Web Scraping Tables using Pandas

Estimated Effort: 5 mins

The Pandas library in Python contains a function read_html() that can be used to extract tabular information from any web page.

Consider the following example:

Let us assume we want to extract the list of the largest banks in the world by market capitalization, from the following link:

- 1. URL = 'https://en.wikipedia.org/wiki/List of largest banks'

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We may use pandas.read_html() function in python to extract all the tables in the web page directly.

A snapshot of the webpage is shown below.



List of largest banks

Contents [hide] (Top) By market capitalization By total assets Banks by country or territory See also References

Article Talk

From Wikipedia, the free encyclopedia The following are lists of the largest banks in the world, as measured by market capitalization

By market capitalization [edit]

The list is based on Forbes.com's ranking as of August 2023 based on an analysis of the bar the global economy.[1]

Rank 	Bank name \$	Market cap [hide] (US\$ billion)		
1	JPMorgan Chase	419.25		
2	Bank of America	231.52		
3	Industrial and Commercial Bank of China	194.56		
4	Agricultural Bank of China	160.68		
5	HDFC Bank	157.91		
6	Wells Fargo	155.87		
7	HSBC Holdings PLC	148.90		
8	Morgan Stanley	140.83		
9	China Construction Bank	139.82		
10	Bank of China	136.81		

We can see that the required table is the first one in the web page.

Note: This is a live web page and it may get updated over time. The image shown above has been captured in November 2023. The process of data extraction remains the same.

We may execute the following lines of code to extract the required table from the web page.

- 2. 2 3. 3 4. 4

- import pandas as pd
 URL = 'https://en.wikipedia.org/wiki/List_of_largest_banks'
 tables = pd.read_html(URL)
 df = tables[0]
 print(df)

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1 of 3 28/09/24, 15:34 This will extract the required table as a dataframe of. The output of the print statement would look as shown below.

	Rank	Bank name	Market cap(US\$ billion)
0	1	JPMorgan Chase	419.25
1	2	Bank of America	231.52
2	3	Industrial and Commercial Bank of China	194.56
3	4	Agricultural Bank of China	160.68
4	5	HDFC Bank	157.91
5	6	Wells Fargo	155.87
6	7	HSBC Holdings PLC	148.90
7	8	Morgan Stanley	140.83
8	9	China Construction Bank	139.82
9	10	Bank of China	136.81

Although convenient, this method comes with its own set of limitations.

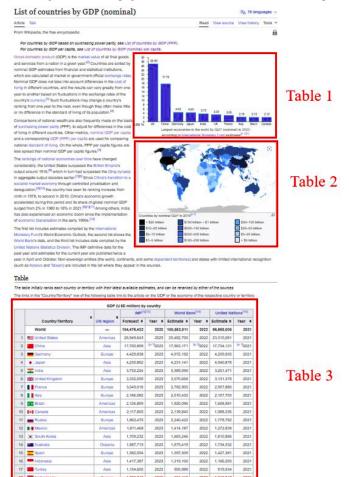
Firstly, web pages may have content saved in them as tables but they may not appear as tables on the web page.

For instance, consider the following URL showing the list of countries by GDP (nominal).

- 1. 1
- $1. \ \ URL = \ \ 'https://en.wikipedia.org/wiki/List_of_countries_by_GDP_(nominal)'$

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The images on the web page are also saved in tabular format. A snapshot of the web page is shared below.



Secondly, the contents of the tables in the web pages may contain elements such as hyperlink text and other denoters, which are also scraped directly using the pandas method. This may lead to a requirement of further cleaning of data.

A closer look at table 3 in the image shown above indicates that there are many hyperlink texts which are also going to be treated as information by the pandas function.

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GDP (USD million) by country									
	Country/Territory \$	UN .	IMF ^{[1][13]}		World Bank [14]		United Nations ^[15]		
		region	Forecast \$	Year ♦	Estimate \$	Year ¢	Estimate \$	Year 4	
	World	-	104,476,432	2023	100,562,011	2022	96,698,005	202	
1	United States	Americas	26,949,643	2023	25,462,700	2022	23,315,081	202	
2	China	Asia	17,700,899	[n 1]2023	17,963,171	[n 3] ₂ 022	17,734,131	[n 1] ₂ 02	
3	Germany	Europe	4,429,838	2023	4,072,192	2022	4,259,935	202	
4	Japan	Asia	4,230,862	2023	4,231,141	2022	4,940,878	202	
5	India India	Asia	3,732,224	2023	3,385,090	2022	3,201,471	202	
6	United Kingdom	Europe	3,332,059	2023	3,070,668	2022	3,131,378	202	
7	France	Europe	3,049,016	2023	2,782,905	2022	2,957,880	202	
8	■ Italy	Europe	2,186,082	2023	2,010,432	2022	2,107,703	202	
9	◆ Brazil	Americas	2,126,809	2023	1,920,096	2022	1,608,981	202	
10	■● Canada	Americas	2,117,805	2023	2,139,840	2022	1,988,336	202	
11	Russia	Europe	1,862,470	2023	2,240,422	2022	1,778,782	202	
12	■•■ Mexico	Americas	1,811,468	2023	1,414,187	2022	1,272,839	202	
13	South Korea	Asia	1,709,232	2023	1,665,246	2022	1,810,966	202	
14	Australia	Oceania	1,687,713	2023	1,675,419	2022	1,734,532	202	
15	Spain	Europe	1,582,054	2023	1,397,509	2022	1,427,381	202	

We can extract the table using the code shown below.

- 1. 1 2. 2 3. 3 4. 4 5. 5

- import pandas as pd
 URL = 'https://en.wikipedia.org/wiki/List_of_countries_by_GDP_(nominal)'
 tables = pd.read_html(URL)
 df = tables(2) # the required table will have index 2
 print(df)

The output of the print statement is shown below.

	Country/Territory	UN region	IMF[1][13]		World Bank [14]		United Nations[15]	
	Country/Territory	UN region	Forecast	Year	Estimate	Year	Estimate	Year
Θ	World		104476432	2023	100562011	2022	96698005	2021
1	United States	Americas	26949643	2023	25462700	2022	23315081	2021
2	China	Asia	17700899	[n 1]2023	17963171	[n 3]2022	17734131	[n 1]2021
3	Germany	Europe	4429838	2023	4072192	2022	4259935	2021
4	Japan	Asia	4230862	2023	4231141	2022	4940878	2021
209	Palau	Oceania	267	2023	-	-	218	2021
210	Kiribati	Oceania	246	2023	223	2022	227	2021
211	Nauru	Oceania	150	2023	151	2022	155	2021
212	Montserrat	Americas	-	-	7	-	72	2021
213	Tuvalu	Oceania	63	2023	60	2022	60	2021

Note that the hyperlink texts have also been retained in the code output.

It is further prudent to point out, that this method exclusively operates only on tabular data extraction. BeautifulSoup library still remains the default method of extracting any kind of information from web pages.

Author(s)

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