# Lab Assignment 2: How to Load CSV, ASCII, and other data into Python

# **DS 6001: Practice and Application of Data Science**

### Instructions

Please answer the following questions as completely as possible using text, code, and the results of code as needed. Format your answers in a Jupyter notebook. To receive full credit, make sure you address every part of the problem, and make sure your document is formatted in a clean and professional way.

There are 11 data files attached to this lab assignment, with different extensions. First, download all of these data files, and save them in the same folder on your local machine. Your task in the following questions is to load each file into Python correctly, so that you can begin the process of data cleaning. If the variable names are included in the file, use those names to name the columns. If the variable names are not included, use these names in order:

If you loaded the data correctly, it will look like data clean.csv, which is also attached to this lab.

### Problem 0

Import the libraries you will need. Then write code to change the working directory to the folder in which you saved the data files, run the code displayed above to create the <code>column\_names</code> list, load <code>data\_clean.csv</code>, and display the output of the <code>.info()</code> method of <code>data\_clean.(1 point)</code>

```
In [104]: # Import the libraries
import os
import numpy as np
import pandas as pd
In [105]: # Changing the working directory
#os.chdir('lab data')
```

```
In [106]: # Check the data directory
          !ls
                                        data5.csv
          data1.csv
                         data2.txt
                                                       data8.dta
          data10.xpt
                         data3.txt
                                        data6.dat
                                                       data9.sav
          data11.txt
                         data4.txt
                                        data7.xlsx
                                                       data_clean.csv
In [107]: # Load clean dataset
          data clean = pd.read_csv('data_clean.csv')
In [108]: | # Create the list of column names
          column names = list(data clean.columns)
          type(column_names)
Out[108]: list
In [109]: data_clean.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 156 entries, 0 to 155
          Data columns (total 11 columns):
               Column
                                                           Non-Null Count Dtype
           0
                                                                           object
               Country
                                                           156 non-null
               Happiness score
                                                           156 non-null
                                                                           float6
           1
           2
               Whisker-high
                                                           156 non-null
                                                                           float6
          4
           3
               Whisker-low
                                                           156 non-null
                                                                           float6
          4
           4
               Dystopia (1.92) + residual
                                                           156 non-null
                                                                           float6
          4
               Explained by: GDP per capita
           5
                                                           156 non-null
                                                                           float6
          4
           6
               Explained by: Social support
                                                           156 non-null
                                                                           float6
           7
               Explained by: Healthy life expectancy
                                                          156 non-null
                                                                           float6
          4
           8
               Explained by: Freedom to make life choices 156 non-null
                                                                           float6
          4
           9
               Explained by: Generosity
                                                           156 non-null
                                                                           float6
           10 Explained by: Perceptions of corruption
                                                           156 non-null
                                                                           float6
          dtypes: float64(10), object(1)
          memory usage: 13.5+ KB
```

Load datal.csv. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (1 point)

Answer 1: after skipping the first 2 rows, Finland is the first observation as expected - this does match the clean template:

```
In [110]: # Loading and inspecting `data1`
            data1 = pd.read csv('data1.csv')
            data1.head(3)
Out[110]:
                    Source: The World Happiness
                   Report (2018), The Sustainable
                                              Unnamed: Unnamed: Unnamed: Unnamed: Ur
                  Development Solutions Network
                                                               2
                                                                         3
                                                                                   4
                                                                                             5
                                      (SDSN)
                                        URL:
                                                   NaN
                                                             NaN
                                                                       NaN
                                                                                 NaN
                                                                                           NaN
               http://worldhappiness.report/ed/2018
                                                                              Dystopia
                                                                                       Explained
                                                                                                 Ε
                                                          Whisker-
                                                                    Whisker-
                                              Happiness
             1
                                      Country
                                                                               (1.92) +
                                                                                        by: GDP
                                                                                                 b
                                                             high
                                                                        low
                                                  score
                                                                               residual
                                                                                       per capita
                                                  7.632
                                                            7.695
                                                                      7.569
                                                                                2.595
                                                                                          1.305
                                      Finland
             2
In [111]: # Need to skip firts 2 rows as it is technical description of the data,
             not the data per se
            data1 = pd.read csv('data1.csv', skiprows=2)
            data1.head(3)
```

#### Out[111]:

	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	b Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

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```
In [112]: data_clean.head(3)
```

Out[112]:

	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	Explaine b Freedo to mal li choice
_	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
_2	2 Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

# **Problem 2**

Load data2.txt. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (1 point)

### Answer 2: again selected rows 0, 1, and 3 needed to be skipped. The result matches the template

```
In [113]: data2 = pd.read_csv('data2.txt')
    data2.head(5)
```

Out[113]:

	Source: The World Happiness Report (2018), The Sustainable Development Solutions Network (SDSN)	Unnamed: 1	Unnamed: 2	Unnamed:	Unnamed: 4	Unnamed: 5	Ur
0	URL: http://worldhappiness.report/ed/2018	NaN	NaN	NaN	NaN	NaN	
1	Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	E b
2	/The following countries comprise the "very ha	NaN	NaN	NaN	NaN	NaN	
3	Finland	7.632	7.695	7.569	2.595	1.305	
4	Norway	7.594	7.657	7.530	2.383	1.456	

```
In [114]: data2 = pd.read_csv('data2.txt', skiprows=[0, 1, 3])
    data2.head(3)
```

Out[114]:

	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	Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

# **Problem 3**

Load data3.txt. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (1 point)

Answer 3: this was tap delimited text file - use sep='\t' and also skip the first 2 rows.

```
In [115]: data3 = pd.read_csv('data3.txt')
    data3.head(3)
```

Out[115]:

Evalaina

0	URL: http://worldhappiness.report/ed/2018\t\t\
1	Country\tHappiness score\tWhisker-high\tWhiske
2	Finland\t7.632\t7.695\t7.569\t2.595\t1.305\t1

```
In [116]: data3 = pd.read_csv('data3.txt', sep='\t', skiprows=2)
    data3.head(3)
```

Out[116]:

	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	Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

Load data4.txt. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (1 point)

Answer 4: different delimiter was used and column names were missing - pass dollar sign to sep= and a list of  $column_names$  to names.

	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	Explaine b Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

# **Problem 5**

Load data5.csv. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (1 point)

Answer 5: this file had extra 2 rows at the bottom - used skip footer and explicitly called original python parsing engine to avoid the warning.

```
In [119]: data5 = pd.read_csv('data5.csv')
  data5.head(3)
```

Out[119]:

	Country	Happiness score	Whisker- high	Whisker- Iow	Dystopia (1.92) + residual	Explained by: GDP per capita	Explained by: Social support	Explained by: Healthy life expectancy	Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

In [120]: data5.columns == column\_names

Out[120]: array([ True, True])

In [121]: len(data5)

Out[121]: 158

In [122]: len(data\_clean)

Out[122]: 156

In [123]: data5.tail()

Out[123]:

	Country	Happiness score	Whisker- high	Whisker- low	Dystopia (1.92) + residual	Explained by: GDP per capita	Expl by: { su
153	South Sudan	3.254	3.385	3.123	1.691	0.337	
154	Central African Republic	3.083	3.227	2.939	2.487	0.024	
155	Burundi	2.905	3.074	2.735	1.752	0.091	
156	Source: The World Happiness Report (2018), The	NaN	NaN	NaN	NaN	NaN	
157	URL: http://worldhappiness.report/ed/2018	NaN	NaN	NaN	NaN	NaN	

```
In [124]: data_clean.tail()
```

Out[124]:

	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	Expla Free to rr cho
151	Yemen	3.355	3.448	3.262	1.106	0.442	1.073	0.343	0
152	Tanzania	3.303	3.414	3.193	0.628	0.455	0.991	0.381	0
153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0
154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0
155	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0

```
In [125]: data5 = pd.read_csv('data5.csv', skipfooter=2, engine='python')
    data5.tail(3)
```

Out[125]:

	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	Expla Free to m
153	South Sudan	3.254	3.385	3.123	1.691	0.337	0.608	0.177	0
154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.000	0.010	0
155	Burundi	2.905	3.074	2.735	1.752	0.091	0.627	0.145	0

```
In [126]: len(data_clean) == len(data5)
```

Out[126]: True

# **Problem 6**

Load data6.dat. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (1 point)

Answer 6: all looks good when using pd.read\_csv() na dreplacing 999 with NaN - checking with comparing output of .info() method.

```
In [127]: data6 = pd.read_csv('data6.dat', na_values=999 )
    data6.head(3)
```

Out[127]:

	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	Explaine b Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	NaN	NaN	NaN	0.68
1	Norway	7.594	7.657	7.530	NaN	NaN	1.582	NaN	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	NaN	0.68

In [128]: data6.tail(3)

Out[128]:

	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	Free to m
153	South Sudan	3.254	NaN	3.123	1.691	0.337	NaN	0.177	0
154	Central African Republic	3.083	3.227	2.939	2.487	0.024	0.0	0.010	0
155	Burundi	2.905	3.074	NaN	1.752	0.091	NaN	0.145	0

In [129]: data6.columns == data\_clean.columns

Out[129]: array([ True, True, True, True, True, True, True, True, True, True])

In [130]: data6.info() == data\_clean.info()

_	reIndex: 156 entries, 0 to 155 columns (total 11 columns): Column	Non-Null Count	Dtype
0	Country	145 non-null	object
1	Happiness score	142 non-null	float6
4 2 4	Whisker-high	135 non-null	float6
3 4	Whisker-low	136 non-null	float6
4	Dystopia (1.92) + residual	145 non-null	float6
5 4	Explained by: GDP per capita	137 non-null	float6
6 4	Explained by: Social support	134 non-null	float6
7 4	Explained by: Healthy life expectancy	142 non-null	float6
8 4	Explained by: Freedom to make life choices	140 non-null	float6
9 4	Explained by: Generosity	145 non-null	float6
10 4	Explained by: Perceptions of corruption	143 non-null	float6
<cla< td=""><td>ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'&gt;</td><td></td><td></td></cla<>	ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'>		
_	reIndex: 156 entries, 0 to 155 columns (total 11 columns):		
Data		Non-Null Count	Dtype
Data #	columns (total 11 columns): Column		
Data #	columns (total 11 columns): Column		
Data #  0 1	columns (total 11 columns): Column Country	156 non-null	object
Data #  0 1 4	columns (total 11 columns): Column Country Happiness score	156 non-null 156 non-null	object
Data #  0 1 4 2 4	columns (total 11 columns): Column Country Happiness score Whisker-high	156 non-null 156 non-null 156 non-null	object float6
Data # 0 1 4 2 4 3 4 4	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low	156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6
Data # 0 1 4 2 4 3 4 4 5	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6
Data # 0 1 4 2 4 3 4 4 5 4 6 4 7	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual Explained by: GDP per capita	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6 float6
Data # 0 1 4 2 4 3 4 4 4 5 4 6 4 7 4 8	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6
Data # 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6
Data # 0 1 4 2 4 3 4 4 6 4 7 4 8 4 9 4 10	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6 float6 float6 float6
Data # 0 1 4 2 4 3 4 4 4 6 4 7 4 8 4 9 4 10 4	columns (total 11 columns): Column 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	156 non-null	object float6 float6 float6 float6 float6 float6 float6 float6 float6

Load data7.xlsx, which is an Excel file. Keep only the sheet named "Data". Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (2 points)

Answer 7: after having used parameter sheet\_name='Data', all looks good as confirmed by .info() method comparison above.

```
In [131]: data7 = pd.read_excel('data7.xlsx', sheet_name='Data')
    data7.head(3)
```

Out[131]:

	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	Explains b Freedo to mal li choics
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

In [132]: data7.info() == data\_clean.info()

_	reIndex: 156 entries, 0 to 155 columns (total 11 columns): Column	Non-Null Count	Dtype
0	Country	156 non-null	object
1	Happiness score	156 non-null	float6
4 2 4	Whisker-high	156 non-null	float6
3 4	Whisker-low	156 non-null	float6
4	Dystopia (1.92) + residual	156 non-null	float6
5 4	Explained by: GDP per capita	156 non-null	float6
6 4	Explained by: Social support	156 non-null	float6
7 4	Explained by: Healthy life expectancy	156 non-null	float6
8 4	Explained by: Freedom to make life choices	156 non-null	float6
9 4	Explained by: Generosity	156 non-null	float6
10 4	Explained by: Perceptions of corruption	156 non-null	float6
<cla< td=""><td>ery usage: 13.5+ KB uss 'pandas.core.frame.DataFrame'&gt; useIndex: 156 entries, 0 to 155</td><td></td><td></td></cla<>	ery usage: 13.5+ KB uss 'pandas.core.frame.DataFrame'> useIndex: 156 entries, 0 to 155		
Data	columns (total 11 columns):		
		Non-Null Count	Dtype
#	columns (total 11 columns): Column		
# 	columns (total 11 columns): Column		
#  0 1	columns (total 11 columns): Column Country	156 non-null	object
#  0 1 4	columns (total 11 columns): Column Country Happiness score	156 non-null 156 non-null	object float6
#  0 1 4 2 4 3	columns (total 11 columns): Column Country Happiness score Whisker-high	156 non-null 156 non-null 156 non-null	object
#  0 1 4 2 4 3 4	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low	156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 4 5	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual Explained by: GDP per capita	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8	Columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual Explained by: GDP per capita Explained by: Social support	156 non-null	object float6 float6 float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9 4 10	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9 4 10	Column Co	156 non-null	object float6

Load data8.dta, which is a Stata 13 file. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (2 points)

Answer 8: after having used  $pd.read\_stata$ , need to specify required column names but providing the list of names  $column\_names$ .

```
In [133]: data8 = pd.read_stata('data8.dta')
    data8.head(2)
```

#### Out[133]:

	country	happinessscore	whiskerhigh	whiskerlow	dystopia192residual	explainedbygdppercapit
0	Finland	7.632	7.695	7.569	2.595	1.30
1	Norway	7.594	7.657	7.530	2.383	1.45

In [134]:	data8.columns = column_names
	data8.head(2)

### Out[134]:

	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	Freedor to mak lif choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68

In [135]: data8.info() == data\_clean.info()

Int6	ss 'pandas.core.frame.DataFrame'> 4Index: 156 entries, 0 to 155 columns (total 11 columns):		
#	Column	Non-Null Count	Dtype
0 1	Country Happiness score	156 non-null 156 non-null	object float3
2 2 2	Whisker-high	156 non-null	float3
3 2	Whisker-low	156 non-null	float3
4 2	Dystopia (1.92) + residual	156 non-null	float3
5 2	Explained by: GDP per capita	156 non-null	float3
6 2	Explained by: Social support	156 non-null	float3
7 2	Explained by: Healthy life expectancy	156 non-null	float3
8 2	Explained by: Freedom to make life choices		float3
9	Explained by: Generosity	156 non-null	float3
10	Explained by: Perceptions of corruption	156 non-null	float3
memo <cla Rang</cla 	es: float32(10), object(1) ry usage: 8.5+ KB ss 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 columns (total 11 columns):		
	Column	Non-Null Count	Dtype
	Country Happiness score	156 non-null 156 non-null	_
2 4	Whisker-high	156 non-null	float6
3 4	Whisker-low	156 non-null	float6
4 4	Dystopia (1.92) + residual	156 non-null	£1+C
		130 HOH-HUII	float6
5 4	Explained by: GDP per capita	156 non-null	float6
	Explained by: GDP per capita  Explained by: Social support		
4 6		156 non-null	float6
4 6 4 7	Explained by: Social support	156 non-null 156 non-null	float6
4 6 4 7 4	Explained by: Social support  Explained by: Healthy life expectancy	156 non-null 156 non-null	float6 float6 float6
4 6 4 7 4 8 4	Explained by: Social support  Explained by: Healthy life expectancy  Explained by: Freedom to make life choices	156 non-null 156 non-null 156 non-null 156 non-null	float6 float6 float6 float6
4 6 4 7 4 8 4 9 4 10 4 dtyp	Explained by: Social support  Explained by: Healthy life expectancy  Explained by: Freedom to make life choices  Explained by: Generosity	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	float6 float6 float6 float6 float6

Switzerland

7.487

Load data9.sav, which is an SPSS file. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (2 points)

Answer 9: I had to install additional library pyreadstat and restart the kernel to refresh pandas import.

```
In [136]:
             #!pip install pyreadstat
In [137]:
             data9 = pd.read_spss('data9.sav')
             data9.head()
Out[137]:
                    country happiness
                                       whiskerhigh whiskerlow dystopia gdpPC
                                                                                 socsupport lifeexp lifechc
                                 7.632
                                             7.695
                                                         7.569
                                                                   2.595
                                                                           1.305
                                                                                       1.592
                                                                                                         0.
              0
                    Finland
                                                                                              0.874
              1
                    Norway
                                 7.594
                                             7.657
                                                         7.530
                                                                   2.383
                                                                           1.456
                                                                                       1.582
                                                                                              0.861
                                                                                                         0.
              2
                   Denmark
                                 7.555
                                             7.623
                                                         7.487
                                                                   2.370
                                                                           1.351
                                                                                       1.590
                                                                                              0.868
                                                                                                         0.
                    Iceland
                                 7.495
                                             7.593
                                                         7.398
                                                                   2.426
                                                                           1.343
                                                                                       1.644
                                                                                              0.914
                                                                                                         0.
              3
```

7.405

2.320

1.420

1.549

0.927

0.

7.570

```
In [138]: data9.info() == data_clean.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 156 entries, 0 to 155
          Data columns (total 11 columns):
           #
                            Non-Null Count
               Column
                                            Dtype
          ___
                                            ____
           0
               country
                            156 non-null
                                            object
               happiness
                            156 non-null
                                            float64
           1
           2
               whiskerhigh 156 non-null
                                            float64
           3
               whiskerlow
                            156 non-null
                                            float64
               dystopia
                            156 non-null
                                            float64
           5
               gdpPC
                            156 non-null
                                            float64
           6
               socsupport
                            156 non-null
                                            float64
           7
               lifeexp
                            156 non-null
                                            float64
               lifechoice
                            156 non-null
           8
                                            float64
           9
               generous
                            156 non-null
                                            float64
           10
               corrupt
                            156 non-null
                                            float64
          dtypes: float64(10), object(1)
          memory usage: 13.5+ KB
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 156 entries, 0 to 155
          Data columns (total 11 columns):
               Column
                                                           Non-Null Count
                                                                           Dtype
                                                            _____
                                                                           ____
           0
               Country
                                                           156 non-null
                                                                           object
           1
               Happiness score
                                                           156 non-null
                                                                           float6
          4
           2
               Whisker-high
                                                           156 non-null
                                                                           float6
          4
           3
                                                           156 non-null
               Whisker-low
                                                                           float6
          4
           4
               Dystopia (1.92) + residual
                                                           156 non-null
                                                                           float6
          4
           5
               Explained by: GDP per capita
                                                           156 non-null
                                                                           float6
          4
           6
               Explained by: Social support
                                                           156 non-null
                                                                           float6
          4
           7
               Explained by: Healthy life expectancy
                                                           156 non-null
                                                                           float6
          4
           8
               Explained by: Freedom to make life choices 156 non-null
                                                                           float6
          4
           9
               Explained by: Generosity
                                                           156 non-null
                                                                           float6
          4
           10
               Explained by: Perceptions of corruption
                                                           156 non-null
                                                                           float6
```

dtypes: float64(10), object(1)
memory usage: 13.5+ KB

Load data10.xpt, which is a SAS file. Use the tools we discussed in class to decide whether the data file loaded correctly, and include that code in your lab report. In one or two sentences, describe how you decided on the right combination of parameters needed to load the data. (If some of the country names display as b'Finland', don't worry aout that.) (2 points)

# Answer 10: I could not pass desired column names to pd.read\_sas() therefore assigning names manually in a separate step.

```
In [139]: data10 = pd.read_sas('data10.xpt')
    data10.head()
```

Out[139]:

	COUNTRY	HAPPINES	WHISKERH	WHISKERL	DYSTOPIA	EXPLAINE	EXPLAIN2	EXPLAIN
0	b'Finland'	7.632	7.695	7.569	2.595	1.305	1.592	0.87
1	b'Norway'	7.594	7.657	7.530	2.383	1.456	1.582	0.86
2	b'Denmark'	7.555	7.623	7.487	2.370	1.351	1.590	0.86
3	b'Iceland'	7.495	7.593	7.398	2.426	1.343	1.644	0.9
4	b'Switzerland'	7.487	7.570	7.405	2.320	1.420	1.549	0.92

In [140]: data10.columns = column\_names

In [141]: data10.info() == data\_clean.info()

_	reIndex: 156 entries, 0 to 155 columns (total 11 columns): Column	Non-Null Count	Dtype
0	Country	156 non-null	object
1	Happiness score	156 non-null	float6
4 2 4	Whisker-high	156 non-null	float6
3 4	Whisker-low	156 non-null	float6
4	Dystopia (1.92) + residual	156 non-null	float6
5 4	Explained by: GDP per capita	156 non-null	float6
6 4	Explained by: Social support	156 non-null	float6
7 4	Explained by: Healthy life expectancy	156 non-null	float6
8 4	Explained by: Freedom to make life choices	156 non-null	float6
9 4	Explained by: Generosity	156 non-null	float6
10 4	Explained by: Perceptions of corruption	156 non-null	float6
<cla< td=""><td>ery usage: 13.5+ KB uss 'pandas.core.frame.DataFrame'&gt; useIndex: 156 entries, 0 to 155</td><td></td><td></td></cla<>	ery usage: 13.5+ KB uss 'pandas.core.frame.DataFrame'> useIndex: 156 entries, 0 to 155		
Data	columns (total 11 columns):		
		Non-Null Count	Dtype
#	columns (total 11 columns): Column		
# 	columns (total 11 columns): Column		
#  0 1	columns (total 11 columns): Column Country	156 non-null	object
#  0 1 4	columns (total 11 columns): Column Country Happiness score	156 non-null 156 non-null	object float6
#  0 1 4 2 4 3	columns (total 11 columns): Column Country Happiness score Whisker-high	156 non-null 156 non-null 156 non-null	object
#  0 1 4 2 4 3 4	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low	156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 4 5	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7	columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual Explained by: GDP per capita	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8	Columns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low Dystopia (1.92) + residual Explained by: GDP per capita Explained by: Social support	156 non-null	object float6 float6 float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9 4 10	Column Co	156 non-null	object float6 float6 float6 float6 float6 float6
# 0 1 4 2 4 3 4 4 5 4 6 4 7 4 8 4 9 4 10	Column Co	156 non-null	object float6

Please load the data11.txt file, which is a fixed width file. The columns are defined as follows:

Variable	Width	Start	End
Country	24	1	24
Happiness score	5	25	29
Whisker-high	5	30	34
Whisker-low	5	35	39
Dystopia (1.92) + residual	5	40	44
Explained by: GDP per capita	5	45	49
Explained by: Social support	5	50	54
Explained by: Healthy life expectancy	5	55	59
Explained by: Freedom to make life choices	5	60	64
Explained by: Generosity	5	65	69
Explained by: Perceptions of corruption	5	70	74

Then save the this loaded data frame as a CSV file on your local machine. Be sure to use a unique filename so as not to overwrite any existing files. (5 points)

# Answer 11: this fixed width text file required widths of the columns that were provided manually via widths list; also names for the columns were provided in column names.

```
In [142]: widths = [24, 5, 5, 5, 5, 5, 5, 5, 5, 5, 5]
data11 = pd.read_fwf('data11.txt', widths=widths, names=column_names)
data11.head(3)
```

Out[142]:

	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	Freedo to mal li choice
0	Finland	7.632	7.695	7.569	2.595	1.305	1.592	0.874	0.68
1	Norway	7.594	7.657	7.530	2.383	1.456	1.582	0.861	0.68
2	Denmark	7.555	7.623	7.487	2.370	1.351	1.590	0.868	0.68

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In [143]: data11.info() == data\_clean.info()

#	columns (total 11 columns): Column	Non-Null Count	Dtype
0	Country	156 non-null	object
1 4	Happiness score	156 non-null	float6
4 2 4	Whisker-high	156 non-null	float6
3 4	Whisker-low	156 non-null	float6
4 4	Dystopia (1.92) + residual	156 non-null	float6
5 4	Explained by: GDP per capita	156 non-null	float6
6 4	Explained by: Social support	156 non-null	float6
7 4	Explained by: Healthy life expectancy	156 non-null	float6
8 4	Explained by: Freedom to make life choices	156 non-null	float6
9	Explained by: Generosity	156 non-null	float6
4 10 4	Explained by: Perceptions of corruption	156 non-null	float6
dtyp memo	es: float64(10), object(1) ery usage: 13.5+ KB		
dtyp memo <cla Rang</cla 			
dtyp memo <cla Rang</cla 	ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155	Non-Null Count	
dtypmemc <class< td=""><td>ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 columns (total 11 columns):</td><td>Non-Null Count 156 non-null</td><td></td></class<>	ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 columns (total 11 columns):	Non-Null Count 156 non-null	
dtypmemc <class< td=""><td>ery usage: 13.5+ KB ess 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 ecolumns (total 11 columns): Column</td><td></td><td> object</td></class<>	ery usage: 13.5+ KB ess 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 ecolumns (total 11 columns): Column		 object
dtypmemc <class< td=""><td>ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 columns (total 11 columns): Column Country</td><td>156 non-null</td><td> object</td></class<>	ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 columns (total 11 columns): Column Country	156 non-null	 object
dtypmemc <class< td=""><td>ery usage: 13.5+ KB ess 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 columns (total 11 columns): Column Country Happiness score</td><td>156 non-null 156 non-null</td><td>object</td></class<>	ery usage: 13.5+ KB ess 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 columns (total 11 columns): Column Country Happiness score	156 non-null 156 non-null	object
dtypmemc <class< td=""><td>ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'&gt; .eIndex: 156 entries, 0 to 155 .columns (total 11 columns): .Column Country Happiness score  Whisker-high</td><td>156 non-null 156 non-null 156 non-null</td><td>object float6</td></class<>	ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'> .eIndex: 156 entries, 0 to 155 .columns (total 11 columns): .Column Country Happiness score  Whisker-high	156 non-null 156 non-null 156 non-null	object float6
dtypmemc <class< td=""><td>ery usage: 13.5+ KB ess 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 ecolumns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low</td><td>156 non-null 156 non-null 156 non-null 156 non-null</td><td>object float6</td></class<>	ery usage: 13.5+ KB ess 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 ecolumns (total 11 columns): Column Country Happiness score Whisker-high Whisker-low	156 non-null 156 non-null 156 non-null 156 non-null	object float6
dtypmemc <class< td=""><td>ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'&gt; .eIndex: 156 entries, 0 to 155 .columns (total 11 columns): .Column Country Happiness score  Whisker-high  Whisker-low  Dystopia (1.92) + residual</td><td>156 non-null 156 non-null 156 non-null 156 non-null 156 non-null</td><td>object float6 float6 float6 float6</td></class<>	ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'> .eIndex: 156 entries, 0 to 155 .columns (total 11 columns): .Column Country Happiness score  Whisker-high  Whisker-low  Dystopia (1.92) + residual	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6
dtypmemc <clarker #="" 0="" 1="" 2="" 3="" 4="" 5="" 7<="" <clarker="" data="" rang="" td=""><td>ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 .columns (total 11 columns): Column Country Happiness score  Whisker-high Whisker-low  Dystopia (1.92) + residual  Explained by: GDP per capita</td><td>156 non-null 156 non-null 156 non-null 156 non-null 156 non-null 156 non-null</td><td>object float6 float6 float6 float6</td></clarker>	ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 .columns (total 11 columns): Column Country Happiness score  Whisker-high Whisker-low  Dystopia (1.92) + residual  Explained by: GDP per capita	156 non-null 156 non-null 156 non-null 156 non-null 156 non-null 156 non-null	object float6 float6 float6 float6
dtypmemc <class< td=""><td>ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'&gt; .eIndex: 156 entries, 0 to 155 .columns (total 11 columns): .Column Country Happiness score  Whisker-high  Whisker-low  Dystopia (1.92) + residual  Explained by: GDP per capita  Explained by: Social support</td><td>156 non-null 156 non-null</td><td>object float6 float6 float6 float6 float6 float6</td></class<>	ry usage: 13.5+ KB .ss 'pandas.core.frame.DataFrame'> .eIndex: 156 entries, 0 to 155 .columns (total 11 columns): .Column Country Happiness score  Whisker-high  Whisker-low  Dystopia (1.92) + residual  Explained by: GDP per capita  Explained by: Social support	156 non-null	object float6 float6 float6 float6 float6 float6
dtypmemc <class< td=""><td>ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'&gt; eIndex: 156 entries, 0 to 155 columns (total 11 columns): Column 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</td><td>156 non-null 156 non-null</td><td>object float6 float6 float6 float6 float6 float6</td></class<>	ry usage: 13.5+ KB ss 'pandas.core.frame.DataFrame'> eIndex: 156 entries, 0 to 155 columns (total 11 columns): Column 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	156 non-null	object float6 float6 float6 float6 float6 float6

Out[143]:	True
In [ ]:	
In [ ]:	