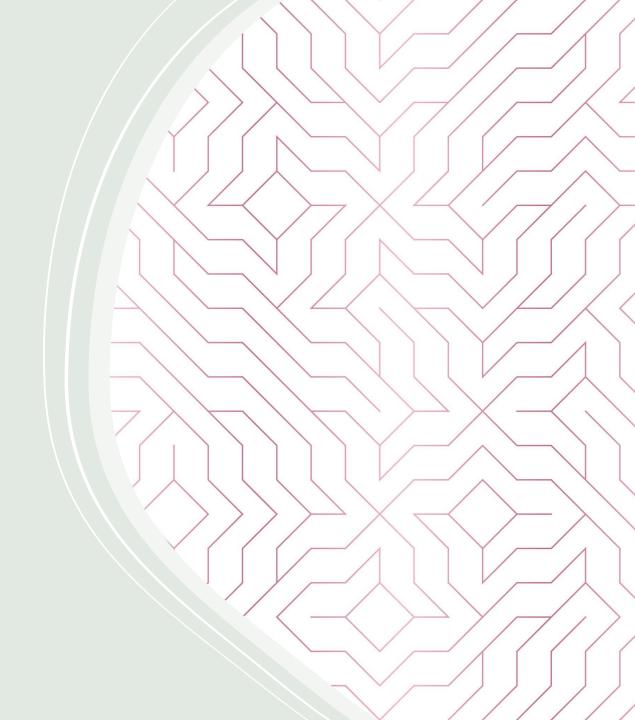
Charming the Snake: an introduction to Python

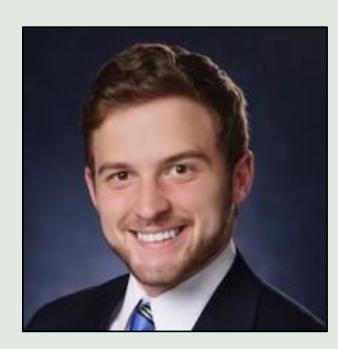
03/09/2022 - Tim Witham

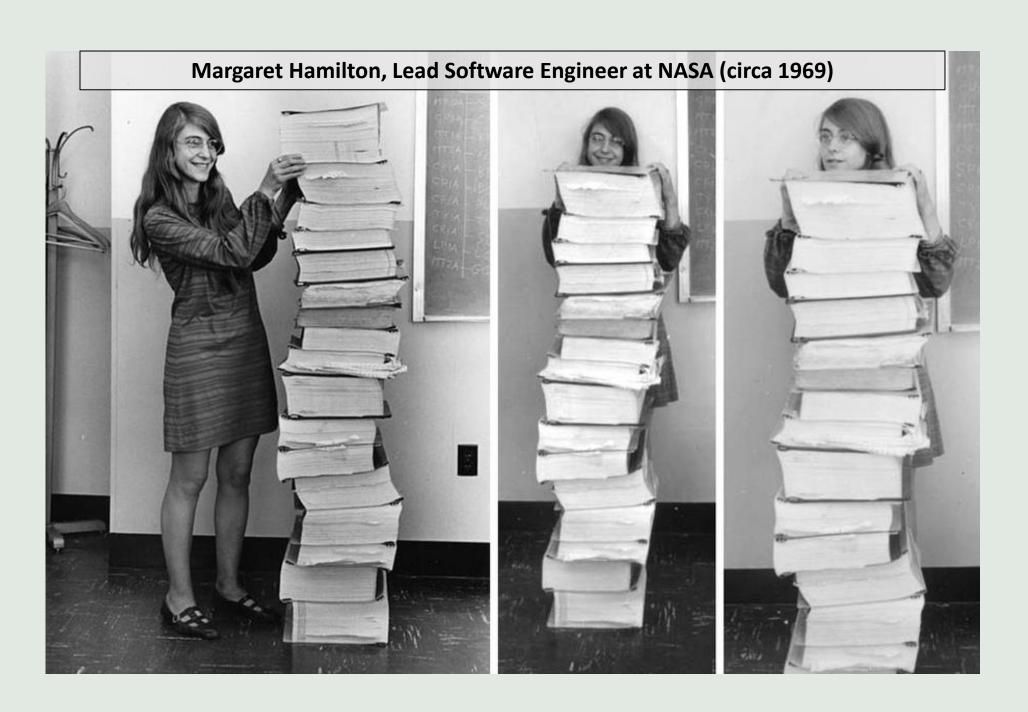


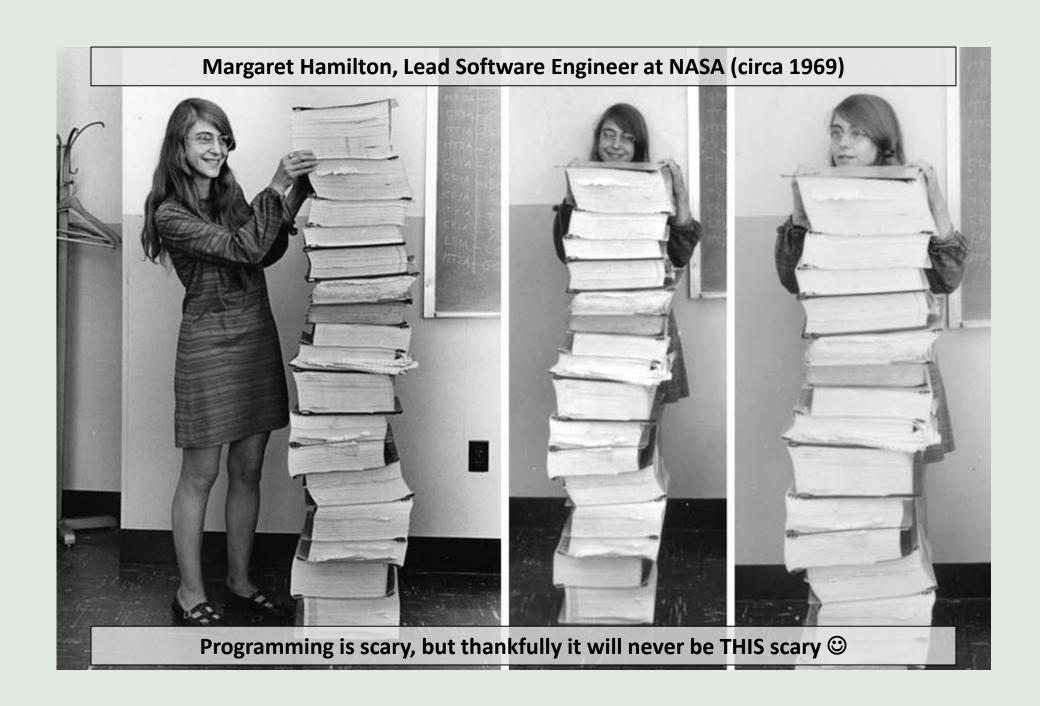
About me:

- B.S. Geosciences University of Tulsa ('17)
 - Geoscience Student of the Year (2016) Tulsa Geological Society
- M.S. Geosciences Penn State ('19)
 - Imperial Barrel Award, 3rd Place (2018) AAPG
- Geologist at Diamondback Energy in Midland, TX since 2019.









Motivation for Learning Python

Feb 2022	Feb 2021	Change	Programming Language	Ratings	Change
1	3	^	🤚 Python	15.33%	+4.47%
2	1	~	G c	14.08%	-2.26%
3	2	•	<u>«</u> Java	12.13%	+0.84%
4	4		○ C++	8.01%	+1.13%
5	5		C#	5.37%	+0.93%
6	6		VB Visual Basic	5.23%	+0.90%
7	7		JS JavaScript	1.83%	-0.45%
8	8		PhP PHP	1.79%	+0.04%
9	10	^	Asm Assembly language	1.60%	-0.06%
10	9	•	SQL SQL	1.55%	-0.18%

The TIOBE scale shows us which programming language is most popular!



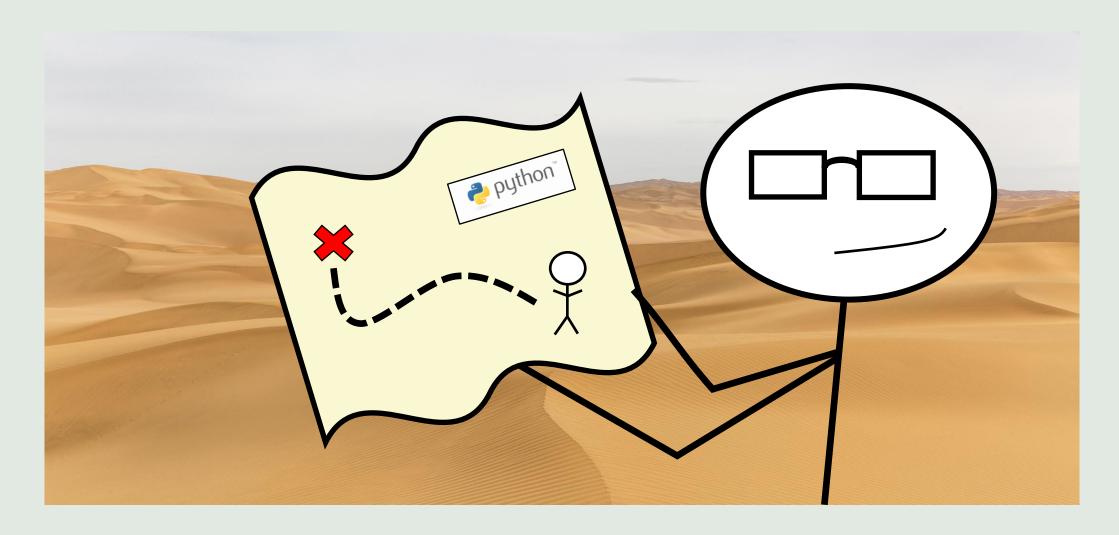
Oscar Wilde – 19th C. Poet

Source: **TIOBE index** (February 2022)

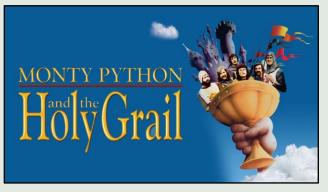
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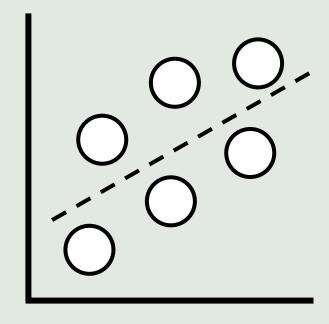




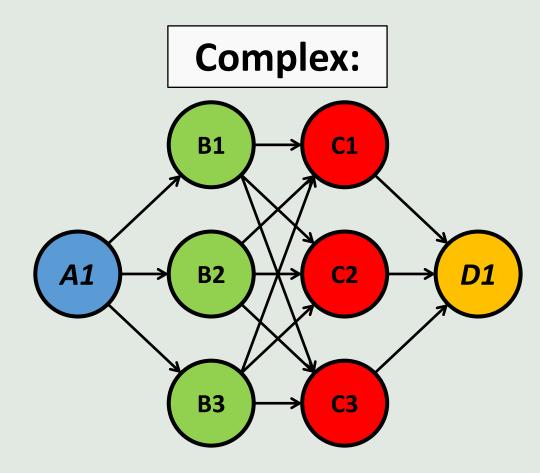
- It is a high level scripting language:
 - We write code down, and the computer executes it line by line.
 - "sentences" rather than big "paragraphs" of code.
- Python helps us automate the boring stuff we do every day.
- Runs tasks way faster than we ever can.
- IT CUTS DOWN ON ALL THE "CLICKY-CLICKY".
- Python is also "open-source", meaning it is FREE!
 - Free is the best type of software according to your future boss.

Learn the basics of data manipulation BEFORE applying complex "Machine Learning" techniques:

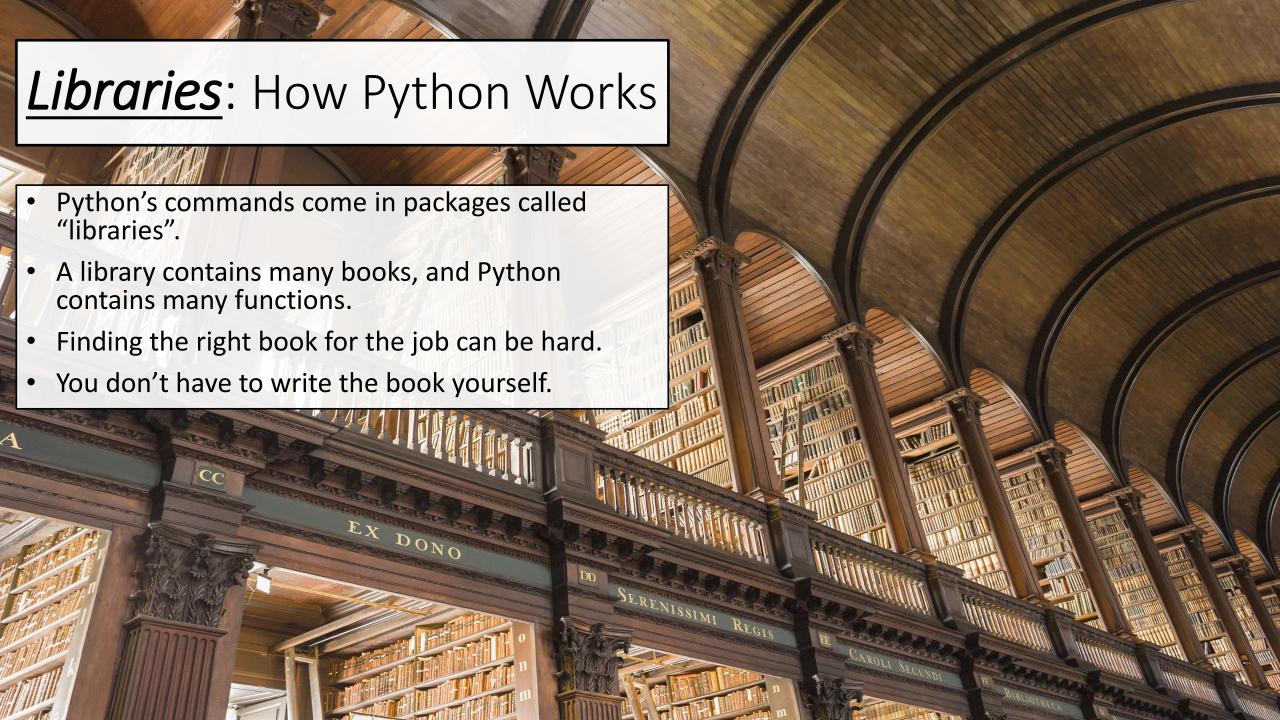
Simple:



Cross-plot



Machine Learning



Goals of our Script:

- 1.) Pull data on median annual household income for the USA from Wikipedia.
- 2.) Plot the data up as a heat-map.
- Python will make maps faster than pointing and clicking.
- Speed is important if you will be making maps every day.

Please feel free to view my Python script and use it on your own! https://github.com/timowith/gis_scripts/blob/main/Wiki_Web_Scr ape_Py.txt

Pulling data off the internet

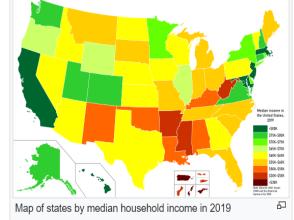
- Called "Web Scraping".
- All websites have background code we can access.
- Right click on a webpage and select "inspect" to view the code.
- Use the web code to access the raw data.
- The Python library we use for this is BeautifulSoup.

States and territories ranked by median household income

Data given in current dollars. Note that tables do not reflect the margin of error in the values. [2]

State Rank	States and Washington, D.C.	2019 💠	2018 \$	2017 \$	2016 💠	2015 ♦	2014 \$	2013 \$	2012 \$	2011 💠	2010 \$	Average annual growth rate (current dollars) in 2010- ♦ 2019, %
	United States	\$65,712	\$63,179	\$60,336	\$57,617	\$55,775	\$53,657	\$52,250	\$51,371	\$50,502	\$50,046	3.07%
_	₩ Washington, D.C.	\$92,266	\$85,203	\$82,372	\$75,506	\$75,628	\$71,648	\$67,572	\$66,583	\$63,124	\$60,903	4.72%
1	Maryland Maryland	\$86,738	\$83,242	\$80,776	\$78,945	\$75,847	\$73,971	\$72,483	\$71,122	\$70,004	\$68,854	2.6%
2	Massachusetts	\$85,843	\$79,835	\$77,385	\$75,297	\$70,628	\$69,160	\$66,768	\$65,339	\$62,859	\$62,072	3.67%
3	New Jersey	\$85,751	\$81,740	\$80,088	\$76,126	\$72,222	\$71,919	\$70,165	\$69,667	\$67,458	\$67,681	2.66%
4	Hawaii	\$83,102	\$80,212	\$77,765	\$74,511	\$73,486	\$69,592	\$68,020	\$66,259	\$61,821	\$63,030	3.12%
5	California	\$80,440	\$75,277	\$71,805	\$67,739	\$64,500	\$61,933	\$60,190	\$58,328	\$57,287	\$57,708	3.76%







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Pulling data off the internet

From Wikipedia, the free encyclopedia

For the list of states by income inequality, see List of U.S. states by Gini coefficient.

This is a list of U.S. states, territories and the District of Columbia by income. Data is given according to the 2019 American Community Survey (ACS) 1-Year Estimates, except for the American Samoa, Guam, the Northern Mariana Islands and the U.S. Virgin Islands, for which the data comes from 2010, as ACS does not operate in these areas. [note 1]

Contents [hide]

- 1 States and territories ranked by median household income
- 2 States and territories ranked by per capita income
- 3 See also
- 4 Notes
- 5 References
- 6 External links

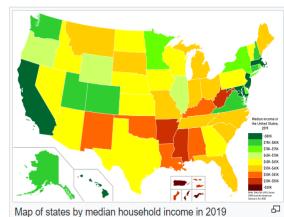
Right click on the web page and select the "Inspect" option

States and territories ranked by median household income

Data given in current dollars. Note that tables do not reflect the margin of error in the values. [2]

State •	States and	2019 \$	2018 \$	2017 \$	2016 \$	2015 \$	2014 \$	2013 \$	2012 \$	_	ew page source spect	ctrl+U >wth rate n 2010- ◆
Rank	Washington, D.C.											2019, %
	United States	\$65,712	\$63,179	\$60,336	\$57,617	\$55,775	\$53,657	\$52,250	\$51,371	\$50,502	\$50,046	3.07%
_	₩ Washington, D.C.	\$92,266	\$85,203	\$82,372	\$75,506	\$75,628	\$71,648	\$67,572	\$66,583	\$63,124	\$60,903	4.72%
1	Maryland	\$86,738	\$83,242	\$80,776	\$78,945	\$75,847	\$73,971	\$72,483	\$71,122	\$70,004	\$68,854	2.6%
2	Massachusetts	\$85,843	\$79,835	\$77,385	\$75,297	\$70,628	\$69,160	\$66,768	\$65,339	\$62,859	\$62,072	3.67%
3	New Jersey	\$85,751	\$81,740	\$80,088	\$76,126	\$72,222	\$71,919	\$70,165	\$69,667	\$67,458	\$67,681	2.66%
4	Hawaii	\$83,102	\$80,212	\$77,765	\$74,511	\$73,486	\$69,592	\$68,020	\$66,259	\$61,821	\$63,030	3.12%
5	California	\$80,440	\$75,277	\$71,805	\$67,739	\$64,500	\$61,933	\$60,190	\$58,328	\$57,287	\$57,708	3.76%
6	Connecticut	\$78,833	\$76,348	\$74,168	\$73,433	\$71,346	\$70,048	\$67,098	\$67,276	\$65,753	\$64,032	2.34%







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Français Bahasa Indonesia

Pulling data off the internet

Web scraping

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For broader coverage of this topic

How the website's background code looks... this is where we scrape raw data from

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.vector-body {

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line-height: 1.6;

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Inherited from div#bodyContent.vector-body

font-size: calc(1em * 0.875);

Web scraping, web harvesting, or web data extraction is data scraping used for extracting data from websites. The web scraping software may directly access the World Wide Web using the Hypertext Transfer Protocol or a web browser. While web scraping can be done manually by a software user, the term typically refers to automated processes implemented using a bot or web crawler. It is a form of copying in which specific data is gathered and copied from the web, typically into a central local database or spreadsheet, for later retrieval or analysis.

Web scraping a web page involves fetching it and extracting from it. Fetching is the downloading of a page (which a browser does when a user views a page). Therefore, web crawling is a main component of web scraping, to fetch pages for later processing. Once fetched, then extraction can take place. The content of a page may be parsed, searched, reformatted, its data copied into a spreadsheet or loaded into a database. Web scrapers typically take something out of a page, to make use of it for another purpose somewhere else. An example would be to find and copy names and telephone numbers, or companies and their URLs, or e-mail addresses to a list (contact scraping).

Web scraping is used for contact scraping, and as a component of applications used for web indexing, web mining and data mining, online price change monitoring and price comparison, product review scraping (to watch the competition), gathering real estate listings, weather data monitoring, website change detection, research, tracking online presence and reputation, web mashup, and web data integration.

Web pages are built using text-based mark-up languages (HTML and XHTML), and frequently contain a wealth of useful data in text form. However, most web pages are designed for human end-users and not for ease of automated use. As a result, specialized tools and software have been developed to facilitate the scraping of web pages.

Newer forms of web scraping involve monitoring data feeds from web servers. For example, JSON is commonly used as a transport storage mechanism between the client and the web server.

There are methods that some websites use to prevent web scraping, such as detecting and disallowing bots from crawling

Elements Console Sources Network <div id="mw-page-base" class="noprint"></div> <div id="mw-head-base" class="noprint"></div> ▼<div id="content" class="mw-body" role="main"> ▶ <div id="siteNotice">...</div> <div class="mw-indicators"> </div> ▶ <h1 id="firstHeading" class="firstHeading">...</h1> ▼ <div id="bodyContent" class="vector-body"> <div id="siteSub" class="noprint">From Wikipedia, the free encyclopedia </div> <div id="contentSub"></div> <div id="contentSub2"></div> <div id="jump-to-nav"></div> Jump to navigation Jump to search ▼<div id="mw-content-text" class="mw-body-content mw-content-ltr" lang="e n" dir="ltr"> ▼<div class="mw-parser-output"> == \$0 <div class="shortdescription nomobile noexcerpt noprint searchaux"</pre> style="display:none">Data scraping used for extracting data from websites</div> ▶ <table class="box-More citations needed plainlinks metadata ambox am box-content ambox-Refimprove" role="presentation">... ▶ <style data-mw-deduplicate="TemplateStyles:r1033289096">...</style> \div role="note" class="hatnote navigation-not-searchable">...</div> >... ... ctor-body div#mw-content-text.mw-body-content.mw-content-ltr div.mw-parser-output Layout Event Listeners DOM Breakpoints Properties Accessibility Filter :hov .cls + element.style { div { user agent stylesheet

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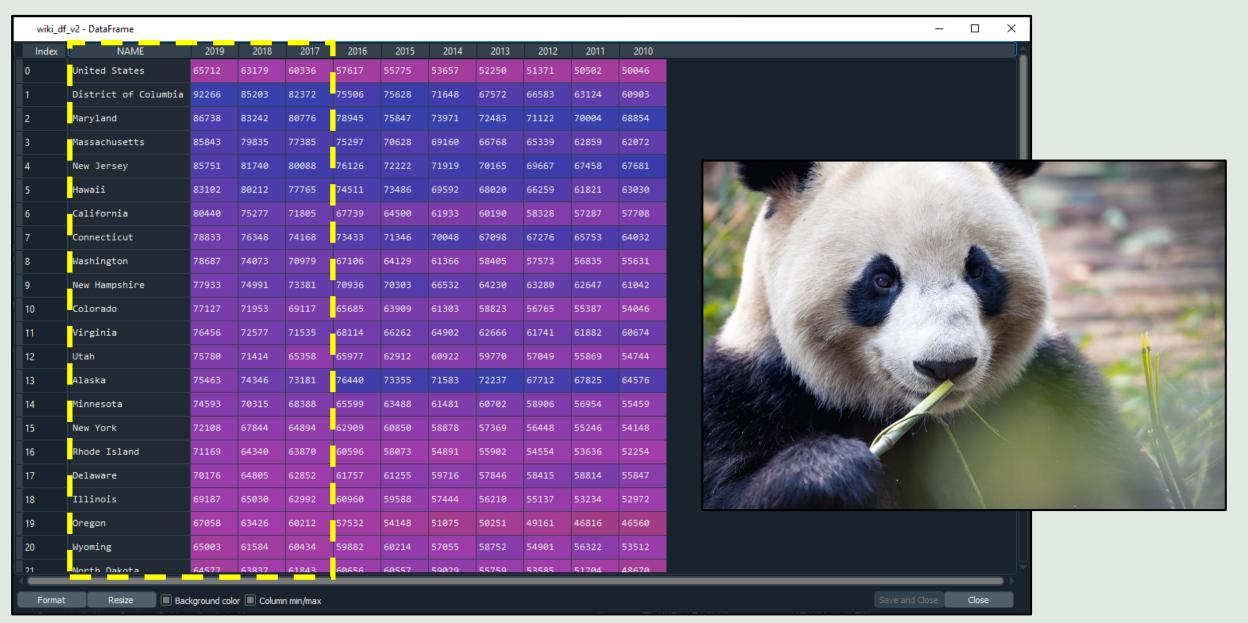
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Storing Data in Python

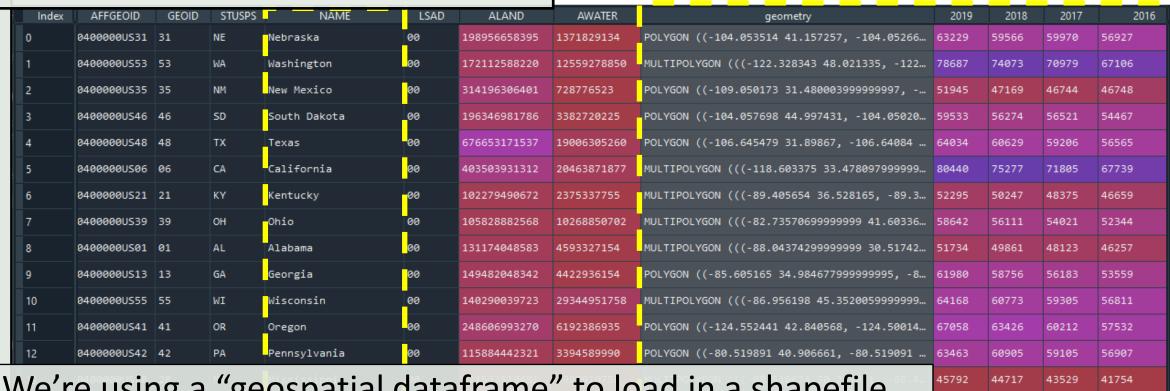


- Python can use many different data structures.
- I recommend "Dataframes".
- Dataframes are like Excel sheets:
 - Rows and columns format.
 - We choose names for the columns.
 - Rows and columns can be selected individually.
 - Calculations can be done over the whole dataframe very fast.
- Linear Algebra is a class that teaches data structure basics.
- The primary Python library for using dataframes is called Pandas.

Storing Data in Python



Plotting geospatial data with Python



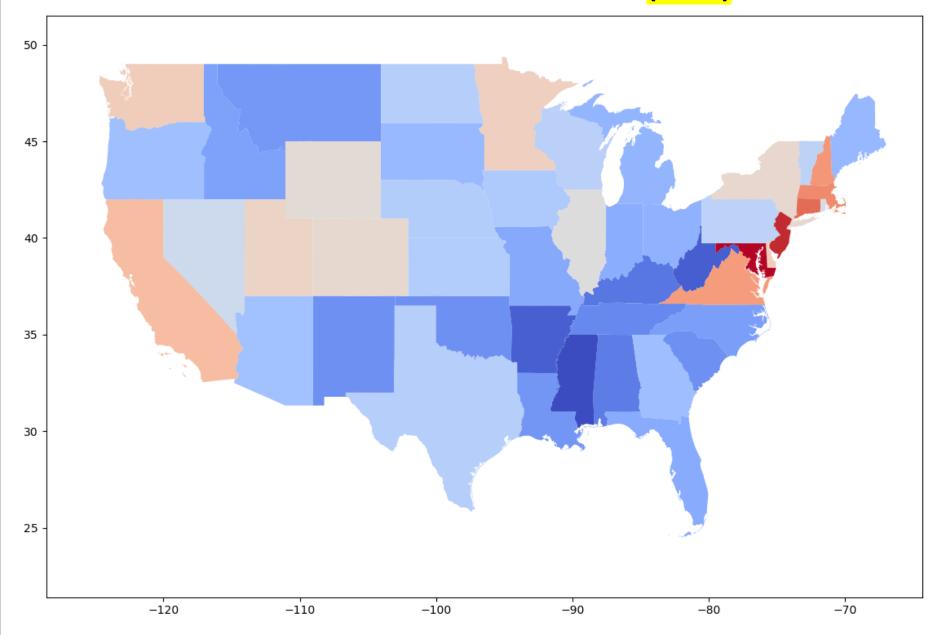
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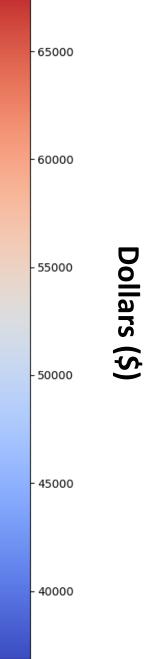
- We're using a "geospatial dataframe" to load in a shapefile.
- This geodataframe behaves just like a regular dataframe.
 - But now it has geospatial information too.

Background color Column min/max

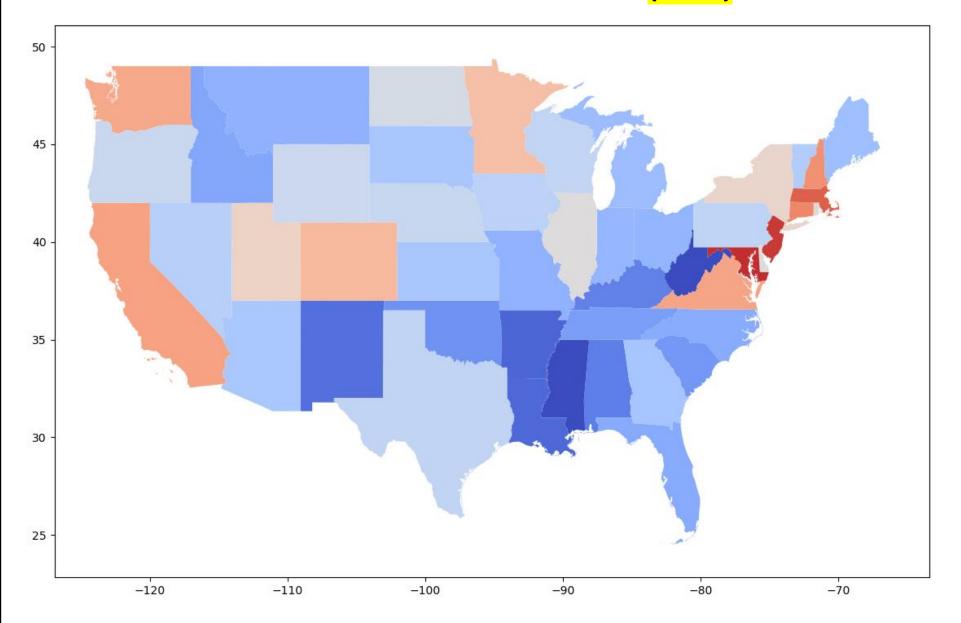
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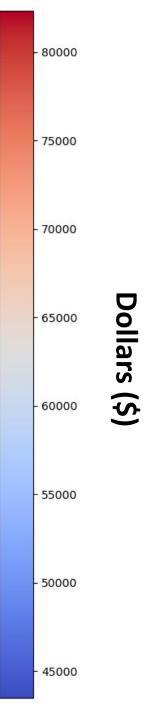
Median income of households in each state (2010)

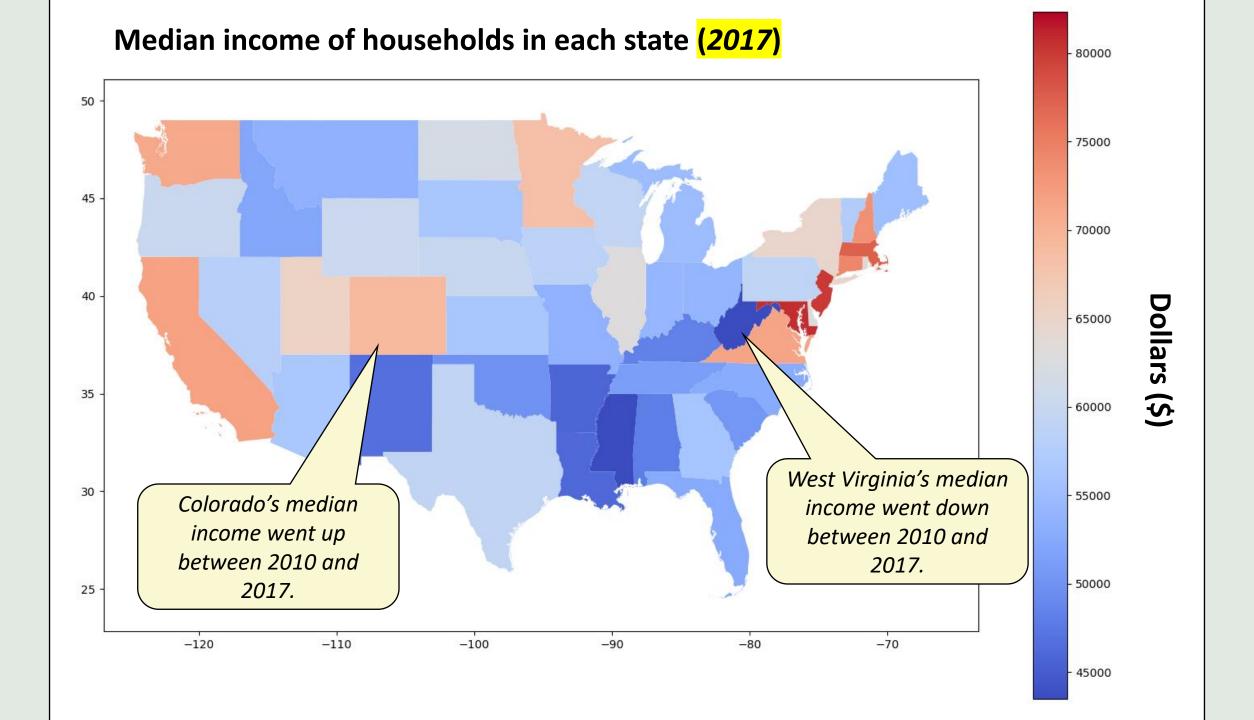




Median income of households in each state (2017)







Conclusion

- Programming skills will make you stand out.
- Choose one of the top 5 most popular languages to learn.
 - Python!
- DataFrames are a very concise way to store your data.

Sources

- https://medium.com/analytics-vidhya/web-scraping-a-wikipedia-table-into-a-dataframe-c52617e1f451
- https://geopandas.org/en/stable/getting_started/install.html
- https://pandas.pydata.org/docs/reference/api/pandas.DataFrame.ht
- https://www.tiobe.com/tiobe-index/

Important Python Libraries

- Data Organization:
 - pandas
- Geographic Mapping:
 - geopandas
- Manipulating and searching through strings:
 - re (short for 'regex')
- General mathematics:
 - Numpy (short for 'Numeric Python')
- General scientific concepts:
 - scipy (short for 'Scientific Python')
- Web scraping:
 - BeautifulSoup
- Pathway and folder manipulation:
 - OS
 - Glob
- Plotting/graphs:
 - matplotlib
- Machine Learning:
 - TensorFlow
 - Keras
 - Scikit-learn

Additional programming resources

Python is a great place to start for learning programming. It's intuitive and open source, meaning that anyone can use it for free. This also means that a lot of people around the are constantly adding new features to it. According to the TIOBE index, which measures how popular different programming languages are, Python is the #3 most used programming language worldwide as of March 2021 (https://www.tiobe.com/tiobe-index/).

For installation, I would recommend downloading "Anaconda", it's a one stop shop program that contains Python and few other programming languages. Once you download "Anaconda", I would recommend coding Python within "Spyder" or "Jupyter", both are Python IDEs included within Anaconda.

Keep in mind that the current release of Python is 3.9.2. So make sure any tutorials you utilize are teaching to Python 3 when possible.

Link to Anaconda download: https://www.anaconda.com/products/individual

Introduction:

- CodeAcademy: This is where I started when I was learning programming. When I used it last, you could perform programming exercises entirely from within your web browser, so you didn't need to download anything. Very useful and quick to get into. Link: https://www.codecademy.com/learn/learn-python-3
- Khan Academy: a multifaceted website with a bunch of different tutorials for a wide range of science and math topics. A lot of their videos are on YouTube. Here's a Python tutorial on YouTube: https://www.voutube.com/watch?v=husPzLE6sZc&list=PLJR1V NHIKrCkswPMULzQFHpYa57ZFGbs.

General Classes/Certificates:

The Code Academy link from earlier in this email has a certificate.

- Coursera.org has a lot of good professional certificates. I've participated in a few of their classes. Depending on the professor teaching the class, the quality will vary. https://www.coursera.org/specializations/python?utm_source=gg&utm_medium=sem&utm_campaign=06-PythonforEverybody-US&utm_content=06-PvthonforEvervbodv-
 - US&campaignid=300366907&adgroupid=34186056677&device=c&keyword=coursera%20python&matchtype=e&network=g&devicemodel=&adpostion=&creativeid=2 73185019772&hide mobile promo&gclid=EAIaIQobChMIzrDv8a M7wIVg5uGCh1b6QnnEAAYASAAEgKIQfD BwE
- General blog post describing the different Python certificates available:

https://www.dataguest.io/blog/python-certification/

They make a good general point that a certificate is only as good as the portfolio of programs you have written. I personally think a certificate is nice, as long as you also have a list of projects you've worked on to supplement the certificate.

Reference Materials:

- StackOverflow: An incredibly helpful website. It's like the Google of programming questions. Whenever I don't know how to do something in Python, I go to this website and type in my question. Other users might've had the same question and other programmers provide their answers and insights for free. Here's an example of what a question and answer on this website looks like (https://stackoverflow.com/questions/27885397/how-do-i-install-a-python-package-with-a-whl-file/27909082#27909082). Just type in the search bar at the top if you have a question.
- List of good libraries to use within Python: https://www.dataguest.io/blog/15-python-libraries-for-data-science/

# 02/28/2022: Writing a webscrape script that takes relevant data out of a website and populates a dataframe	# Import one pure plotting library (will help with making nice looking legends at the end of this script):	discussion if you still need more	# Note to self, what if there were multiple wikitables per url? Would we have to iterate then? Probably worth looking into later	plotting functions in Python require numerical values to be actual	e ### PART 3: GEOPANDAS TIME ###	wiki_df_v2.rename(columns = {'States and Washington, D.C.' : 'NAME'})	USA = USA_shp
# From there, use GeoPandas to plot the scraped data up as a choropleth map.		# https://gis.stackexchange.com/que stions/330840/error-installing- geopandas	# Pandas has some built in functions that can read HTML, use these to make our DF:	# If we try to plot a string, even though that string might be made up of numerical characters, we will get errors.	#You will need to download a shapefile of state polygons, I used the US census website below to	# Now, we are going to do a merge (specifically, a left join) on the two data sets: the geodataframe named "USA shp" & the regular Pandas dataframe named	# pick a year of your choice from this list: 2019,2018,2017,2016,2015,2014,20 13,2012,2011, or 2010. # I graduated from TU in 2017, so
# Checking some background environment stuff: import sys	# I had a LOT of trouble getting the 'geopandas' library installed properly using pip I would recommend downloading the raw "wheel" files for	### PART 1: LOAD THE TABLE FROM THE WEB ###	wiki_df = pd.read_html(str(usa_table)) # convert list to dataframe below (because dataframes have more functionality than lists):	# Said another way: unless the numerals are actual numbers, we will see errors returned by our script whenever we try to run it.	find a good one:	"wiki_df_v2": # NOTE! The Geodataframe should always be the left side of the join, otherwise we will remove the geometry information which is	let's pick that year.
sys.executable import platform platform.architecture()	# the prerequisite "GDAL" and "Fiona" packages from these links (download only the package that matches your system's architecture (32 or 64 bit) and the corresponding Python version):	# Ping the URL we want to extract a wiki table from: URL = "https://en.wikipedia.org/wiki/List of U.S. states and territories by	wiki_df = pd.DataFrame(wiki_df[0]	# We're going to loop through a lis of only the column names we want to remove dollar signs and comma in the numbers from.	t t	most crucial for Python to be able # to draw our map automatically for us.	# !!! Now, Make a map plot of our data !!! #
# Import most critical libraries for this script:	# https://www.lfd.uci.edu/~gohlke/p ythonlibs/#gdal # https://www.lfd.uci.edu/~gohlke/p	income" # random testing next two lines:	### PART 2: REFORMAT THE DATAFRAME SLIGHTLY ### # Looking at our fresh and brand new dataframe, we can see there	# Make a list of the matching column names, this will be used fo column selection in the target column_list = list(wiki_df_v2.columns)	USA_shp = 'gpd_read_file('G:\TWitham\DataRo le\Testing_Python\cb_2018_us_sta te_5m.shp')	# if you would like to read further about attribute joins in geopandas, I recommend reading their documentation here: https://geopandas.org/en/stable/d ocs/user_guide/mergingdata.html	# USA.plot(column = vear of choice):
# Import HTML libraries:	ythonlibs/#fiona	# table_class = "wikitable sortable iquery-tablesorter"	might be some issues with plotting the data on a map.	# Now, loop through the first	# Note that the names of Washington D.C. differ between this shapefile we just loaded in and		# Start by adding a legend using matplotlib:
import requests from bs4 import BeautifulSoup	# Once these are downloaded, type the following in the command line below in Spyder or your Python IDE	2	# First of all, there are some columns we are not interested in for the purposes of this script.	through last items in the target_column_list:	# the equivalent location within the wiki_df_v2. Rename "Washington DC" to "District of Columbia"	USA_shp = USA_shp.merge(wiki_df_v2, on = 'NAME')	# Add a legend:
# Import REGEX library if needed (can stay commented out):	INSTALLED BEFORE FIONA!):	# Save the HTML response: response = requests.get(URL)	# Namely, we don't care about the 'State Rank' and 'Growth Rate' columns, so let's drop them:	list(range(1,(len(target_column_list)),1)):		# Great, the merge above worked	fig, ax = plt.subplots(1,1) USA.plot(column = year of choice,
# import re	# pip install path/to/downloaded/GDALfile.whl			<pre>wiki_df_v2[target_column_list[i] = wiki_df_v2[target_column_list[i]].s tr.replace(',', '')</pre>	wiki_df_v2.at[1,'States and Washington, D.C.'] = 'District of Columbia'	perfectly! We joined the geodataframe (USA_shp) to the regular Wikipedia dataframe	ax = ax, legend = True) # !!! NOTE that the map will appear
# Import dataframe/geospatial dataframe libraries:	# pip install path/to/downloaded/fionafile.whl	# Parse the above raw html table/data response into a BeautifulSoup object (a cleaner way to look at the response):	wiki_df_v2 = wiki_df.drop(columns = ['State Rank','Average annual growth rate (current dollars) in 2010-2019,\xa0%'], axis = 1)		# ^ Little data cleanups like this are often necessary to do when getting data ready for actual analysis.	# All that's left now is to make a	in a separate window down in your taskbar. Click on the icon in your taskbar to view the map. # Make sure to zoom in on the map
import pandas as pd	# Once those are installed, then type the following in the command line and you should be good to go:	BeautifulSoup(response.text.'html.	# Cool, now that we've narrowed	wiki_df_v2[target_column_list[i]] # We also want to rename the	choropleth map of each state, colored by the average household income.	using the "magnifying glass" button on the top left if you want to zoom in on the continental USA.
import geopandas as gpd from shapely.geometry import Point, Polygon	# pip install geopandas		our data frame down to the	wiki_df_v2[target_column_list[i]].a stype(int)	"States" column in wiki_df_v2 to "NAME" so the left join with USA_shp will be as seamless as possible:	# Make a copy of USA shp with a	
		usa_table = soupy.find('table', {'class':"wikitable"})	# We convert from string (str) to		wiki_df_v2 =	shorter name (will make for easier to read code when plotting):	

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# Checking some background environment stuff:
       import sys
       sys.executable
       import platform
      platform.architecture()
       # Import most critical libraries for this script:
       # Import HTML libraries:
      import requests
      from bs4 import BeautifulSoup
      # Import REGEX library if needed (can stay commented out):
       # import re
      # Import dataframe/geospatial dataframe libraries:
       import pandas as pd
      import geopandas as gpd
       from shapely geometry import Point, Polygon
      # Import one pure plotting library (will help with making nice looking legends at the end of this script):
       import matplotlib.pyplot as plt
       # IMPORTANT NOTE to future users:
      # I had a LOT of trouble getting the 'geopandas' library installed properly using pip... I would recommend downloading the raw "wheel" files for
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      # the prerequisite "GDAL" and "Fiona" packages from these links (download only the package that matches your system's architecture (32 or 64 bit) and the corresponding Python version)
          # https://www.lfd.uci.edu/~gohlke/pythonlibs/#gdal
          # https://www.lfd.uci.edu/~gohlke/pythonlibs/#fiona
      # Once these are downloaded, type the following in the command line below in Spyder or your Python IDE of choice (MAKE SURE GDAL IS INSTALLED BEFORE FIONA!):
           # pip install path/to/downloaded/GDALfile.whl
          # pip install path/to/downloaded/fionafile.whl
      # Once those are installed, then type the following in the command line and you should be good to go:
          # pip install geopandas
      # Refer to this GIS StackExchange discussion if you still need more help:
           # https://gis.stackexchange.com/questions/330840/error-installing-geopandas
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# https://gis.stackexchange.com/questions/330840/error-installing-geopandas
# # # --- PART 1: LOAD THE TABLE FROM THE WEB --- # # #
# Ping the URL we want to extract a wiki table from:
URL = "https://en.wikipedia.org/wiki/List of U.S. states and territories by income"
# random testing next two lines:
# page = requests.get(URL)
# table class = "wikitable sortable jquery-tablesorter"
# Save the HTML response:
response = requests.get(URL)
# Parse the above raw html table/data response into a BeautifulSoup object (a cleaner way to look at the response):
soupy = BeautifulSoup(response.text, 'html.parser')
# find the sub-object in that HTML response that is of class 'wikitable':
usa table = soupy.find('table', {'class':"wikitable"})
# Note to self, what if there were multiple wikitables per url? Would we have to iterate then? Probably... worth looking into later.
# Pandas has some built in functions that can read HTML, use these to make our DF:
wiki df = pd.read html(str(usa table))
# convert list to dataframe below (because dataframes have more functionality than lists):
wiki df = pd.DataFrame(wiki df[0])
# # # --- PART 2: REFORMAT THE DATAFRAME SLIGHTLY --- # # #
# Looking at our fresh and brand new dataframe, we can see there might be some issues with plotting the data on a map.
# First of all, there are some columns we are not interested in for the purposes of this script.
# Namely, we don't care about the 'State Rank' and 'Growth Rate' columns, so let's drop them:
wiki_df_v2 = wiki_df.drop(columns = ['State Rank','Average annual growth rate (current dollars) in 2010-2019, \xa0%'], axis = 1)
# Cool, now that we've narrowed our data frame down to the columns we want, let's get rid of the dollar signs and convert all the numbers from str to int.
# We convert from string (str) to integer (int) because a lot of the plotting functions in Python require numerical values to be actual integers or float (decimal) numbers.
# If we try to plot a string, even though that string might be made up of numerical characters, we will get errors.
# Said another way: unless the numerals are actual numbers, we will see errors returned by our script whenever we try to run it.
# We're going to loop through a list of only the column names we want to remove dollar signs and commas in the numbers from.
# Make a list of the matching column names, this will be used for column selection in the
target column list = list(wiki df v2.columns)
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# Now, loop through the first through last items in the target column list :
for i in list(range(1,(len(target column list)),1)):
    wiki_df_v2[target_column_list[i]] = wiki_df_v2[target_column_list[i]].str.replace(',', '')
   wiki df v2[target column list[i]] = wiki df v2[target column list[i]].str.replace(r'$', '')
    wiki df v2[target column list[i]] = wiki df v2[target column list[i]].astype(int)
# # # --- PART 3: GEOPANDAS TIME --- # # #
# You will need to download a shapefile of state polygons, I used the US census website below to find a good one:
# https://www.census.gov/geographies/mapping-files/time-series/geo/carto-boundary-file.html
USA shp = gpd.read file('G:\TWitham\DataRole\Testing Python\cb 2018 us state 5m.shp')
# Note that the names of Washington D.C. differ between this shapefile we just loaded in and
# the equivalent location within the wiki df v2. Rename "Washington DC" to "District of Columbia"
wiki df v2.at[1,'States and Washington, D.C.'] = 'District of Columbia'
# ^ Little data cleanups like this are often necessary to do when getting data ready for actual analysis.
# We also want to rename the "States" column in wiki df v2 to "NAME" so the left join with USA shp will be as seamless as possible:
wiki df v2 = wiki df v2.rename(columns = {'States and Washington, D.C.' : 'NAME'})
# Now, we are going to do a merge (specifically, a left join) on the two data sets: the geodataframe named "USA shp" & the regular Pandas dataframe named "wiki df v2":
# NOTE! The Geodataframe should always be the left side of the join, otherwise we will remove the geometry information which is most crucial for Python to be able
# to draw our map automatically for us.
# if you would like to read further about attribute joins in geopandas, I recommend reading their documentation here: https://geopandas.org/en/stable/docs/user guide/mergingdata.html
USA shp = USA shp.merge(wiki df v2, on = 'NAME')
# Great, the merge above worked perfectly! We joined the geodataframe (USA shp) to the regular Wikipedia dataframe (wiki df v2).
# All that's left now is to make a choropleth map of each state, colored by the average household income.
# Make a copy of USA shp with a shorter name (will make for easier to read code when plotting):
USA = USA shp
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# the equivalent location within the wiki df v2. Rename "Washington DC" to "District of Columbia"
wiki df v2.at[1, 'States and Washington, D.C.'] = 'District of Columbia'
# ^ Little data cleanups like this are often necessary to do when getting data ready for actual analysis.
# We also want to rename the "States" column in wiki df v2 to "NAME" so the left join with USA shp will be as seamless as possible:
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USA shp = USA shp.merge(wiki df v2, on = 'NAME')
# Great, the merge above worked perfectly! We joined the geodataframe (USA shp) to the regular Wikipedia dataframe (wiki df v2).
# All that's left now is to make a choropleth map of each state, colored by the average household income.
# Make a copy of USA shp with a shorter name (will make for easier to read code when plotting):
USA = USA shp
# pick a year of your choice from this list: 2019,2018,2017,2016,2015,2014,2013,2012,2011, or 2010.
# I graduated from TU in 2017, so let's pick that year.
year of choice = '2017'
# !!! Now, Make a map plot of our data !!! #
# default, basic plot is below, but let's make it prettier.
# USA.plot(column = year of choice);
# Start by adding a legend using matplotlib:
# Add a legend:
fig, ax = plt.subplots(1,1)
USA.plot(column = year of choice, ax = ax, legend = True)
# !!! NOTE that the map will appear in a separate window down in your taskbar. Click on the icon in your taskbar to view the map.
# Make sure to zoom in on the map using the "magnifying glass" button on the top left if you want to zoom in on the continental USA.
```