

Exploring Weather Trends

June 23, 2020

1 Exploring Weather Trends

1.1 Outline

1.1.1 Tools

- Step 1 Data extraction I use the following SQL code to export the dataset from the server SQL

```
SELECT    global_data.year,                global_data.avg_temp AS g_temp,  
city_data.avg_temp AS atl_temp            FROM    global_data    LEFT  
JOIN city_data    ON            city_data.year = global_data.year    AND  
city_data.city = 'Atlanta';
```
- Step 2 Data process I use Python for data analysis and visualization.

1.1.2 Moving Average

- There are a few NA data points. To calculate the moving average for all periods, I first replace those NA values with their last period values.
- Then I simply calculate the arithmetic mean of last n period:

$$temp_t^{ma,n} = \sum_{\tau=t-n+1}^t \frac{temp_{\tau}}{n}$$

1.1.3 Visualization

- Since I need to highlight the trend, I choose a line plot.

1.2 Line Chart

```
[1]: import numpy as np  
import pandas as pd  
import seaborn as sb  
import matplotlib.pyplot as plt  
!ls
```

Exploring Weather Trends.ipynb results.csv

```
[2]: df = pd.read_csv("./results.csv", sep=',')

[3]: # df.info()

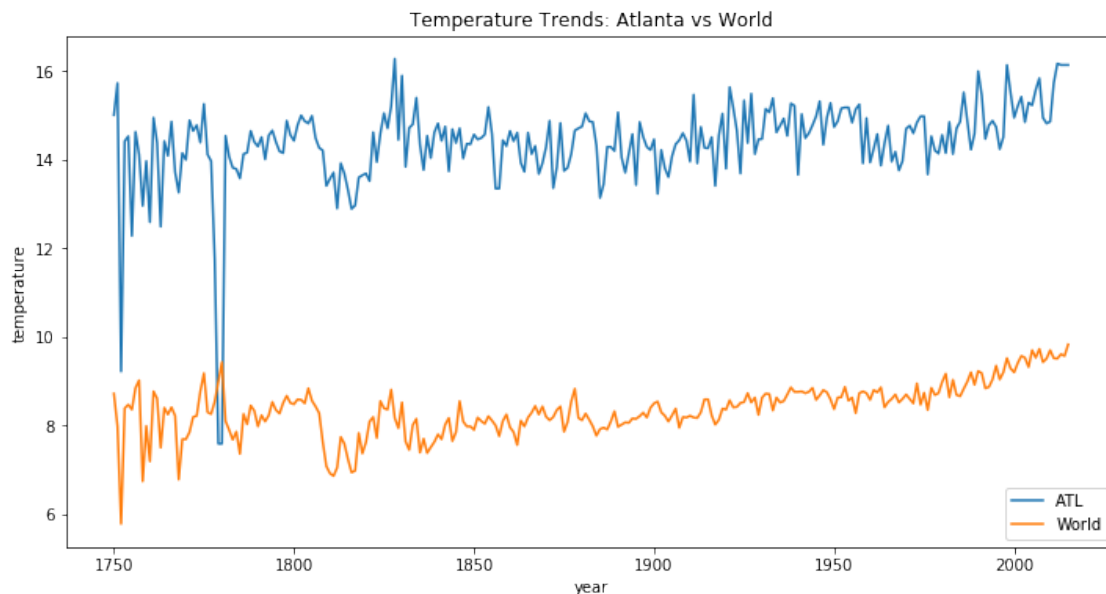
[4]: # df.head(10)

[5]: # df[df['atl_temp'].isna()]

[6]: df = df.fillna(method='ffill')

[7]: df2 = pd.DataFrame(index=df['year'], data=df[['atl_temp', 'g_temp']].values,
    ↪ columns=['ATL', 'World'])
    # df2.head()

[8]: ax = df2.plot(figsize=(12,6))
    ax.set(ylabel='temperature', title='Temperature Trends: Atlanta vs World');
```



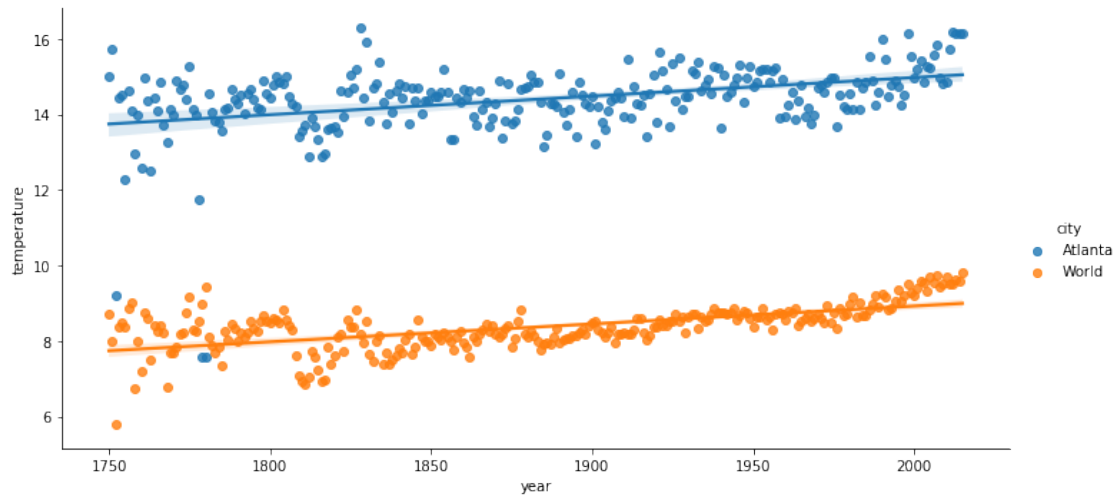
```
[9]: # reshape the data for plotting with Seaborn
df_atl = df.iloc[:, [0,1]]
df_atl = df_atl.rename(columns = {'atl_temp': 'temperature'})
df_atl['city'] = 'Atlanta'

df_glo = df.iloc[:, [0,2]]
df_glo = df_glo.rename(columns = {'g_temp': 'temperature'})
df_glo['city'] = 'World'

df3 = pd.concat([df_atl, df_glo], axis=0)
```

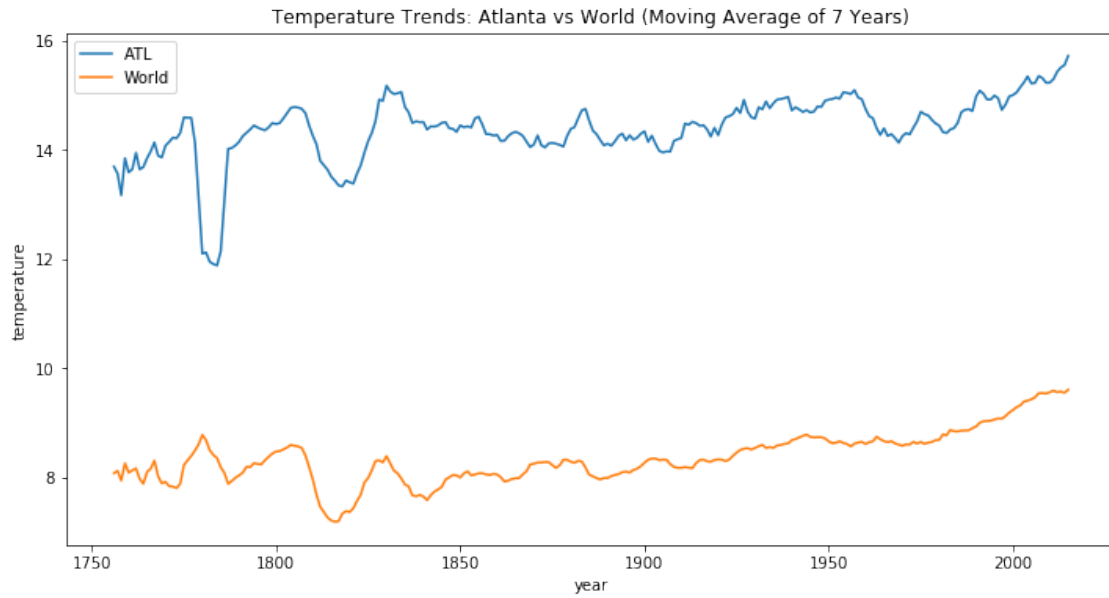
```
df3.reset_index(drop=True);
```

```
[10]: sb.lmplot(x='year', y="temperature", hue='city', data=df3, height=5, aspect=2);
```

