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Check data:

In [3]:
```
#import
df1 = pd.read_csv(the_path + "Assignment 2 - USA_AL_Auburn-Opelika.AP.722284_TMY3_BASE.csv")

df2 = pd.read_csv(the_path + "Assignment 2 - new.app4.csv")
```

In [4]:
```
list(df1.columns)
```

Out[4]:
```
['Date/Time',
 'Electricity:Facility [kW](Hourly)',
 'Gas:Facility [kW](Hourly)',
 'Heating:Electricity [kW](Hourly)',
 'Heating:Gas [kW](Hourly)',
 'Cooling:Electricity [kW](Hourly)',
 'HVACFan:Fans:Electricity [kW](Hourly)',
 'Electricity:HVAC [kW](Hourly)',
 'Fans:Electricity [kW](Hourly)',
 'General:InteriorLights:Electricity [kW](Hourly)',
 'General:ExteriorLights:Electricity [kW](Hourly)',
 'Appl:InteriorEquipment:Electricity [kW](Hourly)',
 'Misc:InteriorEquipment:Electricity [kW](Hourly)',
 'Water Heater:WaterSystems:Electricity [kW](Hourly) ']
```

In [5]:
```
list(df2.columns)
```

Out[5]:
```
['Unnamed: 0', 'time', 'W_min']
```

Transform 2 data source into same format:

In [8]:
```
#data info display
df22 = df2.copy()  # Make copy

df22['time'] = pd.to_datetime(df22['time'])
# Drop the year component
df22['time'] = df22['time'].dt.strftime('%m/%d %H:%M')
df22 >> head(5)
```

Out[8]:

	Unnamed: 0	time	W_min
0	1	06/07 11:04	1142.919571
1	2	06/07 11:05	371.239567
2	3	06/07 11:06	367.887333
3	4	06/07 11:07	702.714100
4	5	06/07 11:08	1655.944450

Double check the data range:

```
#double check the data accuracy
start_time = df22['time'].min()
end_time = df22['time'].max()

"""extract the dataframe timeline"""
print("Starts from: " + str(start_time) + "\nEnds at: " + str(end_time))
```

```
Starts from: 06/07 11:04
Ends at: 09/17 23:10
```

Due to the formatting issues, some data needs to be recoded and transformed into DATETIME, such as 24:00 to 00:00:

And I summarize the minute usage into hourly usage in W/hour:

```
# Group and sum the data by date and hour
df22_hour = df22.groupby(['date', 'hour'])['W_min'].sum()
# Reset index to make 'date' and 'hour' columns back to regular columns
df22_hour = df22_hour.reset_index()

#rename columns
df22_hour.rename(columns={'W_min': 'W_hour'}, inplace=True)
df22_hour >> head(5)
```

	date	hour	W_hour
0	06/07	11	57388.943382
1	06/07	12	27227.961318
2	06/07	13	111476.298141
3	06/07	14	109021.960420
4	06/07	15	5773.963306

While checking the data before merging, I found out that, there are about 65 hours of data are still missing, and this might result in unreasonable trend variation.

Out[14]:

	time	W_hour
0	06/07 11:00:00	57388.943382
1	06/07 12:00:00	27227.961318
2	06/07 13:00:00	111476.298141
3	06/07 14:00:00	109021.960420
4	06/07 15:00:00	5773.963306

double check with calculation, 10846 data / 60 = 180hours.
however dataset includes 245 hours, therefore not all of minute usages are recorded.

After checking the data missing, I merged the data and transfer the Watt into kilowatts and form the new df:

```
#transfer W_min to kW_min
merged2_out['kW_hour'] = merged2_out['W_hour'] / 1000
merged2_out = merged2_out >> drop(['W_hour'])

#sum
merged2_out['Sum'] = merged2_out.drop('time', axis=1).sum(axis=1)
merged2_out >> select(['time', 'Sum']) >> head(5)
```

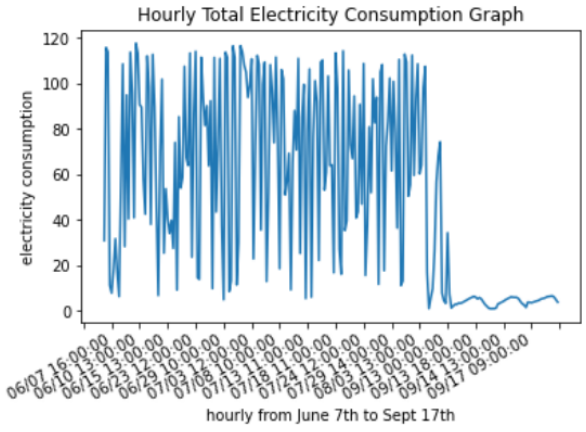
l6]:

	time	Sum
0	06/07 11:00:00	57.388943
1	06/07 12:00:00	31.065016
2	06/07 13:00:00	115.828105
3	06/07 14:00:00	113.919095
4	06/07 15:00:00	11.407785

There are 2 merge methods, and I decided to keep it with inner merging since there will be too much missing data for many electricity consumption sources in the first csv file.

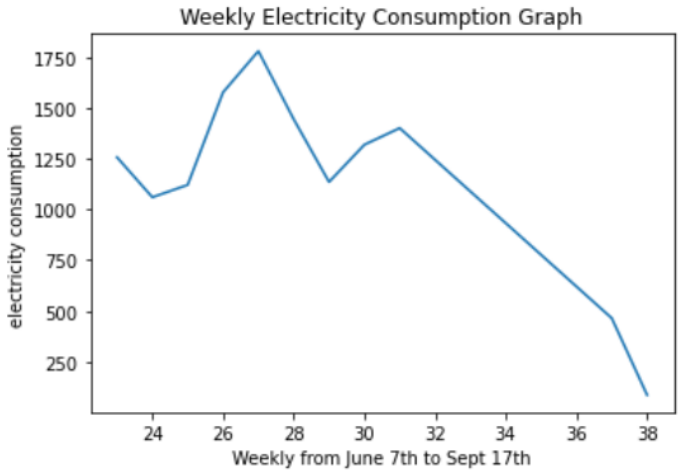
By merging, I am able to produce the Hourly Total Electricity Consumption graph:

```
plt.plot(merged2['time'], merged2['Sum'])

# Format X-label
plt.gca().xaxis.set_major_locator(mdates.DayLocator(interval=15)) # Show ticks every x days

plt.xlabel('hourly from June 7th to Sept 17th')
plt.ylabel('electricity consumption')
plt.title('Hourly Total Electricity Consumption Graph')
plt.gcf().autofmt_xdate()

plt.show()
```



By merging the data of hourly consumption again, I got the daily data, and then I collected the weekly data and generated the Weekly Total Electricity Consumption Graph:

In [23]:
```
plt.plot(merge_weekly['week_num'], merge_weekly['Sum'])
plt.xlabel('Weekly from June 7th to Sept 17th')
plt.ylabel('electricity consumption')
plt.title('Weekly Electricity Consumption Graph')
plt.show()
```



June 7th to Sept 17th, which is from 24th week to 38th week.

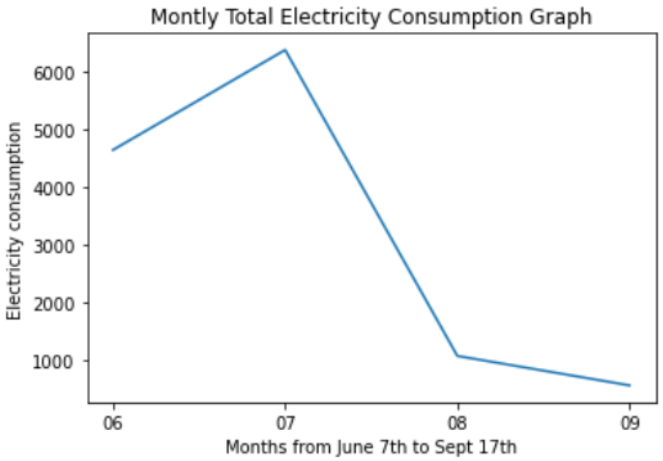
Then, I put the daily data into the month, collected the total consumption for each month from June to September:

:
```
merge_Month = merge_day.copy()
# Group and sum the data by date, for merged2
merge_Month[['month', 'date']] = merge_day['date'].str.split(pat='/', n=1, expand=True)
# Group and sum the data by date and hour
merge_Month = merge_Month.groupby(['month'])['Sum'].sum()
# Reset index
merge_Month = merge_Month.reset_index()
merge_Month >> head(5)
```

[24]:

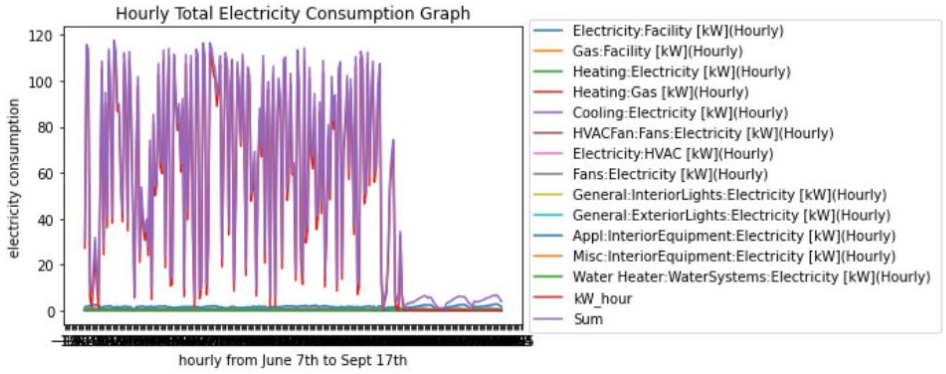
	month	Sum
0	06	4646.109817
1	07	6384.893747
2	08	1063.754975
3	09	551.780369

In [25]:
```
plt.plot(merge_Month['month'], merge_Month['Sum'])
plt.xlabel('Months from June 7th to Sept 17th')
plt.ylabel('Electricity consumption')
plt.title('Montly Total Electricity Consumption Graph')
plt.show()
```



The consumption decreased very strangely, and I decided to further develop the reason behind each electricity consumptions:

In [26]:
```
ax = merged2.plot()

plt.gca().xaxis.set_major_locator(mdates.DayLocator(interval=1))  # Show ticks every x days
plt.xlabel('hourly from June 7th to Sept 17th')
plt.ylabel('electricity consumption')
plt.title('Hourly Total Electricity Consumption Graph')
plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))  # Legend to top right

plt.show()
```



As we could observe from the map, the total consumption is highly correlated to the New Added appliance's electricity consumption, and it takes more than 80% of house consumption. And this might be the reason why the total consumption suddenly decrease in the last few days.