_week	80.44	2.759	210.896761	7.808
17	1	04-06-2010	1615524.71	Non_holiday_week	80.69	2.705	211.176428	7.808

In [55]:

```
dfk
```

Out[55]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment
0	1	05-02-2010	1643690.90	Non_holiday_week	42.31	2.572	211.096358	8.106
1	1	12-02-2010	1641957.44	Holiday_week	38.51	2.548	211.242170	8.106
2	1	19-02-2010	1611968.17	Non_holiday_week	39.93	2.514	211.289143	8.106
3	1	26-02-2010	1409727.59	Non_holiday_week	46.63	2.561	211.319643	8.106
4	1	05-03-2010	1554806.68	Non_holiday_week	46.50	2.625	211.350143	8.106
...
6430	45	28-09-2012	713173.95	Non_holiday_week	64.88	3.997	192.013558	8.684
6431	45	05-10-2012	733455.07	Non_holiday_week	64.89	3.985	192.170412	8.667
6432	45	12-10-2012	734464.36	Non_holiday_week	54.47	4.000	192.327265	8.667
6433	45	19-10-2012	718125.53	Non_holiday_week	56.47	3.969	192.330854	8.667
6434	45	26-10-2012	760281.43	Non_holiday_week	58.85	3.882	192.308899	8.667

6435 rows × 8 columns

In [181]:

```
desk=dfk.describe()
desk
```

Out[181]:

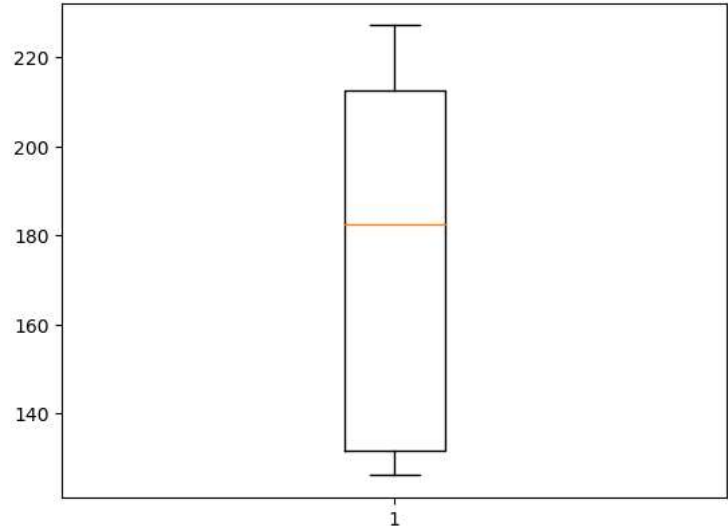
	Store	Weekly_Sales	Temperature	Fuel_Price	CPI	Unemployment
count	6435.000000	6.435000e+03	6435.000000	6435.000000	6435.000000	6435.000000
mean	23.000000	1.046965e+06	60.663782	3.358607	171.578394	7.999151
std	12.988182	5.643666e+05	18.444933	0.459020	39.356712	1.875885
min	1.000000	2.099862e+05	-2.060000	2.472000	126.064000	3.879000
25%	12.000000	5.533501e+05	47.460000	2.933000	131.735000	6.891000
50%	23.000000	9.607460e+05	62.670000	3.445000	182.616521	7.874000
75%	34.000000	1.420159e+06	74.940000	3.735000	212.743293	8.622000
max	45.000000	3.818686e+06	100.140000	4.468000	227.232807	14.313000

In [185]:

```
plt.boxplot(dfk['CPI'])
```

Out[185]:

```
{'whiskers': [<matplotlib.lines.Line2D at 0x1df8e1c8a00>,
<matplotlib.lines.Line2D at 0x1df8e1c8cd0>],
'caps': [<matplotlib.lines.Line2D at 0x1df8e1c8fa0>,
<matplotlib.lines.Line2D at 0x1df8e1d72b0>],
'boxes': [<matplotlib.lines.Line2D at 0x1df8e1c8700>],
'medians': [<matplotlib.lines.Line2D at 0x1df8e1d7580>],
'fliers': [<matplotlib.lines.Line2D at 0x1df8e1d7850>],
'means': []}
```

