## Problem 1.

https://github.com/meesuny/lab2

## Problem 2.

```
!pip install openpyxl # for Excel
Collecting openpyxl
 Downloading openpyxl-3.0.10-py2.py3-none-any.whl (242 kB)
                                ----- 242.1/242.1 kB 4.3 MB/s eta 0:00:00a 0:00:01
Collecting et-xmlfile
 Downloading et xmlfile-1.1.0-py3-none-any.whl (4.7 kB)
Installing collected packages: et-xmlfile, openpyxl
Successfully installed et-xmlfile-1.1.0 openpyxl-3.0.10
WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a
virtual environment instead: https://pip.pypa.io/warnings/venv
!pip install pyreadstat # for SPSS
Collecting pyreadstat
 Downloading pyreadstat-1.1.9-cp310-cp310-manylinux 2 17 x86 64.manylinux2014 x86 64.whl (2.7 MB)
                                           - 2.7/2.7 MB 10.1 MB/s eta 0:00:0000:0100:01
Requirement already satisfied: pandas>=1.2.0 in /usr/local/lib/python3.10/site-packages (from pyreadstat) (1.4.4)
Requirement already satisfied: numpy>=1.21.0 in /usr/local/lib/python3.10/site-packages (from pandas>=1.2.0->pyreadstat) (1.23.3)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/site-packages (from pandas>=1.2.0->pyreadstat) (2022.4)
Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.10/site-packages (from pandas>=1.2.0->pyreadstat) (2.8.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/site-packages (from python-dateutil>=2.8.1->pandas>=1.2.0->pyreadstat) (1.16.0)
Installing collected packages: pyreadstat
Successfully installed pyreadstat-1.1.9
WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a
 virtual environment instead: https://pip.pypa.io/warnings/venv
!pip install pystata
```

```
Collecting pystata
          Downloading pystata-0.0.1-py3-none-any.whl (21 kB)
        Requirement already satisfied: pandas in /usr/local/lib/python3.10/site-packages (from pystata) (1.4.4)
        Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/site-packages (from pandas->pystata) (2022.4)
        Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.10/site-packages (from pandas->pystata) (2.8.2)
        Requirement already satisfied: numpy>=1.21.0 in /usr/local/lib/python3.10/site-packages (from pandas->pystata) (1.23.3)
        Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/site-packages (from python-dateutil>=2.8.1->pandas->pystata) (1.16.0)
        Installing collected packages: pystata
        Successfully installed pystata-0.0.1
        WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a
         virtual environment instead: https://pip.pypa.io/warnings/venv
In [4]: import pandas as pd
        import numpy as np
        import requests
        import ison
        import os
        import openpyxl
        from bs4 import BeautifulSoup
        import collections
        collections.Callable = collections.abc.Callable
In [5]: data1 = pd.read_csv('data/data1.csv', header=2)
        data1
```

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Out	12	

]:	Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Explained by: Freedom to make life choices	Explained by: Generosity	Explained by: Perceptions of corruption
	<b>0</b> Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	1 Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	<b>2</b> Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3 Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4 Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	<b></b>				<b></b>						
15	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
15	<b>T</b> anzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
15	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106
1!	Central <b>54</b> African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	0.218	0.038
15	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0.065	0.149	0.076

```
In [6]: data2 = pd.read_csv('data/data2.txt', header=2, comment = '/')
     data2
```

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		Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Explained by: Freedom to make life choices	Explained by: Generosity	Explained by: Perceptions of corruption
	0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3	Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4	Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	•••											
1	51	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
1	52	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
	53	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106
	54	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	0.218	0.038
1	55	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0.065	0.149	0.076

In [7]: data3 = pd.read\_csv('data/data3.txt', header=2, sep='\t')
 data3

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]:	Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Explained by: Freedom to make life choices	Explained by: Generosity	Explained by: Perceptions of corruption
	<b>0</b> Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	<b>1</b> Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	<b>2</b> Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3 Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4 Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
			•••								
15	<b>1</b> Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
15	<b>2</b> Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
15	3 South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106
15	Central 4 African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	0.218	0.038
15	<b>5</b> Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0.065	0.149	0.076

In [8]: data4 = pd.read\_csv('data/data4.txt', header=None, sep='\$')
 data4

# Add headers on this one?

Out[8]:		0	1	2	3	4	5	6	7	8	9	10
	0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3	Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4	Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	•••											
	151	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
	152	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
	153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106
	154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	0.218	0.038
	155	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0.065	0.149	0.076

In [9]: data5 = pd.read\_csv('data/data5.csv', nrows=156)
 data5

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]:		Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Explained by: Freedom to make life choices	Explained by: Generosity	Explained by: Perceptions of corruption
	0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3	Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4	Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	•••											
•	151	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
	152	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
•	153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106
1	154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	0.218	0.038
	155	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0.065	0.149	0.076

In [10]: data6 = pd.read\_csv('data/data6.dat', na\_values = '999.000')
 data6

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•	Country	Happiness score			Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Explained by: Freedom to make life choices	Explained by: Generosity	Explained by: Perceptions of corruption
	<b>0</b> Finland	7.632	7.695	7.569	2.595	NaN	NaN	NaN	0.681	0.192	0.393
	1 Norway	7.594	7.657	7.530	NaN	NaN	1.582	NaN	0.686	0.286	0.340
	2 Denmark	7.555	7.623	7.487	2.370	1.351	1.590	NaN	0.683	0.284	0.408
	3 Iceland	7.495	7.593	NaN	2.426	1.343	1.644	0.914	0.677	0.353	NaN
	4 Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	•••										
1	<b>51</b> NaN	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	NaN	0.064
1	<b>52</b> Tanzania	NaN	NaN	3.193	0.628	NaN	0.991	0.381	0.481	0.270	0.097
1	53 South Sudan	3.254	NaN	3.123	1.691	0.337	NaN	0.177	0.112	0.224	0.106
1	Central <b>54</b> African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	NaN	0.038
1	55 Burundi	2.905	3.074	NaN	1.752	0.091	NaN	0.145	0.065	0.149	0.076

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		Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Explained by: Freedom to make life choices	Explained by: Generosity	Explained by: Perceptions of corruption
	0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3	Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4	Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	•••			•••								
1	151	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
1	52	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
1	153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106
1	154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0.305	0.218	0.038
1	155	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0.065	0.149	0.076

In [12]: data8 = pd.read\_stata('data/data8.dta')
 data8

# STATA pip install

Out[12]:		country	happinessscore	whiskerhigh	whiskerlow	dystopia 192 residual	explained by gdpper capita	explained by social support	explained by healthy life expectancy	explained by freedom tor
	0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	
	1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	
	2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	
	3	Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	
	4	Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	
	•••									
	151	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	
	152	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	
	153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	
	154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	
	155	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	



Out[13]:		country	happiness	whiskerhigh	whiskerlow	dystopia	gdpPC	socsupport	lifeexp	lifechoice	generous	corrupt
	0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.681	0.192	0.393
	1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.686	0.286	0.340
	2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.683	0.284	0.408
	3	Iceland	7.495	7.593	7.398	2.426	1.343	1.644	0.914	0.677	0.353	0.138
	4	Switzerland	7.487	7.570	7.405	2.320	1.420	1.549	0.927	0.660	0.256	0.357
	•••											
	151	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0.244	0.083	0.064
	152	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0.481	0.270	0.097
	153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0.112	0.224	0.106

2.939

2.735

2.487

1.752

0.024

0.091

0.010

0.627 0.145

0.000

0.218

0.149

0.038

0.076

0.305

0.065

156 rows × 11 columns

155

**154** Central African Republic

3.083

2.905

3.227

3.074

In [14]: data10 = pd.read\_sas('data/data10.xpt')
 data10

Burundi

Out[14]:		COUNTRY	HAPPINES	WHISKERH	WHISKERL	DYSTOPIA	EXPLAINE	EXPLAIN2	EXPLAIN3	EXPLAIN4	EXPLAIN5	EXPLAIN6
	0	b'Finland'	7.632	7.695	7.569	2.595	1.305	1.592000e+00	0.874	0.681	0.192	0.393
	1	b'Norway'	7.594	7.657	7.530	2.383	1.456	1.582000e+00	0.861	0.686	0.286	0.340
	2	b'Denmark'	7.555	7.623	7.487	2.370	1.351	1.590000e+00	0.868	0.683	0.284	0.408
	3	b'Iceland'	7.495	7.593	7.398	2.426	1.343	1.644000e+00	0.914	0.677	0.353	0.138
	4	b'Switzerland'	7.487	7.570	7.405	2.320	1.420	1.549000e+00	0.927	0.660	0.256	0.357
	•••											
	151	b'Yemen'	3.355	3.448	3.262	1.106	0.442	1.073000e+00	0.343	0.244	0.083	0.064
	152	b'Tanzania'	3.303	3.414	3.193	0.628	0.455	9.910000e-01	0.381	0.481	0.270	0.097
	153	b'South Sudan'	3.254	3.385	3.123	1.691	0.337	6.080000e-01	0.177	0.112	0.224	0.106
	154	b'Central African Republic'	3.083	3.227	2.939	2.487	0.024	5.397605e-79	0.010	0.305	0.218	0.038
	155	b'Burundi'	2.905	3.074	2.735	1.752	0.091	6.270000e-01	0.145	0.065	0.149	0.076

## Problem 3.

```
headers = headers)
         <Response [200]>
Out[17]:
In [18]: myjson = json.loads(r.text)
          near earth objects1 = pd.json normalize(myjson, record path = ['near earth objects','2022-10-01'])
          near_earth_objects2 = pd.json_normalize(myjson, record_path = ['near_earth_objects','2022-10-02'])
          near earth objects3 = pd.json normalize(myjson, record path = ['near earth objects','2022-10-03'])
          near earth objects4 = pd.json normalize(myjson, record path = ['near earth objects','2022-10-04'])
          near_earth_objects5 = pd.json_normalize(myjson, record_path = ['near_earth_objects','2022-10-05'])
          near earth objects6 = pd.json normalize(myjson, record path = ['near earth objects','2022-10-06'])
          near earth objects7 = pd.json normalize(myjson, record path = ['near earth objects','2022-10-07'])
          near_earth_objects_all = pd.concat([near_earth_objects1, near_earth_objects2, near_earth_objects3, near_earth_objects4,
                                            near earth objects5, near earth objects6, near earth objects7], ignore index=True)
          #near earth objects all
          # This adds "close approach date" as the very last column of the resulting data frame.
          df = pd.DataFrame(data=near earth objects all)
          df['close approach date'] = [d[0]['close approach date'] for d in df['close approach data']]
          near earth objects all
```

Out[18]:		id	neo_reference_id	name	nasa_jpl_url	absolute_magnitude_h	$is\_potentially\_haz ardous\_asteroid$	close_approach_data	is_sentry_object	
	0	3429684	3429684	(2008 TZ)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3429684	25.40	False	[{'close_approach_date': '2022-10-01', 'close	False	http://api.nasa.gov/neo
	1	3548667	3548667	(2010 TX54)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3548667	20.98	False	[{'close_approach_date': '2022-10-01', 'close	False	http://api.nasa.gov/neo
	2	3650847	3650847	(2013 TN127)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3650847	26.20	False	[{'close_approach_date': '2022-10-01', 'close	False	http://api.nasa.gov/neo
	3	3772755	3772755	(2017 FT90)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3772755	26.20	False	[{'close_approach_date': '2022-10-01', 'close	False	http://api.nasa.gov/neo
	4	3782015	3782015	(2017 SZ20)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3782015	24.50	False	[{'close_approach_date': '2022-10-01', 'close	False	http://api.nasa.gov/neo
		•••								
	75	3342645	3342645	(2006 SG7)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3342645	22.90	False	[{'close_approach_date': '2022-10-07', 'close	False	http://api.nasa.gov/neo
	76	3384488	3384488	(2007 RD20)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3384488	21.00	False	[{'close_approach_date': '2022-10-07', 'close	False	http://api.nasa.gov/neo
	77	3625706	3625706	(2013 BD74)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3625706	24.20	False	[{'close_approach_date': '2022-10-07', 'close	False	http://api.nasa.gov/neo
	78	3648879	3648879	(2013 TJ6)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=3648879	25.20	False	[{'close_approach_date': '2022-10-07', 'close	False	http://api.nasa.gov/neo
	79	54055076	54055076	(2020 TE2)	http://ssd.jpl.nasa.gov/sbdb.cgi? sstr=54055076	26.30	False	[{'close_approach_date': '2022-10-07', 'close	False	http://api.nasa.gov/neo/





## Problem 4.

a) Zillow.com disallows the following pages and subpages to be crawled, which suggests that they do not allow pulling the current listed prices for houses for sale in Charlottesville from zillow.com.

- Disallow: /homes/comps\_for\_sale/
- Disallow: /homes/comps/
- Disallow: /homes/for\_sale/\*\_agent\_list/
- Disallow: /homes/for\_sale/1\_favorite\_list/1\_rs/1\_fr/
- **b)** Google.com allows both /m/finance and /finance to be crawled, so scraping the current stock prices off of google.com/finance is permitted.
- c) Twitter disallows /\*/followers to be crawled, so we wouldn't be able to copy the list of twitter accounts that each NBA player follows.
- **d)** Genius.com disallows crawling /songs/\*/metadata, and metadata presumably includes the lyrics, so it doesn't allow getting lyrics from their site. This means that we can't get the lyrics to Lizzo's Good as Hell from https://genius.com/Lizzo-good-as-hell-lyrics.

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