

In [1]:

```

1 #import libraries
2 import pandas as pd
3 import numpy as np
4 import seaborn as sns
5 import matplotlib.pyplot as plt
6 import datetime
7

```

In [2]:

```

1 #importing sales dataset
2 sales_train_df = pd.read_csv('D:/Data Science for Business Package/3. Sales Department Data/train.csv')
3 sales_train_df.head()

```

C:\Users\Pradeep\AppData\Local\Temp\ipykernel_24732\3788187673.py:2: Dtype Warning: Columns (7) have mixed types. Specify dtype option on import or set low_memory=False.

```
sales_train_df = pd.read_csv('D:/Data Science for Business Package/3. Sales Department Data/train.csv')
```

Out[2]:

| | Store | DayOfWeek | Date | Sales | Customers | Open | Promo | StateHoliday | SchoolHoliday |
|---|-------|-----------|------------|-------|-----------|------|-------|--------------|---------------|
| 0 | 1 | 5 | 2015-07-31 | 5263 | 555 | 1 | 1 | 0 | 1 |
| 1 | 2 | 5 | 2015-07-31 | 6064 | 625 | 1 | 1 | 0 | 1 |
| 2 | 3 | 5 | 2015-07-31 | 8314 | 821 | 1 | 1 | 0 | 1 |
| 3 | 4 | 5 | 2015-07-31 | 13995 | 1498 | 1 | 1 | 0 | 1 |
| 4 | 5 | 5 | 2015-07-31 | 4822 | 559 | 1 | 1 | 0 | 1 |

In [3]:

```

1 # Id: transaction ID (combination of Store and date)
2 # Store: unique store Id
3 # Sales: sales/day, this is the target variable
4 # Customers: number of customers on a given day
5 # Open: Boolean to say whether a store is open or closed (0 = closed, 1 = open)
6 # Promo: describes if store is running a promo on that day or not
7 # StateHoliday: indicate which state holiday (a = public holiday, b = Easter holiday)
8 # SchoolHoliday: indicates if the (Store, Date) was affected by the closure of public schools

```

In [4]:

```
1 sales_train_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1017209 entries, 0 to 1017208
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Store            1017209 non-null  int64
1   DayOfWeek        1017209 non-null  int64
2   Date             1017209 non-null  object
3   Sales            1017209 non-null  int64
4   Customers        1017209 non-null  int64
5   Open            1017209 non-null  int64
6   Promo            1017209 non-null  int64
7   StateHoliday     1017209 non-null  object
8   SchoolHoliday    1017209 non-null  int64
dtypes: int64(7), object(2)
memory usage: 69.8+ MB
```

In [5]:

```
1 #checking for NULL values
2 sales_train_df.isnull().sum()
```

Out[5]:

```
Store            0
DayOfWeek        0
Date             0
Sales            0
Customers        0
Open            0
Promo            0
StateHoliday     0
SchoolHoliday    0
dtype: int64
```

In [6]:

```
1 sales_train_df.describe()
```

Out[6]:

| | Store | DayOfWeek | Sales | Customers | Open | Promo |
|-------|--------------|--------------|--------------|--------------|--------------|--------------|
| count | 1.017209e+06 | 1.017209e+06 | 1.017209e+06 | 1.017209e+06 | 1.017209e+06 | 1.017209e+06 |
| mean | 5.584297e+02 | 3.998341e+00 | 5.773819e+03 | 6.331459e+02 | 8.301067e-01 | 3.815145e-01 |
| std | 3.219087e+02 | 1.997391e+00 | 3.849926e+03 | 4.644117e+02 | 3.755392e-01 | 4.857586e-01 |
| min | 1.000000e+00 | 1.000000e+00 | 0.000000e+00 | 0.000000e+00 | 0.000000e+00 | 0.000000e+00 |
| 25% | 2.800000e+02 | 2.000000e+00 | 3.727000e+03 | 4.050000e+02 | 1.000000e+00 | 0.000000e+00 |
| 50% | 5.580000e+02 | 4.000000e+00 | 5.744000e+03 | 6.090000e+02 | 1.000000e+00 | 0.000000e+00 |
| 75% | 8.380000e+02 | 6.000000e+00 | 7.856000e+03 | 8.370000e+02 | 1.000000e+00 | 1.000000e+00 |
| max | 1.115000e+03 | 7.000000e+00 | 4.155100e+04 | 7.388000e+03 | 1.000000e+00 | 1.000000e+00 |

In [7]:

```
1 #importing store info dataset
2 store_info_df = pd.read_csv('D:/Data Science for Business Package/3. Sales Department/store_info.csv')
3 store_info_df.head()
```

Out[7]:

| | Store | StoreType | Assortment | CompetitionDistance | CompetitionOpenSinceMonth | CompetitionOpenSinceYear |
|---|-------|-----------|------------|---------------------|---------------------------|--------------------------|
| 0 | 1 | c | a | 1270.0 | 9.0 | 2013 |
| 1 | 2 | a | a | 570.0 | 11.0 | 2013 |
| 2 | 3 | a | a | 14130.0 | 12.0 | 2013 |
| 3 | 4 | c | c | 620.0 | 9.0 | 2013 |
| 4 | 5 | a | a | 29910.0 | 4.0 | 2013 |

In [8]:

```
1 # StoreType: categorical variable to indicate type of store (a, b, c, d)
2 # Assortment: describes an assortment level: a = basic, b = extra, c = extended
3 # CompetitionDistance (meters): distance to closest competitor store
4 # CompetitionOpenSince [Month/Year]: provides an estimate of the date when competition started
5 # Promo2: Promo2 is a continuing and consecutive promotion for some stores (0 = no promotion, 1 = promotion)
6 # Promo2Since [Year/Week]: date when the store started participating in Promo2
7 # PromoInterval: describes the consecutive intervals Promo2 is started, naming the month and week
8
9
```

In [9]:

```
1 store_info_df.isnull().sum()
```

Out[9]:

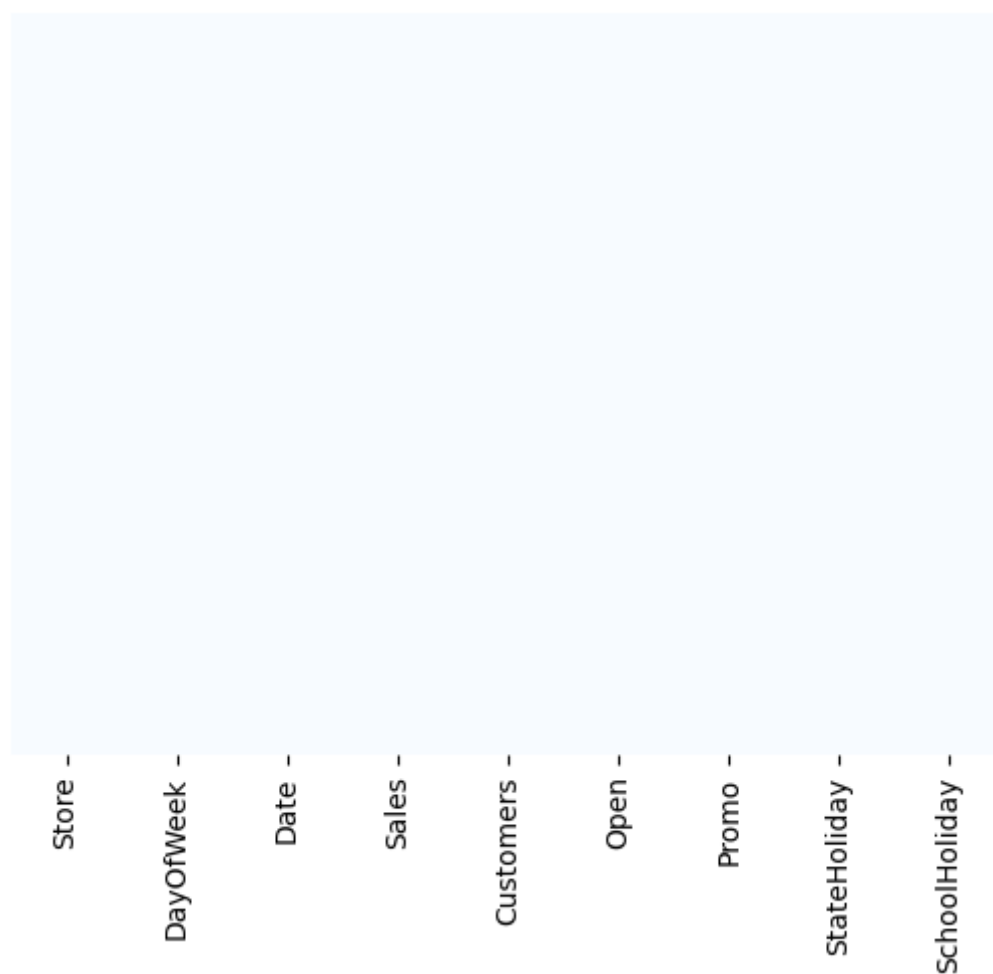
```
Store                0
StoreType            0
Assortment           0
CompetitionDistance  3
CompetitionOpenSinceMonth  354
CompetitionOpenSinceYear  354
Promo2              0
Promo2SinceWeek     544
Promo2SinceYear     544
PromoInterval       544
dtype: int64
```

In [10]:

```
1 #visualizing the null values if any
2 sns.heatmap(sales_train_df.isnull(), yticklabels = False, cbar = False, cmap="Blues")
3
```

Out[10]:

<AxesSubplot: >

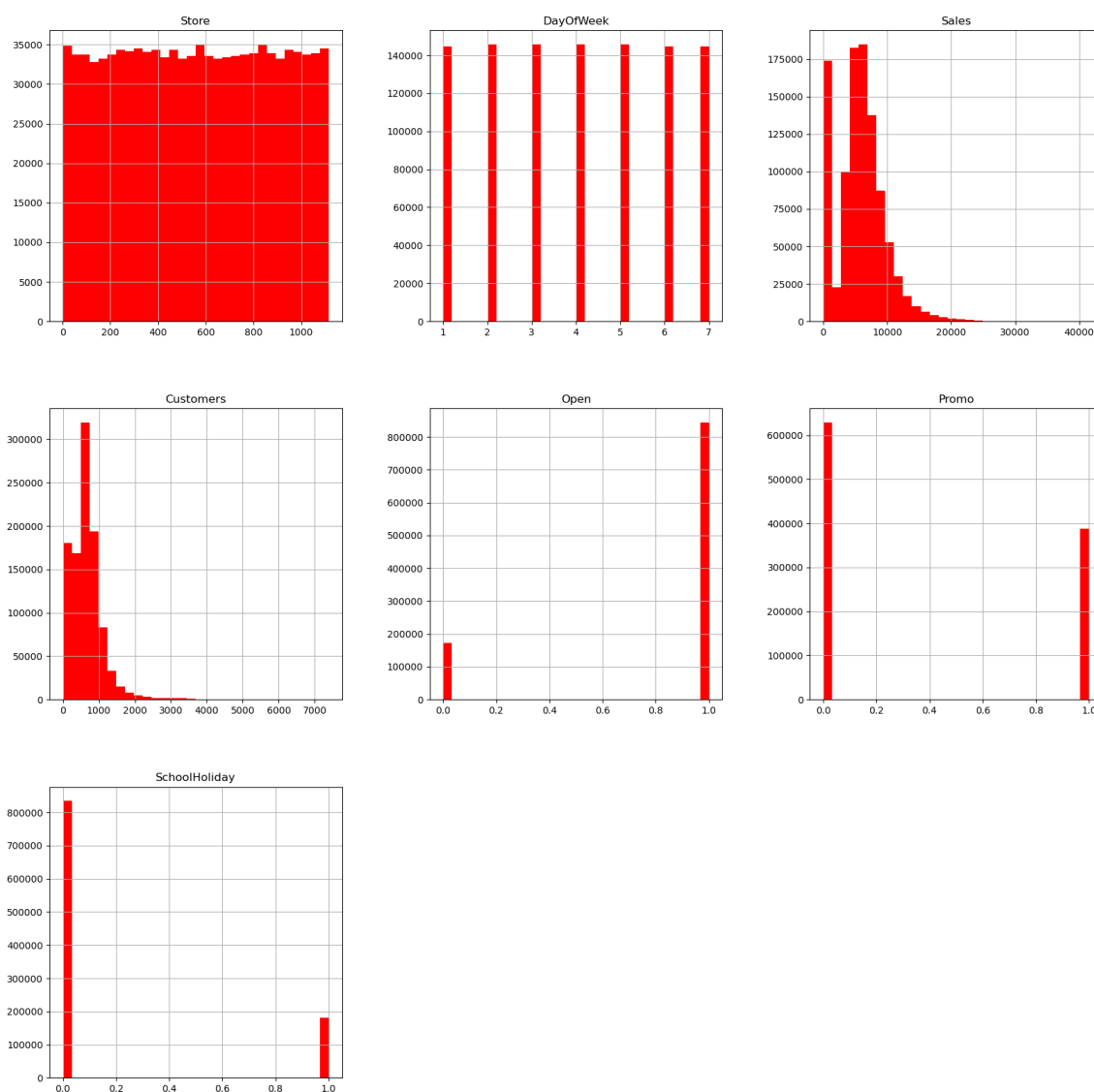


In [11]:

```
1 #visualizing numerical column
2 sales_train_df.hist(bins = 30, figsize = (20,20), color = 'r')
```

Out[11]:

```
array([[<AxesSubplot: title={'center': 'Store'}>,
        <AxesSubplot: title={'center': 'DayOfWeek'}>,
        <AxesSubplot: title={'center': 'Sales'}>],
       [<AxesSubplot: title={'center': 'Customers'}>,
        <AxesSubplot: title={'center': 'Open'}>,
        <AxesSubplot: title={'center': 'Promo'}>],
       [<AxesSubplot: title={'center': 'SchoolHoliday'}>,
        <AxesSubplot: >, <AxesSubplot: >]], dtype=object)
```



In [12]:

```

1 #stores that are open and closed
2 closed_train_df = sales_train_df[sales_train_df['Open'] == 0]
3 open_train_df = sales_train_df[sales_train_df['Open'] == 1]
4
5 print("Total =", len(sales_train_df))
6 print("Number of closed stores =", len(closed_train_df))
7 print("Number of open stores =", len(open_train_df))
8

```

Total = 1017209

Number of closed stores = 172817

Number of open stores = 844392

In [13]:

```

1 #removing closed stores
2 sales_train_df = sales_train_df[sales_train_df['Open'] == 1]
3 print(len(sales_train_df))
4 sales_train_df.head()

```

844392

Out[13]:

| | Store | DayOfWeek | Date | Sales | Customers | Open | Promo | StateHoliday | SchoolHoliday |
|---|-------|-----------|------------|-------|-----------|------|-------|--------------|---------------|
| 0 | 1 | 5 | 2015-07-31 | 5263 | 555 | 1 | 1 | 0 | 1 |
| 1 | 2 | 5 | 2015-07-31 | 6064 | 625 | 1 | 1 | 0 | 1 |
| 2 | 3 | 5 | 2015-07-31 | 8314 | 821 | 1 | 1 | 0 | 1 |
| 3 | 4 | 5 | 2015-07-31 | 13995 | 1498 | 1 | 1 | 0 | 1 |
| 4 | 5 | 5 | 2015-07-31 | 4822 | 559 | 1 | 1 | 0 | 1 |

In [14]:

```

1 # Let's drop the open column since it has no meaning now
2 sales_train_df.drop(['Open'], axis=1, inplace=True)
3 sales_train_df.head()

```

Out[14]:

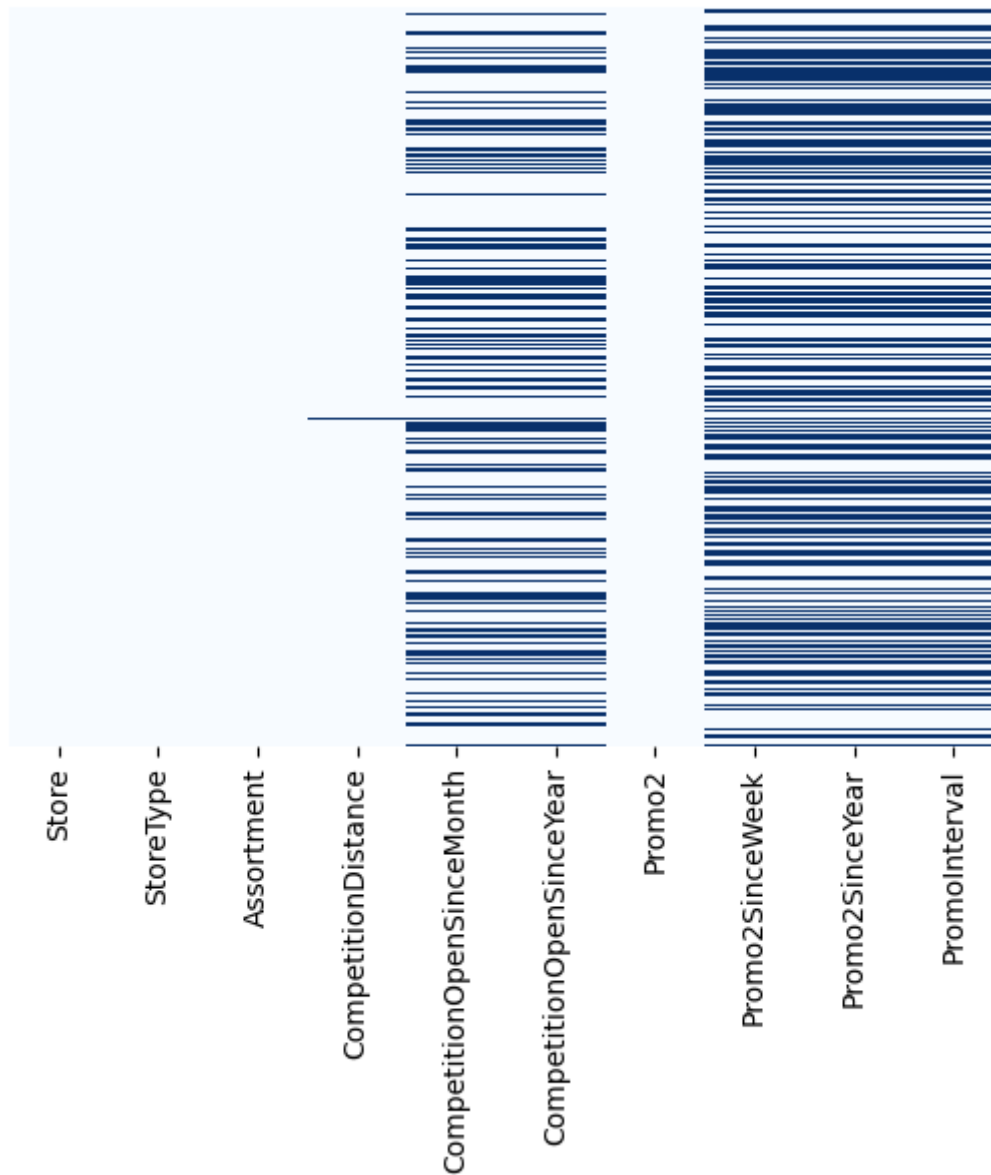
| | Store | DayOfWeek | Date | Sales | Customers | Promo | StateHoliday | SchoolHoliday |
|---|-------|-----------|------------|-------|-----------|-------|--------------|---------------|
| 0 | 1 | 5 | 2015-07-31 | 5263 | 555 | 1 | 0 | 1 |
| 1 | 2 | 5 | 2015-07-31 | 6064 | 625 | 1 | 0 | 1 |
| 2 | 3 | 5 | 2015-07-31 | 8314 | 821 | 1 | 0 | 1 |
| 3 | 4 | 5 | 2015-07-31 | 13995 | 1498 | 1 | 0 | 1 |
| 4 | 5 | 5 | 2015-07-31 | 4822 | 559 | 1 | 0 | 1 |

In [15]:

```
1 #visualizing NULL values for store dataset
2 sns.heatmap(store_info_df.isnull(), yticklabels = False, cbar = False, cmap="Blues")
3
```

Out[15]:

<AxesSubplot: >



In [16]:

```
1 #checkig NULL Values
2 store_info_df.isnull().sum()
```

Out[16]:

```
Store          0
StoreType      0
Assortment     0
CompetitionDistance    3
CompetitionOpenSinceMonth  354
CompetitionOpenSinceYear  354
Promo2         0
Promo2SinceWeek  544
Promo2SinceYear  544
PromoInterval   544
dtype: int64
```

In [17]:

```
1 #missing values in the 'CompetitionDistance'
2 store_info_df[store_info_df['CompetitionDistance'].isnull()]
```

Out[17]:

| | Store | StoreType | Assortment | CompetitionDistance | CompetitionOpenSinceMonth | Comp |
|-----|-------|-----------|------------|---------------------|---------------------------|------|
| 290 | 291 | d | a | NaN | NaN | |
| 621 | 622 | a | c | NaN | NaN | |
| 878 | 879 | d | a | NaN | NaN | |



In [18]:

```
1 #missing values in the 'CompetitionOpenSinceMonth'
2 store_info_df[store_info_df['CompetitionOpenSinceMonth'].isnull()]
```

Out[18]:

| | Store | StoreType | Assortment | CompetitionDistance | CompetitionOpenSinceMonth | CompetitionOpenSinceYear |
|------|-------|-----------|------------|---------------------|---------------------------|--------------------------|
| 11 | 12 | a | c | 1070.0 | NaN | 2013 |
| 12 | 13 | d | a | 310.0 | NaN | 2013 |
| 15 | 16 | a | c | 3270.0 | NaN | 2013 |
| 18 | 19 | a | c | 3240.0 | NaN | 2013 |
| 21 | 22 | a | a | 1040.0 | NaN | 2013 |
| ... | ... | ... | ... | ... | ... | ... |
| 1095 | 1096 | a | c | 1130.0 | NaN | 2013 |
| 1099 | 1100 | a | a | 540.0 | NaN | 2013 |
| 1112 | 1113 | a | c | 9260.0 | NaN | 2013 |
| 1113 | 1114 | a | c | 870.0 | NaN | 2013 |
| 1114 | 1115 | d | c | 5350.0 | NaN | 2013 |

354 rows × 10 columns

In [19]:

```
1 store_info_df[ store_info_df['Promo2'] == 0]
```

Out[19]:

| | Store | StoreType | Assortment | CompetitionDistance | CompetitionOpenSinceMonth | CompetitionOpenSinceYear |
|------|-------|-----------|------------|---------------------|---------------------------|--------------------------|
| 0 | 1 | c | a | 1270.0 | 9.0 | 2013 |
| 3 | 4 | c | c | 620.0 | 9.0 | 2013 |
| 4 | 5 | a | a | 29910.0 | 4.0 | 2013 |
| 5 | 6 | a | a | 310.0 | 12.0 | 2013 |
| 6 | 7 | a | c | 24000.0 | 4.0 | 2013 |
| ... | ... | ... | ... | ... | ... | ... |
| 1107 | 1108 | a | a | 540.0 | 4.0 | 2013 |
| 1109 | 1110 | c | c | 900.0 | 9.0 | 2013 |
| 1111 | 1112 | c | c | 1880.0 | 4.0 | 2013 |
| 1112 | 1113 | a | c | 9260.0 | NaN | 2013 |
| 1113 | 1114 | a | c | 870.0 | NaN | 2013 |

544 rows × 10 columns

In [20]:

```
1 #if 'promo2' is zero, 'promo2SinceWeek', 'Promo2SinceYear', and 'PromoInterval' info
2 str_cols = ['Promo2SinceWeek', 'Promo2SinceYear', 'PromoInterval', 'CompetitionOpenS
3
4 for str in str_cols:
5     store_info_df[str].fillna(0, inplace = True)
```

In [21]:

```
1 #checkig NULL Values
2 store_info_df.isnull().sum()
```

Out[21]:

```
Store                0
StoreType            0
Assortment           0
CompetitionDistance  3
CompetitionOpenSinceMonth  0
CompetitionOpenSinceYear  0
Promo2              0
Promo2SinceWeek     0
Promo2SinceYear     0
PromoInterval       0
dtype: int64
```

In [22]:

```
1 #average values of the 'CompetitionDistance' column
2 store_info_df['CompetitionDistance'].fillna(store_info_df['CompetitionDistance'].mea
```

In [23]:

```
1 #checkig NULL Values
2 store_info_df.isnull().sum()
```

Out[23]:

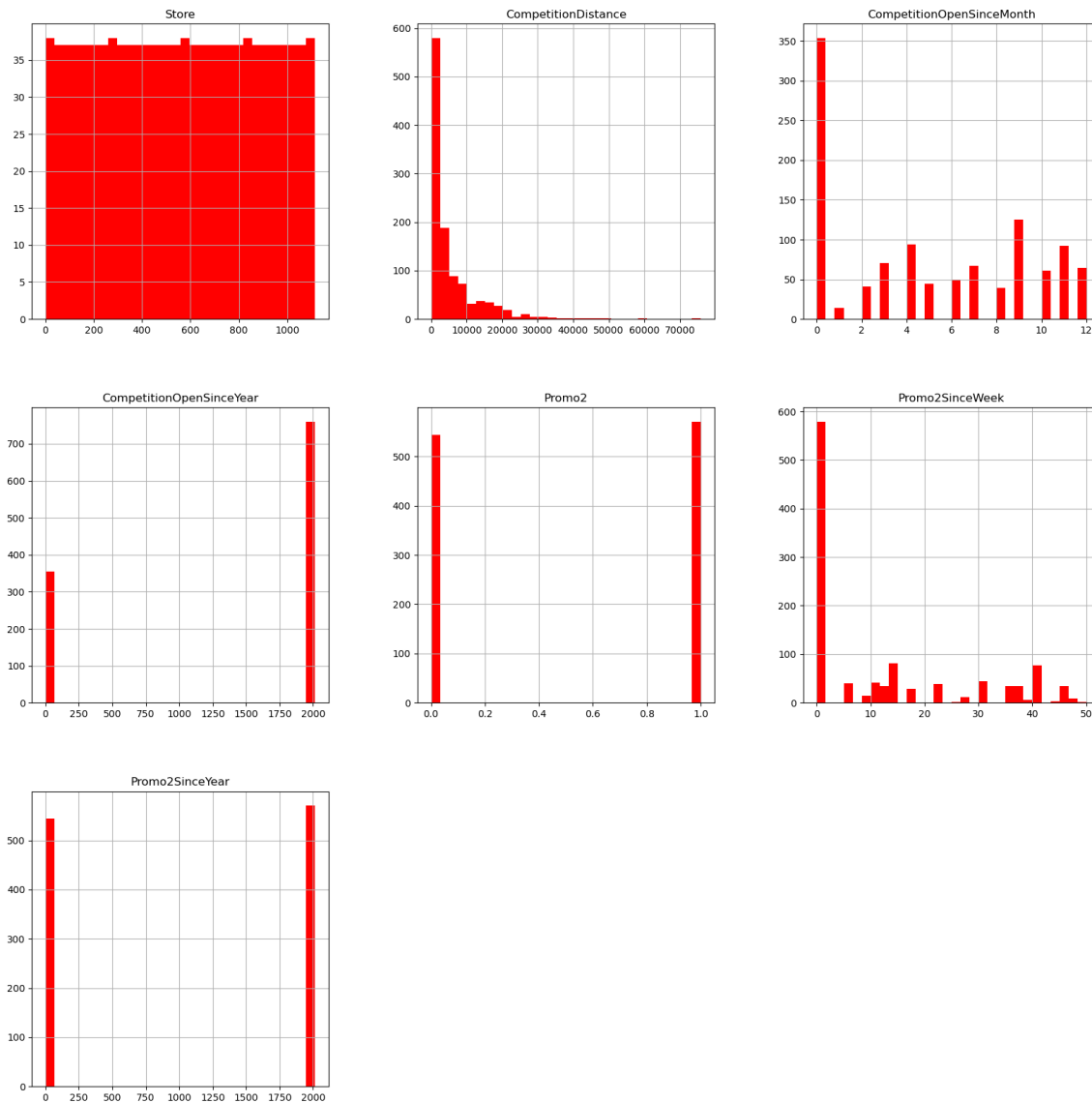
```
Store                0
StoreType            0
Assortment           0
CompetitionDistance  0
CompetitionOpenSinceMonth  0
CompetitionOpenSinceYear  0
Promo2              0
Promo2SinceWeek     0
Promo2SinceYear     0
PromoInterval       0
dtype: int64
```

In [24]:

```
1 store_info_df.hist(bins = 30, figsize = (20,20), color = 'r')
```

Out[24]:

```
array([[<AxesSubplot: title={'center': 'Store'}>,
        <AxesSubplot: title={'center': 'CompetitionDistance'}>,
        <AxesSubplot: title={'center': 'CompetitionOpenSinceMonth'}>],
       [<AxesSubplot: title={'center': 'CompetitionOpenSinceYear'}>,
        <AxesSubplot: title={'center': 'Promo2'}>,
        <AxesSubplot: title={'center': 'Promo2SinceWeek'}>],
       [<AxesSubplot: title={'center': 'Promo2SinceYear'}>,
        <AxesSubplot: >, <AxesSubplot: >]], dtype=object)
```



In [25]:

```
1 # merge both data frames together based on 'store'
2 sales_train_all_df = pd.merge(sales_train_df, store_info_df, how = 'inner', on = 'St
```

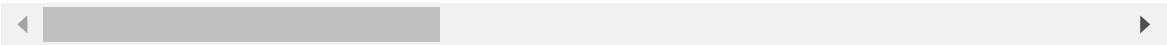
In [26]:

```
1 sales_train_all_df
```

Out[26]:

| | Store | DayOfWeek | Date | Sales | Customers | Promo | StateHoliday | SchoolHoliday | S |
|--------|-------|-----------|------------|-------|-----------|-------|--------------|---------------|-----|
| 0 | 1 | 5 | 2015-07-31 | 5263 | 555 | 1 | 0 | 1 | |
| 1 | 1 | 4 | 2015-07-30 | 5020 | 546 | 1 | 0 | 1 | |
| 2 | 1 | 3 | 2015-07-29 | 4782 | 523 | 1 | 0 | 1 | |
| 3 | 1 | 2 | 2015-07-28 | 5011 | 560 | 1 | 0 | 1 | |
| 4 | 1 | 1 | 2015-07-27 | 6102 | 612 | 1 | 0 | 1 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 844387 | 292 | 1 | 2013-01-07 | 9291 | 1002 | 1 | 0 | 0 | |
| 844388 | 292 | 6 | 2013-01-05 | 2748 | 340 | 0 | 0 | 0 | |
| 844389 | 292 | 5 | 2013-01-04 | 4202 | 560 | 0 | 0 | 1 | |
| 844390 | 292 | 4 | 2013-01-03 | 4580 | 662 | 0 | 0 | 1 | |
| 844391 | 292 | 3 | 2013-01-02 | 5076 | 672 | 0 | 0 | 1 | |

844392 rows × 17 columns



In [27]:

```

1 #correlation
2 correlations = sales_train_all_df.corr()
3 f, ax = plt.subplots(figsize = (20, 20))
4 sns.heatmap(correlations, annot = True)
5 # Customers/Prmo2 and sales are strongly correlated

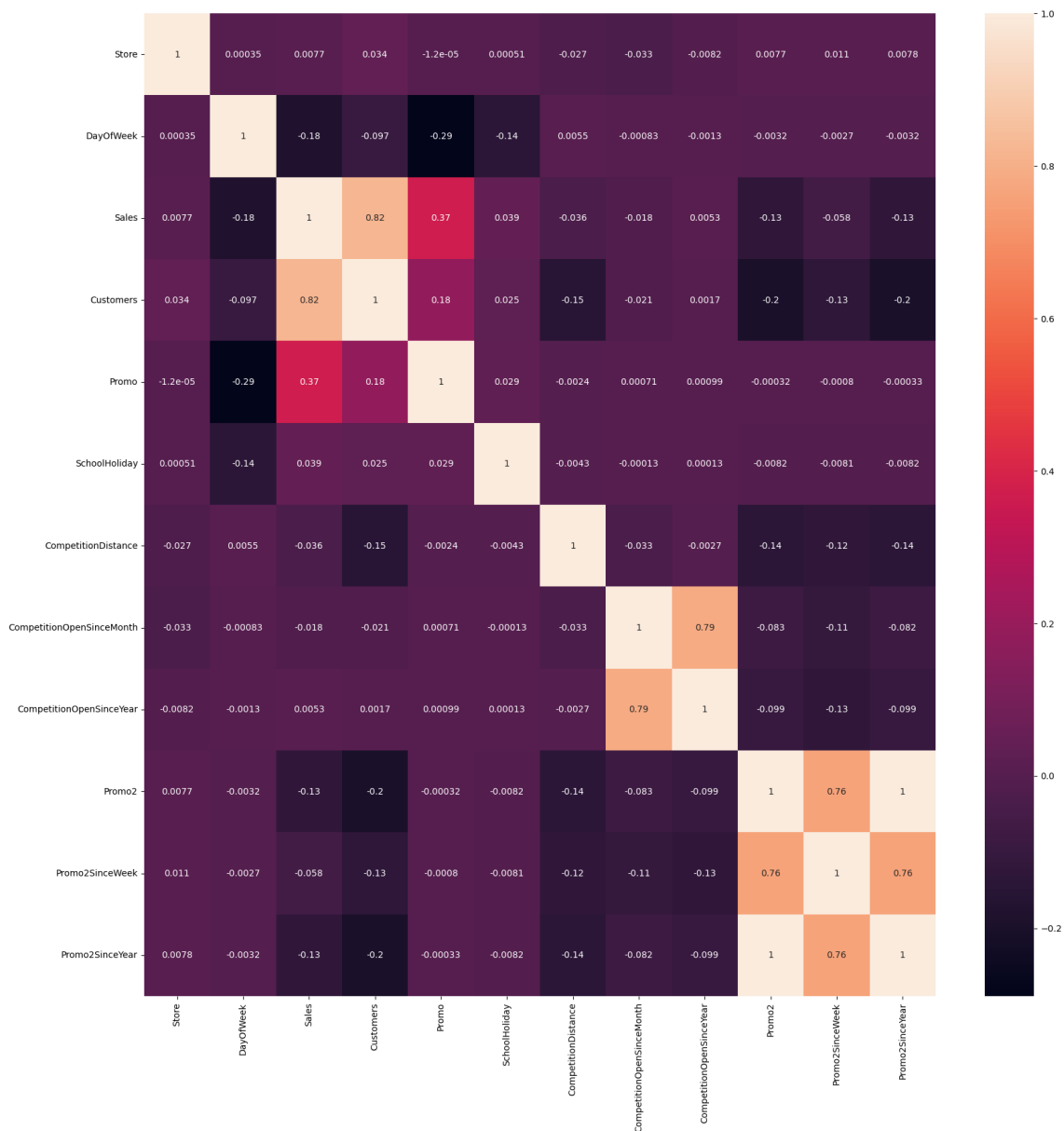
```

C:\Users\Pradeep\AppData\Local\Temp\ipykernel_24732\1938977837.py:2: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.

```
correlations = sales_train_all_df.corr()
```

Out[27]:

<AxesSubplot: >



In [28]:

```
1 correlations = sales_train_all_df.corr()['Sales'].sort_values()  
2 correlations
```

C:\Users\Pradeep\AppData\Local\Temp\ipykernel_24732\1950390705.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.

```
correlations = sales_train_all_df.corr()['Sales'].sort_values()
```

Out[28]:

| | |
|---------------------------|-----------|
| DayOfWeek | -0.178736 |
| Promo2SinceYear | -0.127621 |
| Promo2 | -0.127596 |
| Promo2SinceWeek | -0.058476 |
| CompetitionDistance | -0.036343 |
| CompetitionOpenSinceMonth | -0.018370 |
| CompetitionOpenSinceYear | 0.005266 |
| Store | 0.007710 |
| SchoolHoliday | 0.038617 |
| Promo | 0.368145 |
| Customers | 0.823597 |
| Sales | 1.000000 |

Name: Sales, dtype: float64

In [29]:

```
1 # customers and promo are positively correlated with the sales  
2 # Promo2 does not seem to be effective at all
```

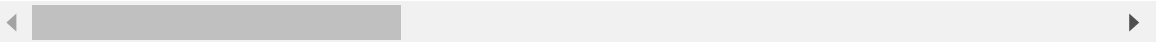
In [30]:

```
1 #separating the year,month and day put it into a separate column
2 sales_train_all_df['Year'] = pd.DatetimeIndex(sales_train_all_df['Date']).year
3 sales_train_all_df['Month'] = pd.DatetimeIndex(sales_train_all_df['Date']).month
4 sales_train_all_df['Day'] = pd.DatetimeIndex(sales_train_all_df['Date']).day
5 sales_train_all_df
```

Out[30]:

| | Store | DayOfWeek | Date | Sales | Customers | Promo | StateHoliday | SchoolHoliday | S |
|--------|-------|-----------|------------|-------|-----------|-------|--------------|---------------|-----|
| 0 | 1 | 5 | 2015-07-31 | 5263 | 555 | 1 | 0 | 1 | |
| 1 | 1 | 4 | 2015-07-30 | 5020 | 546 | 1 | 0 | 1 | |
| 2 | 1 | 3 | 2015-07-29 | 4782 | 523 | 1 | 0 | 1 | |
| 3 | 1 | 2 | 2015-07-28 | 5011 | 560 | 1 | 0 | 1 | |
| 4 | 1 | 1 | 2015-07-27 | 6102 | 612 | 1 | 0 | 1 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 844387 | 292 | 1 | 2013-01-07 | 9291 | 1002 | 1 | 0 | 0 | |
| 844388 | 292 | 6 | 2013-01-05 | 2748 | 340 | 0 | 0 | 0 | |
| 844389 | 292 | 5 | 2013-01-04 | 4202 | 560 | 0 | 0 | 1 | |
| 844390 | 292 | 4 | 2013-01-03 | 4580 | 662 | 0 | 0 | 1 | |
| 844391 | 292 | 3 | 2013-01-02 | 5076 | 672 | 0 | 0 | 1 | |

844392 rows × 20 columns

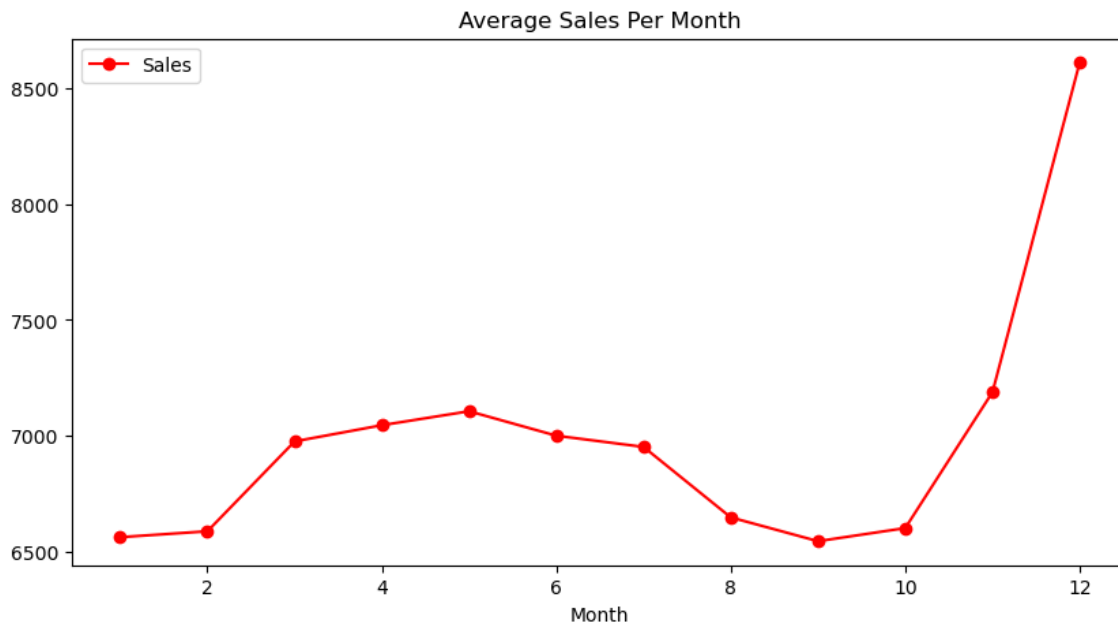


In [31]:

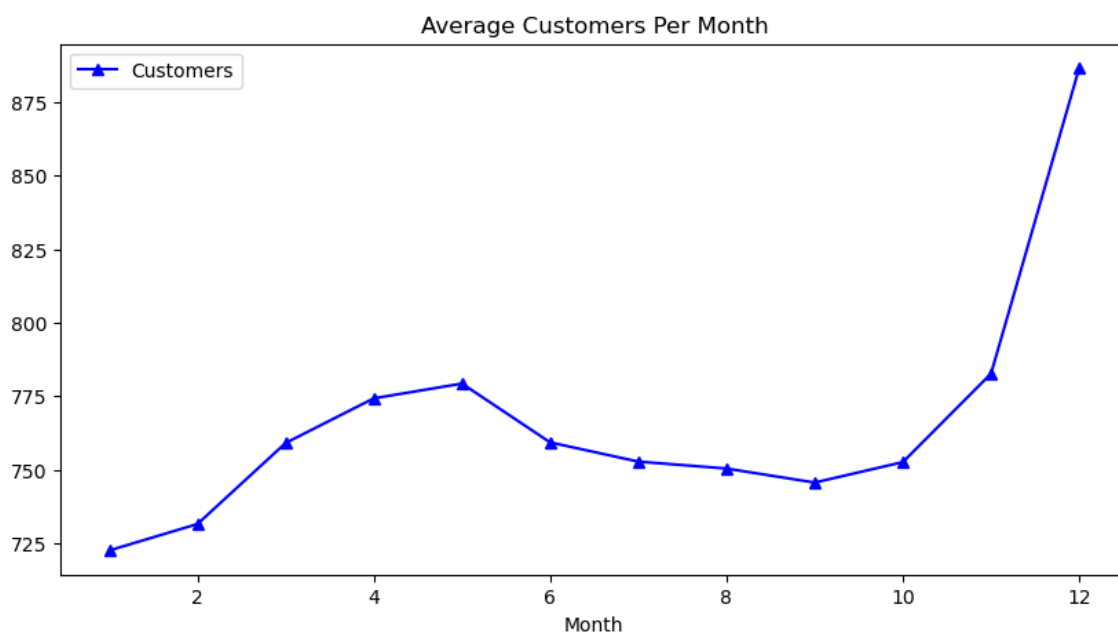
```
1 #average sales and customers per month
2 axis = sales_train_all_df.groupby('Month')[['Sales']].mean().plot(figsize = (10,5),
3 axis.set_title('Average Sales Per Month')
4
5 plt.figure()
6 axis = sales_train_all_df.groupby('Month')[['Customers']].mean().plot(figsize = (10,
7 axis.set_title('Average Customers Per Month')
8
```

Out[31]:

Text(0.5, 1.0, 'Average Customers Per Month')



<Figure size 640x480 with 0 Axes>

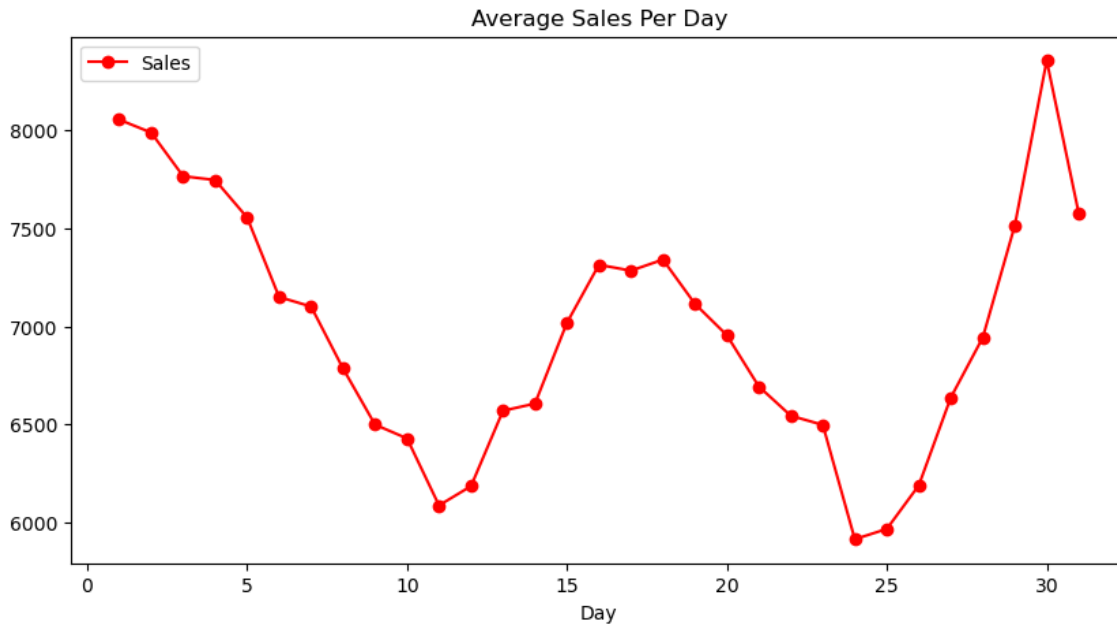


In [32]:

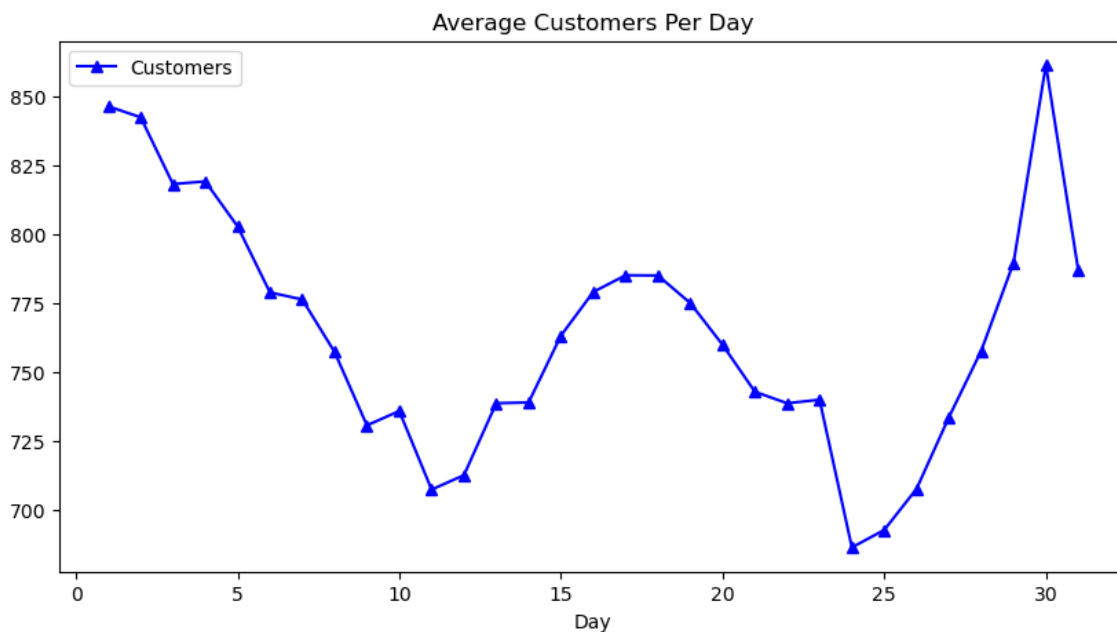
```
1 #average sales and customers per day
2 axis = sales_train_all_df.groupby('Day')[['Sales']].mean().plot(figsize = (10,5), ma
3 axis.set_title('Average Sales Per Day')
4
5 plt.figure()
6 axis = sales_train_all_df.groupby('Day')[['Customers']].mean().plot(figsize = (10,5)
7 axis.set_title('Average Customers Per Day')
8
```

Out[32]:

Text(0.5, 1.0, 'Average Customers Per Day')



<Figure size 640x480 with 0 Axes>

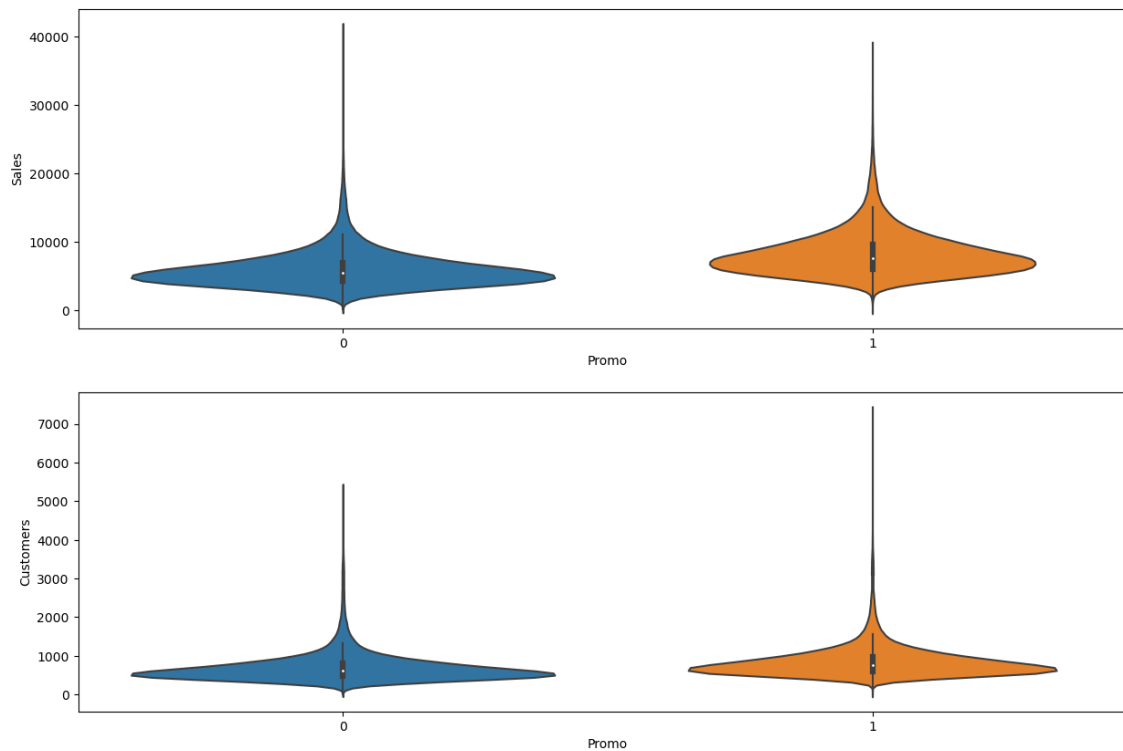


In [33]:

```
1 #violin plot
2 plt.figure(figsize=[15,10])
3
4 plt.subplot(211)
5 sns.violinplot(x = 'Promo', y = 'Sales', data = sales_train_all_df)
6
7 plt.subplot(212)
8 sns.violinplot(x = 'Promo', y = 'Customers', data = sales_train_all_df)
```

Out[33]:

<AxesSubplot: xlabel='Promo', ylabel='Customers'>



In [34]:

```
1 #training model
2 from prophet import Prophet
```

In [35]:

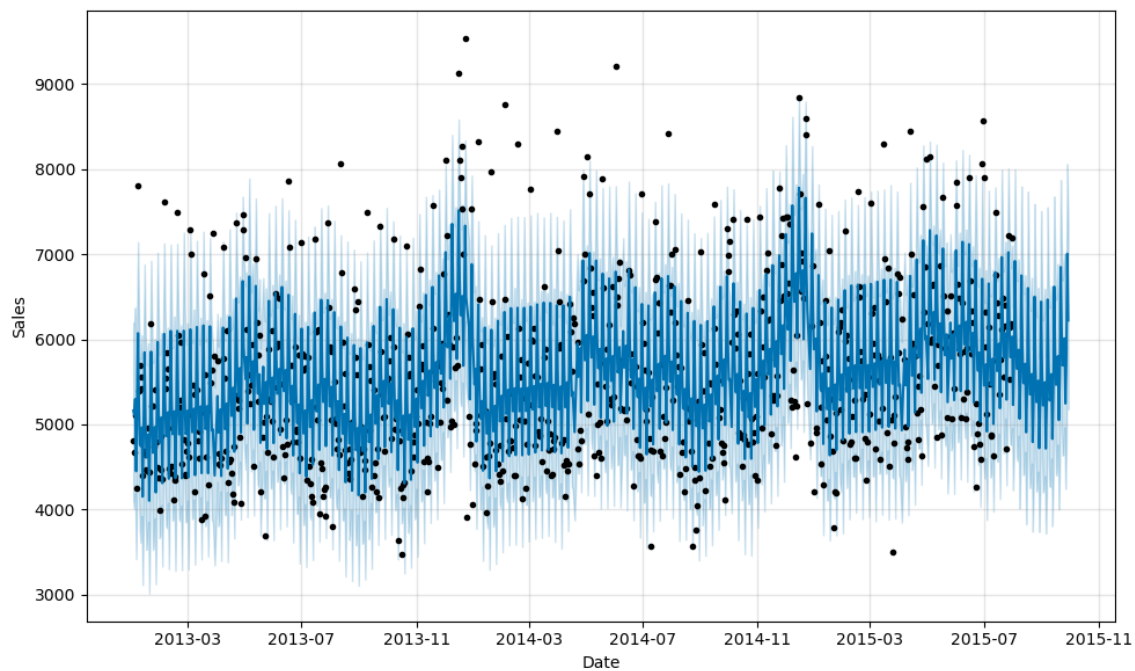
```
1 def sales_prediction(Store_ID, sales_df, periods):
2     # Function that takes in the data frame, storeID, and number of future period fore
3     # The function then generates date/sales columns in Prophet format
4     # The function then makes time series predictions
5
6     sales_df = sales_df[ sales_df['Store'] == Store_ID ]
7     sales_df = sales_df[['Date', 'Sales']].rename(columns = {'Date': 'ds', 'Sales':'y'})
8     sales_df = sales_df.sort_values('ds')
9     #print(sales_df)
10    model = Prophet()
11    model.fit(sales_df)
12    future = model.make_future_dataframe(periods=periods)
13
14    forecast = model.predict(future)
15    figure = model.plot(forecast, xlabel='Date', ylabel='Sales')
16    figure2 = model.plot_components(forecast)
```

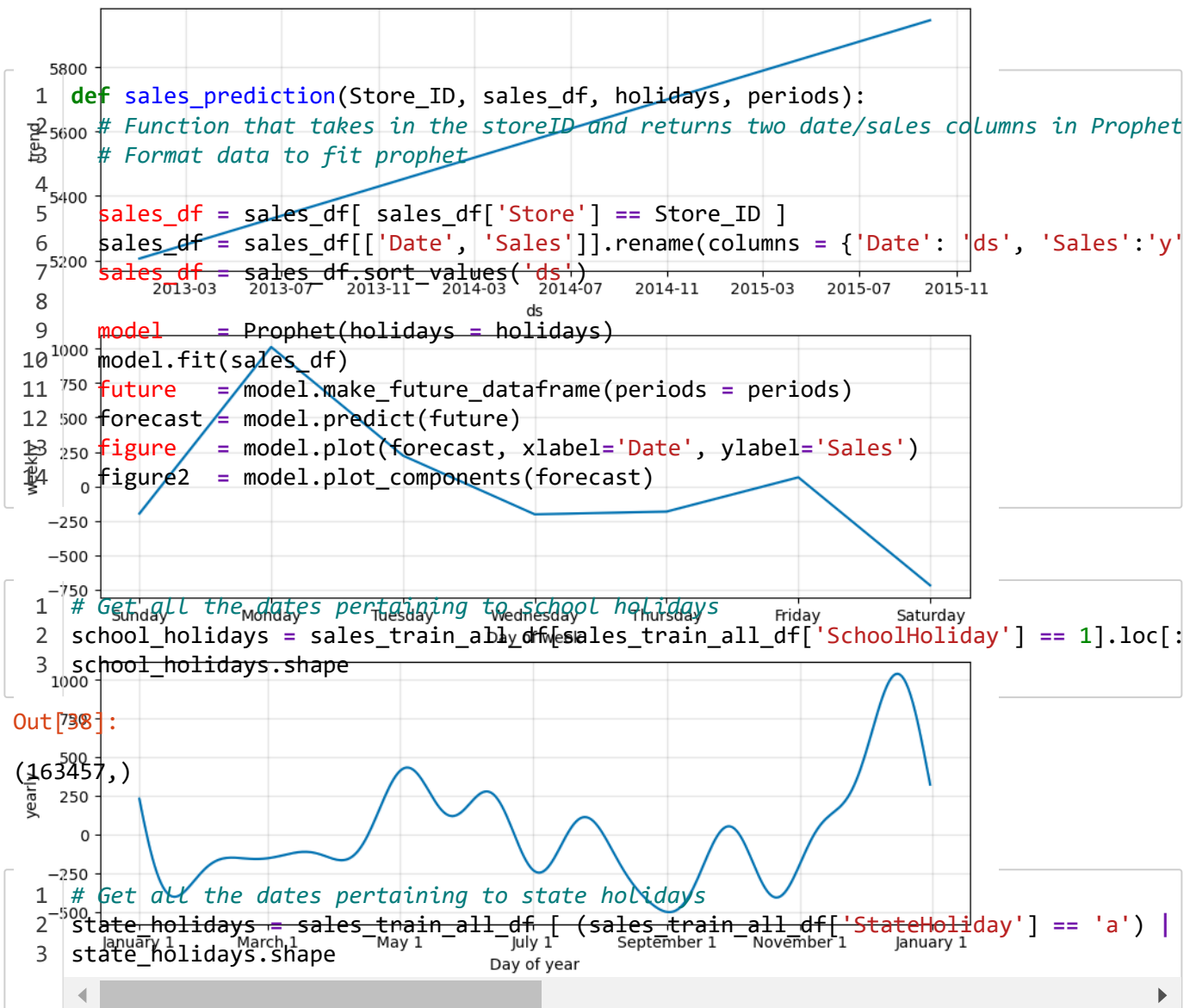
In [36]:

```
1 sales_prediction(10, sales_train_all_df, 60)
```

18:24:35 - cmdstanpy - INFO - Chain [1] start processing

18:24:35 - cmdstanpy - INFO - Chain [1] done processing





| | ds | holiday |
|-----|------------|---------------|
| 0 | 2014-10-03 | state_holiday |
| 1 | 2013-10-03 | state_holiday |
| 2 | 2015-06-04 | state_holiday |
| 3 | 2014-06-19 | state_holiday |
| 4 | 2013-05-30 | state_holiday |
| .. | ... | ... |
| 905 | 2013-04-01 | state_holiday |
| 906 | 2013-08-15 | state_holiday |
| 907 | 2015-06-04 | state_holiday |
| 908 | 2014-06-19 | state_holiday |
| 909 | 2013-05-30 | state_holiday |

[910 rows x 2 columns]

In [41]:

```
1 school_holidays = pd.DataFrame({'ds': pd.to_datetime(school_holidays),
2                                'holiday': 'school_holiday'})
3 print(school_holidays)
```

| | ds | holiday |
|--------|------------|----------------|
| 0 | 2015-07-31 | school_holiday |
| 1 | 2015-07-30 | school_holiday |
| 2 | 2015-07-29 | school_holiday |
| 3 | 2015-07-28 | school_holiday |
| 4 | 2015-07-27 | school_holiday |
| ... | ... | ... |
| 163452 | 2013-02-05 | school_holiday |
| 163453 | 2013-02-04 | school_holiday |
| 163454 | 2013-01-04 | school_holiday |
| 163455 | 2013-01-03 | school_holiday |
| 163456 | 2013-01-02 | school_holiday |

[163457 rows x 2 columns]

In [42]:

```
1 # concatenate both school and state holidays
2 school_state_holidays = pd.concat((state_holidays, school_holidays))
3
4 school_state_holidays.groupby('holiday').holiday.count()
```

Out[42]:

```
holiday
school_holiday    163457
state_holiday      910
Name: holiday, dtype: int64
```

In [43]:

```
1 # Let's make predictions using holidays for a specific store  
2 sales_prediction(6, sales_train_all_df, school_state_holidays, 60)
```

18:24:53 - cmdstanpy - INFO - Chain [1] start processing

18:24:53 - cmdstanpy - INFO - Chain [1] done processing

