project Zero - Exploring Weather Trends

September 17, 2020

1 Exploring Weather Trends - Project Instructions

1.1 Summary

In this project, you will analyze local and global temperature data and compare the temperature trends where you live to overall global temperature trends.

1.2 Goal

My goal will be to create a visualization and prepare a write up describing the similarities and differences between global temperature trends and temperature trends in the closest big city to where I live.

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2 Project workflow

2.1 Extract the data from the database

I will extract necessary data from database. I will export the temperature data for the world as well as for the closest big city to where I live and that is Dhaka, capital of Bangladesh. I have written SQL query for that.

Firstly, to get the nearest city near me, I checked for all the cities for my country using this command:

```
SELECT * from city_list WHERE country = 'Bangladesh'
```

- **the city level data:** I collected the data for City Dhaka using the following command

```
SELECT * from city_data WHERE city = 'Dhaka'
```

and saved it as a csv with name **city_Data.csv**

- **the global data:** for the global data I ran the command

```
SELECT * from global_data
```

and saved it as csv with name **global_Data.csv**

2.2 Open up the CSV

To open up the CSV file, I have used panda library and used matplotlib for plotting.

```
In [1]: import warnings
        warnings.filterwarnings('ignore')

        import pandas as pd
        import matplotlib.pyplot as plt
```

```
In [2]: city_data = pd.read_csv("city_data.csv")
        global_data = pd.read_csv("global_data.csv")
        city_list = pd.read_csv("city_list.csv")
```

```
In [3]: city_data.info()
        city_data.head()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 218 entries, 0 to 217
Data columns (total 4 columns):
 #   Column    Non-Null Count  Dtype
---  ------    --------------  -----
 0   year      218 non-null    int64
 1   city      218 non-null    object
 2   country   218 non-null    object
 3   avg_temp  213 non-null    float64
dtypes: float64(1), int64(1), object(2)
memory usage: 6.9+ KB
```

Out[3]:

	year	city	country	avg_temp
0	1796	Dhaka	Bangladesh	25.35
1	1797	Dhaka	Bangladesh	26.36
2	1798	Dhaka	Bangladesh	25.22
3	1799	Dhaka	Bangladesh	25.61
4	1800	Dhaka	Bangladesh	25.54

```
In [4]: global_data.info()
        global_data.head()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 266 entries, 0 to 265
Data columns (total 2 columns):
 #   Column    Non-Null Count  Dtype
---  ------    --------------  -----
 0   year      266 non-null    int64
 1   avg_temp  266 non-null    float64
dtypes: float64(1), int64(1)
memory usage: 4.3 KB
```

```
Out[4]:    year  avg_temp
        0  1750      8.72
        1  1751      7.98
        2  1752      5.78
        3  1753      8.39
        4  1754      8.47
```

```
In [5]: #select the cities from Bangladesh out of the city_list dataframe
        city_list[city_list["country"] == "Bangladesh"]
```

```
Out[5]:        city     country
        88    Dhaka  Bangladesh
        152  Khulna  Bangladesh
```

```
In [6]: #create a new reference to the city_data dataframe for the city "Dhaka"
        df_dhaka = city_data[city_data["city"] == "Dhaka"]
        df_dhaka.head()
```

```
Out[6]:    year   city     country  avg_temp
        0  1796  Dhaka  Bangladesh     25.35
        1  1797  Dhaka  Bangladesh     26.36
        2  1798  Dhaka  Bangladesh     25.22
        3  1799  Dhaka  Bangladesh     25.61
        4  1800  Dhaka  Bangladesh     25.54
```

```
In [7]: #create a new reference to the city_data dataframe for the city "Khulna"
        df_khulna = city_data[city_data["city"] == "Khulna"]
        df_khulna.head()
```

```
Out[7]: Empty DataFrame
        Columns: [year, city, country, avg_temp]
        Index: []
```

Here I can see, for Khulna there is no data available. So I will work with Dhaka's data only.

```
In [8]: print('The sum of the missing value in the Dhaka dataframe in the column "avg_temp" is',

        #show missing values in the Dhaka
        df_dhaka[df_dhaka["avg_temp"].isna()]
```

```
The sum of the missing value in the Dhaka dataframe in the column "avg_temp" is 5
```

```
Out[8]:     year   city     country  avg_temp
        12  1808  Dhaka  Bangladesh       NaN
        13  1809  Dhaka  Bangladesh       NaN
        14  1810  Dhaka  Bangladesh       NaN
        15  1811  Dhaka  Bangladesh       NaN
        16  1812  Dhaka  Bangladesh       NaN
```

2.3 Creating line chart

A **line chart** or line plot or line graph or curve chart is a type of chart which displays information as a series of data points called 'markers' connected by straight line segments.

Here I have to create a line chart that compares my city's temperatures with the global temperatures. I have plotted the moving average rather than the yearly averages in order to smooth out the lines, making trends more observable.

```
In [9]: # joining the two csvs data on year column
        combine_data= global_data.merge(city_data, how='left', on= "year",
                                        suffixes=('_global', '_city')).drop(['city', 'country'],

In [10]: combine_data.set_index('year', inplace= True)

In [11]: combine_data.head()
```

Out[11]:

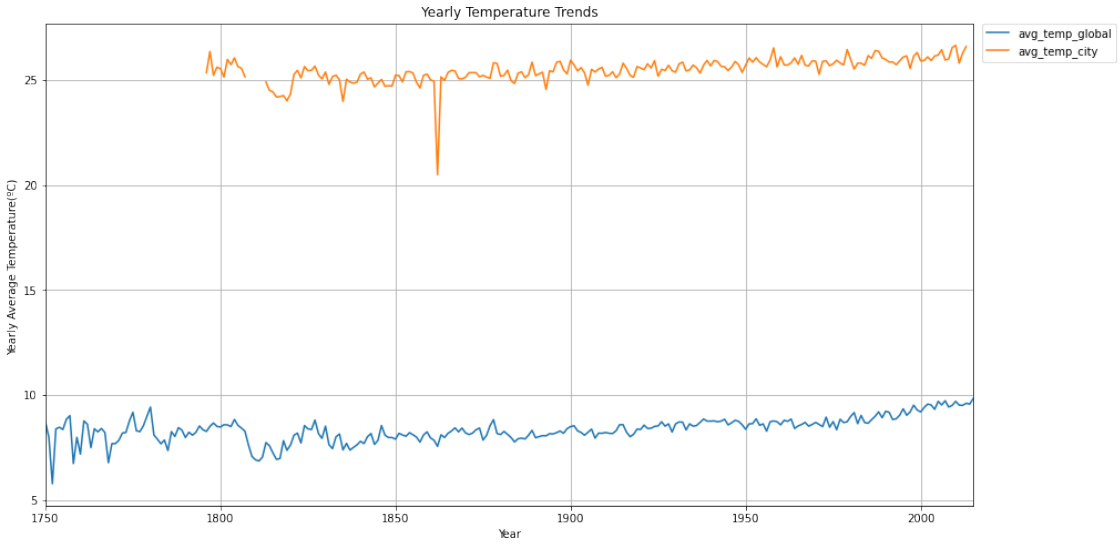
year	avg_temp_global	avg_temp_city
1750	8.72	NaN
1751	7.98	NaN
1752	5.78	NaN
1753	8.39	NaN
1754	8.47	NaN

2.4 Exploring Trends

```
In [12]: title='Yearly Temperature Trends'
         ylabel='Yearly Average Temperature(žC) '
         xlabel='Year'

         fig, ax = plt.subplots()
         combine_data.plot(figsize=(15,8), ax=ax , title= title).autoscale(axis='x',tight=True)
         ax.legend(["Global Average Temperatures", "Dhaka Average Temperatures"]);
         ax.set(xlabel=xlabel, ylabel=ylabel);

         plt.legend(bbox_to_anchor=(1.01, 1), loc='upper left', borderaxespad=0.)
         plt.grid()
         plt.show()
```



2.5 Moving Averages

Moving averages are used to smooth out data to make it easier to observe long term trends and not get lost in daily fluctuations.

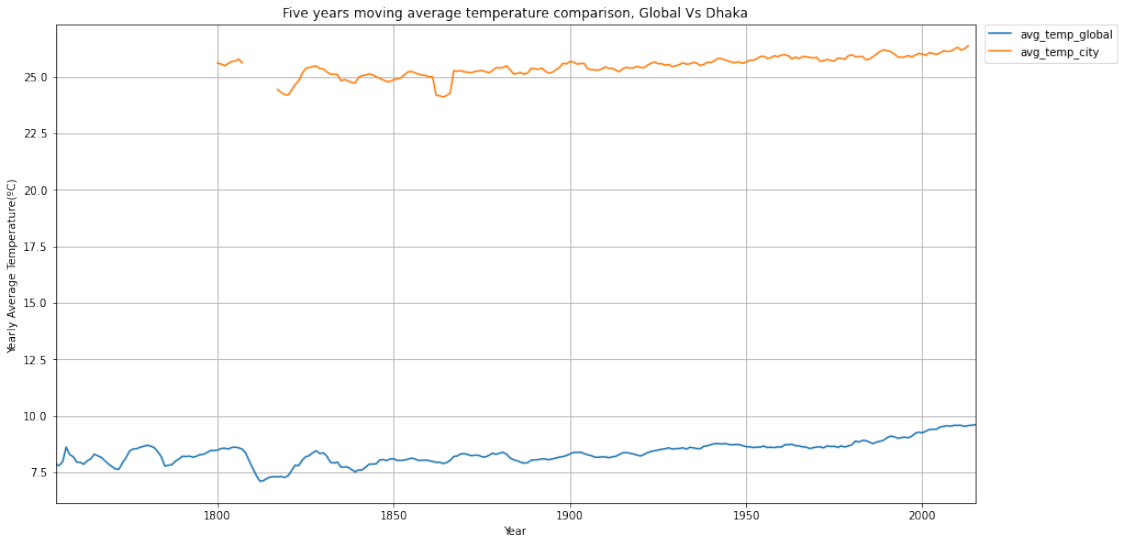
(For example, let's say you wanted to visualize the sales trend at a clothing retail store. You start with daily data, and your chart looks too volatile to interpret because more people shop on the weekends, so sales spike on those days.)

2.5.1 Five Years Moving Average

In [13]:
```
title='Five years moving average temperature comparison, Global Vs Dhaka'
ylabel='Yearly Average Temperature(žC) '
xlabel='Year'

fig, ax = plt.subplots()
combine_data.rolling(window=5).mean().plot(figsize=(15,8), ax=ax , title= title).autosc
#combine_data.rolling(window=5).mean().plot(figsize=(15,8), ax=ax , title= title).autos
ax.legend(["Global Average Temperatures", "City Average Temperatures"]);
ax.set(xlabel=xlabel, ylabel=ylabel);

plt.legend(bbox_to_anchor=(1.01, 1), loc='upper left', borderaxespad=0.)
plt.grid()
plt.show()
```

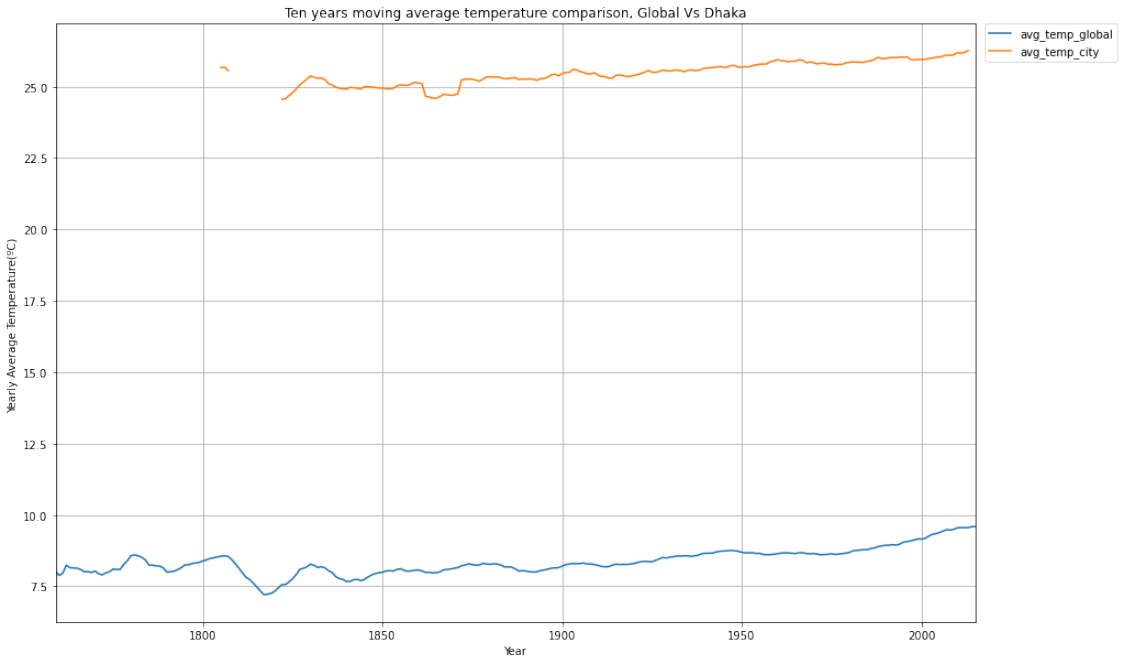


2.5.2 Ten Years Moving Average

In [14]:
```
title='Ten years moving average temperature comparison, Global Vs Dhaka'
ylabel='Yearly Average Temperature(žC) '
xlabel='Year'

fig, ax = plt.subplots()
combine_data.rolling(window=10).mean().plot(figsize=(15,10), ax=ax, title= title).autos
ax.legend(["Global Average Temperatures", "City Average Temperatures"]);
ax.set(xlabel=xlabel, ylabel=ylabel);

plt.legend(bbox_to_anchor=(1.01, 1), loc='upper left', borderaxespad=0.)
plt.grid()
plt.show()
```




```
In [15]: #calculate the mininmal values of the "year" and "avg_temp" column in the global_data d
         print(global_data["year"].min())
         print(global_data["avg_temp"].min())
```

```
1750
5.78
```

```
In [16]: #calculate the maximal values of the "year" and "avg_temp" column in the global_data da
         print(global_data["year"].max())
         print(global_data["avg_temp"].max())
```

```
2015
9.83
```

So here I can see, for in 1750 gobal temparature was the lowest, and in 2015 it is the highest. The lowest temparature was 5.78 and the highest was 9.83

```
In [17]: print(f'The global average temperature (1750 - 2013) is: {global_data["avg_temp"].mean(
```

```
The global average temperature (1750 - 2013) is: 8.37 řC
```

```
In [18]: #calculate the mininmal values of the "year" and "avg_temp" for Dhaka
         print(city_data["year"].min())
         print(city_data["avg_temp"].min())
```

```
1796
20.49
```

```
In [19]: #calculate the maximal values of the "year" and "avg_temp" for Dhaka
         print(city_data["year"].max())
         print(city_data["avg_temp"].max())
```

```
2013
26.65
```

So here I can see, for in 1796 Dhaka's temparature was the lowest, and in 2013 it is the highest. The lowest temparature was 20.49 and the highest was 26.65.

```
In [20]: print(f'Dhakas average temperature (1750 - 2013) is: {city_data["avg_temp"].mean():.2f}
```

```
Dhakas average temperature (1750 - 2013) is: 25.47 řC
```

2.6 Observation of the chart line

Is your city hotter or cooler on average compared to the global average? Has the difference been consistent over time?

- Dhaka is a hotter city than the global average, as the temperature is almost more than double.
- Dhaka's temperature suddenly dropped at 1862, and it is gradually increasing.

How do the changes in your city's temperatures over time compare to the changes in the global average?

- Since 1850 the global temperature has benn increasing quite constantly, while in Dhaka it fluctuated
- Before 1850 the temperature fluctuated more in Dhaka and in the world.

What does the overall trend look like? Is the world getting hotter or cooler? Has the trend been consistent over the last few hundred years?

- the line is not fully constant because from year 1808 to 1812, data is missing.
- Even thought for Global data, its available from 1750, but for Bangladesh, Dhaka it is available from 1796.

If we compare to Dhaka's temparature with global, we can see Dhaka's avarage temparature is very high compared to global temparature.