

## Exploring Weather Trends

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### SQL queries for retrieving data.

1. For global data :  
`select * from global_data;`
2. For city data :  
`select * from city_data  
where city ='Singapore' & country ='Singapore';`
3. Both global data and city data have same column names 'avg\_temp'. Hence we rename the column names in both the data sets.

```
alter table city_data  
rename column avg_temp to city_avg_temp;
```

```
alter table global_data  
rename column avg_temp to global_avg_temp;
```

4. Joined both the tables and saved it in a new table called 'new\_weather'.  
`select city_data.year,city_data.city_avg_temp,global_data.global_avg_temp into  
new_weather  
from global_data  
join city_data on  
global_data.year=city_data.year  
where country='Singapore'`
5. Checked for missing values  
`select * from new_weather  
where country='Singapore' and avg_temp IS NULL;`
6. After the data is fetched from new table, I observed that there were many missing values for avg\_temp column, Singapore. Hence, I took the **mean (avg\_temp)** of Singapore and substituted it in place of the missing values.

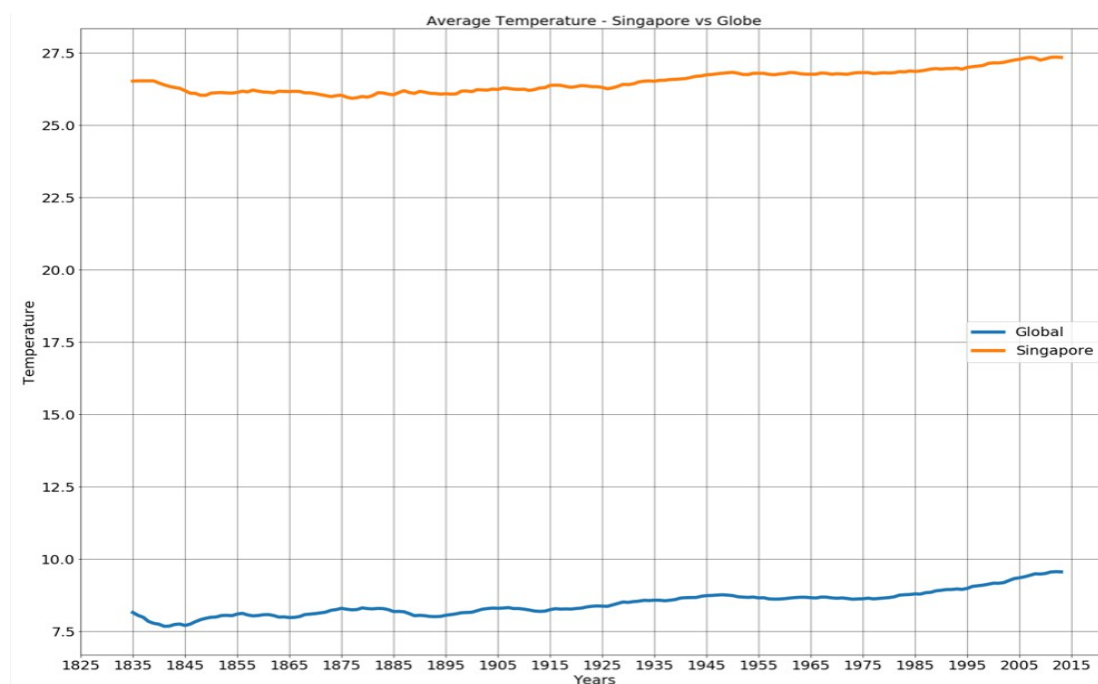
```
select year,coalesce(city_avg_temp,26.52)as city_avg_temp,global_avg_temp from  
new_weather;
```

### Excel for calculating Moving Average columns:

- ➔ Created two columns **MA\_10\_city** and **MA\_10\_global** for smoothing out trends in line chart
- ➔ Calculated the moving average for **10 years**. Command used = **Average(B2:B11)**
- ➔ Calculated correlation coefficients for city\_avg\_temp and global\_avg\_temp. Formula used = **correl(B2:B190;C2:C190)**  
correlation coefficient observed = 0.804897

## Visualization using Python matplotlib.lib:

```
import pandas as pd # for data related queries
import matplotlib.pyplot as plt # for visualizations
temp_df=pd.read_csv("Results.csv")
# for generating a line plot between MA_10_global and year
plt.plot(temp_df['year'],temp_df['MA_10_global'],label='Global',linewidth=5.0)
# for generating a line plot between MA_10_city and year
plt.plot(temp_df['year'],temp_df['MA_10_city'],label='Singapore',linewidth=5.0)
plt.legend(loc='center right',fontsize=20)
plt.xlabel("Years", fontsize=20)
plt.ylabel("Temperature", fontsize=20)
plt.title('Average Temperature - Singapore vs Globe', fontsize=20)
plt.rc('xtick', labels=20)
plt.rc('ytick', labels=20)
# for setting my own interval of year intervals on x axis (Interval of 10 years)
plt.xticks(np.arange(min(temp_df['year']), max(temp_df['year'])+10,10))
# for displaying x and y grids
plt.grid(which='major', linestyle='-', linewidth='0.5', color='black')
# Setting the desired size of figure
fig_size = plt.rcParams["figure.figsize"]
fig_size[0] = 24
fig_size[1] = 18
plt.rcParams["figure.figsize"] = fig_size
```



## Observations

- ➔ Global Moving Average temperature (10 years) between the years 1825 and 2015 varies between 7.6 and 9.5 degrees with the average temperature being 8.5 degrees
- ➔ Singapore on the other hand recorded higher moving average temperatures(10 years) ranging between 26 and 27.3 degrees with average temperature being 26.5 degrees
- ➔ A large difference in temperatures observed between Singapore and the globe.

- ➔ Singapore is hotter in comparison with the average global temperature
- ➔ A positive correlation observed between the global average temperature and Singapore's. Hence, indicating that as global average temperature increases, Singapore's average temperature is also increasing.

### **Conclusion**

Both Singapore and global average temperatures are increasing every year making them hotter

source: google, github