

**By: Tien Duong**

## Exploring Weather Trends

### Overview

As we all know that global temperature is getting warmer as ever. In this project, the data set was given to me as part of the [Udacity](https://classroom.udacity.com/nanodegrees/nd002/parts/93426fc7-0e68-4957-b16b-9fde38776c26/modules/e8455c07-092a-4b76-ba12-018cb53d0526/lessons/d551938c-d004-4801-a269-4b8dd784cc3b/concepts/530f21c0-2f37-4390-aaab-3ce440e56d80) (<https://classroom.udacity.com/nanodegrees/nd002/parts/93426fc7-0e68-4957-b16b-9fde38776c26/modules/e8455c07-092a-4b76-ba12-018cb53d0526/lessons/d551938c-d004-4801-a269-4b8dd784cc3b/concepts/530f21c0-2f37-4390-aaab-3ce440e56d80>) program. The average temperature of global weather trends file is to be extracted from the SQL database. extract the data and export it into two files `global weather.csv` and `local weather`. The `global weather.csv` consists of 2 columns, `year` and `avg_temp` which recorded the average temperature each year in Celsius. The `local weather.csv` contains 4 columns, `year`, `city`, `country`, `avg_temp`, which also recorded the average temperature in Celsius of San Jose which is also known as Silicon Valley.

### Objective Overview

- Extracting data from the dataset by using SQL
- Using pandas for data manipulation
- Using matplotlib for data visualization
- Making observation by analyzing the visualization

## SQL query used to extract the data from

### Query for extracting data from local weather trend dataframe.

```
SELECT * FROM city_data WHERE city='San Jose'
```

### Query for extracting data from global weather trend dataframe.

```
SELECT * FROM global_data
```

```
In [1]: 1 # Importing pandas library for data manipulation
        2 # and matplotlib as for data visualization
        3 import pandas as pd
        4 import matplotlib.pyplot as plt
        5 %matplotlib inline
```

## Global weather

```
In [2]: 1 # Import global temperature yearly average dataset
2 global_ = pd.read_csv('project weather.csv')
3 global_.head()
```

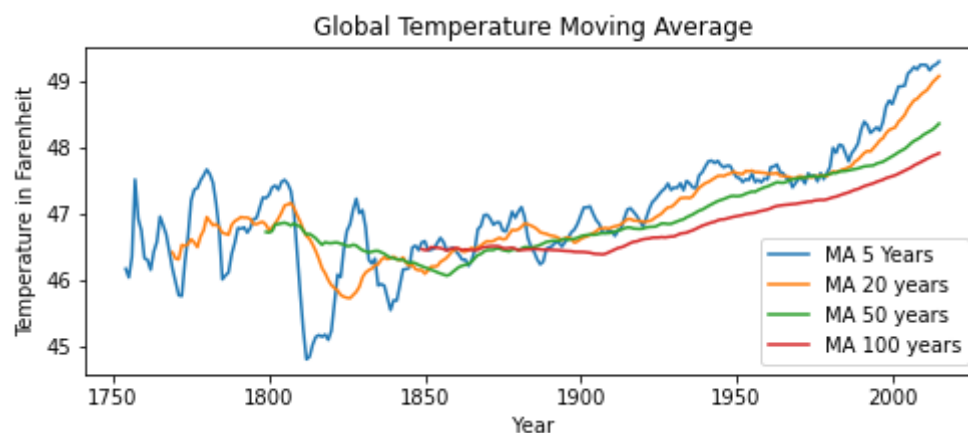
Out[2]:

|   | year | avg_temp |
|---|------|----------|
| 0 | 1750 | 8.72     |
| 1 | 1751 | 7.98     |
| 2 | 1752 | 5.78     |
| 3 | 1753 | 8.39     |
| 4 | 1754 | 8.47     |

```
In [3]: 1 # Function to convert average temperature from Celcius to farenhei
2 def Fconverter(x):
3     """Converting function formulor from Celcius to Farenheit"""
4     return (x * 1.8) + 32
```

```
In [4]: 1 # Set year as index and convert average temperature from celcius t
2 new_global = global_.set_index('year')
3 new_global['avg_temp'] = new_global['avg_temp'].apply(Fconverter)
```

```
In [5]: 1 # Global Temperature moving average plot
2 plt.figure(figsize= [8,3])
3 plt.plot(new_global['avg_temp'].rolling(5).mean(), label='MA 5 Yea
4 plt.plot(new_global['avg_temp'].rolling(20).mean(), label= 'MA 20
5 plt.plot(new_global['avg_temp'].rolling(50).mean(), label= 'MA 50
6 plt.plot(new_global['avg_temp'].rolling(100).mean(), label= 'MA 10
7
8 plt.legend(loc='best')
9 plt.title('Global Temperature Moving Average')
10 plt.xlabel(' Year ')
11 plt.ylabel('Temperature in Farenheit');
```



## Graph interpretaion

According to the graph, the average shift in temperature isn't visually appear to be significant from the graph, we can see that the global temperature is steadily inclining. This can really show the insight of global warming, the earth is getting warmer. And to truly have real investigate further in the climate changing, we must shift our focus onto the more relevant perspective. The average temperature in global weather still consider to be cooler from a personal perspective for a lot of people. For the next step, we'll analyze the weather average in my home town, Silicon Valley.

## Local weather

```
In [6]: 1 # Import the local temperature average dataset
        2 local_ = pd.read_csv('san_jose.csv')
        3 local_.head()
```

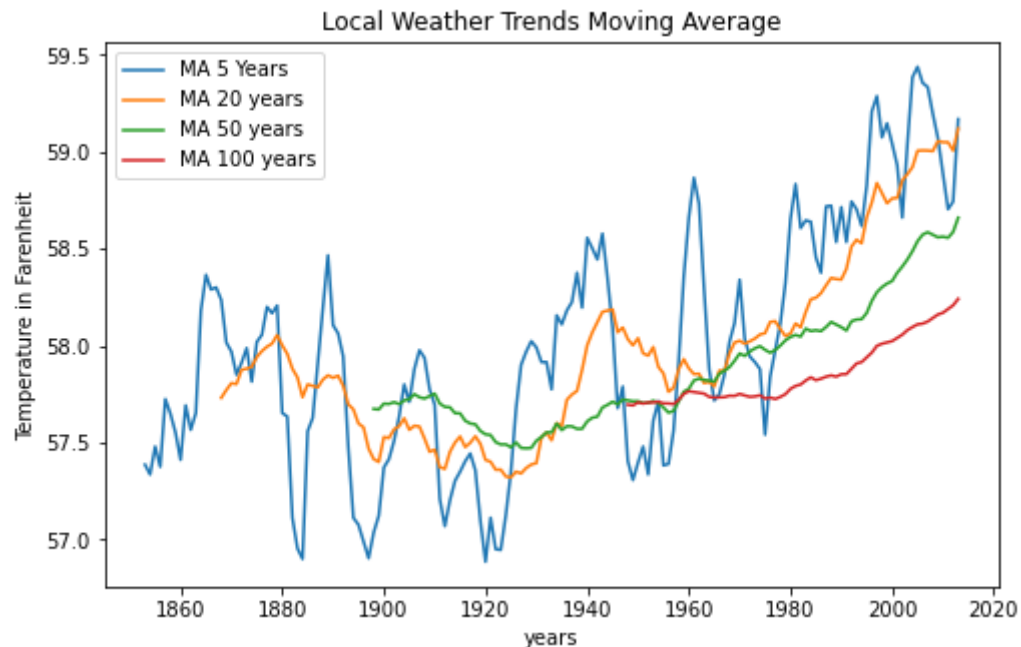
Out[6]:

|   | year | city     | country       | avg_temp |
|---|------|----------|---------------|----------|
| 0 | 1849 | San Jose | United States | 14.12    |
| 1 | 1850 | San Jose | United States | 13.80    |
| 2 | 1851 | San Jose | United States | 14.39    |
| 3 | 1852 | San Jose | United States | 13.81    |
| 4 | 1853 | San Jose | United States | 14.40    |

```

In [7]: 1 # Setting year as index and apply converter
2 local_ = local_[['year', 'avg_temp']].set_index('year')
3 local_['avg_temp'] = local_['avg_temp'].apply(Fconverter)
4
5 # plotting the global weather trends moving average
6 plt.figure(figsize=[8,5])
7 plt.plot(local_['avg_temp'].rolling(5).mean(), label='MA 5 Years')
8 plt.plot(local_['avg_temp'].rolling(20).mean(), label='MA 20 year')
9 plt.plot(local_['avg_temp'].rolling(50).mean(), label='MA 50 year')
10 plt.plot(local_['avg_temp'].rolling(100).mean(), label='MA 100 ye')
11
12 plt.title('Local Weather Trends Moving Average')
13 plt.xlabel('years')
14 plt.ylabel('Temperature in Farenheit')
15 plt.legend(loc='best');

```



## Correlation coefficient of average temperature between global and local weather trend

The correlation coefficient of average weather trends between global and Silicon Valley local weather is 0.54. Yet the correlation coefficient value is significant enough between two variables but to truly measure how much one variable influences another is to square the R values which is 0.29. In terms of prediction, we can determine that by observing the average local weather trends, we can make 29% prediction the relationship from average global weather trends. We are not too promising but predicting weather is another field of study that is way out of this project's scope of study.

```

In [8]: 1 # Use pandas built in method to find correlation
2 R = new_global['avg_temp'].corr(local_['avg_temp'])
3 R, R**2

```

Out[8]: (0.5360381430194767, 0.287336890771769)

## Visualization chart of global and Silicon Valley local weather trends average temperature

To truly satisfy our understanding of the average weather temperature between our two variables will be conducting a plot that visualizes the steady trends. The plot will contain global and local average weather trends of moving average by 10. In terms of compatibility, the local weather record of average temperature didn't record until the year 1849.

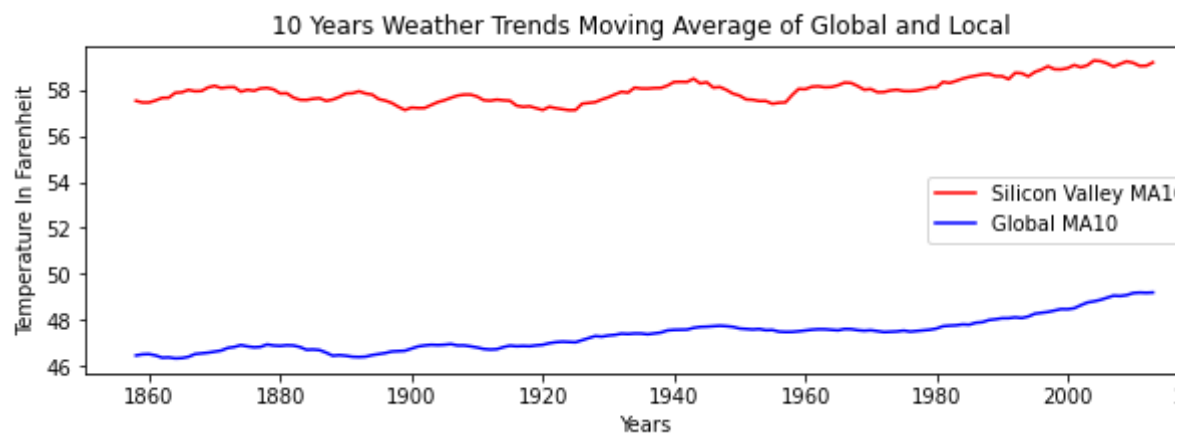
```
In [12]: 1 # First to check for the Dataset's shape
          2 local_.shape, global_.shape
```

```
Out[12]: ((165, 1), (266, 2))
```

```
In [13]: 1 # look like the local weather tracking system started in later and
          2 # We will modify this Dataset to fit into the other
          3 edit_global = new_global.query('year >=1849 & 2013 >= year')
          4 # Quick view comparison of the finalized global and local weather trends
          5 edit_global.head(), local_.head()
```

```
Out[13]: (      avg_temp
year
1849      46.364
1850      46.220
1851      46.724
1852      46.580
1853      46.472,
      avg_temp
year
1849      57.416
1850      56.840
1851      57.902
1852      56.858
1853      57.920)
```

```
In [14]: 1 plt.figure(figsize=[10,3])
2 plt.plot(local_['avg_temp'].rolling(10).mean(), label='Silicon Val
3 plt.plot(edit_global['avg_temp'].rolling(10).mean(), label='Global
4 plt.legend(loc='best')
5
6 plt.title('10 Years Weather Trends Moving Average of Global and Lc
7 plt.xlabel('Years')
8 plt.ylabel('Temperature In Farenheit');
```



## Conclusion

1. According to the graphs in this project shown that the steadily upward trends from both global and local weather temperatures.
2. There is correlation in the weather trends between Silicon valley and global temperature
3. The local average temperature is higher than global average temperature
4. GLOBAL WARMING IS HAPPENING