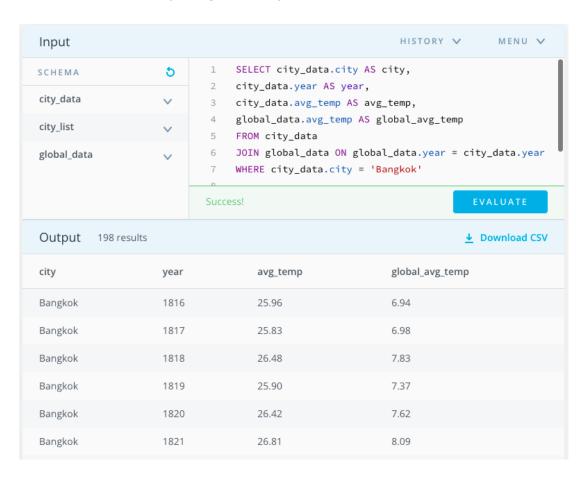
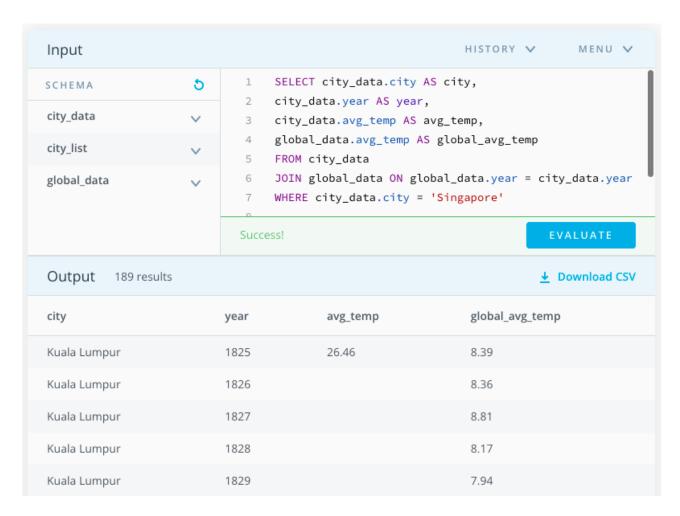
# **Exploring Weather Trends**

# 1. Extracting Data

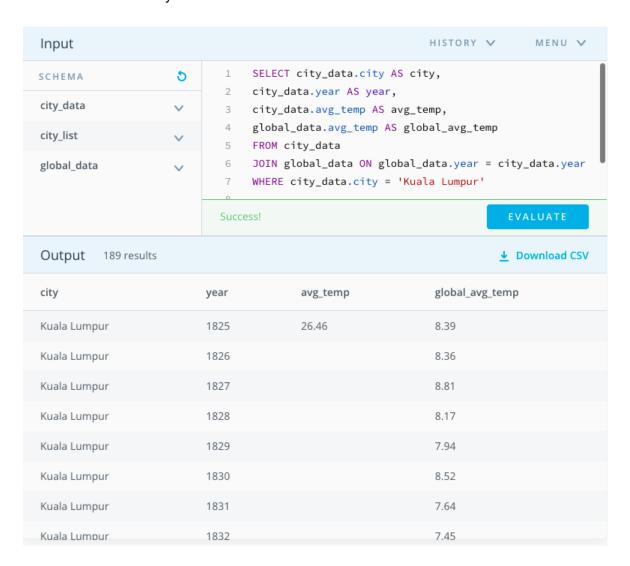
- Extracting the weather data by using SQL software installed in the Udacity website
- Select local cities such as Bangkok and the neighbouring cities of Singapore and Kuala Lumpur to analyze the trend and see whether they are the same as Bangkok.
  - a) Use SQL to extract Bangkok average temperature data as well as Global average temperature data and join these two tables together by using JOIN on year data of both tables.



 Use SQL to extract Singapore average temperature data as well as global average temperature data joining the two tables together by year data



 Use SQL to extract Kuala Lumpur average temperature data as well as global average temperature data joining the two tables together by year data



# 2. Open the saved CSV of these 3 cities data and visualize the trend in Python

- a) Import the Bangkok CSV file and explore its data
  - Check the head and tail of the data of Bangkok and see that there is an increase of average temperature of Bangkok from 25.96 in 1816 to 28.98 in 2013. Global average temperature has also increased over the 197 years period from 6.94 to 9.61.

### Import and Explore Data In [307]: import pandas as pd import seaborn as sns import matplotlib.pyplot as plt BKK = pd.read\_csv('/Users/atikarnpattamavichaiporn/Downloads/BKK.csv') In [308]: BKK.head() Out[308]: city year avg\_temp global\_avg\_temp 0 Bangkok 1816 25.96 6.94 1 Bangkok 1817 25.83 6.98 2 Bangkok 1818 26.48 7.83 7.37 3 Bangkok 1819 25.90 4 Bangkok 1820 7.62 26.42 In [309]: BKK.tail() Out[309]: city year avg\_temp global\_avg\_temp 193 Bangkok 2009 9.51 27.79 194 Bangkok 2010 28.54 9.70 195 Bangkok 2011 9.52 196 Bangkok 2012 28.48 9.51

```
In [310]: BKK.info()
          <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 198 entries, 0 to 197
         Data columns (total 4 columns):
          # Column
                               Non-Null Count Dtype
          0 city
                               198 non-null
                                               object
                               198 non-null
          1
              year
                                               int64
                                               float64
          2 avg_temp
                               190 non-null
          3 global_avg_temp 198 non-null
                                               float64
          dtypes: float64(2), int64(1), object(1)
         memory usage: 6.3+ KB
```

197 Bangkok 2013

28.98

9.61

- b) Visualize the trend of the average temperature in Bangkok compared to the world average temperature.
  - Slight upward trend can be identified from the graph of both global and Bangkok average temperature.
  - Bangkok temperatures were approximately 20 degrees hotter than the world average

# Visualize the trend of temperatures In [311]: sns.lineplot(data=BKK, y='avg\_temp', x='year') sns.lineplot(data=BKK, y='global\_avg\_temp', x='year') plt.legend(['Bangkok k, 'global]') plt.title('Bangkok & Global temperature trend ') Out[311]: Text(0.5, 1.0, 'Bangkok & Global temperature trend ') Bangkok & Global temperature trend 25 Bangkok & Global temperature trend 25 Bangkok & Global temperature trend 26 Bangkok & Global temperature trend 27 Bangkok & Global temperature trend 28 Bangkok & Global temperature trend

# 3. Calculate the Moving Average of Bangkok and Global Average Temperature over the period of 10 years, 15 years, and 20 years

a) Calculate the Moving Average

### Calculate Moving Average of 10, 15, and 20 years period

```
In [312]: # calculate Moving Average of BKK average temperature over 10 years
BKK['BKK_MA_10'] = BKK.avg_temp.rolling(10, min_periods=1).mean()
# calculate Moving Average of global average temperature over 10 years
BKK['global_MA_10'] = BKK.global_avg_temp.rolling(10, min_periods=1).mean()
# calculate Moving Average of BKK average temperature over 15 years
BKK['BKK_MA_15'] = BKK.avg_temp.rolling(15, min_periods=1).mean()
# calculate Moving Average of global average temperature over 15 years
BKK['global_MA_15'] = BKK.global_avg_temp.rolling(15, min_periods=1).mean()
# calculate Moving Average of BKK average temperature over 20 years
BKK['BKK_MA_20'] = BKK.avg_temp.rolling(20, min_periods=1).mean()
# calculate Moving Average of global average temperature over 20 years
BKK['global_MA_20'] = BKK.global_avg_temp.rolling(20, min_periods=1).mean()
```

- b) Visualize the moving average of 10, 15 and 20 years for Bangkok average temperature with Global average temperature
  - From the charts, the moving average over the 20 years period smoothen out the noise of the data the most, making the chart the most smooth when comparing the 10 years moving average and 15 years moving average graph. The chart of moving average over 20 years help see the trend clearer.

### Visualize the BKK Moving Average with Global trend

```
In [313]: sns.lineplot(data=BKK, y='BKK_MA_10', x='year')
    sns.lineplot(data=BKK, y='global_MA_10', x='year')
    plt.legend(['Bangkok ', 'global'])
    plt.title('Bangkok & Global Moving Average 10 years')
```

Out[313]: Text(0.5, 1.0, 'Bangkok & Global Moving Average 10 years')

```
Bangkok & Global Moving Average 10 years

25

Bangkok & Global Moving Average 10 years

25

Bangkok global

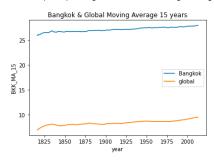
15

10

1825 1850 1875 1900 1925 1950 1975 2000
```

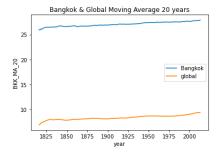
```
In [314]: sns.lineplot(data=BKK, y='BKK_MA_15', x='year')
    sns.lineplot(data=BKK, y='global_MA_15', x='year')
    plt.legend(['Bangkok', 'global'])
    plt.title('Bangkok & Global Moving Average 15 years')
```

Out[314]: Text(0.5, 1.0, 'Bangkok & Global Moving Average 15 years')



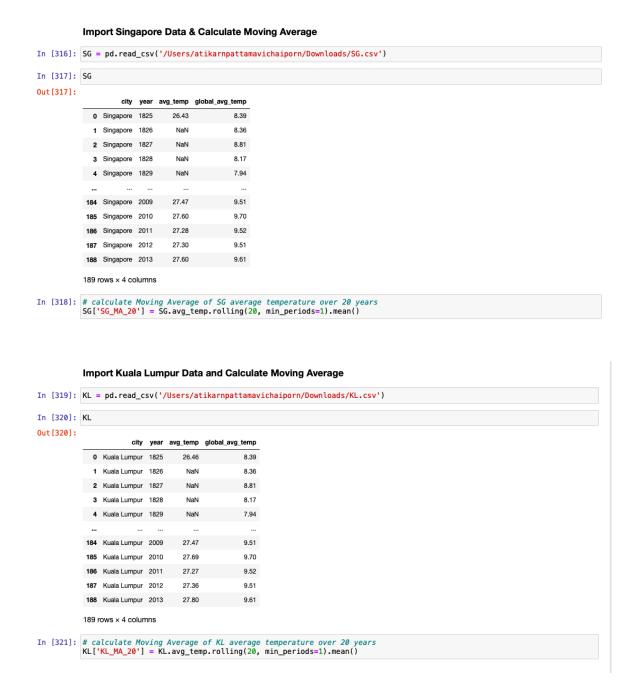
```
In [315]: sns.lineplot(data=BKK, y='BKK_MA_20', x='year')
sns.lineplot(data=BKK, y='global_MA_20', x='year')
plt.legend(['Bangkok', 'global'])
plt.title('Bangkok & Global Moving Average 20 years')
```

Out[315]: Text(0.5, 1.0, 'Bangkok & Global Moving Average 20 years')



# 4. Import Singapore and Kuala Lumpur CSV data and calculate the moving average of 20 years period

Since moving average of 20 years period smoothen out the trend the most



# 5. Visualize and compare the average temperature trend of each city vs. global average temperature

Singapore and Kuala Lumpur have similar trends like Bangkok, foreseeing some upward trend of temperature from 1816 to 2013. Both cities also have the same gap of temperature differences between the city temperature and global temperature, which is around 20 degrees.

### Compare BKK, SG, and KL trend with Global trend

15 10

1825

1850

1875

1900

1925



Global

2000

1950

1975

# 6. Find and make the heatmap chart of correlation of Bangkok data

- From the chart, the closer the correlation coefficient is to 1, the higher the correlation between those 2 columns.
- From the correlation coefficient in the chart, all columns are highly correlated to other columns, all coefficients are higher than or equal to 0.8

### Correlation of BKK dataset In [323]: BKK\_corr = BKK.corr() In [324]: sns.heatmap(BKK\_corr, annot=True) plt.title('Correlation heatmap') Out[324]: Text(0.5, 1.0, 'Correlation heatmap') Correlation heatmap - 1.000 year - 1 0.8 0.84 0.96 0.92 0.98 0.94 0.98 0.975 avg\_temp - 0.8 1 0.84 0.82 0.79 0.82 0.8 0.82 0.8 - 0.950 global\_avg\_temp - 0.84 0.84 1 0.85 0.9 0.85 0.88 0.85 0.88 - 0.925 BKK\_MA\_10 - 0.96 0.82 0.85 1 0.96 0.98 0.96 0.98 0.97 global\_MA\_10 - 0.92 0.79 0.9 0.96 1 0.94 0.99 0.93 0.98 - 0.900 BKK\_MA\_15 - 0.98 0.82 0.85 0.98 0.94 1 0.96 0.99 0.97 - 0.875 global MA 15 - 0.94 0.8 0.88 0.96 0.99 0.96 1 0.96 0.99 - 0.850 BKK\_MA\_20 - 0.98 0.82 0.85 0.98 0.93 0.99 0.96 1 0.97 - 0.825 0.8 0.88 0.97 0.98 0.97 0.99 0.97 1 global\_MA\_20 - 0.95 0.800 BKK\_MA\_10 BKK\_MA\_15 -global\_MA\_15 -BKK\_MA\_20 -global\_MA\_20 global MA 10

# 7. Predict Bangkok temperature from Global temperature using Linear Regression model

```
Predict BKK temp based on Global temp
In [325]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    import numpy as np
In [347]: x = np.array(BKK['global_avg_temp'])
y = np.array(BKK['avg_temp'])
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)
In [343]: BKK=BKK.dropna(how='any')
In [344]: BKK.info()
                  <class 'pandas.core.frame.DataFrame'>
Int64Index: 190 entries, 0 to 197
Data columns (total 10 columns):
# Column Non-Null Count
                                                         Non-Null Count Dtype
                    0 city
                                                         190 non-null
                                                                                      object
                                                         190 non-null
190 non-null
                                                                                       int64
float64
                  1 year 190 non-null ir 2 avg_temp 190 non-null fl 3 global_avg_temp 190 non-null fl 4 BKK_MA_10 190 non-null fl 5 global_MA_15 190 non-null fl 6 BKK_MA_15 190 non-null fl 8 BKK_MA_20 190 non-null fl 8 BKK_MA_20 190 non-null fl dtypes: float64(8), int64(1), object(1) memory usage: 16.3+ KB
                                                                                       float64
                                                                                      float64
                                                                                      float64
float64
                                                                                       float64
In [348]: model = LinearRegression()
model.fit(x_train.reshape(-1,1), y_train)
Out[348]: LinearRegression()
In [365]: bkk_temp = np.array([12]).reshape(-1,1)
print(model.predict(bkk_temp))
In [366]: y_pred = model.predict(x_test.reshape(-1,1))
In [367]: BKK_temp_pred = pd.DataFrame({'Global Temp': x_test, 'Bangkok Temp': y_pred})
```

- a) Importing the relevant libraries need for implementing Linear Regression prediction
- b) Specity global\_avg\_temp as x and Bangkok avg\_temp as y, then proceed to split the data to x\_train, x\_test, y\_train, y\_test
- c) Drop NA values presented since NAN values are presented in the data making the model unable to train the x\_train data
- d) Predict the temp:
  - i) Specify the bkk\_temp and plug into model.predict
  - Plug x\_test into model.predict to make prediction stored the y\_value as y\_pred, then create data frame of x\_test (Global Temp) and y\_pred (Bangkok Temp)

In [364]: BKK\_temp\_pred

Out[364]:

	Global Temp	Bangkok Temp
0	9.29	27.885559
1	9.32	27.911977
2	8.52	27.207510
3	8.32	27.031393
4	8.53	27.216316
5	7.63	26.423790
6	8.01	26.758412
7	7.80	26.573490
8	8.53	27.216316
9	8.75	27.410044
10	8.34	27.049005
11	8.38	27.084228
12	8.02	26.767218
13	9.20	27.806307
14	8.18	26.908112
15	8.85	27.498103
16	8.63	27.304374
17	7.97	26.723189
18	8.12	26.855276
19	8.29	27.004976
20	8.21	26.934529
21	7.98	26.731995
22	8.73	27.392433
23	8.57	27.251539
24	8.07	26.811247
25	8.63	27.304374
26	8.80	27.454073
27	8.17	26.899306
28	8.43	27.128257