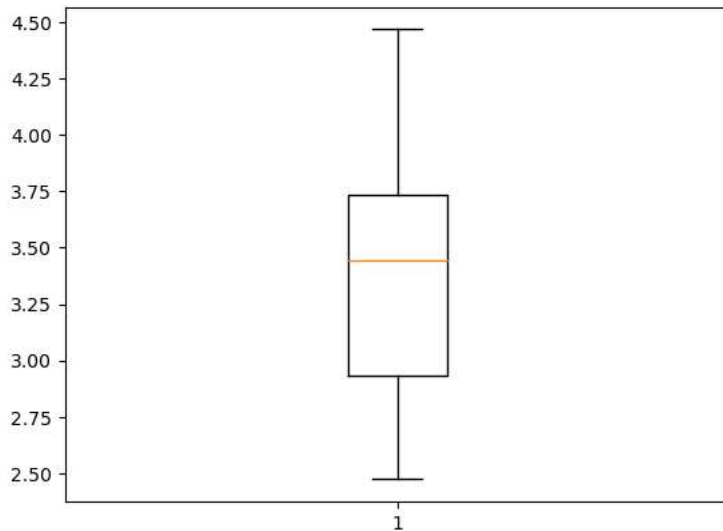


In [186]:

```
plt.boxplot(dfk['Fuel_Price'])
```

Out[186]:

```
{'whiskers': [<matplotlib.lines.Line2D at 0x1df8e230520>,\n             <matplotlib.lines.Line2D at 0x1df8e230820>],\n 'caps': [<matplotlib.lines.Line2D at 0x1df8e230af0>,\n          <matplotlib.lines.Line2D at 0x1df8e230dc0>],\n 'boxes': [<matplotlib.lines.Line2D at 0x1df8e230250>],\n 'medians': [<matplotlib.lines.Line2D at 0x1df8e23e0d0>],\n 'fliers': [<matplotlib.lines.Line2D at 0x1df8e23e3a0>],\n 'means': []}
```

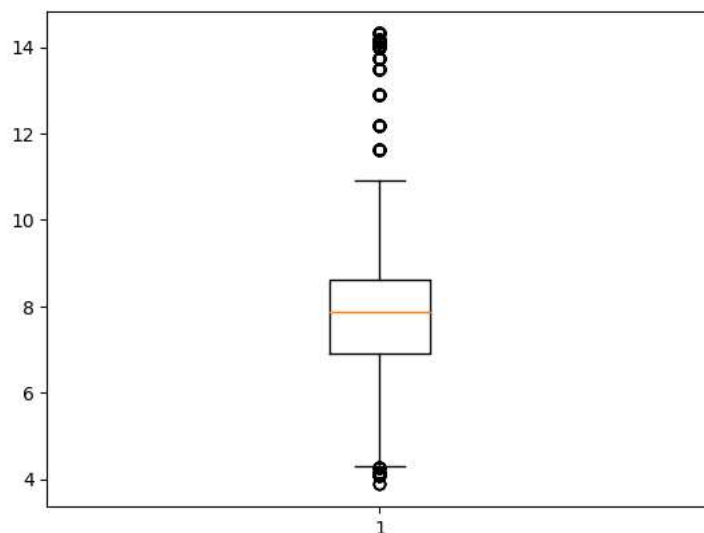


In [188]:

```
plt.boxplot(dfk['Unemployment'])
```

Out[188]:

```
{'whiskers': [<matplotlib.lines.Line2D at 0x1df8e47b7c0>,\n             <matplotlib.lines.Line2D at 0x1df8e47bac0>],\n 'caps': [<matplotlib.lines.Line2D at 0x1df8e47bd90>,\n          <matplotlib.lines.Line2D at 0x1df8e30a0a0>],\n 'boxes': [<matplotlib.lines.Line2D at 0x1df8e47b4f0>],\n 'medians': [<matplotlib.lines.Line2D at 0x1df8e30a3a0>],\n 'fliers': [<matplotlib.lines.Line2D at 0x1df8e30a670>],\n 'means': []}
```

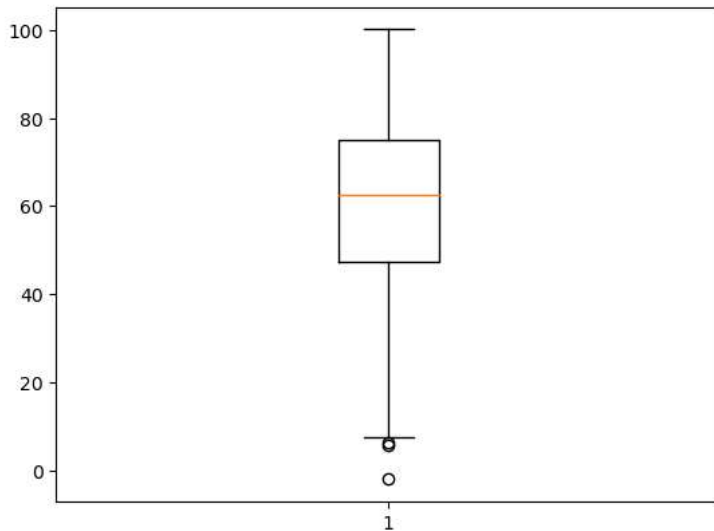


In [189]:

```
plt.boxplot(dfk['Temperature'])
```

Out[189]:

```
{'whiskers': [<matplotlib.lines.Line2D at 0x1df8e3607c0>,
<matplotlib.lines.Line2D at 0x1df8e360a90>],
'caps': [<matplotlib.lines.Line2D at 0x1df8e360d60>,
<matplotlib.lines.Line2D at 0x1df8e36c070>],
'boxes': [<matplotlib.lines.Line2D at 0x1df8e3604f0>],
'medians': [<matplotlib.lines.Line2D at 0x1df8e36c340>],
'fliers': [<matplotlib.lines.Line2D at 0x1df8e36c610>],
'means': []}
```



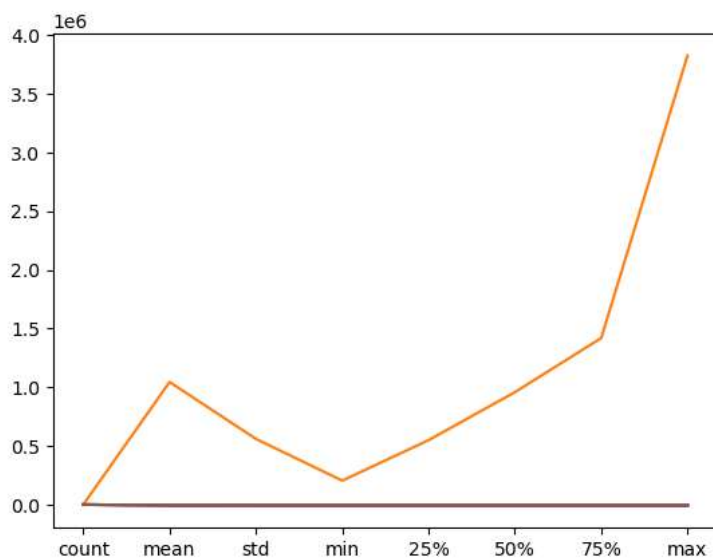
In []:

In [177]:

```
plt.plot(desk)
```

Out[177]:

```
[<matplotlib.lines.Line2D at 0x1df8d0f3490>,
<matplotlib.lines.Line2D at 0x1df8d0f3430>,
<matplotlib.lines.Line2D at 0x1df8d0f35e0>,
<matplotlib.lines.Line2D at 0x1df8d0f3700>,
<matplotlib.lines.Line2D at 0x1df8d0f3820>,
<matplotlib.lines.Line2D at 0x1df8d0f3940>]
```



```
# relations items
dfk.corr()
```

	Store	Weekly_Sales	Temperature	Fuel_Price	CPI	Unemployment
Store	1.000000	-0.335332	-0.022659	0.060023	-0.209492	0.223531
Weekly_Sales	-0.335332	1.000000	-0.063810	0.009464	-0.072634	-0.106176
Temperature	-0.022659	-0.063810	1.000000	0.144982	0.176888	0.101158
Fuel_Price	0.060023	0.009464	0.144982	1.000000	-0.170642	-0.034684
CPI	-0.209492	-0.072634	0.176888	-0.170642	1.000000	-0.302020
Unemployment	0.223531	-0.106176	0.101158	-0.034684	-0.302020	1.000000

```
col=dfk.columns
dfk[col].isnull().sum()
```

```
Store      0
Date       0
Weekly_Sales  0
Holiday_Flag  0
Temperature  0
Fuel_Price  0
CPI        0
Unemployment 0
dtype: int64
```

[illegible]

```
dfk.drop_duplicates(inplace=True)
```

In [64]:

```
dfk.duplicated()
```

Out[64]:

```
0      False
1      False
2      False
3      False
4      False
...
6430   False
6431   False
6432   False
6433   False
6434   False
Length: 6435, dtype: bool
```

In [65]:

```
dfmax1=dfk.groupby('Store')['Weekly_Sales'].sum()
dfmax1
```

Out[65]:

```
Store
1      2.224028e+08
2      2.753824e+08
3      5.758674e+07
4      2.995440e+08
5      4.547569e+07
6      2.237561e+08
7      8.159828e+07
8      1.299512e+08
9      7.778922e+07
10     2.716177e+08
11     1.939628e+08
12     1.442872e+08
13     2.865177e+08
14     2.889999e+08
15     8.913368e+07
16     7.425243e+07
17     1.277821e+08
18     1.551147e+08
19     2.066349e+08
20     3.013978e+08
21     1.081179e+08
22     1.470756e+08
23     1.987506e+08
24     1.940160e+08
25     1.010612e+08
26     1.434164e+08
27     2.538559e+08
28     1.892637e+08
29     7.714155e+07
30     6.271689e+07
31     1.996139e+08
32     1.668192e+08
33     3.716022e+07
34     1.382498e+08
35     1.315207e+08
36     5.341221e+07
37     7.420274e+07
38     5.515963e+07
39     2.074455e+08
40     1.378703e+08
41     1.813419e+08
42     7.956575e+07
43     9.056544e+07
44     4.329309e+07
45     1.123953e+08
Name: Weekly_Sales, dtype: float64
```

In [66]:

```
print("index {} and max_value {}".format(dfmax1.idxmax(),dfmax1.max()))
```

```
index 20 and max_value 301397792.46
```

In [67]:

```
dfmax2=dfk.groupby('Store')['Weekly_Sales'].std()  
dfmax2
```

Out[67]:

```
Store  
1      155980.767761  
2      237683.694682  
3       46319.631557  
4      266201.442297  
5       37737.965745  
6      212525.855862  
7      112585.469220  
8      106280.829881  
9       69028.666585  
10     302262.062504  
11     165833.887863  
12     139166.871880  
13     265506.995776  
14     317569.949476  
15     120538.652043  
16       85769.680133  
17     112162.936087  
18     176641.510839  
19     191722.638730  
20     275900.562742  
21     128752.812853  
22     161251.350631  
23     249788.038068  
24     167745.677567  
25     112976.788600  
26     110431.288141  
27     239930.135688  
28     181758.967539  
29       99120.136596  
30     22809.665590  
31     125855.942933  
32     138017.252087  
33       24132.927322  
34     104630.164676  
35     211243.457791  
36       60725.173579  
37       21837.461190  
38       42768.169450  
39     217466.454833  
40     119002.112858  
41     187907.162766  
42       50262.925530  
43       40598.413260  
44       24762.832015  
45     130168.526635  
Name: Weekly_Sales, dtype: float64
```

In [68]:

```
print("index {} and max_value {}".format(dfmax2.idxmax(),dfmax2.max()))
```

index 14 and max_value 317569.9494755081

In [69]:

```
dfk.groupby('Holiday_Flag')['Weekly_Sales'].mean().to_frame().reset_index()
```

Out[69]:

	Holiday_Flag	Weekly_Sales
0	Holiday_week	1.122888e+06
1	Non_holiday_week	1.041256e+06

In [70]:

```
# filter
holiday=dfk[dfk.Holiday_Flag=='Holiday_week']
holiday
```

Out[70]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment
	1	12-02-2010	1641957.44	Holiday_week	38.51	2.548	211.242170	8.106
	31	10-09-2010	1507460.69	Holiday_week	78.69	2.565	211.495190	7.787
	42	1 26-11-2010	1955624.11	Holiday_week	64.52	2.735	211.748433	7.838
	47	1 31-12-2010	1367320.01	Holiday_week	48.43	2.943	211.404932	7.838
	53	1 11-02-2011	1649614.93	Holiday_week	36.39	3.022	212.936705	7.742

	6375	45 09-09-2011	746129.56	Holiday_week	71.48	3.738	186.673738	8.625
	6386	45 25-11-2011	1170672.94	Holiday_week	48.71	3.492	188.350400	8.523
	6391	45 30-12-2011	869403.63	Holiday_week	37.79	3.389	189.062016	8.523
	6397	45 10-02-2012	803657.12	Holiday_week	37.00	3.640	189.707605	8.424
	6427	45 07-09-2012	766512.66	Holiday_week	75.70	3.911	191.577676	8.684

450 rows × 8 columns

In [71]:

```
# filter
non_holiday=dfk[dfk.Holiday_Flag=='Non_holiday_week']
non_holiday
```

Out[71]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment
	0	1 05-02-2010	1643690.90	Non_holiday_week	42.31	2.572	211.096358	8.106
	2	1 19-02-2010	1611968.17	Non_holiday_week	39.93	2.514	211.289143	8.106
	3	1 26-02-2010	1409727.59	Non_holiday_week	46.63	2.561	211.319643	8.106
	4	1 05-03-2010	1554806.68	Non_holiday_week	46.50	2.625	211.350143	8.106
	5	1 12-03-2010	1439541.59	Non_holiday_week	57.79	2.667	211.380643	8.106

	6430	45 28-09-2012	713173.95	Non_holiday_week	64.88	3.997	192.013558	8.684
	6431	45 05-10-2012	733455.07	Non_holiday_week	64.89	3.985	192.170412	8.667
	6432	45 12-10-2012	734464.36	Non_holiday_week	54.47	4.000	192.327265	8.667
	6433	45 19-10-2012	718125.53	Non_holiday_week	56.47	3.969	192.330854	8.667
	6434	45 26-10-2012	760281.43	Non_holiday_week	58.85	3.882	192.308899	8.667

5985 rows × 8 columns

In [72]:

```
holidays_higher_sales=holiday[(holiday.Weekly_Sales)>(non_holiday.Weekly_Sales.mean())]
holidays_higher_sales
```

Out[72]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment
	1	12-02-2010	1641957.44	Holiday_week	38.51	2.548	211.242170	8.106
	31	10-09-2010	1507460.69	Holiday_week	78.69	2.565	211.495190	7.787
	42	1 26-11-2010	1955624.11	Holiday_week	64.52	2.735	211.748433	7.838
	47	1 31-12-2010	1367320.01	Holiday_week	48.43	2.943	211.404932	7.838
	53	1 11-02-2011	1649614.93	Holiday_week	36.39	3.022	212.936705	7.742

	5819	41 30-12-2011	1264014.16	Holiday_week	34.12	3.119	196.358610	6.759
	5825	41 10-02-2012	1238844.56	Holiday_week	22.00	3.103	196.919506	6.589
	5855	41 07-09-2012	1392143.82	Holiday_week	67.41	3.596	198.095048	6.432
	6334	45 26-11-2010	1182500.16	Holiday_week	46.15	3.039	182.783277	8.724
	6386	45 25-11-2011	1170672.94	Holiday_week	48.71	3.492	188.350400	8.523

220 rows × 8 columns

In [73]:

```
dfk['year']=dfk['Date'].apply(lambda x:x[6:])
dfk['month']=dfk['Date'].apply(lambda x:x[3:5])
dfk['day']=dfk['Date'].apply(lambda x:x[0:2])
```

In [79]:

```
dfk
```

Out[79]:

	Store	Date	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price	CPI	Unemployment	year	month	day
0	1	05-02-2010	1643690.90	Non_holiday_week	42.31	2.572	211.096358	8.106	2010	02	05
1	1	12-02-2010	1641957.44	Holiday_week	38.51	2.548	211.242170	8.106	2010	02	12
2	1	19-02-2010	1611968.17	Non_holiday_week	39.93	2.514	211.289143	8.106	2010	02	19
3	1	26-02-2010	1409727.59	Non_holiday_week	46.63	2.561	211.319643	8.106	2010	02	26
4	1	05-03-2010	1554806.68	Non_holiday_week	46.50	2.625	211.350143	8.106	2010	03	05
...
6430	45	28-09-2012	713173.95	Non_holiday_week	64.88	3.997	192.013558	8.684	2012	09	28
6431	45	05-10-2012	733455.07	Non_holiday_week	64.89	3.985	192.170412	8.667	2012	10	05
6432	45	12-10-2012	734464.36	Non_holiday_week	54.47	4.000	192.327265	8.667	2012	10	12
6433	45	19-10-2012	718125.53	Non_holiday_week	56.47	3.969	192.330854	8.667	2012	10	19
6434	45	26-10-2012	760281.43	Non_holiday_week	58.85	3.882	192.308899	8.667	2012	10	26

6435 rows × 11 columns

In []:

```
# df=df.drop('Date' , axis=1)
```

In [159]:

```
dfa=dfk.groupby(['month' , 'year'])['Weekly_Sales'].sum().to_frame().reset_index()  
dfa
```

Out[159]:

	month	year	Weekly_Sales
0	01	2011	1.637040e+08
1	01	2012	1.688945e+08
2	02	2010	1.903330e+08
3	02	2011	1.863313e+08
4	02	2012	1.920636e+08
5	03	2010	1.819198e+08
6	03	2011	1.793564e+08
7	03	2012	2.315097e+08
8	04	2010	2.314124e+08
9	04	2011	2.265265e+08
10	04	2012	1.889209e+08
11	05	2010	1.867109e+08
12	05	2011	1.816482e+08
13	05	2012	1.887665e+08
14	06	2010	1.922462e+08
15	06	2011	1.897734e+08
16	06	2012	2.406103e+08
17	07	2010	2.325801e+08
18	07	2011	2.299114e+08
19	07	2012	1.875095e+08
20	08	2010	1.876401e+08
21	08	2011	1.885993e+08
22	08	2012	2.368508e+08
23	09	2010	1.772679e+08
24	09	2011	2.208477e+08
25	09	2012	1.806455e+08
26	10	2010	2.171618e+08
27	10	2011	1.832613e+08
28	10	2012	1.843617e+08
29	11	2010	2.028534e+08
30	11	2011	2.101624e+08
31	12	2010	2.887605e+08
32	12	2011	2.880781e+08

In [103]:

```
# filter  
year2010=dfa[dfa.year!='2010']  
year2010
```

Out[103]:

	month	year	Weekly_Sales
2	02	2010	1.903330e+08
5	03	2010	1.819198e+08
8	04	2010	2.314124e+08
11	05	2010	1.867109e+08
14	06	2010	1.922462e+08
17	07	2010	2.325801e+08
20	08	2010	1.876401e+08
23	09	2010	1.772679e+08
26	10	2010	2.171618e+08
29	11	2010	2.028534e+08
31	12	2010	2.887605e+08

In [104]:

```
# filter
year2011=dfa[dfa.year=='2011']
year2011
```

Out[104]:

	month	year	Weekly_Sales
0	01	2011	1.637040e+08
3	02	2011	1.863313e+08
6	03	2011	1.793564e+08
9	04	2011	2.265265e+08
12	05	2011	1.816482e+08
15	06	2011	1.897734e+08
18	07	2011	2.299114e+08
21	08	2011	1.885993e+08
24	09	2011	2.208477e+08
27	10	2011	1.832613e+08
30	11	2011	2.101624e+08
32	12	2011	2.880781e+08

In [105]:

```
# filter
year2012=dfa[dfa.year=='2012']
year2012
```

Out[105]:

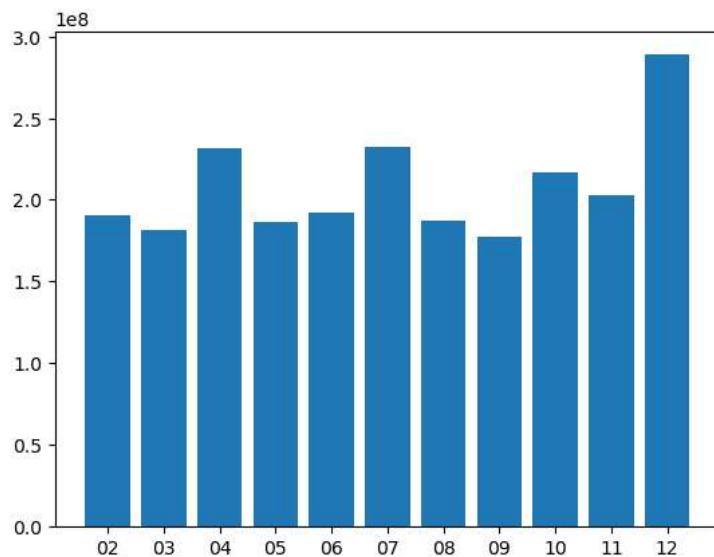
	month	year	Weekly_Sales
1	01	2012	1.688945e+08
4	02	2012	1.920636e+08
7	03	2012	2.315097e+08
10	04	2012	1.889209e+08
13	05	2012	1.887665e+08
16	06	2012	2.406103e+08
19	07	2012	1.875095e+08
22	08	2012	2.368508e+08
25	09	2012	1.806455e+08
28	10	2012	1.843617e+08

In [106]:

```
plt.bar(year2010['month'],year2010['Weekly_Sales'])
#one month notfound 12month is higher sales
```

Out[106]:

<BarContainer object of 11 artists>

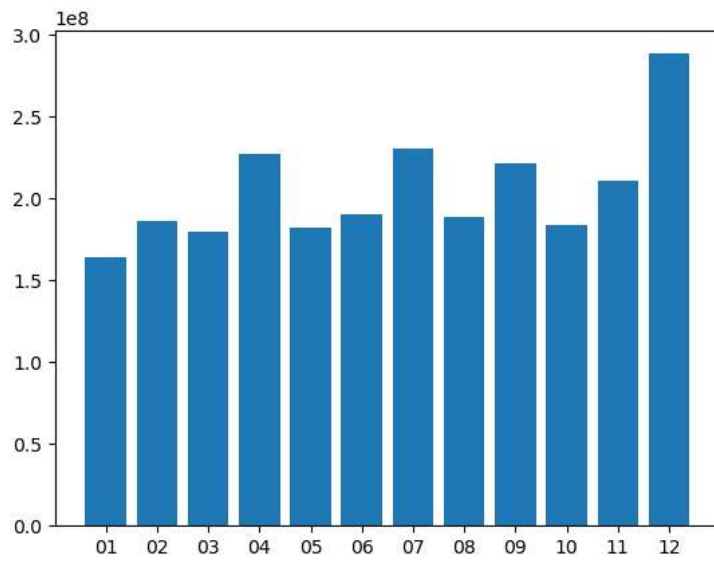


In [107]:

```
plt.bar(year2011['month'],year2011['Weekly_Sales'])
# 12month is higher sales
```

Out[107]:

<BarContainer object of 12 artists>

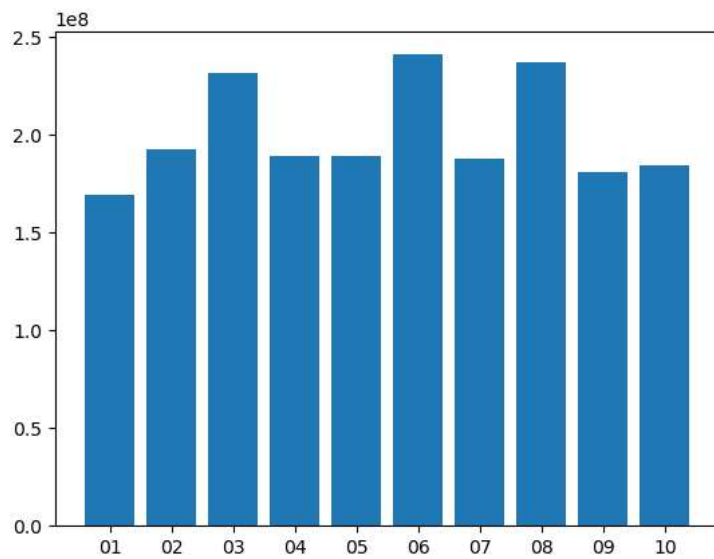


In [108]:

```
plt.bar(year2012['month'],year2012['Weekly_Sales'])
# 11 and 12 months not found and 6month is higher sales
```

Out[108]:

<BarContainer object of 10 artists>



In [109]:

```
dfak=dfk.groupby('year')['Weekly_Sales'].sum().to_frame().reset_index()
dfak
```

Out[109]:

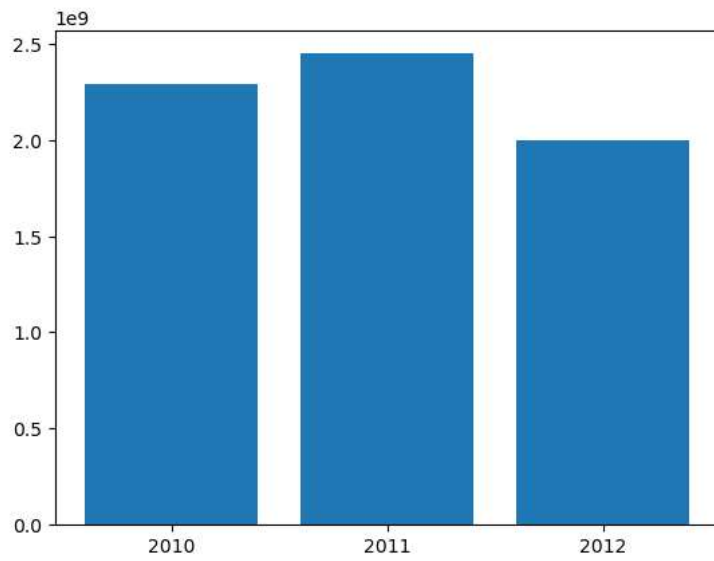
	year	Weekly_Sales
0	2010	2.288886e+09
1	2011	2.448200e+09
2	2012	2.000133e+09

In [98]:

```
plt.bar(dfak['year'],dfak['Weekly_Sales'])
```

Out[98]:

<BarContainer object of 3 artists>



In [161]:

```
sales_and_temp=dfk.groupby('Store').agg({  
    'Weekly_Sales':lambda sales :sales.sum(),  
    'Temperature':lambda temp :temp.mean()  
}).reset_index()  
sales_and_temp
```

Out[161]:

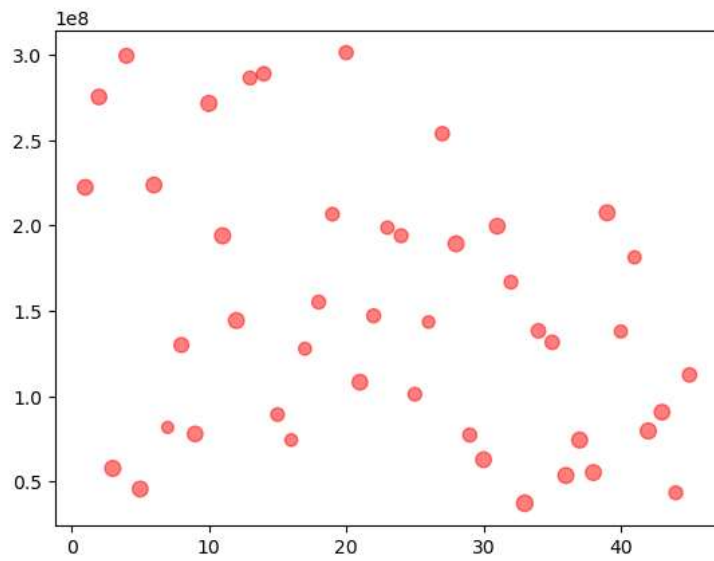
	Store	Weekly_Sales	Temperature
0	1	2.224028e+08	68.306783
1	2	2.753824e+08	68.216364
2	3	5.758674e+07	71.434196
3	4	2.995440e+08	62.253357
4	5	4.547569e+07	69.410140
5	6	2.237561e+08	69.700000
6	7	8.159828e+07	39.720280
7	8	1.299512e+08	62.513986
8	9	7.778922e+07	67.775175
9	10	2.716177e+08	72.241189
10	11	1.939628e+08	72.480769
11	12	1.442872e+08	70.262797
12	13	2.865177e+08	53.697133
13	14	2.889999e+08	57.790979
14	15	8.913368e+07	51.833846
15	16	7.425243e+07	45.030070
16	17	1.277821e+08	46.387203
17	18	1.551147e+08	53.371259
18	19	2.066349e+08	52.295035
19	20	3.013978e+08	55.451399
20	21	1.081179e+08	68.847622
21	22	1.470756e+08	54.897133
22	23	1.987506e+08	48.805105
23	24	1.940160e+08	54.030000
24	25	1.010612e+08	52.138392
25	26	1.434164e+08	43.658252
26	27	2.538559e+08	57.311119
27	28	1.892637e+08	70.262797
28	29	7.714155e+07	54.897133
29	30	6.271689e+07	68.847622
30	31	1.996139e+08	68.847622
31	32	1.668192e+08	52.747552
32	33	3.716022e+07	76.728182
33	34	1.382498e+08	58.495874
34	35	1.315207e+08	57.311119
35	36	5.341221e+07	71.160350
36	37	7.420274e+07	71.160350
37	38	5.515963e+07	70.262797
38	39	2.074455e+08	70.597343
39	40	1.378703e+08	47.674545
40	41	1.813419e+08	48.410350
41	42	7.956575e+07	72.241189
42	43	9.056544e+07	68.877692
43	44	4.329309e+07	53.697133
44	45	1.123953e+08	57.790979

In [169]:

```
plt.scatter(sales_and_temp['Store'],sales_and_temp['Weekly_Sales'],s=sales_and_temp['Temperature'],color='red',alpha=.5)  
# temperature is not effect in thr weekly_sales
```

Out[169]:

<matplotlib.collections.PathCollection at 0x1df8ccdbfa0>



In []:

In [162]:

```
sales_and_cpi=dfk.groupby('Store').agg({
    'Weekly_Sales':lambda sales :sales.sum(),
    'CPI':lambda cpi :cpi.mean()
}).reset_index()
sales_and_cpi
```

Out[162]:

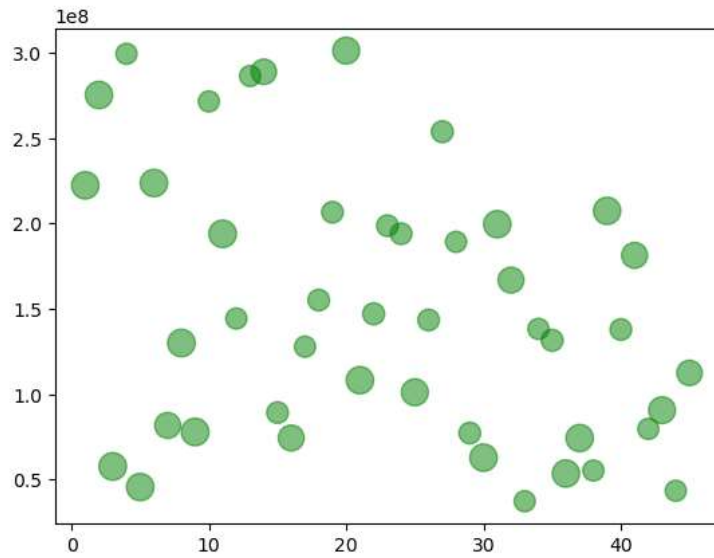
	Store	Weekly_Sales	CPI
0	1	2.224028e+08	215.996892
1	2	2.753824e+08	215.646311
2	3	5.758674e+07	219.391531
3	4	2.995440e+08	128.679669
4	5	4.547569e+07	216.565581
5	6	2.237561e+08	217.553197
6	7	8.159828e+07	193.664243
7	8	1.299512e+08	219.439026
8	9	7.778922e+07	219.626689
9	10	2.716177e+08	128.679669
10	11	1.939628e+08	219.391531
11	12	1.442872e+08	128.679669
12	13	2.865177e+08	128.679669
13	14	2.889999e+08	186.285678
14	15	8.913368e+07	135.092607
15	16	7.425243e+07	193.664243
16	17	1.277821e+08	128.679669
17	18	1.551147e+08	135.092607
18	19	2.066349e+08	135.092607
19	20	3.013978e+08	209.038131
20	21	1.081179e+08	215.646311
21	22	1.470756e+08	139.011284
22	23	1.987506e+08	135.092607
23	24	1.940160e+08	135.092607
24	25	1.010612e+08	209.038131
25	26	1.434164e+08	135.092607
26	27	2.538559e+08	139.011284
27	28	1.892637e+08	128.679669
28	29	7.714155e+07	135.092607
29	30	6.271689e+07	215.646311
30	31	1.996139e+08	215.646311
31	32	1.668192e+08	193.664243
32	33	3.716022e+07	128.679669
33	34	1.382498e+08	128.679669
34	35	1.315207e+08	139.011284
35	36	5.341221e+07	214.729069
36	37	7.420274e+07	214.729069
37	38	5.515963e+07	128.679669
38	39	2.074455e+08	214.729069
39	40	1.378703e+08	135.092607
40	41	1.813419e+08	193.664243
41	42	7.956575e+07	128.679669
42	43	9.056544e+07	207.735162
43	44	4.329309e+07	128.679669
44	45	1.123953e+08	186.285678

In [170]:

```
plt.scatter(sales_and_cpi['Store'],sales_and_cpi['Weekly_Sales'],s=sales_and_cpi['CPI'],color='g',alpha=.5)  
# CPI is effect in the weekly_sales  
# increase the CPI leads to increase the weewkly_sales
```

Out[170]:

<matplotlib.collections.PathCollection at 0x1df8ce3af40>



In []:

THE FINAL PROJEGT DATA SCIENCE METHODOLOGE

In []: