1_Exploring_Weather_Trends

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1 Exploring Weather Trends

For the Udacity Data Analyst Nanodegree By Ken Norton, 2019

```
In [1]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt

        %config InlineBackend.figure_format = 'retina'
In [2]: plt.style.use('fivethirtyeight')
        plt.style.use('seaborn-poster')
```

1.1 Extract the data from the database

Schema:

- city_list This contains a list of cities and countries in the database. Look through them in order to find the city nearest to you.
- city_data This contains the average temperatures for each city by year (řC).
- global_data This contains the average global temperatures by year (řC).

First I checked to see how granular the data was:

```
SELECT
  city
FROM
  city_list
WHERE
  country = 'United States';
```

I determined that the database only includes major U.S. cities. San Francisco is my nearest city in the database.

```
SELECT
*
FROM
```

```
city_data
WHERE
  city = 'San Francisco';

I downloaded the results for SF into a CSV called sanfrancisco.csv

SELECT
  *
FROM
  global_data;

I downloaded the results for global temperatures into a CSV called global.csv
```

1.2 Import the data

```
In [3]: # Import data from the CSVs into dataframes
        sf_temp = pd.read_csv('data/sanfrancisco.csv')
        global_temp = pd.read_csv('data/global.csv')
In [4]: sf_temp.describe()
Out [4]:
                               avg_temp
                      year
        count
                165.000000 165.000000
               1931.000000
                              14.450788
        mean
        std
                 47.775517
                              0.504885
        min
               1849.000000
                              13.220000
        25%
               1890.000000
                              14.120000
        50%
               1931.000000
                              14.390000
        75%
               1972.000000
                              14.760000
        max
               2013.000000
                              16.230000
In [5]: global_temp.describe()
Out[5]:
                      year
                               avg_temp
                266.000000 266.000000
        count
               1882.500000
                               8.369474
        mean
        std
                 76.931788
                               0.584747
        min
               1750.000000
                               5.780000
        25%
               1816.250000
                               8.082500
        50%
                               8.375000
               1882.500000
        75%
               1948.750000
                               8.707500
               2015.000000
                               9.830000
        max
```

I note that the global data goes back to 1750 whereas the San Francisco data only goes back to 1849. I decide to merge them into a single dataframe to make comparisons more convenient.

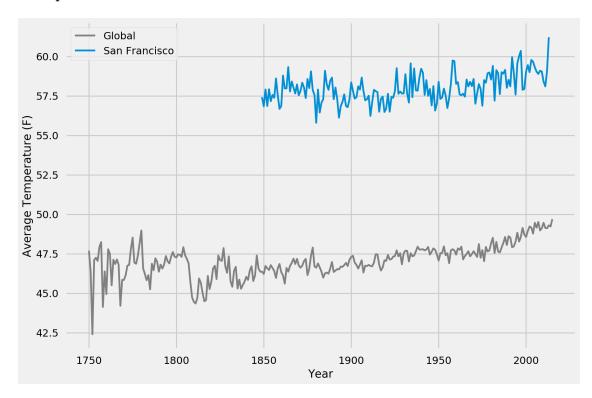
Since I'm based in the US, I find Fahrenheit measurements easier to understand than Celsius. I'm going to convert the data to Fahrenheit using a quick-and-dirty function.

I'm going to run the c2f function over both temperature columns and update them in place, effectively converting the entire dataframe to Fahrenheit.

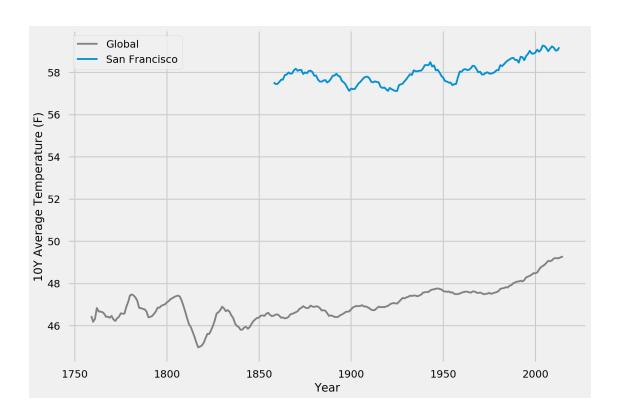
```
In [9]: temps['avg_temp_glob'] = c2f(temps['avg_temp_glob'])
        temps['avg_temp_sf'] = c2f(temps['avg_temp_sf'])
In [10]: temps.head()
Out[10]:
               avg_temp_glob avg_temp_sf
        year
        1750
                      47.696
                                      NaN
        1751
                      46.364
                                      NaN
                      42.404
        1752
                                      NaN
        1753
                      47.102
                                      NaN
        1754
                     47.246
                                      NaN
```

1.3 Create a line chart

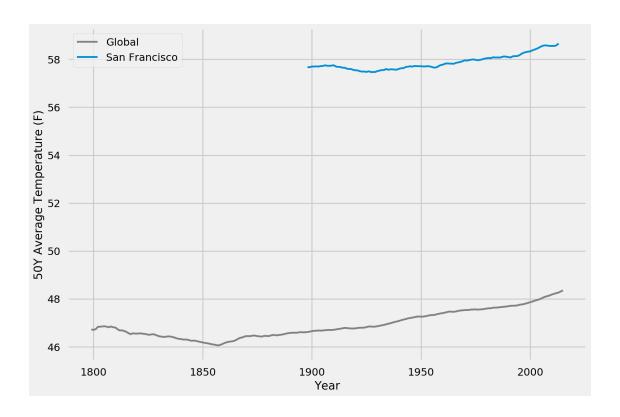
That compares your city's temperatures with the global temperatures. Make sure to plot the moving average rather than the yearly averages in order to smooth out the lines, making trends more observable (the last concept in the previous lesson goes over how to do this in a spreadsheet).



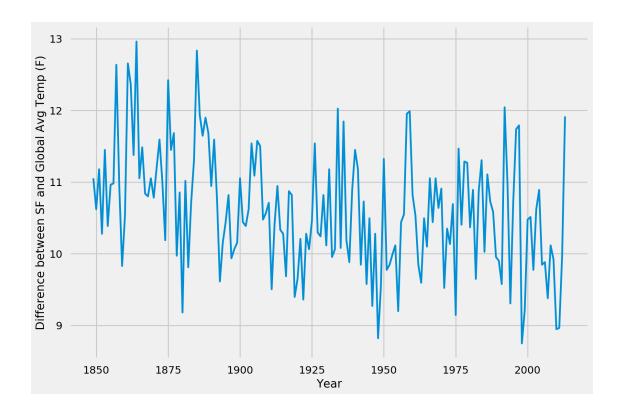
1.3.1 Weather trends: ten year moving average



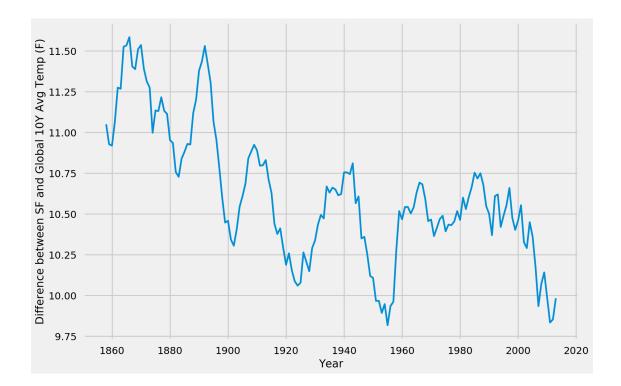
1.3.2 Weather trends: fifty year moving average



1.3.3 San Francisco versus the global average



We can see that SF is hotter than the global average, varying from between ~9-13 degrees Fahrenheit hotter. Let's use a 10-year average to smooth the lines a bit.



Looking at the ten-year average, we can see that the trend is narrowing. Although both SF and global temperature averages are increasing, the global average seems to be increasing faster and therefore the gap between the two is narrowing.

What's the correlation coefficient for SF vs. global temperature?

```
In [16]: temps['avg_temp_sf'].corr(temps['avg_temp_glob'])
Out[16]: 0.5360381430194766
```

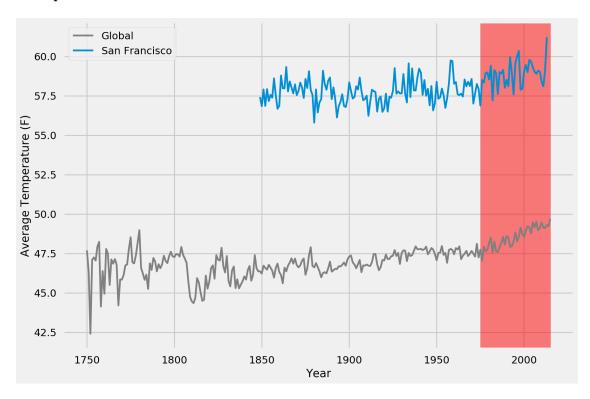
1.3.4 How does San Francisco compare to global since 1975?

```
In [17]: temps.tail()
Out [17]:
                avg_temp_glob
                                 avg_temp_sf
         year
          2011
                        49.136
                                      58.100
          2012
                        49.118
                                      59.090
                        49.298
          2013
                                      61.214
          2014
                        49.226
                                          NaN
         2015
                        49.694
                                          NaN
```

I checked to see how recent the data is. I see that global is only through 2015 and SF is only through 2013. That means the 10-year averages I was plotting above for SF actually cover the years 2003-13, making them somewhat misleading.

When I examine the annual trends I can see both the global uptick since 1975 and an increase for SF, especially in the most recent year in the data set (2015) which is the hottest measured.

```
In [18]: plt.plot(temps['avg_temp_glob'], color='gray', label='Global')
    plt.plot(temps['avg_temp_sf'], label='San Francisco')
    plt.xlabel('Year')
    plt.ylabel('Average Temperature (F)')
    plt.axvspan(1975, 2015, color='red', alpha=0.5)
    plt.legend()
    plt.show()
```



1.4 Observations

- 1. Is your city hotter or cooler on average compared to the global average? Has the difference been consistent over time? San Francisco is hotter on average than the global average. The difference has ranged from 9-13 degrees Fahrenheit, but this gap is narrowing in recent years to less than 10 degrees.
- 2. How do the changes in your city's temperatures over time compare to the changes in the global average? San Francisco is getting hotter, but at a slower rate than the global average.
- 3. What does the overall trend look like? Is the world getting hotter or cooler? Has the trend been consistent over the last few hundred years? The world is getting hotter according to my plots. Since 1850, the global 50-year moving average has increased more than 2 degrees Fahrenheit. According to an ongoing temperature analysis conducted by scientists at NASA's Goddard Institute for Space Studies (GISS), the average global temperature on Earth has increased by about 0.8 degrees Celsius (1.4 degree Fahrenheit) since 1880. Two-thirds of the warming has occurred since 1975, at a rate of roughly 0.15-0.20 degree Celsius per decade. (Source: NASA) The global trend is consistent with what I observed.

- 4. **What is the correlation coefficient?** *The correlation coefficient between SF and Global temperatures is 0.54. There is a medium positive correlation between the two values.*
- 5. Has most of San Francisco's temperature increase happened since 1975? I was intrigued that "two-thirds of the warming has occurred since 1975" and was curious if the same phenomenon was present in SF. I investigated that in the last section above. In my last 10-year average plot, it seemed that SF's average temperature has actually declined since 1975. After digging deeper, I discovered that (a) SF data only goes through 2013, and (b) the ten year moving average disguised the effects of the significant increases in the past 3 years of the data set. This demonstrates one of the weaknesses of using moving averages: although it can make it easier to spot longer-term trends, it can also disguise dramatic increases or decreases in the data.