${\bf Exploring_Weather_Trends}$

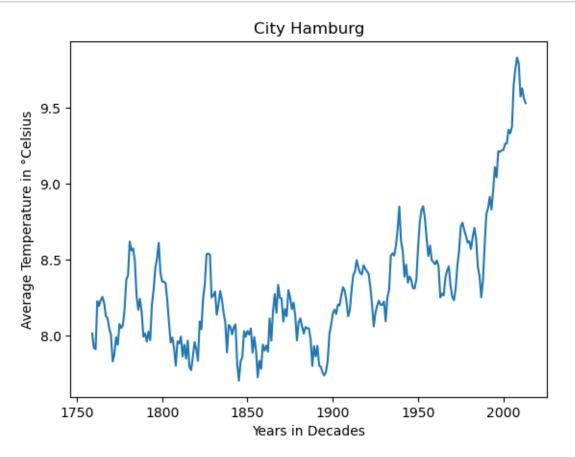
February 20, 2023

```
[1]: import pandas as pd
    import matplotlib.pyplot as plt
    %matplotlib inline
    0.1 Data Gathering
[2]: city = pd.read_csv('results_city.csv')
[3]: city.head()
                city country avg_temp
[3]:
       year
    0 1743 Hamburg Germany
                                   6.23
    1 1744 Hamburg Germany
                                   9.63
    2 1745 Hamburg Germany
                                   1.25
    3 1746 Hamburg Germany
                                    NaN
    4 1747 Hamburg Germany
                                    NaN
[4]: city_decade = city.avg_temp.rolling(10).mean()
    city_decade[10:20]
[4]: 10
            NaN
    11
            NaN
    12
            NaN
    13
            NaN
    14
            NaN
    15
            NaN
    16
          8.016
    17
          7.921
    18
          7.913
    19
          8.229
    Name: avg_temp, dtype: float64
```

0.2 Plotting City Data

```
[5]: plt.figure()
  plt.plot(city.year, city_decade)
  plt.xlabel("Years in Decades")
  plt.ylabel("Average Temperature in °Celsius")
  plt.title("City Hamburg")

plt.show()
```



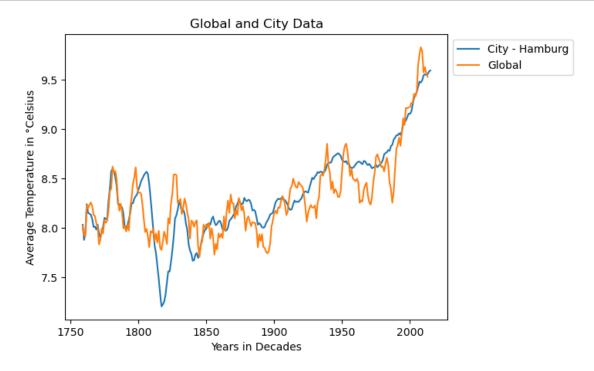
```
[8]: globe = pd.read_csv('results_global.csv')
[9]:
    globe.head()
[9]:
        year
              avg_temp
       1750
                  8.72
     1
        1751
                  7.98
                  5.78
     2
       1752
     3
        1753
                  8.39
     4
                  8.47
        1754
```

```
[10]: globe_decade = globe.avg_temp.rolling(10).mean()
      globe_decade[10:20]
[10]: 10
            7.877
            7.956
      11
      12
            8.239
      13
            8.150
            8.143
      14
      15
            8.132
            8.088
      16
      17
            8.008
            8.012
      18
      19
            7.982
      Name: avg_temp, dtype: float64
```

0.3 Plotting City and Global Data

```
plt.figure()
  plt.plot(globe.year, globe_decade, label='Global')
  plt.plot(city.year, city_decade, label='City')
  d ={'City - Hamburg': city_decade, 'Global': globe_decade}
  plt.legend(d, bbox_to_anchor=(1.0, 1.0))
  plt.xlabel("Years in Decades")
  plt.ylabel("Average Temperature in "Celsius")
  plt.title("Global and City Data")

plt.show()
```



0.4 Outline

The datasets where downloaded in csv-format from a database provided by Udacity with SQL queries, for analysis the data where gathered in a jupyter notebook with pandas, after calculating the rolling averages the data where plotted with matplotlib line chart functionality. Considering the timespan of the dataset a rolling average of 10 years was chosen, the final plot lies the data for the city and the global data together to allow direct comparison and discussion below in the conclusion.

0.5 Conclusion

Given the duration in regards to time for weather data, it was appropriate to set the rolling average to 10 years, hence plotted above are rolling averages and the closest city to my residence has shown a steady rise of temperature since the start of the industrialization, with at steep increase in the last 2-3 decades. This development is in line with the global increase in temperature, which however shows a steader increase after 1840-2013, compared to the global development which is shows a similar increase but with higher swings during this period of time. As mentioned above the overall trend is a steep increase in temperature for both, the City of Hamburg and the globe alike.