Lab6

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1 STOR 320: Introduction to Data Science

1.1 Lab 6

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```
[]: from datetime import datetime
from bs4 import BeautifulSoup
from io import StringIO
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import requests
```

[]: %pip install html5lib

```
Requirement already satisfied: html5lib in
```

/Library/Frameworks/Python.framework/Versions/3.12/lib/python3.12/site-packages (1.1)

Requirement already satisfied: six>=1.9 in

/Users/ivynangalia/Library/Python/3.12/lib/python/site-packages (from html5lib) (1.16.0)

Requirement already satisfied: webencodings in

/Users/ivynangalia/Library/Python/3.12/lib/python/site-packages (from html5lib) (0.5.1)

Note: you may need to restart the kernel to use updated packages.

2 Scraping, Merging, and Analyzing Datasets for Countries (25 points)

Background: Many times in data science, your data will be split between many different sources, some of which may be online. In this analysis assignment, we will webscrape country level data from multiple websites, clean the data individually, and merge the data. The website Worldometers contains very interesting country level data that when connected may allow us to learn interesting things about the wonderful world in which we exist.

2.1 0. GDP by Country (7 Points)

Information at Worldometer GDP contains GDP data from 2022 published by the world bank. GDP is the monetary value of goods and services produced within a country over a period of time. On this website, GDP is presented in dollars.

2.1.1 0.0 Scraping the Data

We will walk through webscraping the data from https://www.worldometers.info/gdp/gdp-by-country/ using Pandas into a DataFrame called GDP. You should end up with a new object called GDP which is a DataFrame with 177 observations and 8 variables.

```
[]: URL_GDP = "https://www.worldometers.info/gdp/gdp-by-country/"

# Send a GET request to the URL
response = requests.get(URL_GDP)

# Parse the HTML content
soup = BeautifulSoup(response.content, 'html.parser')

# Find all tables and read into pandas DataFrame
tables = soup.find_all('table')

table_IO = StringIO(str(tables))
GDP = pd.read_html(table_IO, flavor='bs4', header=0)[0] # Read the first table
GDP.shape
```

[]: (177, 8)

```
[]: GDP.head(5)
```

```
[]:
        #
                 Country GDP
                               (nominal, 2022)
                                                        (abbrev.) GDP growth \
     0
        1
           United States
                           $25,462,700,000,000
                                                 $25.463 trillion
                                                                        2.06%
     1
        2
                    China
                           $17,963,200,000,000
                                                 $17.963 trillion
                                                                        2.99%
     2
        3
                            $4,231,140,000,000
                                                  $4.231 trillion
                                                                        1.03%
                    Japan
     3
        4
                 Germany
                            $4,072,190,000,000
                                                  $4.072 trillion
                                                                        1.79%
     4 5
                            $3,385,090,000,000
                                                                        7.00%
                    India
                                                  $3.385 trillion
```

	Population	(2022)	GDP	per	capita	Share	of	World	GDP
0	341534046		\$74,554					25.	32%
1	142	5179569		9	\$12,604			17.	86%
2	12	4997578		5	\$33,850			4.	21%
3	8-	4086227		9	\$48,429			4.	05%
4	142	5423212			\$2,375			3.	37%

2.1.2 0.1 Cleaning the Data (7 points)

Now that we scraped our data into a DataFrame, we need to clean it up. Perform the following tasks:

- 1. Remove the first ('#') and fourth ('GDP (abbrev.)') columns from the DataFrame.
- 2. Rename the columns 'GDP (nominal, 2022)', 'GDP growth', 'Population (2022)', 'GDP per capita', and 'Share of World GDP' to 'GDP', 'Growth', 'Population', 'PerCapita', and 'Share', respectively.
- 3. Remove all dollar signs, percent signs, and commas from 'GDP', 'Growth', 'PerCapita', and 'Share'.
- 4. Update column data type of "Country" to be a string dtype and the remaining columns to be numeric. Hint: use pd.to_numeric
- 5. Rewrite over the original 'GDP' variable with a new variable called 'GDP' that is in trillions of dollars rather than in actual dollars. Rewrite over the original 'Population' variable with a new variable of the same name that is in millions of people rather than in actual people. You are scaling the original variables to change the units without changing the variable names.

Be careful of the formatting and spacing in the original column names! Display the first five rows of the cleaned GDP DataFrame and the dtype info for GDP.

```
[]: # Code Solution Here
     # 1
     GDP = GDP.drop('#', axis=1)
     GDP = GDP.drop('GDP (abbrev.)', axis=1)
     # 2
     GDP = GDP.rename(columns={
         'GDP (nominal, 2022)': 'GDP',
         'GDP growth': 'Growth',
         'Population (2022)': 'Population',
         'GDP per capita': 'PerCapita',
         'Share of World GDP': 'Share'
     })
     # 3
     columns_to_clean = ['GDP', 'Growth', 'PerCapita', 'Share']
     GDP[columns_to_clean] = GDP[columns_to_clean].replace({'\$': '', ',': '', '\%':
     →''}, regex=True)
     # 4
     GDP['Country'] = GDP['Country'].astype(str)
     GDP[columns_to_clean] = GDP[columns_to_clean].apply(pd.to_numeric)
     # 5
     GDP['GDP'] = GDP['GDP'] / 1000000000000
     GDP['Population'] = GDP['Population'] / 1000000
```

```
GDP
<>:18: SyntaxWarning: invalid escape sequence '\$'
<>:18: SyntaxWarning: invalid escape sequence '\$'
/var/folders/nw/5zcrqdxs7c57b12ptv8284p80000gn/T/ipykernel_36199/1712976710.py:1
8: SyntaxWarning: invalid escape sequence '\$'
  GDP[columns_to_clean] = GDP[columns_to_clean].replace({'\$': '', ',': '', '\%':
''}, regex=True)
/var/folders/nw/5zcrqdxs7c57b12ptv8284p80000gn/T/ipykernel_36199/1712976710.py:1
8: SyntaxWarning: invalid escape sequence '\$'
  GDP[columns_to_clean] = GDP[columns_to_clean].replace({'\$': '', ',': '', '%':
''}, regex=True)
 KeyError
                                            Traceback (most recent call last)
 Cell In[6], line 4
       1 # Code Solution Here
       2
       3 # 1
 ----> 4 GDP = GDP.drop('#', axis=1)
       5 GDP = GDP.drop('GDP (abbrev.)', axis=1)
       7 # 2
 File ~/Library/Python/3.12/lib/python/site-packages/pandas/core/frame.py:5344,u
  → in DataFrame.drop(self, labels, axis, index, columns, level, inplace, errors)
    5196 def drop(
    5197
             self,
    5198
             labels: IndexLabel | None = None,
    (\dots)
    5205
             errors: IgnoreRaise = "raise",
    5206 ) -> DataFrame | None:
    5207
    5208
             Drop specified labels from rows or columns.
    5209
    (...)
    5342
                     weight 1.0
                                      0.8
    5343
 -> 5344
             return super().drop(
    5345
                  labels=labels,
    5346
                  axis=axis,
    5347
                  index=index,
                  columns=columns,
    5348
    5349
                 level=level,
```

inplace=inplace,

errors=errors,

53505351

5352

```
File ~/Library/Python/3.12/lib/python/site-packages/pandas/core/generic.py:4711
 →in NDFrame.drop(self, labels, axis, index, columns, level, inplace, errors)
   4709 for axis, labels in axes.items():
            if labels is not None:
   4710
                obj = obj._drop_axis(labels, axis, level=level, errors=errors)
-> 4711
   4713 if inplace:
   4714
            self. update inplace(obj)
File ~/Library/Python/3.12/lib/python/site-packages/pandas/core/generic.py:4753
 →in NDFrame._drop_axis(self, labels, axis, level, errors, only_slice)
                new_axis = axis.drop(labels, level=level, errors=errors)
   4751
   4752
            else:
-> 4753
                new_axis = axis.drop(labels, errors=errors)
            indexer = axis.get_indexer(new_axis)
   4756 # Case for non-unique axis
   4757 else:
File ~/Library/Python/3.12/lib/python/site-packages/pandas/core/indexes/base.py
 ⇔7000, in Index.drop(self, labels, errors)
   6998 if mask.any():
            if errors != "ignore":
   6999
-> 7000
                raise KeyError(f"{labels[mask].tolist()} not found in axis")
            indexer = indexer[~mask]
   7001
   7002 return self.delete(indexer)
KeyError: "['#'] not found in axis"
```

2.2 1. Education Index Data by Country (3 Points)

Check out the Wikipedia page, which contains the education index for all countries from 1990 to 2019.

2.2.1 1.0 Scraping the Education Index Data

The code provided scrapes the data from (https://en.wikipedia.org/wiki/Education_Index) into a data frame called EDU.

```
[]: # URL to fetch data from
URL_EDU = "https://en.wikipedia.org/wiki/Education_Index"

# Fetch the HTML content
response = requests.get(URL_EDU)
soup = BeautifulSoup(response.content, 'html.parser')

# Find the table and read it into a DataFrame
table = soup.find_all('table')[0] # Assuming the first table is the one we want
table_IO = StringIO(str(table))
```

```
EDU = pd.read_html(table_IO, flavor='bs4', header=0)[0]
     EDU_preclean = EDU.copy()
     EDU.head(5)
[]:
             Country
                        1990
                               1991
                                       1992
                                               1993
                                                      1994
                                                              1995
                                                                     1996
                                                                             1997
                                                                                     1998
     0
        Afghanistan
                      0.122
                              0.133
                                      0.145
                                             0.156
                                                     0.168
                                                             0.179
                                                                    0.190
                                                                            0.202
                                                                                    0.213
     1
             Albania
                      0.583
                              0.588
                                      0.557
                                             0.542
                                                     0.528
                                                             0.550
                                                                    0.557
                                                                            0.569
                                                                                    0.579
     2
             Algeria
                      0.385
                              0.395
                                      0.405
                                             0.414
                                                     0.424
                                                             0.431
                                                                    0.443
                                                                            0.458
                                                                                    0.473
     3
             Andorra
                        NaN
                                NaN
                                        NaN
                                               NaN
                                                       NaN
                                                               NaN
                                                                      NaN
                                                                              NaN
                                                                                      NaN
     4
              Angola
                                        NaN
                                                       NaN
                        NaN
                                NaN
                                               NaN
                                                               NaN
                                                                      NaN
                                                                              NaN
                                                                                      NaN
             2010
                    2011
                            2012
                                    2013
                                           2014
                                                   2015
                                                          2016
                                                                  2017
                                                                          2018
                                                                                  2019
           0.372
                   0.374
                           0.390
     0
                                  0.398
                                          0.403
                                                  0.405
                                                         0.406
                                                                 0.408
                                                                         0.413
                                                                                0.414
           0.671
                   0.714
                           0.739
                                  0.749
                                          0.758
                                                  0.753
                                                         0.745
                                                                         0.743
     1
                                                                 0.747
                                                                                0.746
     2
           0.626
                   0.644
                           0.639
                                  0.639
                                          0.652
                                                  0.659
                                                         0.660
                                                                 0.665
                                                                         0.668
                                                                                0.672
     3
           0.670
                   0.671
                           0.724
                                  0.714
                                          0.725
                                                  0.718
                                                         0.722
                                                                 0.713
                                                                         0.720
                                                                                0.720
            0.398
                   0.423
                           0.435
                                  0.447
                                          0.460
                                                 0.472
                                                         0.487
                                                                 0.498
                                                                        0.500
                                                                                0.500
```

[5 rows x 31 columns]

[]:

2.2.2 1.1 Cleaning the Education Data (3 points)

Perform the following tasks to clean the EDU DataFrame:

- 1. Modify the resulting DataFrame EDU to only keep 2 variables: 1) the country's name and 2) its education index from 2019.
- 2. Rename the variable named "2019" to "EDIndex".
- 3. Update the dtype of 'Country' to a string.

Display the first 5 rows of EDU and the info of EDU after making these changes.

```
[]: # Code Solution Here

# 1
EDU = EDU[['Country', '2019']]

# 2
EDU = EDU.rename(columns={'2019': 'EDIndex'})

# 3
EDU['Country'] = EDU['Country'].astype(str)
EDU.head(5)
```

```
[]:
            Country
                      EDIndex
        Afghanistan
     0
                        0.414
     1
            Albania
                        0.746
     2
            Algeria
                        0.672
            Andorra
     3
                        0.720
     4
              Angola
                        0.500
```

[]: EDU.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 189 entries, 0 to 188
Data columns (total 2 columns):
    Column
             Non-Null Count
                             Dtype
             _____
 0
    Country
             189 non-null
                             object
 1
    EDIndex 189 non-null
                             float64
dtypes: float64(1), object(1)
memory usage: 3.1+ KB
```

2.3 2: Merging the Datasets (8 points)

Now, we are going to merge the datasets for maximum gains. Make sure you carefully read the instructions for each question. Be very careful in this part of the assignment.

2.3.1 2.0 Joining GDP and EDU (2 Points)

The dataset GDP is our primary dataset. Create a new DataFrame GDP_EDU that brings the the education data from EDU into the dataset GDP using a left join only. Display the first 12 rows of GDP_EDU.

```
[]: # Code Solution Here

GDP_EDU = pd.merge(GDP,EDU, how="left")

GDP_EDU
```

[]:		Country	GDP	Growth	Population	PerCapita	Share	\
	0	United States	25.462700	2.06	341.534046	74554	25.32	
	1	China	17.963200	2.99	1425.179569	12604	17.86	
	2	Japan	4.231140	1.03	124.997578	33850	4.21	
	3	Germany	4.072190	1.79	84.086227	48429	4.05	
	4	India	3.385090	7.00	1425.423212	2375	3.37	
		•••	•••		•••	•••		
	172	Sao Tome & Principe	0.000547	0.93	0.226305	2416	0.00	
	173	Micronesia	0.000427	-0.62	0.523477	816	0.00	
	174	Marshall Islands	0.000280	1.50	0.040077	6978	0.00	
	175	Kiribati	0.000223	1.56	0.130469	1712	0.00	
	176	Tuvalu	0.000060	0.68	0.009992	6040	0.00	

```
EDIndex 0.900
```

```
1
        0.862
2
        0.851
3
        0.943
4
        0.555
172
          NaN
173
          NaN
174
        0.707
175
        0.594
176
          NaN
```

[177 rows x 7 columns]

2.3.2 2.1 Missing Education Index (2 Points)

How many countries in GDP_EDU have missing values for Education Index? Show code that can be used to answer this question and then write your answer in complete sentences.

```
[]:  # Code Solution Here

missing_edu = GDP_EDU['EDIndex'].isnull().sum()
missing_edu
```

[]: 19

Answer: There are 19 missing values for Education Index. This was found by slicing the GDP_EDU dataframe by the EDIndex and summing all the null (NaN) values.

2.3.3 2.2 Data Inspection (3 Points)

Closely inspect the original datasets and answer the following questions about GDP_EDU in complete sentences. You can use the code if needed, but it is not required. Please show all work. If you don't reference the appropriate dataset or you are not specific in your answers, you will get 0 points.

2.2.0 Why is there no education index for Iran in the dataset G_EDU? (1 Point)

```
[]: # assuming G_EDU is referring to GDP_EDU

EDU_preclean[EDU_preclean['Country'] == 'Iran']
```

```
[]: Empty DataFrame
```

```
Columns: [Country, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997, 1998, 1999, 2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019]
Index: []
```

[0 rows x 31 columns]

```
[]: EDU_preclean[EDU_preclean['Country'] == 'Iran (Islamic Republic of)']
```

```
[]:
                             Country
                                                       1992
                                                               1993
                                        1990
                                                1991
                                                                      1994
                                                                              1995
         Iran (Islamic Republic of)
                                       0.397
                                               0.414
                                                      0.432
                                                              0.449
                                                                     0.466
                                                                             0.483
          1996
                                           2011
                                                   2012
                  1997
                         1998
                                    2010
                                                          2013
                                                                  2014
                                                                          2015
                                                                                 2016
                0.504
                                   0.662
     79
        0.493
                        0.514
                                         0.684
                                                 0.731
                                                         0.738
                                                                0.738
                                                                        0.739
                                                                                0.747
          2017
                 2018
                        2019
                       0.756
        0.749
                0.75
     [1 rows x 31 columns]
    Answer: Looking at the original dataset, Iran has an education index of 0.756 in 2019, but it's
    listed as "Iran (Islamic Republic of)", which likely caused the error.
    2.2.1 Why is there no education index for State of Palestine in the dataset GDP_EDU?
    (1 Point)
[]: EDU preclean[EDU preclean['Country'] == 'State of Palestine']
[]: Empty DataFrame
     Columns: [Country, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997, 1998, 1999,
     2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012,
     2013, 2014, 2015, 2016, 2017, 2018, 2019]
     Index: []
     [0 rows x 31 columns]
     EDU_preclean[EDU_preclean['Country'] == 'Palestine, State of']
[]:
                                                                       1996
                                                                                    \
                       Country
                                 1990
                                       1991
                                              1992
                                                    1993
                                                           1994
                                                                 1995
                                                                              1997
     131
          Palestine, State of
                                  NaN
                                        NaN
                                               NaN
                                                     NaN
                                                            NaN
                                                                  NaN
                                                                        NaN
                                                                               NaN
          1998
                     2010
                            2011
                                    2012
                                          2013
                                                  2014
                                                          2015
                                                                2016
                                                                       2017
                                                                               2018
                                                                              0.676
     131
                    0.656
                           0.662 0.671 0.66
                                                0.663
                                                        0.668
                                                                0.67
                                                                      0.675
           2019
```

[1 rows x 31 columns]

0.678

131

Answer: There is no "State of Palestine" in the original dataset. There is, however, a "Palestine, State of" value with an education index of 0.678 in 2019.

2.2.2 Why is there no education index for Laos in the dataset GDP_EDU? (1 point) Answer: Same reason as the other two, it's listed as "Lao People's Democratic Republic" in the original dataset, and so since the names don't match up the join did not work.

2.3.4 2.2 Removing NA Values (1 point)

Instead of replacing or dropping all the countries with missing values by hand, we will just drop all rows that are missing the Education Index to move forward with the analysis portion. Drop all rows from GDP_EDU that are null for EDIndex.

```
[]: # Code Solution Here

GDP_EDU = GDP_EDU.dropna(subset=['EDIndex'])
```

2.4 3. Analyzing the Merged Dataset (12 points)

In these questions, find the answer using code, and then answer the question using complete sentences below the code.

2.4.1 3.0 Above Average GDP PerCapita (2 Points)

How many countries have a GDP per capita above the global average?

```
[]: # Code Solution Here
```

Answer:

3.1 Highest GDP Growth Rate (4 Points)

- Of the countries that have above average GDP PerCapita, what country has the highest GDP growth rate?
- Of the countries that have below average GDP PerCapita, what country has the highest GDP growth rate?

```
[]: # Code Solution Here
```

Answer:

3.2 Lowest Education Index (4 Points)

- Of the countries that have above average GDP PerCapita, what country has the lowest education index?
- Of the countries that have below average GDP PerCapita, what country has the lowest education index?

```
[]: # Code Solution Here
```

Answer:

3.3 Critical Thinking (2 points) State two additional questions you could answer with the merged dataset. Be creative. You do not need to find the answer, but are welcome to if you are curious.

Answer: