	Homework 2 PSTAT134 API's & Webscraping
	Part One: Analyzing the Weather In this section, you will gain more practice working with public APIs, this time using a public weather API, WeatherAPI. The first thing you'll need to access this is an API key which you can sign up for here.
	Exercise 1 Use the http://api.weatherapi.com/v1/current.json URL to access the API and obtain real-time weather data. Note that you will want to specify three query parameters, at least - key, which should be set to your individual API key, q, which should equal the latitude and
	longitude of a specified location - for example q = "34.432961, -119.856870" - and aqi, which indicates whether you want to obtain air quality data ("yes" or "no"). Obtain current real-time weather data for fifty randomly selected locations. Use the cities within the cities.csv file.
In [166	<pre>import api_key_holder #Reference to api_key so we don't publicize import numpy as np import pandas as pd import matplotlib.pyplot as plt</pre>
In [154	<pre>#Location list locations_df = pd.read_csv('data/cities.csv') locations = locations_df['names'].tolist() #Prep key and url ref. for request base_url = "http://api.weatherapi.com/v1/current.json" api key = api key holder.weatherapi key</pre>
	Exercise 2 Write code in Python to extract and store the following data for each location (see orig.)
In [212	<pre>weather_data = [] #init for loc in locations: parameters = { 'key': api_key, 'q': loc,</pre>
	<pre>'aqi': 'yes' } response = requests.get(base_url, parameters) data = response.json() entry_current = {"City_Name": data['location']['name'],</pre>
	"Daytime": data['current']['is_day'], "Temperature (f)": data['current']['temp_f'], "Humidity": data['current']['humidity'], "Weather_Desc": data['current']['condition']['text'], "Wind Speed (mph)": data['current']['wind_mph'], "Precipitation (mm)": data['current']['precip_mm'], "US EPA AQI": data['current']['air quality']['us-epa-index']}
	weather_data.append (entry_current) Exercise 3 Create a scatterplot of temperature vs. humidity. Add a linear regression line to the plot. What are the estimated intercept and slope
In [215	values for this linear regression? Does there appear to be a significant relationship between temperature and humidity?
In [217	<pre>plt.scatter(x = df_weather['Humidity'] + intercept plt.scatter(x = df_weather['Humidity'] + intercept plt.plot(df_weather['Humidity'] + intercept plt.plot(df_weather['Humidity'], regression_line, color='red', label="Regression_Line") plt.xlabel("Humidity"), plt.ylabel("Temperature (F)");</pre>
	90 -
	80 - (L) 70
	60 -
	20 40 60 80 100
	Humidity Given the linear regression line, we see an intercept of 87.23 degrees F, and a slope of -0.2795. This suggests that as humidity increases, temperature tends to drop (however general domain knowledge of physics would indicated that as temperature drops, the relative humidity increases as colder air is able to hold onto less moisture). Meanwhile, with an
	r_value of -0.4499, this suggests that there is a weak-moderate negative relationship between temperature and humidity as mentioned earlier. Exercise 4
In [280	Create a bar chart of the EPA air quality index values. What does the distribbution of air quality look like? Identify the locations(s) with the best and worst air quality. plt.figure(figsize=(13,5)), plt.bar(df_weather['City_Name'], df_weather['US_EPA_AQI']), plt.xlabel('Location'), plt.ylabel('US_EPA_Air_Quality_Index'), plt.xticks(rotation=45, ha='right');
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	OS EPA Air Quality Index
	Y W 2 - 1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 -
In [262 Out[262]:	<pre>bad_aqi = df_weather[df_weather["US_EPA_AQI"] > 3] bad_aqi[['City_Name','Country', 'US_EPA_AQI']]</pre> City_Name Country US_EPA_AQI
:	1 Chongqing China 5 5 Tangshan China 5 8 Zunyi China 4 13 Pyongyang North Korea 4
	24 Busan Pakistan 4 29 Hyderabad India 4 33 Riyadh Saudi Arabia 4
In [270	39 Tianjin China 5 41 Lucknow India 4 46 Lanzhou China 4 good_aqi = df_weather[df_weather["US_EPA_AQI"] < 2]
Out[270]:	good_aqi[['City_Name','Country', 'US_EPA_AQI']] City_Name
	12 Daegu South Korea 1 16 Fes-Saïss Airport Morocco 1 22 Basra Iran 1
	Toronto Canada 1 Porto Alegre Brazil 1 Havana Cuba 1 Brisbane Australia 1
	36AucklandNew Zealand137QuitoEcuador138Guatemala CityGuatemala143AccraGhana1
	48 Kyoto Japan 1 49 Cape Town South Africa 1 Due to the sheer amount of locations displayed on the graph, it may be a bit difficult to interpret accurately so I've created
	the above dataframes as well. We see a slightly sporadic distribution of air qualities with AQI values of 1-2 being the most common while AQI values of 4-5 were quite rare. We see locations in China had the worst AQI with only Chinese cities like Chongqing, Tangshan, and Tianjin reaching values of 5. India, Saudi Arabia, North Korea, and Pakistan were the only other countries that had a city/cities with AQI scores over 3. Meanwhile, for the locations with the best air quality we see
	Brisbane, Guatemala City, Kyoto, and Basra to name a few. Note, these current values of AQI are subject to change and the interpretations from this document are only appropriate at the given time in which this was queried (Oct 31st 9:45 PCT). Exercise 5
In [319	Create a bar chart of the current weather description. Which conditions are the most common? Which are the least? weather_counts = df_weather["Weather_Desc"].value_counts() weather_counts.plot(kind = 'bar', color = 'lightblue') plt.title("Barchart of Weather Desc. Distribution of 50 Cities")
	Barchart of Weather Desc. Distribution of 50 Cities 10 -
	8 -
	6 - 6 - 6 - 6 - 6 - 6 - 6 - 6 - 6 - 6 -
	Sunny Clear Partly cloudy Patchy rain nearby Mist Partly Cloudy Overcast Light rain Light rain Light rain Fog Fog
	Based on the chart above we see that Sunny, Clear, and Partly cloudly are the most common conditions, while Fog, Light rain shower, and moderate rain were the least common city conditions at the time of this query.
	Part Two: Scraping Books
	In this section, we'll practice our web scraping skills by experimenting with a fictional online bookstore located here. Exercise 9
	 Scrape the first 20 resultes from this site. Create a data frame that stores the following for each book: Title Price (excluding tax) Star rating Whether the book is in stock.
-	<pre>• Whether the book is in stock from bs4 import BeautifulSoup url = "https://books.toscrape.com/"</pre>
	<pre>response = requests.get(url) soup = BeautifulSoup(response.content, "html.parser") book_pods = soup.find_all("article", class_ = "product_pod") book_df = [] titles = [] prices = []</pre>
In [530	<pre>ratings = [] availability = [] for book in book_pods: #book titles title = book.h3.a['title'] titles.append(title)</pre>
	<pre>#book prices price = book.find('p', class_= "price_color").text.strip().replace("f", "") prices.append(price) #book ratings rating = book.p['class'][1] rating_map = {"One": 1, "Two": 2, "Three": 3, "Four": 4, "Five": 5}</pre>
	<pre>ratings.append(rating_map.get(rating, 0)) #book availability status = book.find("p", class_="instock availability").text.strip() availability.append(status) book_df = pd.DataFrame({"Title": titles,</pre>
	<pre>"Rating": ratings,</pre>
In [532 Out[532]:	Title Price Rating Availability O A Light in the Attic 51.77 3 In stock Tipping the Velvet 53.74 1 In stock
	Soumission 50.10 1 In stock Sharp Objects 47.82 4 In stock Sapiens: A Brief History of Humankind 54.23 5 In stock The Requiem Red 22.65 1 In stock
	 The Dirty Little Secrets of Getting Your Dream 33.34 4 In stock The Coming Woman: A Novel Based on the Life of 17.93 3 In stock The Boys in the Boat: Nine Americans and Their 22.60 4 In stock
	The Black Maria 52.15 1 In stock Starving Hearts (Triangular Trade Trilogy, #1) 13.99 2 In stock Shakespeare's Sonnets 20.66 4 In stock Set Me Free 17.46 5 In stock
	13 Scott Pilgrim's Precious Little Life (Scott Pi 52.29 5 In stock 14 Rip it Up and Start Again 35.02 5 In stock 15 Our Band Could Be Your Life: Scenes from the A 57.25 3 In stock 16 Olio 23.88 1 In stock
	17 Mesaerion: The Best Science Fiction Stories 18 37.59 1 In stock 18 Libertarianism for Beginners 51.33 2 In stock 19 It's Only the Himalayas 45.17 2 In stock
In [537	Exercise 10 Create a histogram of prices for these 20 books. What is the average price? book_df['Price'] = pd.to_numeric(book_df['Price'])
	<pre>plt.hist(book_df['Price'], bins = 10, color = 'lightblue', edgecolor = 'black'); average_price = book_df['Price'].mean() print(f"Average Price: £{average_price:.2f}") Average Price: £38.05</pre>
	5 - 4 -
	3 -
	0 20 30 40 50
In [542	Exercise 11 Create a bar chart of star rating for these 20 books. Find the book(s) with the highest and lowest star ratings. rating_counts = book_df['Rating'].value_counts().sort_index() rating_counts.plot(kind = 'bar', color = 'lightblue')
	plt.title("Barplot of Book Ratings"), plt.xlabel("Star Rating"); Barplot of Book Ratings 6 -
	5 - 4 -
	3 - 2 -
	1-
In [546	#Finding lowest rated books one_star = book_df[book_df['Rating'] == 1]
Out[546]:	Title Price Rating Availability 1 Tipping the Velvet 53.74 1 In stock
	5 The Requiem Red 22.65 1 In stock 9 The Black Maria 52.15 1 In stock 16 Olio 23.88 1 In stock
In [548	<pre>17 Mesaerion: The Best Science Fiction Stories 18 37.59 1 In stock #Finding highest rated books five_star = book_df[book_df['Rating'] == 5] five_star</pre>
Out[548]:	Title Price Rating Availability 4 Sapiens: A Brief History of Humankind 54.23 5 In stock 12 Set Me Free 17.46 5 In stock 13 Scott Pilgrim's Precious Little Life (Scott Pi 52.29 5 In stock
	Rip it Up and Start Again 35.02 5 In stock