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#!/usr/bin/env python
# coding: utf-8
# In[1]:
# Create Latitude and Longitude Combinations
# Import the dependencies.
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
# In[2]:
# Create a set of random latitude and longitude combinations, size 1500.
lats = np.random.uniform(low=-90.000, high=90.000, size=1500)
lngs = np.random.uniform(low=-180.000, high=180.000, size=1500)
lat_lngs = zip(lats, lngs)
lat lngs
# In[3]:
# Add the latitudes and longitudes to a list.
coordinates = list(lat_lngs)
coordinates
# In[4]:
# Use the tuple() function to display the latitude and longitude combinations.
for coordinate in coordinates:
    print(coordinate[0], coordinate[1])
# In[5]:
# Use the citipy module to determine city based on latitude and longitude.
from citipy import citipy
# In[6]:
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# Use the fifteen hundred pairs of latitudes and longitudes we used from our zip
practice to get a city and country code from the citipy module
# Use the tuple() function to display the latitude and longitude combinations.
for coordinate in coordinates:
    print(citipy.nearest_city(coordinate[0], coordinate[1]).city_name,
          citipy.nearest_city(coordinate[0], coordinate[1]).country_code)
# In[7]:
# Create a list for holding the cities.
cities = []
# Identify the nearest city for each latitude and longitude combination.
for coordinate in coordinates:
    city = citipy.nearest city(coordinate[0], coordinate[1]).city name
    print(coordinate[0], coordinate[1])
    if city not in cities:
        cities.append(city)
# Print the city count to confirm sufficient count.
len(cities)
# In[8]:
city = citipy.nearest_city(25.12903645, -67.59741259)
city.city name
# In[9]:
# Create a list for holding the cities.
cities = []
# Identify the nearest city for each latitude and longitude combination.
for coordinate in coordinates:
    city = citipy.nearest_city(coordinate[0], coordinate[1]).city_name
    # If the city is unique, then we will add it to the cities list.
    if city not in cities:
        cities.append(city)
# Print the city count to confirm sufficient count.
```

```
len(cities)
# In[11]:
# Import the requests library.
import import_ipynb
import requests
import json
# Import the API key.
from config import weather_api_key
# In[12]:
# Starting URL for Weather Map API Call.
url = "http://api.openweathermap.org/data/2.5/weather?units=Imperial&APPID=" +
weather_api_key
type(url)
# In[13]:
# Create an endpoint URL for a city.
city_url = url + "&q=" + "Boston"
# In[14]:
print(requests.get(city_url))
# In[15]:
# Using json()
requests.get(city_url).json()
# In[16]:
# Make a 'Get' request for the city weather.
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city_weather = requests.get(city_url)
city weather
# In[17]:
# Get the text of the 'Get' request.
city weather.text
# In[18]:
# Get the JSON text of the 'Get' request.
city_weather.json()
# In[19]:
# Handle Request Errors
# Create an endpoint URL for a city.
city_url = url + "&q=" + "Boston"
city weather = requests.get(city url)
if city_weather.status_code == 200:
    print(f"City Weather found.")
else:
    print(f"City weather not found.")
# In[20]:
# Mining the JSON file to retrieve specific weather data for each city and add it
to a DataFrame
# 1. Let's look at Boston city first: correct the spelling for the city of Boston
# Create an endpoint URL for a city.
city url = url + "&q=" + "Boston"
city_weather = requests.get(city_url)
city_weather.json()
# In[21]:
# Get the JSON data.
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boston_data = city_weather.json()
# Get the corresponding value using sys
boston_data["sys"]
# In[22]:
# Get the corresponding value using sys & country
boston_data["sys"]["country"]
# In[23]:
# Retrieve the date in the weather data using dt
boston_data["dt"]
# In[24]:
boston_data["main"]["temp_max"]
# In[25]:
# Get the time of day, the latitude, longitude, maximum temperature, humidity,
percent cloudiness, and wind speed
lat = boston data["coord"]["lat"]
lng = boston_data["coord"]["lon"]
max_temp = boston_data["main"]["temp_max"]
humidity = boston_data["main"]["humidity"]
clouds = boston_data["clouds"]["all"]
wind = boston_data["wind"]["speed"]
print(lat, lng, max_temp, humidity, clouds, wind)
# In[26]:
# Convert the Date Timestamp
# Import the datetime module from the datetime library.
from datetime import datetime
# Get the date from the JSON file.
date = boston_data["dt"]
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# Convert the UTC date to a date format with year, month, day, hours, minutes,
and seconds.
datetime.utcfromtimestamp(date)
# In[27]:
# Convert this datetime format to 2019-10-21 17:24:35 using the Python string
format method
# strftime() and adding how we want the string to look inside the parentheses
datetime.utcfromtimestamp(date).strftime('%Y-%m-%d %H:%M:%S')
# In[28]:
# 1. Get the 500 City Weather Data
# 1.1. Import Dependencies, and Initialize an Empty List and Counters
# Import the time library and the datetime module from the datetime library
import time
from datetime import datetime
# In[29]:
# Create an empty list to hold the weather data.
city_data = []
# Print the beginning of the logging.
print("Beginning Data Retrieval ")
print("-----")
# Create counters.
record count = 1
set_count = 1
# In[31]:
# Loop through all the cities in the list.But, instead of using two for loops,
# we can use the enumerate() method as an alternative way to iterate through the
list of cities and retrieve both the index,
# and the city from the list.
for i, city in enumerate(cities):
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# Group cities in sets of 50 for logging purposes.
    if (i \% 50 == 0 \text{ and } i >= 50):
        set count += 1
        record count = 1
        time.sleep(60)
    # Create endpoint URL with each city.
    city url = url + "&q=" + city.replace(" ","+")
    # Log the URL, record, and set numbers and the city.
    print(f"Processing Record {record count} of Set {set count} | {city}")
    # Add 1 to the record count.
    record count += 1
# In[32]:
# Handle API Request Errors with try-except Blocks
for i, city in enumerate(cities):
    # Group cities in sets of 50 for logging purposes.
    if (i \% 50 == 0 \text{ and } i >= 50):
        set_count += 1
        record count = 1
        time.sleep(60)
    # Create endpoint URL with each city.
    city_url = url + "&q=" + city.replace(" ","+")
    # Log the URL, record, and set numbers and the city.
    print(f"Processing Record {record_count} of Set {set_count} | {city}")
    # Add 1 to the record count.
    record count += 1
# Run an API request for each of the cities.
    try:
        # Parse the JSON and retrieve data.
        city weather = requests.get(city url).json()
        # Parse out the needed data.
        city lat = city weather["coord"]["lat"]
        city_lng = city_weather["coord"]["lon"]
        city_max_temp = city_weather["main"]["temp_max"]
        city humidity = city weather["main"]["humidity"]
        city clouds = city weather["clouds"]["all"]
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city_wind = city_weather["wind"]["speed"]
       city_country = city_weather["sys"]["country"]
       # Convert the date to ISO standard.
       city date = datetime.utcfromtimestamp(city weather["dt"]).strftime('%Y-
%m-%d %H:%M:%S')
       # Append the city information into city_data list.
       city_data.append({"City": city.title(),
                         "Lat": city_lat,
                         "Lng": city lng,
                         "Max Temp": city_max_temp,
                         "Humidity": city humidity,
                         "Cloudiness": city clouds,
                         "Wind Speed": city_wind,
                         "Country": city country,
                         "Date": city_date})
# If an error is experienced, skip the city.
   except:
       print("City not found. Skipping...")
# Indicate that Data Loading is complete.
print("----")
print("Data Retrieval Complete
print("-----")
# In[33]:
# Create a DataFrame of City Weather Data
# Convert the array of dictionaries to a Pandas DataFrame.
import pandas as pd
city data df = pd.DataFrame(city data)
city_data_df.head(10)
# In[34]:
new_column_order = ["City", "Country", "Date","Lat", "Lng", "Max Temp",
"Humidity", "Cloudiness", "Wind Speed"]
# In[35]:
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city_data_df = city_data_df[new_column_order]
city data df.head(10)
# In[38]:
# Create the output file (CSV).
output_data_file = "cities.csv"
# Export the City Data into a CSV.
city_data_df.to_csv(output_data_file, index_label="City_ID")
# In[39]:
# Extract relevant fields from the DataFrame for plotting.
lats = city_data_df["Lat"]
max_temps = city_data_df["Max Temp"]
humidity = city data df["Humidity"]
cloudiness = city_data_df["Cloudiness"]
wind_speed = city_data_df["Wind Speed"]
# In[40]:
# Import the time module.
import time
# Get today's date in seconds.
today = time.time()
today
# In[41]:
today = time.strftime("%x")
today
# In[43]:
# Now, we can add time.strftime("%x") to our plt.title() function in our scatter
plot.
# Import time module
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```
import time
from matplotlib import pyplot as plt
# Build the scatter plot for latitude vs. max temperature.
plt.scatter(lats,
            max_temps,
            edgecolor="black", linewidths=1, marker="o",
            alpha=0.8, label="Cities")
# Incorporate the other graph properties.
plt.title(f"City Latitude vs. Max Temperature "+ time.strftime("%x"))
plt.ylabel("Max Temperature (F)")
plt.xlabel("Latitude")
plt.grid(True)
# Save the figure.
plt.savefig("Fig1.png")
# Show plot.
plt.show()
# In[44]:
# Build the scatter plot for latitude vs. Humidity.
plt.scatter(lats,
            humidity,
            edgecolor="black", linewidths=1, marker="o",
            alpha=0.8, label="Cities")
# Incorporate the other graph properties.
plt.title(f"City Latitude vs. Humidity "+ time.strftime("%x"))
plt.ylabel("Humidity (%)")
plt.xlabel("Latitude")
plt.grid(True)
# Save the figure.
plt.savefig("Fig2.png")
# Show plot.
plt.show()
# In[45]:
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# Build the scatter plot for latitude vs. cloudiness.
plt.scatter(lats,
            cloudiness,
            edgecolor="black", linewidths=1, marker="o",
            alpha=0.8, label="Cities")
# Incorporate the other graph properties.
plt.title(f"City Latitude vs. cloudiness (%)"+ time.strftime("%x"))
plt.ylabel("cloudiness (%)")
plt.xlabel("Latitude")
plt.grid(True)
# Save the figure.
plt.savefig("Fig3.png")
# Show plot.
plt.show()
# In[46]:
# Build the scatter plot for latitude vs. cloudiness.
plt.scatter(lats,
            wind_speed,
            edgecolor="black", linewidths=1, marker="o",
            alpha=0.8, label="Cities")
# Incorporate the other graph properties.
plt.title(f"City Latitude vs. Wind speed "+ time.strftime("%x"))
plt.ylabel("Wind speed (mph)")
plt.xlabel("Latitude")
plt.grid(True)
# Save the figure.
plt.savefig("Fig4.png")
# Show plot.
plt.show()
# In[47]:
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# Import necessary libraries
# Import linregress
from scipy.stats import linregress
# Import matplotlib
import matplotlib.pyplot as plt
# In[48]:
# Create a Linear Regression Function
# Create a function to create perform linear regression on the weather data
# and plot a regression line and the equation with the data.
def plot_linear_regression(x_values, y_values, title, y_label, text_coordinates):
    # Run regression on hemisphere weather data.
    (slope, intercept, r value, p value, std err) = linregress(x values,
y values)
    # Calculate the regression line "y values" from the slope and intercept.
    regress values = x values * slope + intercept
    # Get the equation of the line.
    line_eq = "y = " + str(round(slope,2)) + "x + " + str(round(intercept,2))
    # Create a scatter plot and plot the regression line.
    plt.scatter(x_values,y_values)
    plt.plot(x values, regress values, "r")
    # Annotate the text for the line equation.
    plt.annotate(line_eq, text_coordinates, fontsize=15, color="red")
    plt.title(title)
    plt.xlabel('Latitude')
    plt.ylabel(y label)
    plt.show()
# For this code there will be no output until we call the function with five
parameters.
# In[49]:
# Create the Hemisphere DataFrames
# Create Northern and Southern Hemisphere DataFrames.
northern hemi df = city data df.loc[(city data df["Lat"] >= 0)]
southern hemi df = city data df.loc[(city data df["Lat"] < 0)]</pre>
# In[50]:
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```
# Perform Linear Regression on the Maximum Temperature for the Northern
Hemisphere
# Linear regression on the Northern Hemisphere
x values = northern hemi df["Lat"]
y_values = northern_hemi_df["Max Temp"]
# Call the function.
plot linear_regression(x_values, y_values,
                       'Linear Regression on the Northern Hemisphere \
                       for Maximum Temperature', 'Max Temp', (30,90))
r = np.corrcoef(x values, y values)
# In[51]:
# Perform Linear Regression on the Maximum Temperature for the Southern
Hemisphere
# Linear regression on the Southern Hemisphere
x values = southern hemi df["Lat"]
y_values = southern_hemi_df["Max Temp"]
# Call the function.
plot linear regression(x values, y values,
                       'Linear Regression on the Southern Hemisphere \
                       for Maximum Temperature', 'Max Temp',(-30,40))
 = np.corrcoef(x_values, y_values)
# In[52]:
# Create the linear equation and scatter plot of the latitude and percent
humidity for the Northern and Southern Hemispheres.
# Perform Linear Regression on the percent humidity for the Northern Hemisphere
# Linear regression on the Northern Hemisphere
x values = northern hemi df["Lat"]
y_values = northern_hemi_df["Humidity"]
# Call the function.
plot_linear_regression(x_values, y_values,
                       'Linear Regression on the Northern Hemisphere \
                       for percent humidity', 'Humidity', (40,10))
```

```
r = np.corrcoef(x values, y values)
# In[53]:
# Perform Linear Regression on the Percent Humidity for the Southern Hemisphere
# Perform Linear Regression on the percent humidity for the Southern Hemisphere
# Linear regression on the Southern Hemisphere
x values = southern hemi df["Lat"]
y_values = southern_hemi_df["Humidity"]
# Call the function.
plot_linear_regression(x_values, y_values,
                       'Linear Regression on the Southern Hemisphere \
                       for percent humidity', 'Humidity',(-25,20))
r = np.corrcoef(x values, y values)
# In[54]:
# Perform Linear Regression on the Percent Cloudiness for the Northern Hemisphere
# Perform Linear Regression on the percent Cloudiness for the Northern Hemisphere
# Linear regression on the Northern Hemisphere
x values = northern_hemi_df["Lat"]
y_values = northern_hemi_df["Cloudiness"]
# Call the function.
plot_linear_regression(x_values, y_values,
                       'Linear Regression on the Northern Hemisphere \
                       for percent cloudiness', 'Cloudiness',(10,60))
r = np.corrcoef(x_values, y_values)
# In[55]:
# Perform Linear Regression on the percent Cloudiness for the Southern Hemisphere
# Linear regression on the Southern Hemisphere
x values = southern hemi df["Lat"]
v values = southern hemi df["Cloudiness"]
```

```
# Call the function.
plot linear regression(x values, y values,
                       'Linear Regression on the Southern Hemisphere \
                        for % Cloudiness', '% Cloudiness',(-50,60))
r = np.corrcoef(x_values, y_values)
# Perform Linear Regression on the Wind Speed for the Northern Hemisphere
# Linear regression on the Northern Hemisphere
x_values = northern_hemi_df["Lat"]
y_values = northern_hemi_df["Wind Speed"]
# Call the function.
plot_linear_regression(x_values, y_values,
                       'Linear Regression on the Northern Hemisphere \
                        for Wind Speed', 'Wind Speed', (10,40))
r = np.corrcoef(x values, y values)
# In[57]:
# Perform Linear Regression on the Wind Speed for the Southern Hemisphere
# Linear regression on the Southern Hemisphere
x values = southern_hemi_df["Lat"]
y_values = southern_hemi_df["Wind Speed"]
# Call the function.
plot linear_regression(x_values, y_values,
                       'Linear Regression on the Southern Hemisphere \
                        for Wind Speed', 'Wind Speed', (-40,30))
r = np.corrcoef(x_values, y_values)
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