

explore_weather_trends

August 15, 2018

1 Explore Weather Trends

```
In [10]: # import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import warnings
import seaborn as sns
sns.set()
warnings.filterwarnings('ignore')
%matplotlib inline
plt.rcParams['figure.figsize'] = [16, 10]
```

```
In [11]: # read dataset
tokyo_df = pd.read_csv('dataset/results_tokyo.csv')
global_df = pd.read_csv('dataset/results_global.csv')
```

```
In [12]: tokyo_df.head()
```

```
Out[12]:
```

| | year | city | country | avg_temp |
|---|------|-------|---------|----------|
| 0 | 1845 | Tokyo | Japan | 11.95 |
| 1 | 1846 | Tokyo | Japan | 12.40 |
| 2 | 1847 | Tokyo | Japan | 12.21 |
| 3 | 1848 | Tokyo | Japan | 12.14 |
| 4 | 1849 | Tokyo | Japan | 12.14 |

```
In [13]: tokyo_df.describe()
```

```
Out[13]:
```

| | year | avg_temp |
|-------|-------------|------------|
| count | 169.000000 | 169.000000 |
| mean | 1929.000000 | 12.565740 |
| std | 48.930222 | 0.649692 |
| min | 1845.000000 | 11.180000 |
| 25% | 1887.000000 | 12.140000 |
| 50% | 1929.000000 | 12.450000 |
| 75% | 1971.000000 | 12.920000 |
| max | 2013.000000 | 16.120000 |

```
In [14]: global_df.describe()
```

```
Out[14]:
```

| | year | avg_temp |
|-------|-------------|------------|
| count | 266.000000 | 266.000000 |
| mean | 1882.500000 | 8.369474 |
| std | 76.931788 | 0.584747 |
| min | 1750.000000 | 5.780000 |
| 25% | 1816.250000 | 8.082500 |
| 50% | 1882.500000 | 8.375000 |
| 75% | 1948.750000 | 8.707500 |
| max | 2015.000000 | 9.830000 |

```
In [15]: global_df.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 [16]: # calculate moving average
```

```
window_size = 10  
tokyo_df['mva_temp'] = tokyo_df['avg_temp'].rolling(window=window_size).mean()  
global_df['mva_temp'] = global_df['avg_temp'].rolling(window=window_size).mean()
```

```
In [17]: tokyo_df.head(30)
```

```
Out[17]:
```

| | year | city | country | avg_temp | mva_temp |
|----|------|-------|---------|----------|----------|
| 0 | 1845 | Tokyo | Japan | 11.95 | NaN |
| 1 | 1846 | Tokyo | Japan | 12.40 | NaN |
| 2 | 1847 | Tokyo | Japan | 12.21 | NaN |
| 3 | 1848 | Tokyo | Japan | 12.14 | NaN |
| 4 | 1849 | Tokyo | Japan | 12.14 | NaN |
| 5 | 1850 | Tokyo | Japan | 11.71 | NaN |
| 6 | 1851 | Tokyo | Japan | 11.76 | NaN |
| 7 | 1852 | Tokyo | Japan | 11.73 | NaN |
| 8 | 1853 | Tokyo | Japan | 11.86 | NaN |
| 9 | 1854 | Tokyo | Japan | 12.24 | 12.014 |
| 10 | 1855 | Tokyo | Japan | 12.56 | 12.075 |
| 11 | 1856 | Tokyo | Japan | 12.06 | 12.041 |
| 12 | 1857 | Tokyo | Japan | 11.94 | 12.014 |
| 13 | 1858 | Tokyo | Japan | 12.23 | 12.023 |
| 14 | 1859 | Tokyo | Japan | 12.15 | 12.024 |
| 15 | 1860 | Tokyo | Japan | 11.72 | 12.025 |
| 16 | 1861 | Tokyo | Japan | 11.79 | 12.028 |
| 17 | 1862 | Tokyo | Japan | 11.73 | 12.028 |

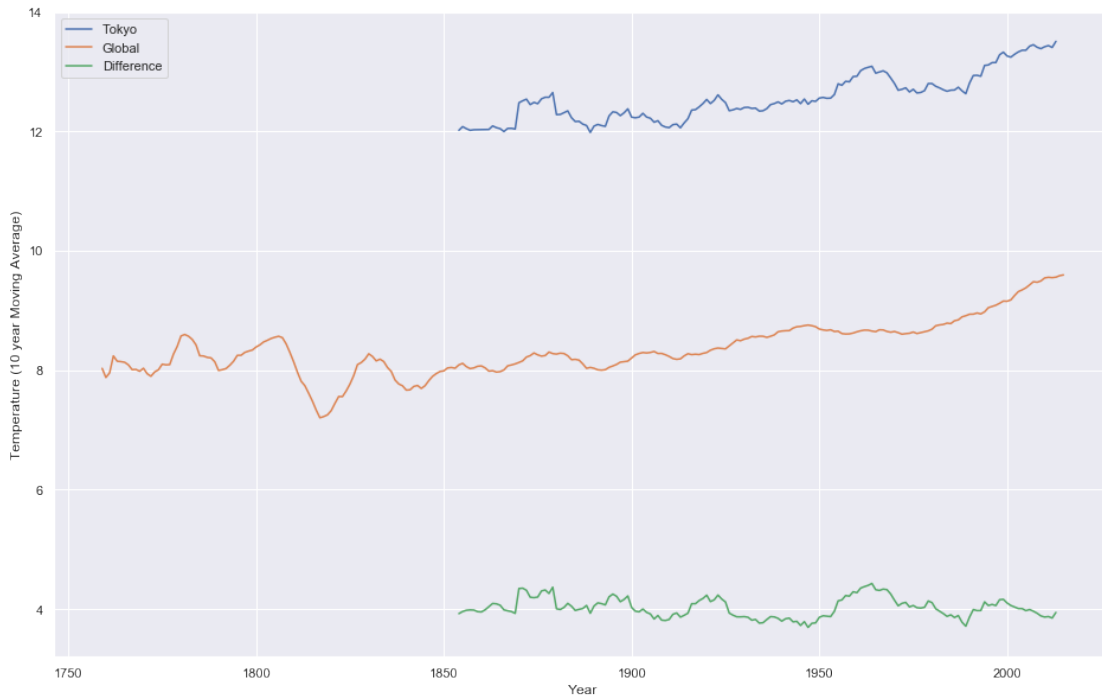
| | | | | | |
|----|------|-------|-------|-------|--------|
| 18 | 1863 | Tokyo | Japan | 12.45 | 12.087 |
| 19 | 1864 | Tokyo | Japan | 11.95 | 12.058 |
| 20 | 1865 | Tokyo | Japan | 12.39 | 12.041 |
| 21 | 1866 | Tokyo | Japan | 11.58 | 11.993 |
| 22 | 1867 | Tokyo | Japan | 12.46 | 12.045 |
| 23 | 1868 | Tokyo | Japan | 12.25 | 12.047 |
| 24 | 1869 | Tokyo | Japan | 12.03 | 12.035 |
| 25 | 1870 | Tokyo | Japan | 16.12 | 12.475 |
| 26 | 1871 | Tokyo | Japan | 12.12 | 12.508 |
| 27 | 1872 | Tokyo | Japan | 12.00 | 12.535 |
| 28 | 1873 | Tokyo | Japan | 11.53 | 12.443 |
| 29 | 1874 | Tokyo | Japan | 12.33 | 12.481 |

2 Plot

```
In [18]: # plot the results
tokyo_year = tokyo_df.year.values[window_size - 2:]
tokyo_temp = tokyo_df.mva_temp.values[window_size - 2:]
global_year = global_df.year.values[window_size - 2:]
global_temp = global_df.mva_temp.values[window_size - 2:]

# calculate the difference
tokyo_year_min = np.min(tokyo_year)
tokyo_year_max = np.max(tokyo_year)
idx = np.where((global_year >= tokyo_year_min) & (global_year <= tokyo_year_max))
diff = tokyo_temp - global_temp[idx]

plt.plot(tokyo_year, tokyo_temp, label='Tokyo')
plt.plot(global_year, global_temp, label='Global')
plt.plot(tokyo_year, diff, label='Difference')
plt.xlabel('Year')
plt.ylabel('Temperature (10 year Moving Average)')
plt.legend()
plt.show()
```



3 Observations

Is your city hotter or cooler on average compared to the global average? My city (Tokyo) is hotter than the global average.

Has the difference been consistent over time? The difference (green line) seems consistent over time.

What does the overall trend look like? The overall trends are similar.

Is the world getting hotter or cooler? Yes, it is. Tokyo is also getting hotter.

4 Outline

What tools did you use for each step? (Python, SQL, Excel, etc)

- SQL to extract data
- Pandas to calculate the moving average
- Matplotlib to plot the results

How did you calculate the moving average? I used Pandas' function.

What were your key considerations when deciding how to visualize the trends? I plot the results so that it's easy to compare the differences and trends.