

weather_trend

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1 Data Analyst Nanodegree Project 0

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

Step 1: Fetching data and converting it to CSV Downloaded all the 3 datasets using the following SQL Queries.

```
SELECT * FROM city_data;
```

```
SELECT * FROM global_data;
```

```
SELECT * FROM city_list;
```

All the *.csv are added to the workspace "Project0" folder.

Step 2: Importing the data Used Pandas, matplotlib and seaborn for data analysis.

```
[72]: #importing packages
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%pdb
%matplotlib inline
%config InlineBackend.figure_format = 'retina'#to have high resolution image
```

Automatic pdb calling has been turned ON

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

Step 3: Preview Data

```
[53]: global_data.head()
```

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

```
[54]: city_data.head()
```

```
[54]:   year   city      country  avg_temp
      0  1849  Abidjan  Côte D'Ivoire    25.58
      1  1850  Abidjan  Côte D'Ivoire    25.52
      2  1851  Abidjan  Côte D'Ivoire    25.67
      3  1852  Abidjan  Côte D'Ivoire     NaN
      4  1853  Abidjan  Côte D'Ivoire     NaN
```

```
[55]: city_list.head()
```

```
[55]:   city      country
      0  Abidjan  Côte D'Ivoire
      1  Abu Dhabi  United Arab Emirates
      2   Abuja      Nigeria
      3  Accra      Ghana
      4  Adana      Turkey
```

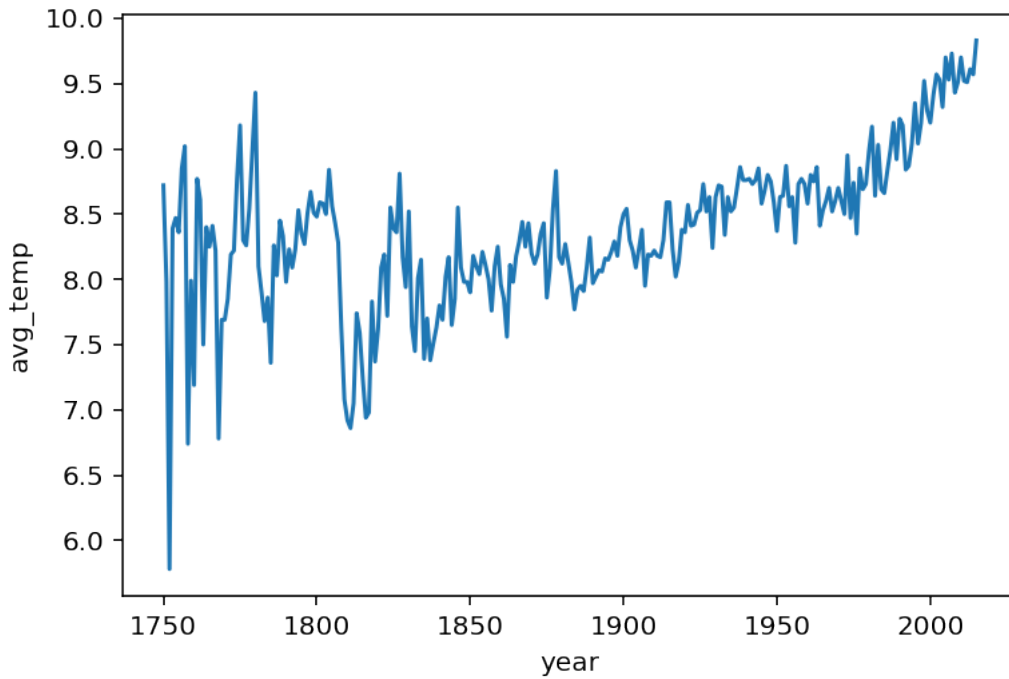
```
[56]: # filltering banglore data from city data
      city_blr = city_data[city_data["city"] == "Bangalore"]
      city_blr.head()
```

```
[56]:   year   city country  avg_temp
      6367  1796  Bangalore  India    24.49
      6368  1797  Bangalore  India    25.18
      6369  1798  Bangalore  India    24.65
      6370  1799  Bangalore  India    24.81
      6371  1800  Bangalore  India    24.85
```