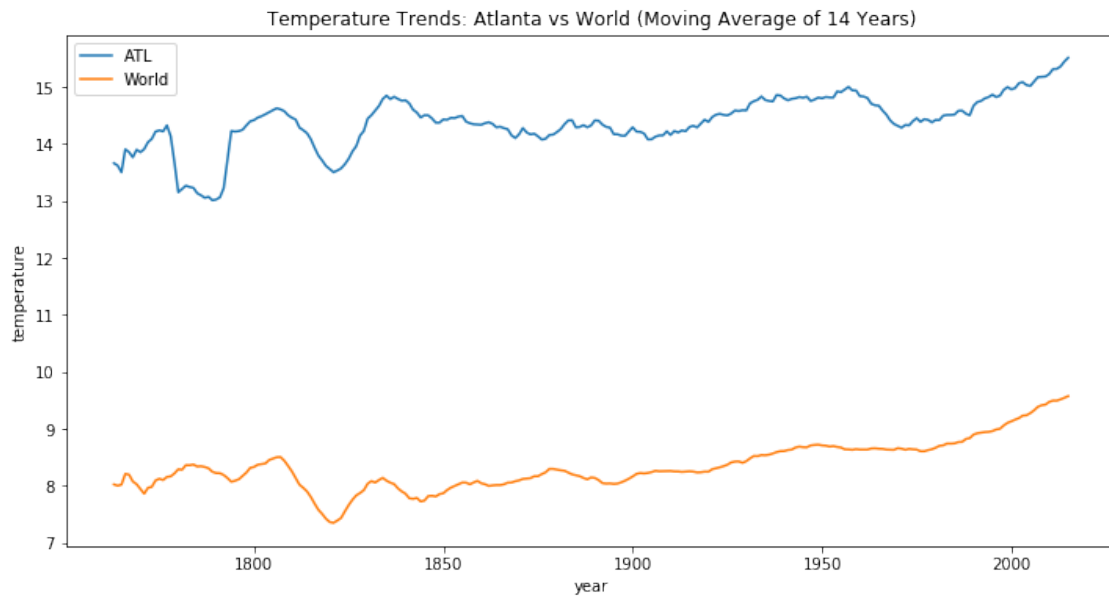


```
[16]: def plot_ma(df, period=7):  
    df_ma = df.copy()  
    df_ma['ATL'] = df['ATL'].rolling(window=period).mean()  
    df_ma['World'] = df['World'].rolling(window=period).mean()  
    ax = df_ma.plot(figsize=(12,6))  
    ax.set(ylabel='temperature',  
           title='Temperature Trends: Atlanta vs World (Moving Average of {}  
↪Years)'.format(period))  
    return None
```

```
[17]: plot_ma(df2, period=7)
```



```
[18]: plot_ma(df2, period=14)
```



1.3 Observations

1. From all plots, we can see that the trends for Atlanta and the world are very similar. The regression lines, which represent the linear trend, are basically parallel.
2. Both are steadily going up, which can probably serve as the evidence of global warming.

3. On average, Atlanta is about 6 degrees warmer than the global average. This gap is pretty stable.
4. For most of the time, the two trends move together.
5. There are some period when the two trend diverge. For example, during the time between 1750 and 1800, the temperature of Atlanta dropped a lot but not the world average. However, this seems to be driven by one or two extreme years.