

# Exploring Weather Trends

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## Extracting the data

I have extracted data from the given database, using SQL queries.

2 datasets extracted

- Global Temperature ('global\_data.csv') USING:

```
SELECT *  
FROM global_data
```

- City Temperature ('delhi\_temp.csv') As I live in New Delhi, India.

```
SELECT *  
FROM city_data  
WHERE Country = 'India' AND City = 'Delhi'
```

In [1]:

```
# Impoting the python libraries and the dataset.  
import pandas as pd  
import matplotlib.pyplot as plt  
import numpy as np  
import seaborn as sb  
%matplotlib inline  
  
df_global = pd.read_csv('global_data.csv')  
df_delhi = pd.read_csv('delhi_temp.csv')
```

## Exploring the datasets

In [2]:

```
df_global.sample()
```

Out[2]:

|    | year | avg_temp |
|----|------|----------|
| 73 | 1823 | 7.72     |

In [3]:

```
df_delhi.sample()
```

Out[3]:

|    | year | city  | country | avg_temp |
|----|------|-------|---------|----------|
| 76 | 1872 | Delhi | India   | 24.94    |

In [4]:

```
df_global.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 266 entries, 0 to 265  
Data columns (total 2 columns):  
year          266 non-null int64  
avg_temp      266 non-null float64  
dtypes: float64(1), int64(1)  
memory usage: 4.2 KB
```

In [5]:

```
df_delhi.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 218 entries, 0 to 217  
Data columns (total 4 columns):  
year          218 non-null int64  
city          218 non-null object  
country       218 non-null object  
avg_temp      201 non-null float64  
dtypes: float64(1), int64(1), object(2)  
memory usage: 6.9+ KB
```

In [6]:

```
pd.isna(df_delhi.avg_temp).sum()
```

Out[6]:

17

In [7]:

```
df_delhi[df_delhi['avg_temp'].isnull()]
```

Out[7]:

|    | year | city  | country | avg_temp |
|----|------|-------|---------|----------|
| 12 | 1808 | Delhi | India   | NaN      |
| 13 | 1809 | Delhi | India   | NaN      |
| 14 | 1810 | Delhi | India   | NaN      |
| 15 | 1811 | Delhi | India   | NaN      |
| 16 | 1812 | Delhi | India   | NaN      |
| 62 | 1858 | Delhi | India   | NaN      |
| 63 | 1859 | Delhi | India   | NaN      |
| 64 | 1860 | Delhi | India   | NaN      |
| 65 | 1861 | Delhi | India   | NaN      |
| 66 | 1862 | Delhi | India   | NaN      |
| 67 | 1863 | Delhi | India   | NaN      |
| 68 | 1864 | Delhi | India   | NaN      |
| 69 | 1865 | Delhi | India   | NaN      |
| 70 | 1866 | Delhi | India   | NaN      |
| 71 | 1867 | Delhi | India   | NaN      |
| 72 | 1868 | Delhi | India   | NaN      |
| 73 | 1869 | Delhi | India   | NaN      |

## Assessment :

- 17 null values in delhi dataset.
- Between years **1808** and **1869** average temperature values are missing at various places.

## Solution :

- We know that null values will create problems later.
- We will consider data in both the datasets after **1869** to prevent future problems.

In [8]:

```
df_delhi = df_delhi.query('year > 1869')
df_delhi.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 144 entries, 74 to 217
Data columns (total 4 columns):
year          144 non-null int64
city          144 non-null object
country       144 non-null object
avg_temp      144 non-null float64
dtypes: float64(1), int64(1), object(2)
memory usage: 5.6+ KB
```

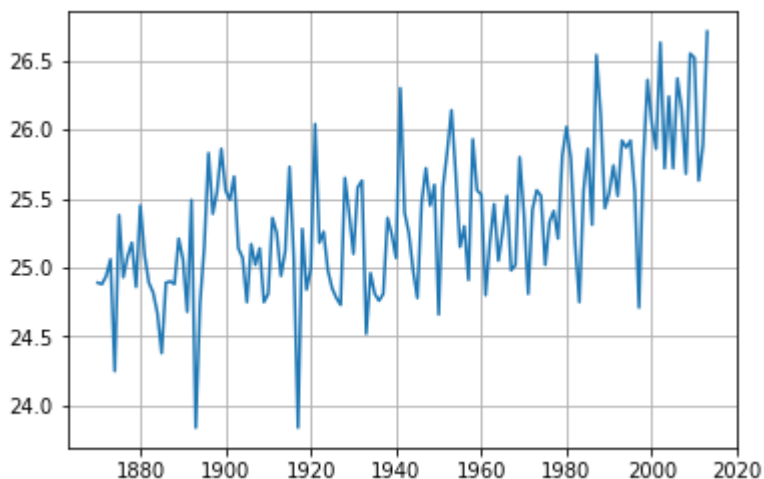
In [9]:

```
df_global = df_global.query('year > 1869')
df_global.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 146 entries, 120 to 265
Data columns (total 2 columns):
year          146 non-null int64
avg_temp      146 non-null float64
dtypes: float64(1), int64(1)
memory usage: 3.4 KB
```

In [10]:

```
plt.plot(df_delhi['year'], df_delhi['avg_temp'], label = "Delhi")
plt.grid(True)
plt.show()
```



## Assessment :

- As we plot average annual temperature over time, it creates graphs that is not easy to read.

## Solution :

- To create moving average of average temperature to smooth out the graph lines.

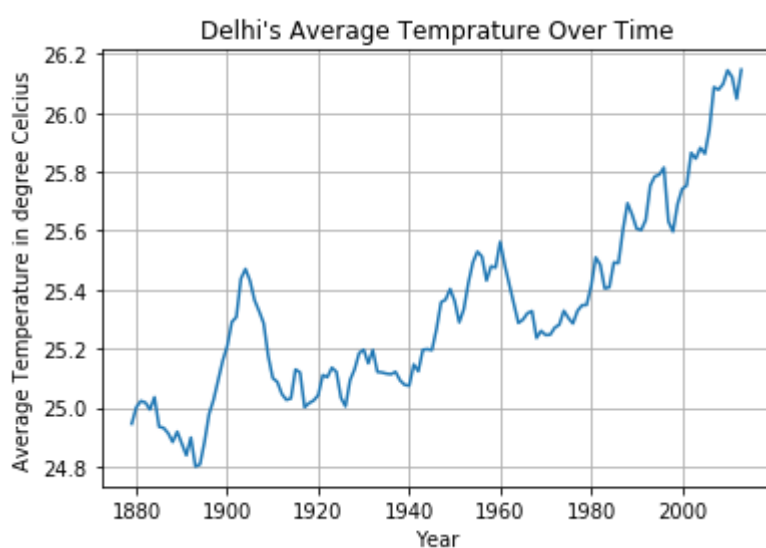
In [11]:

```
global_mavg = df_global['avg_temp'].rolling(10).mean()  
delhi_mavg = df_delhi['avg_temp'].rolling(10).mean()
```

## Visualization

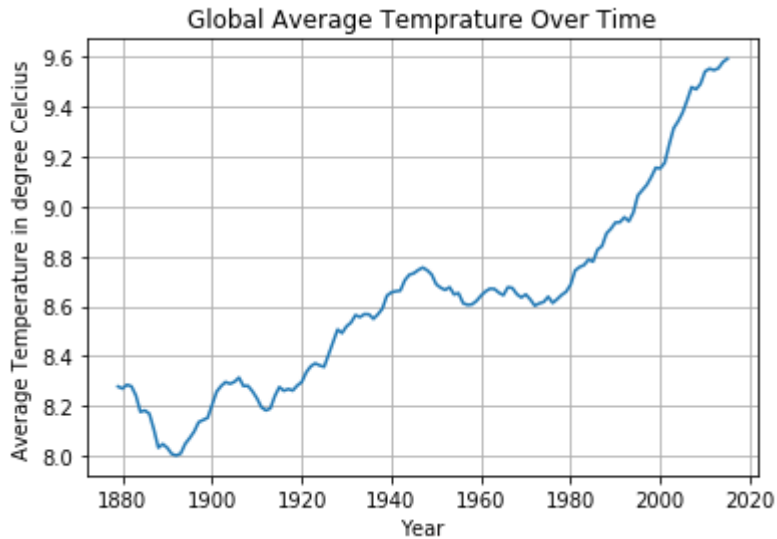
In [12]:

```
plt.plot(df_delhi['year'], delhi_mavg, label = "Delhi")  
plt.title("Delhi's Average Temperature Over Time")  
plt.xlabel('Year')  
plt.ylabel('Average Temperature in degree Celcius')  
plt.grid(True)  
plt.show()
```



In [13]:

```
plt.plot(df_global['year'], global_mavg, label = "Global Temperatures")
plt.title('Global Average Temperature Over Time')
plt.xlabel('Year')
plt.ylabel('Average Temperature in degree Celcius')
plt.grid(True)
plt.show()
```



## OBSERVATION

- Both Delhi's and Global temperature has increased overtime.
- In the last few decades the rate of increase of temperature is considerable larger.

In [14]:

```
df_delhi.avg_temp.mean() - df_global.avg_temp.mean()
```

Out[14]:

16.700742960426176

In [15]:

```
df_delhi.avg_temp.mean() , df_global.avg_temp.mean()
```

Out[15]:

```
(25.3392361111111106, 8.63849315068493)
```

In [16]:

```
float(df_delhi.query('year == "2013"').avg_temp.values) - float(df_delhi.query('year == "1870"').avg_temp.values)
```

Out[16]:

```
1.8200000000000003
```

In [17]:

```
# temperature difference in Delhi in the first and last decade.
del_last_decade = df_delhi.query('year >=2004 and year <=2013')['avg_temp'].mean()
del_first_decade = df_delhi.query('year >= 1870 and year < 1880')['avg_temp'].mean()
print("Delhi's average temperature between 1870 to 1880: ",del_first_decade,"\nDelhi's
      average temperature between 2004 to 2013: ",del_last_decade)
print('Difference :',del_last_decade - del_first_decade)
```

```
Delhi's average temperature between 1870 to 1880:  24.945000000000004
Delhi's average temperature between 2004 to 2013:  26.145999999999997
Difference : 1.2009999999999934
```

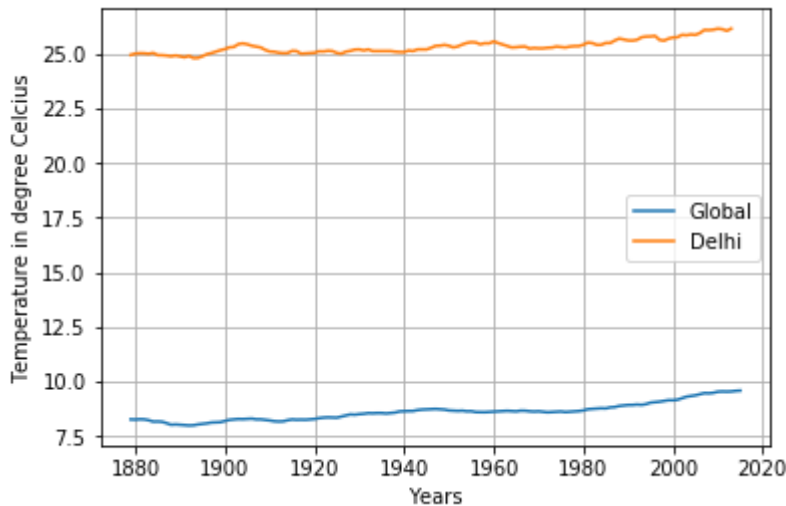
In [18]:

```
# temperature difference in the World in the first and the last decade.
glb_last_decade = df_global.query('year >=2006 and year <=2015')['avg_temp'].mean()
glb_first_decade = df_global.query('year >= 1870 and year <1880')['avg_temp'].mean()
print("World's average temperature between 1870 to 1880: ",glb_first_decade,"\nWorld's
      average temperature between 2006 to 2015: " ,glb_last_decade)
print("Difference :",glb_last_decade - glb_first_decade)
```

```
World's average temperature between 1870 to 1880:  8.277
World's average temperature between 2006 to 2015:  9.593999999999998
Difference : 1.3169999999999984
```

In [19]:

```
# Comparing the two
plt.plot(df_global['year'], global_mavg, label='Global')
plt.plot(df_delhi['year'], delhi_mavg, label = 'Delhi')
plt.grid(True)
plt.legend()
plt.xlabel('Years')
plt.ylabel('Temperature in degree Celcius')
plt.show()
```



## FINAL OBSERVATIONS :

- Both Delhi's and Global temperature has increased overtime .
- In the last few decades the rate of increase of temperature is accelerated.
- Since 1975 Global average temperature is raising without any stops.
- The difference between year **1870** and **2015** in temperature is more than 1.32 °C in the Global average chart.
- The difference between year **1870** and **2015** in temperature is more than 1.2 °C in the National Capital of India, Delhi.
- From the data it looks like world has gotten much hotter than Delhi from **1870** to **2015**.
- Delhi is on average 16.7 °C hotter than the world.

## CONCLUSION:

There are statistically significant evidences suggesting that the global temperature is raising over the years which support the case of climate change.



## **TOOLS USED:**

- SQL for extraction of data from given database.
- Python libraries to convert the raw data into meaningful, understandable pieces of knowledge by the help of numeric data points, graphs and conclusions. (Pandas, Numpy, Matplotlib).