Step 4 : Visualizing Data

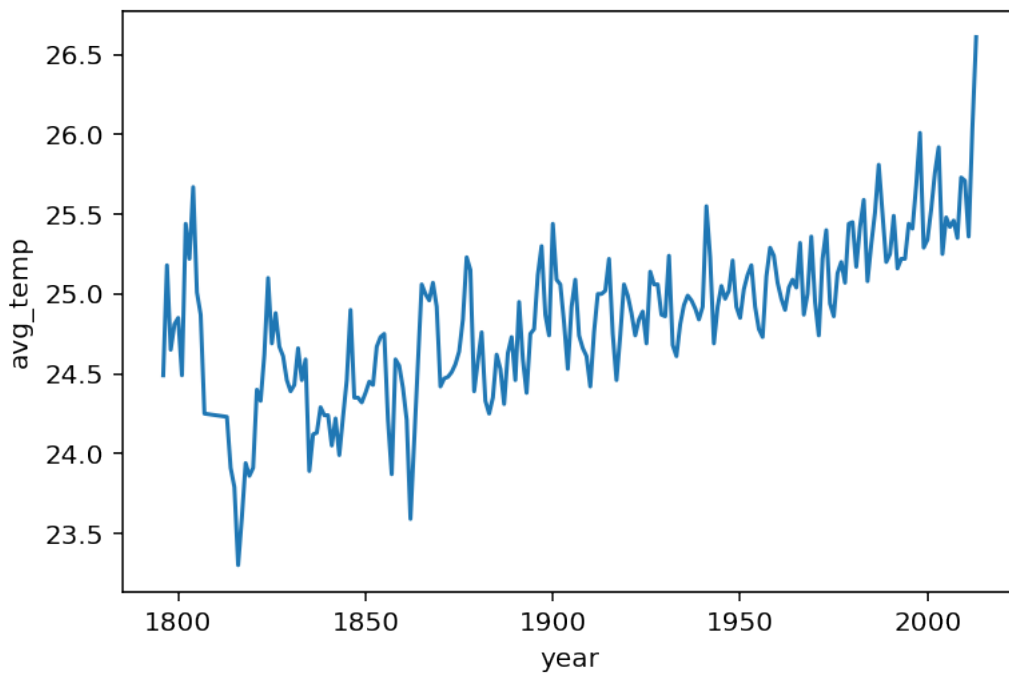
```
[57]: # plotting graph for global data
      sns.lineplot(x = "year",y = "avg_temp",data = global_data)
```

```
[57]: <matplotlib.axes._subplots.AxesSubplot at 0x1dd1bddeb70>
```



```
[58]: # plotting graph for Bangalore city  
sns.lineplot(x = "year", y = "avg_temp", data = city_blr)
```

```
[58]: <matplotlib.axes._subplots.AxesSubplot at 0x1dd190007f0>
```



```
[59]: # Calculated rolling average for global data
global_data=global_data.assign(rolling_gavg=global_data['avg_temp'].rolling(5).
    →mean())
global_data.head()
```

```
[59]:   year  avg_temp  rolling_gavg
0  1750      8.72          NaN
1  1751      7.98          NaN
2  1752      5.78          NaN
3  1753      8.39          NaN
4  1754      8.47      7.868
```

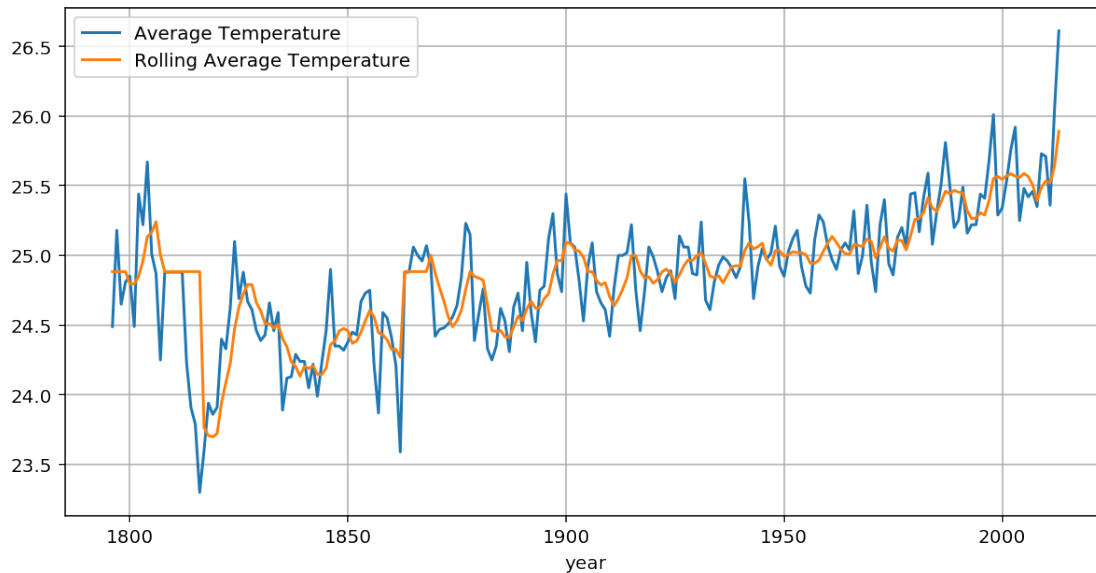
```
[60]: # Calculated rolling average for Bangalore data
city_blr=city_blr.assign(rolling_gavg=city_blr['avg_temp'].rolling(5).mean())
city_blr.head()
```

```
[60]:   year      city country  avg_temp  rolling_gavg
6367  1796  Bangalore   India    24.49          NaN
6368  1797  Bangalore   India    25.18          NaN
6369  1798  Bangalore   India    24.65          NaN
6370  1799  Bangalore   India    24.81          NaN
6371  1800  Bangalore   India    24.85      24.796
```

```
[62]: #finding missing values in local data for average and rolling average
    →temperatures and filling it with respective median
sum(city_blr["avg_temp"].isna())
median=city_blr["avg_temp"].median()
city_blr["avg_temp"].fillna(median,inplace=True)
median1=city_blr["rolling_gavg"].median()
city_blr["rolling_gavg"].fillna(median1,inplace=True)
```

```
[63]: #plotting graph for average temperature vs rolling average temperature of
    →bangalore data
city_blr.plot(x="year", y=
    →['avg_temp','rolling_gavg'],figsize=(10,5),grid=True,label = ['Average
    →Temperature','Rolling Average Temperature'])
```

```
[63]: <matplotlib.axes._subplots.AxesSubplot at 0x1dd190c39b0>
```

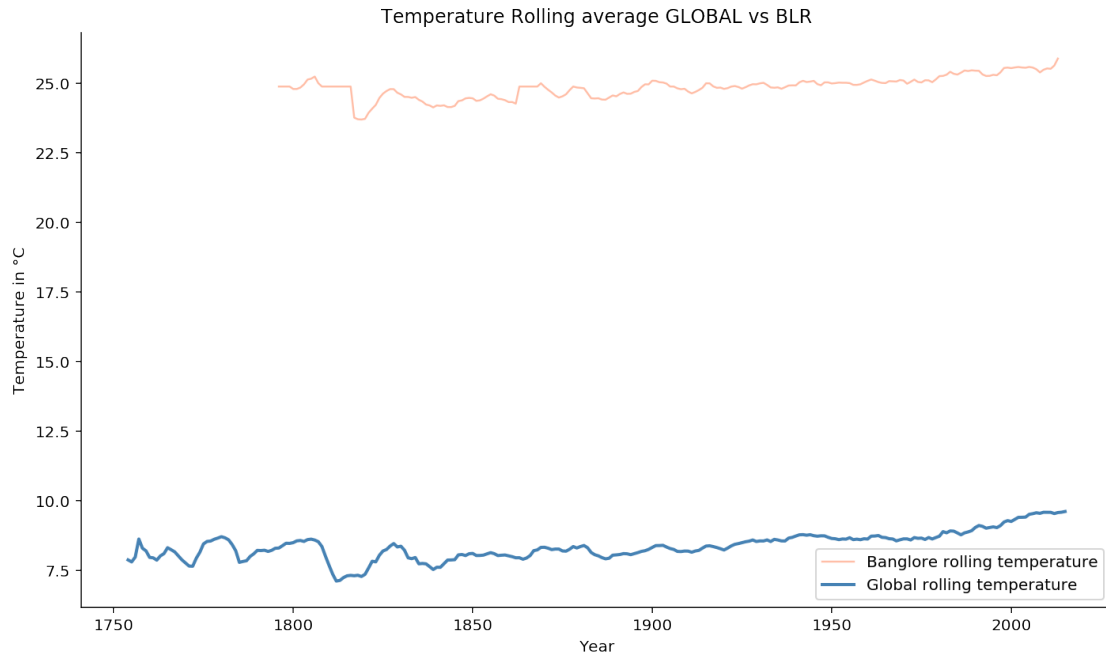


```
[70]: #graph for rolling average temperature for global and bangalore data
#create a matplotlib figure and axes
fig, ax = plt.subplots(figsize = (10,6))

#plot the global data with a lineplot
sns.lineplot(x = "year", y = "rolling_gavg", data = city_blr , label = "Bangalore rolling temperature", alpha = 0.5,
             color = "coral", lw = 1.2)
sns.lineplot(x = "year", y = "rolling_gavg", data = global_data , label = "Global rolling temperature",
             color = "steelblue", lw = 2)

#set the y label of the plot
ax.set_ylabel("Temperature in °C")
#set the x label of the plot
ax.set_xlabel("Year")
#set the title of the plot
ax.set_title("Temperature Rolling average GLOBAL vs BLR")
#disable the right and top spine for better look
ax.spines['right'].set_visible(False)
ax.spines['top'].set_visible(False)
#tighten the plot layout
plt.tight_layout()

#show the plot (basically not necessary for jupyter but I always put it at the end)
plt.show()
```



1.1 Summary

- Bangalore average temperature is hotter than the global temperature and difference is consistent over years Bangalore is about 13.5 degrees hotter than the global temperature
- Bangalore temperature always varied around 23 degree to 26 degrees whereas global temperature varied between 6 degree to 10 degree However both the temperatures are consistent in their variations
- It can be observed that there is a sudden dip in global and local temperatures at the beginning of 19th century where both trends decreased by about 1.5 degrees
- To summarise, there is an overall increase in temperature over years in both globally and locally

[]: