# **Explore Weather Trends**

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

## **Extracting data from database**

- · select \* from global\_data;
- Select year,avg\_temp from city\_data where city='Ranchi' AND country='India';
- Select year, avg\_temp from city\_data where city='Delhi' AND country='India';

## **Collecting Data:**

In [1]:

**217** 2013

25.98

```
import pandas as pd
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
In [2]:
                                                                                             H
df ranchi=pd.read csv('ranchi use.csv')
df_globaldata=pd.read_csv('global_data.csv')
df_delhi=pd.read_csv('delhi.csv')
Exploring City Data
In [3]:
                                                                                             H
df_ranchi.head(1)
Out[3]:
   year
        avg_temp
0 1796
            24.01
In [4]:
                                                                                             Ы
df ranchi.tail(1)
Out[4]:
```

M

The above two operations conclude that city data is given for the years 1849-2013.

```
In [5]:
                                                                                             Ы
df_ranchi.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 218 entries, 0 to 217
Data columns (total 2 columns):
               Non-Null Count Dtype
     Column
 0
     year
               218 non-null
                                int64
     avg_temp 211 non-null
                                float64
1
dtypes: float64(1), int64(1)
memory usage: 3.5 KB
Checking Global Data
                                                                                             H
In [6]:
df_globaldata.head(1)
Out[6]:
   year avg_temp
0 1750
             8.72
In [7]:
df globaldata.tail(1)
Out[7]:
     year avg_temp
265 2015
               9.83
The above two operations conclude that global data is given for the years 1750-2015.
In [8]:
df globaldata.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 266 entries, 0 to 265
Data columns (total 2 columns):
               Non-Null Count Dtype
     Column
0
               266 non-null
                                int64
     year
     avg_temp 266 non-null
                                float64
dtypes: float64(1), int64(1)
memory usage: 4.2 KB
```

In [9]:

```
#For better comparison the global and local dataframes should lie within same period of obs
#Creating a new dataframe for this:
df_global=df_globaldata.query('year > 1848 & year < 2014')
```

# Calculating Rolling mean (or Moving average)

### And handling missing data:

(with a window size of 10 years for better visualisation of data)

```
#setting window size
size=10
#calculating rolling mean
df_global["movingtemp"] = df_globaldata["avg_temp"].rolling(window = size).mean()
```

#### Handling missing data for Ranchi:

```
In [11]:

mean_ranchi= df_ranchi.avg_temp.mean()
mean_ranchi
```

#### Out[11]:

24.17175355450237

```
In [12]:

df_ranchi.avg_temp.fillna(mean_ranchi,inplace=True)
```

```
In [13]:

#setting the window size
size=10
#calculate the rolling mean
```

df\_ranchi['movingtemp']=df\_ranchi['avg\_temp'].rolling(window=size).mean()

#resetting the index of Ranchi dataframe for accurate results:
df\_ranchi.reset\_index(inplace = True, drop = True)

```
In [14]:
                                                                                          H
df ranchi.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 218 entries, 0 to 217
Data columns (total 3 columns):
                 Non-Null Count Dtype
     Column
 #
     ----
                 -----
                                 ----
                 218 non-null
                                 int64
 0
     year
 1
     avg_temp
                 218 non-null
                                 float64
                                 float64
     movingtemp 209 non-null
 2
dtypes: float64(2), int64(1)
memory usage: 5.2 KB
Handling missing data for Delhi:
                                                                                          H
In [15]:
mean_delhi= df_delhi.avg_temp.mean()
mean delhi
Out[15]:
25.16626865671642
In [16]:
                                                                                          H
df_delhi.avg_temp.fillna(mean_delhi,inplace=True)
                                                                                          H
In [17]:
#setting the window size
size=10
#calculate the rolling mean
df_delhi['movingtemp']=df_delhi['avg_temp'].rolling(window=size).mean()
#resetting the index of Delhi dataframe for accurate results:
df_delhi.reset_index(inplace = True, drop = True)
In [18]:
                                                                                          H
df delhi.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 218 entries, 0 to 217
Data columns (total 3 columns):
                 Non-Null Count Dtype
 #
     Column
     -----
                 -----
                                 int64
                 218 non-null
 0
     year
 1
     avg_temp
                 218 non-null
                                 float64
                                 float64
     movingtemp 209 non-null
dtypes: float64(2), int64(1)
memory usage: 5.2 KB
```

# **Plotting Line Graph**

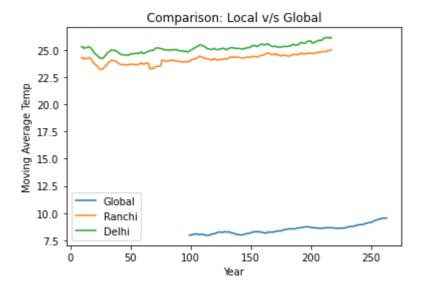
## Comparing local and global trend:

```
In [19]:

plt.plot(df_global.movingtemp,label="Global")
plt.plot(df_ranchi.movingtemp,label="Ranchi")
plt.plot(df_delhi.movingtemp,label="Delhi")
# Set the x axis label of the current axis.
plt.xlabel('Year')
# Set the y axis label of the current axis.
plt.ylabel('Moving Average Temp')
# Set a title of the current axes.
plt.title('Comparison: Local v/s Global')
plt.legend()
```

#### Out[19]:

<matplotlib.legend.Legend at 0x11d349e8>



### Plotting temperature difference:

```
In [20]:

df_ranchi['diff_G']=df_ranchi['avg_temp']-df_global['avg_temp']
df_delhi['diff_G']=df_delhi['avg_temp']-df_global['avg_temp']
df_delhi['diff_R']=df_delhi['avg_temp']-df_ranchi['avg_temp']
```

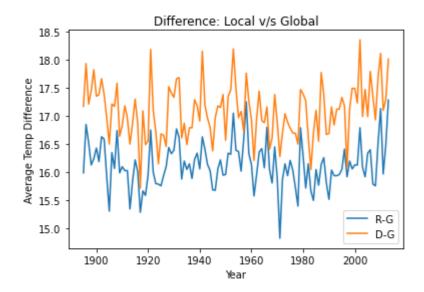
#### Plotting Local-Global temperature difference:

In [21]: ▶

```
plt.plot(df_ranchi.year,df_ranchi['diff_G'],label='R-G')
plt.plot(df_delhi.year,df_delhi['diff_G'],label='D-G')
# Set the x axis label of the current axis.
plt.xlabel('Year')
# Set the y axis label of the current axis.
plt.ylabel('Average Temp Difference')
# Set a title of the current axes.
plt.title('Difference: Local v/s Global')
plt.legend()
```

#### Out[21]:

<matplotlib.legend.Legend at 0x12d979d0>



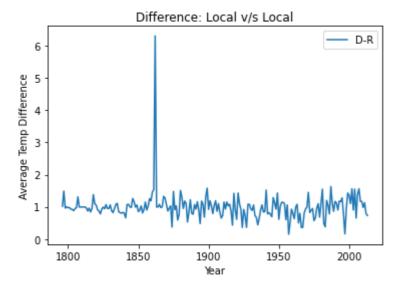
#### Plotting local-local temperature difference:

In [22]:

```
plt.plot(df_delhi.year,df_delhi['diff_R'],label='D-R')
# Set the x axis label of the current axis.
plt.xlabel('Year')
# Set the y axis label of the current axis.
plt.ylabel('Average Temp Difference')
# Set a title of the current axes.
plt.title('Difference: Local v/s Local')
plt.legend()
```

#### Out[22]:

<matplotlib.legend.Legend at 0x12dc8c58>



## **Calculating correlation coefficient:**

Global vs Ranchi:

```
In [23]:
df_global['avg_temp'].corr(df_ranchi['avg_temp'])
Out[23]:
0.3810337379721006
Global vs Delhi:
In [24]:
                                                                                             M
df_global['avg_temp'].corr(df_delhi['avg_temp'])
Out[24]:
0.36352144051290736
Ranchi vs Delhi:
In [25]:
                                                                                             H
df_ranchi['avg_temp'].corr(df_delhi['avg_temp'])
Out[25]:
0.7281892619502837
```

### **Observations:**

- The average temperature for Global and local have shown a linear increase in the observed years.
- The moving average temperature for Ranchi and Delhi is much greater than Global.
- Delhi and Ranchi show very similar trend for moving average temperature
- The average temperature difference for : Ranchi and Global is around 17 degrees. Delhi and Global is around 16 degrees. Ranchi and Delhi is around 1 degree.
- The correalation coefficient for global vs local (Ranchi and Delhi) is around 0.37 and for Delhi vs Ranchi is around 0.73