WebScrape_Cosmetics

September 24, 2023

1 Web Scraping using BeautifulSoup

1.1 Import Libraries

```
[]: from bs4 import BeautifulSoup import requests import pandas as pd
```

1.2 Read webpage

Top 100 buyers of beauty and related skincare imports. Ranked in descending order.

2020-1: change in international sales from 2020 to 2021.

%Total: percentage of the world's total purchases of imported cosmetics in 2021

US\$: Annual sales amount

```
[ ]: url = 'https://importedconsumerproducts.com/beauty-cosmetics/'
    page = requests.get(url)
    page
```

[]: <Response [200]>

1.3 Load webpage content

```
[]: soup = BeautifulSoup(page.content, 'html.parser')

[]: table = soup.find_all("table", class_ = "tablepress tablepress-id-39")
    table

[]: [
        <thead>

              RankImporterUs$%Total2020-1
```

```
</thead>
1China<td
class="column-3">$20,287,362,00027.2%td
class="column-5">+17.1%
2Hong Kong<td
class="column-3">$7,280,376,0009.8%td
class="column-5">+0.6%
3United States<td
class="column-3">$5,557,871,0007.5%td
class="column-5">+23.9%
4Macao<td
class="column-3">\$3,221,960,000\4.3\%<td
class="column-5">+36.6%
5Singapore<td
class="column-3">$3,087,914,0004.1%td
class="column-5">+6.7%
6Germany<td</pre>
class="column-3">$2,629,789,0003.5%td
class="column-5">+17%
7United Kingdom<td
class="column-3">$2,082,046,0002.8%td><td
class="column-5">+2.6%
8France<td
class="column-3">$1,906,145,0002.6%td
class="column-5">+11.4%
9Canada<td
class = "column - 3" > \$1,575,771,000   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%   2.1%  
class="column-5">+11.9%
```

```
10Japan<td
class="column-3">$1,466,193,0002.0%<td
class="column-5">+5.6%
11Belgium<td
class="column-3">$1,395,477,0001.9%td
class="column-5">+7.8%
12Netherlands<td
class="column-3">$1,326,661,0001.8%*td><td
class="column-5">+33.5%
13Russiatd
class="column-3">$1,291,515,0001.7%td
class="column-5">+27.2%
14Poland<td
class="column-3">$1,171,673,0001.6%td
class="column-5">+14.8%
15Italy<td</pre>
class="column-3">$1,159,766,0001.6%td
class="column-5">+12%
16Spain<td
class = "column - 3" > \$1,149,378,000   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%   1.5\%  
class="column-5">+18.6%
17UAE<td
class="column-3">$966,135,0001.3%<td
class="column-5">+21.2%
18South Korea<td
class="column-3">$910,848,0001.2%<td
class="column-5">+3.1%
```

```
19Czech Republic<td
class="column-3">$906,189,0001.2%<td
class="column-5">+10.4%
20Australia<td
class="column-3">$883,136,0001.2%<td
class="column-5">+12%
21Taiwan<td
class="column-3">$867,511,0001.2%<td
class="column-5">+1.9%
22Mexico<td
class="column-3">$622,153,0000.8%<td
class="column-5">+28.3%
23Saudi Arabia<td
class="column-3">$620,513,0000.8%<td
class="column-5">+4%
24Switzerland<td</td>
class="column-3">$613,775,0000.8%<td
class="column-5">+15.5%
25Thailand<td
class="column-3">$526,418,0000.7%<td
class="column-5">-14.4%
26Sweden<td
class="column-3">$524,211,0000.7%<td
class="column-5">+19.9%
27Austria<td
class="column-3">$507,869,0000.7%<td
class="column-5">+17.5%
</t.r>
28Ireland<td
class="column-3">$450,083,0000.6%<td
```

```
class="column-5">+49.6%
29Denmark<td
class="column-3">$409,357,0000.5%<td
class="column-5">+29.6%
30Norway<td
class="column-3">$393,447,0000.5%<td
class="column-5">+30.6%
31Malaysia<td
class="column-3">$384,600,0000.5%<td
class="column-5">+10.1%
32Indonesia<td
class="column-3">$309,480,0000.4%<td
class="column-5">+20.4%
33Chile<td
class="column-3">$307,677,0000.4%<td
class="column-5">+71.5%
34Kuwait<td
class="column-3">$303,105,0000.4%<td
class="column-5">+18.9%
35Turkey<td
class="column-3">$292,394,0000.4%<td
class="column-5">+14.4%
36Romania<td
class="column-3">$269,253,0000.4%<td
class="column-5">+16.8%
37Ukraine<td
class="column-3">$265,571,0000.4%<td
class="column-5">+25.6%
```

```
38India<td
class="column-3">$262,646,0000.4%<td
class="column-5">+90%
39Portugal<td
class="column-3">$261,975,0000.4%<td
class="column-5">+15.6%
40Hungary<td
class="column-3">$244,499,0000.3%<td
class="column-5">+6.2%
41Vietnam<td
class="column-3">$236,350,0000.3%<td
class="column-5">+34.6%
42New Zealand<td</pre>
class="column-3">$231,220,0000.3%<td
class="column-5">+18.9%
43Slovakia<td
class="column-3">$228,795,0000.3%<td
class="column-5">+26.2%
44Lithuania<td
class="column-3">$228,762,0000.3%<td
class="column-5">+16.9%
45Greece<td
class="column-3">$222,781,0000.3%<td
class="column-5">+4.1%
46South Africa<td
class="column-3">$193,885,0000.3%<td
class="column-5">+25.3%
47Croatia<td
```

```
class="column-3">$177,886,0000.2%<td
class="column-5">+16.3%
48Brazil<td
class="column-3">$172,449,0000.2%<td
class="column-5">+5.7%
49Israel<td
class="column-3">$163,768,0000.2%<td
class="column-5">+31.5%
50Colombia<td
class = "column - 3" > $156,756,000   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%  
class="column-5">+26.4%
<span class="ezoic-autoinsert-video ezoic-</pre>
longer_content"></span><!-- ezoic_video_placeholder-</pre>
longer_content-640x360-999994-clearholder --><!-- ezoic_video_placeholder-
longer_content-640x360-999994-nonexxxclearxxxblock --><!--
ezoic video placeholder-longer content-426x240-999994-clearholder --><!--
ezoic_video_placeholder-longer_content-426x240-999994-nonexxxclearxxxblock
--><!-- ezoic video placeholder-longer content-384x216-999994-clearholder
--><!-- ezoic video placeholder-
longer content-384x216-999994-nonexxxclearxxxblock -->
51Finland<td
class="column-3">$150,173,0000.2%<td
class="column-5">+9.5%
52Kazakhstan<td
class = "column - 3" > $149,140,000   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%   0.2%  < < td class = "column - 4" > 0.2%  < < td class = "column - 4" > 0.2%  < < td class = "column - 4" > 0.2%  < < td class = "column - 4" > 0.2%  < < td class = "column - 4" > 0.2%  < < td class = "column - 4" > 0.2%  
class="column-5">+32.8%
53Iraq<td
class="column-3">$146,273,0000.2%<td
class="column-5">+20%
54Philippines<td
class="column-3">$139,259,0000.2%<td
class="column-5">+7.2%
```

```
55Qatar<td
class="column-3">$139,229,0000.2%<td
class="column-5">+18.4%
56Peru<td
class="column-3">$131,957,0000.2%<td
class="column-5">+40.9%
57Argentina<td
class="column-3">$113,013,0000.2%<td
class="column-5">+15.2%
58Slovenia<td
class="column-3">$109,651,0000.1%<td
class="column-5">+18.2%
59Morocco<td
class="column-3">$108,699,0000.1%<td
class="column-5">+24.3%
60Bulgaria<td
class="column-3">$105,585,0000.1%<td
class="column-5">+31.5%
61Panamatd
class="column-3">$101,043,0000.1%<td
class="column-5">+11.6%
62Oman<td
class="column-3">$92,874,0000.1%<td
class="column-5">+11.5%
63Ecuador<td
class="column-3">$88,525,0000.1%<td
class="column-5">+28.3%
</t.r>
64Belarus<td
class="column-3">$88,364,0000.1%<td
```

```
class="column-5">+10.1%
65Luxembourg<td
class="column-3">$81,925,0000.1%<td
class="column-5">+6.8%
66Serbia<td
class="column-3">$74,904,0000.1%<td
class="column-5">+10.5%
67Latvia<td
class = "column - 3" > \$70,928,000   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%   0.1\%  
class="column-5">+3.1%
68Estonia<td
class="column-3">$70,927,0000.1%<td
class="column-5">-0.9%
69Guatemala<td
class="column-3">$64,953,0000.1%<td
class="column-5">+19%
70Cyprus<td
class="column-3">$61,739,0000.1%<td
class="column-5">+23.3%
71Costa Rica<td
class="column-3">$58,507,0000.1%<td
class="column-5">+12.8%
72Jordan<td
class="column-3">$57,801,0000.1%<td
class="column-5">+18.1%
73Bosnia/Herzegovina<td
class="column-3">$50,201,0000.1%<td
class="column-5">+26.2%
```

```
74Algeria<td
class="column-3">$48,641,0000.1%<td
class="column-5">-18.9%
75Egypt<td
class="column-3">$48,275,0000.1%<td
class="column-5">+30.9%
76Dominican Republic<td
class="column-3">$47,039,0000.1%<td
class="column-5">+14.1%
77Bangladesh<td
class="column-3">$44,628,0000.1%<td
class="column-5">+43.5%
78Bahrain<td
class="column-3">$44,586,0000.1%<td
class="column-5">+15.5%
79El Salvador<td
class="column-3">$44,416,0000.06%<td
class="column-5">+42.6%
80Nepal<td
class="column-3">$43,439,0000.06%<td
class="column-5">+67.8%
81Libya<td
class="column-3">$40,699,0000.05%<td
class="column-5">+41%
82Iran<td
class="column-3">$40,565,0000.05%<td
class="column-5">+183.8%
83Georgia<td
```

```
class="column-3">$40,256,0000.05%<td
class="column-5">+27%
84Azerbaijan<td</pre>
class="column-3">$36,028,0000.05%<td
class="column-5">+37.2%
85Andorra<td
class="column-3">\$35,707,0000.05%<td
class="column-5">+48%
86Pakistan<td
class="column-3">$31,191,0000.04%<td
class="column-5">+60.3%
87Namibia<td
class="column-3">$31,189,0000.04%<td
class="column-5">+6.3%
88Botswana<td
class="column-3">$30,939,0000.04%<td
class="column-5">+11.8%
89Lebanon<td
class="column-3">$30,584,0000.04%<td
class="column-5">+16.3%
90Iceland<td
class="column-3">$29,730,0000.04%<td
class="column-5">+26%
91Myanmar<td
class="column-3">$29,071,0000.04%<td
class="column-5">-28.1%
92Moldova<td
class="column-3">$28,335,0000.04%<td
class="column-5">+36.5%
```

```
93Tunisia<td
class="column-3">$28,070,0000.04%<td
class="column-5">-14.1%
94Bolivia<td
class="column-3">$27,549,0000.04%<td
class="column-5">+16%
95Malta<td
class="column-3">$25,571,0000.03%<td
class="column-5">+93%
96Mongolia<td
class = "column - 3" > $24,861,000   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%   0.03\%  
class="column-5">+46.2%
97Honduras<td
class="column-3">$23,818,0000.03%<td
class="column-5">+37%
98Albania<td
class="column-3">$23,726,0000.03%<td
class="column-5">+26%
99North Macedonia<td
class="column-3">$22,232,0000.03%<td
class="column-5">+31%
<span class="ezoic-autoinsert-video ezoic-</pre>
longest_content"></span><!-- ezoic_video_placeholder-
longest_content-640x360-999993-clearholder --><!-- ezoic_video_placeholder-
longest content-640x360-999993-nonexxxclearxxxblock --><!--
ezoic_video_placeholder-longest_content-426x240-999993-clearholder --><!--
ezoic video placeholder-longest content-426x240-999993-nonexxxclearxxxblock
--><!-- ezoic_video_placeholder-longest_content-384x216-999993-clearholder
--><!-- ezoic video placeholder-
longest_content-384x216-999993-nonexxxclearxxxblock -->
100Paraguay<td
class="column-3">$20,871,0000.03%<td
```

```
]
   1.4 Read column headings
[]: x =soup.find_all("th")
    Х
[]: [Rank,
     Importer,
     US$,
     %Total,
     2020-1]
[]: column_nm = [title.text.strip() for title in x]
    column_nm
[]: ['Rank', 'Importer', 'US$', '%Total', '2020-1']
   1.5 Create dataframe and load column names
[]: skincare=[]
    skincare = pd.DataFrame(columns=column_nm)
    skincare
[]: Empty DataFrame
    Columns: [Rank, Importer, US$, %Total, 2020-1]
    Index: []
[]: skincare.info()
   <class 'pandas.core.frame.DataFrame'>
   Index: 0 entries
   Data columns (total 5 columns):
    #
       Column
               Non-Null Count Dtype
       ____
                -----
                            ----
       Rank
                0 non-null
    0
                             object
    1
       Importer 0 non-null
                             object
    2
       US$
                0 non-null
                             object
       %Total
                0 non-null
                             object
               0 non-null
       2020-1
                             object
   dtypes: object(5)
   memory usage: 0.0+ bytes
[1]: #skincare.drop(columns='', inplace=True)
```

class="column-5">+15.9%

1.6 Process HTML table to populate dataframe with rows of data

1.7 Display populated dataframe

[]: skincare

[]:		Rank	Importer	US\$	%Total	2020-1
	0	1	China	\$20,287,362,000	27.2%	+17.1%
	1	2	Hong Kong	\$7,280,376,000	9.8%	+0.6%
	2	3	United States	\$5,557,871,000	7.5%	+23.9%
	3	4	Macao	\$3,221,960,000	4.3%	+36.6%
	4	5	Singapore	\$3,087,914,000	4.1%	+6.7%
		•••	•••		•••	
	95	96	Mongolia	\$24,861,000	0.03%	+46.2%
	96	97	Honduras	\$23,818,000	0.03%	+37%
	97	98	Albania	\$23,726,000	0.03%	+26%
	98	99	North Macedonia	\$22,232,000	0.03%	+31%
	99	100	Paraguay	\$20,871,000	0.03%	+15.9%
	_		a w E columnal			

[100 rows x 5 columns]

1.8 Data Cleaning

```
[]: skincare.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 100 entries, 0 to 99
Data columns (total 5 columns):
    # Column Non-Null Count Dtype
```

```
100 non-null
                              object
0
   Rank
1
   Importer 100 non-null
                              object
2
   US$
              100 non-null
                              object
3
   %Total
              100 non-null
                              object
   2020-1
              100 non-null
                              object
```

dtypes: object(5)
memory usage: 4.7+ KB

1.8.1 Remove % from '%Total' column

```
[]: skincare['%Total']=skincare['%Total'].map(lambda x: x.rstrip('%')) skincare
```

[]:		Rank	${\tt Importer}$	US\$	%Total	2020-1
	0	1	China	\$20,287,362,000	27.2	+17.1%
	1	2	Hong Kong	\$7,280,376,000	9.8	+0.6%
	2	3	United States	\$5,557,871,000	7.5	+23.9%
	3	4	Macao	\$3,221,960,000	4.3	+36.6%
	4	5	Singapore	\$3,087,914,000	4.1	+6.7%
		•••	•••		•••	
	95	96	Mongolia	\$24,861,000	0.03	+46.2%
	96	97	Honduras	\$23,818,000	0.03	+37%
	97	98	Albania	\$23,726,000	0.03	+26%
	98	99	North Macedonia	\$22,232,000	0.03	+31%
	99	100	Paraguay	\$20,871,000	0.03	+15.9%

[100 rows x 5 columns]

1.8.2 Remove dollar sign from 'US\$' column

```
[]: skincare['US$']=skincare['US$'].map(lambda x: x.lstrip('$')) skincare
```

[]:	Rank Importer		Importer	US\$ %Total		2020-1
	0 1 China		China	20,287,362,000	27.2	+17.1%
	1	2 Hong Kong		7,280,376,000	9.8	+0.6%
	2	3	United States	5,557,871,000	7.5	+23.9%
	3	4	Macao	3,221,960,000	4.3	+36.6%
	4 5 Singapore		3,087,914,000	4.1	+6.7%	
		•••	•••		•••	
	95	96	Mongolia	24,861,000	0.03	+46.2%
	96	97	Honduras	23,818,000	0.03	+37%
	97	98	Albania	23,726,000	0.03	+26%
	98	99	North Macedonia	22,232,000	0.03	+31%

[100 rows x 5 columns]

1.8.3 Remove % from '2020-1' column

```
[]: skincare['2020-1']=skincare['2020-1'].map(lambda x: x.rstrip('%')) skincare
```

[]:	Rank Impor		Importer	US\$	%Total	2020-1
	0 1 China		China	20,287,362,000	27.2	+17.1
	1	2	Hong Kong	7,280,376,000	9.8	+0.6
	2 3 United States		United States	5,557,871,000	7.5	+23.9
	3	4	Macao	3,221,960,000	4.3	+36.6
	4 5 Singapore		Singapore	3,087,914,000	4.1	+6.7
		•••	•••		•••	
	95	96	Mongolia	24,861,000	0.03	+46.2
	96	97	Honduras	23,818,000	0.03	+37
	97	98	Albania	23,726,000	0.03	+26
	98	99	North Macedonia	22,232,000	0.03	+31
	99	100	Paraguay	20,871,000	0.03	+15.9

[100 rows x 5 columns]

[]: skincare.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 100 entries, 0 to 99
Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	Rank	100 non-null	object
1	${\tt Importer}$	100 non-null	object
2	US\$	100 non-null	object
3	%Total	100 non-null	object
4	2020-1	100 non-null	object

dtypes: object(5)
memory usage: 4.7+ KB

[]: skincare2 = skincare.copy(deep=True) skincare2

[]:	Rank	Importer	US\$	%Total	2020-1
0	1	China	20,287,362,000	27.2	+17.1
1	2	Hong Kong	7,280,376,000	9.8	+0.6
2	3	United States	5,557,871,000	7.5	+23.9
3	4	Macao	3,221,960,000	4.3	+36.6
4	5	Singapore	3,087,914,000	4.1	+6.7
•		•••		•••	
95	96	Mongolia	24,861,000	0.03	+46.2
96	97	Honduras	23,818,000	0.03	+37
97	7 98	Albania	23,726,000	0.03	+26

```
20,871,000
       99 100
                                                  0.03 + 15.9
                       Paraguay
       [100 rows x 5 columns]
      1.8.4 Convert data type Object to String or Numeric depending on column
[191]: | skincare2['%Total'] = pd.to_numeric(skincare2['%Total'], downcast='float')
[192]: |
       skincare2['2020-1'] = pd.to_numeric(skincare2['2020-1'], downcast='signed')
[193]: skincare2['US$'] = pd.to_numeric(skincare2['US$'].str.replace('[^0-9.]', ''))
      <ipython-input-193-d463c7e98e4e>:1: FutureWarning: The default value of regex
      will change from True to False in a future version.
        skincare2['US$'] = pd.to_numeric(skincare2['US$'].str.replace('[^0-9.]', ''))
[194]: skincare2.info()
      <class 'pandas.core.frame.DataFrame'>
      Int64Index: 100 entries, 0 to 99
      Data columns (total 5 columns):
                     Non-Null Count Dtype
           Column
       0
           Rank
                     100 non-null
                                      object
       1
           Importer 100 non-null
                                      object
       2
           US$
                     100 non-null
                                      int64
                     100 non-null
       3
           %Total
                                     float32
           2020-1
                     100 non-null
                                     float64
      dtypes: float32(1), float64(1), int64(1), object(2)
      memory usage: 4.3+ KB
[212]: skincare2['Rank'] = pd.to_numeric(skincare2['Rank'])
[203]: | skincare2['Importer'] = skincare2['Importer'].astype("string")
      1.9 Data Visualization
  [2]: import seaborn as sns
       import matplotlib.pyplot as plt
       plt.style.use('ggplot')
[231]: | #sns.set(rc={"figure.figsize":(8,6)})
       sns.histplot(skincare2['%Total'])
```

22,232,000

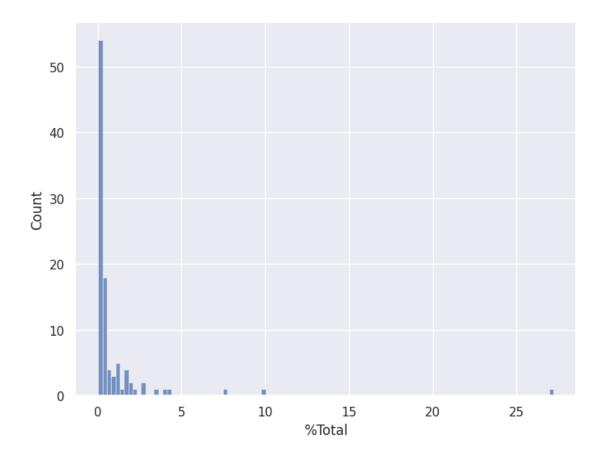
0.03

+31

98

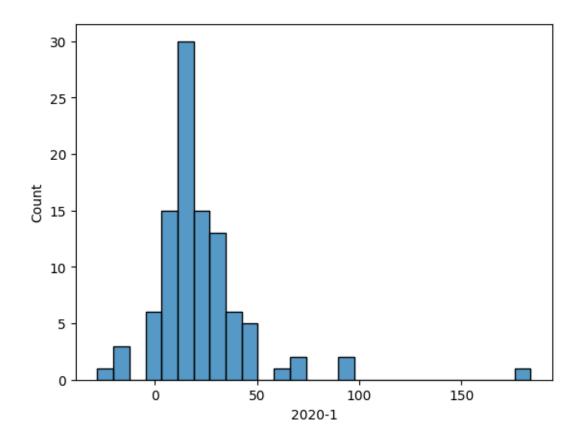
plt.show()

99 North Macedonia



```
[221]: sns.histplot(skincare2['2020-1'])
```

[221]: <Axes: xlabel='2020-1', ylabel='Count'>



1.10 Alternative method: Read HTML table using Pandas

```
[]:
        Rank
                    Importer
                                           US$ %Total
                                                        2020-1
         1.0
                                                27.2%
     0
                       China
                              $20,287,362,000
                                                        +17.1%
     1
         2.0
                               $7,280,376,000
                                                 9.8%
                                                         +0.6%
                  Hong Kong
             United States
                                                 7.5%
     2
         3.0
                               $5,557,871,000
                                                       +23.9%
     3
                               $3,221,960,000
                                                 4.3%
                                                        +36.6%
         4.0
                       Macao
         5.0
                  Singapore
                               $3,087,914,000
                                                 4.1%
                                                         +6.7%
```

[]: df_pandas[0].info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 102 entries, 0 to 101
Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	Rank	100 non-null	float64

1	Importer	100	non-null	object
2	US\$	100	non-null	object
3	%Total	100	non-null	object
4	2020-1	100	non-null	object

dtypes: float64(1), object(4)

memory usage: 4.1+ KB