Module 4: Cleaning data with pandas

Weekly Assignment 4B

Brandan Owens and Loan Pham

Q.1 Download the dataset "adult_income_data.csv"

```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from functools import reduce
from fractions import Fraction
```

In [2]:

(a) read in the dataset with pd.read_csv. Show the first 8 records, what's wrong with
income_data = pd.read_csv("../dataFiles/adult_income_data.csv")
income_data.head(8)

Out[2]:

2]:		39	State- gov	77516	Bachelors	13	Never- married	Adm- clerical	Not-in- family	Male	2174	0	40	United- States	<
	0	50	Self- emp- not- inc	83311	Bachelors	13	Married- civ- spouse	Exec- managerial	Husband	Male	0	0	13	United- States	
	1	38	Private	215646	HS-grad	9	Divorced	Handlers- cleaners	Not-in- family	Male	0	0	40	United- States	,
	2	53	Private	234721	11th	7	Married- civ- spouse	Handlers- cleaners	Husband	Male	0	0	40	United- States	
	3	28	Private	338409	Bachelors	13	Married- civ- spouse	Prof- specialty	Wife	Female	0	0	40	Cuba	•
	4	37	Private	284582	Masters	14	Married- civ- spouse	Exec- managerial	Wife	Female	0	0	40	United- States	•
	5	49	Private	160187	9th	5	Married- spouse- absent	Other- service	Not-in- family	Female	0	0	16	Jamaica	•
	6	52	Self- emp- not- inc	209642	HS-grad	9	Married- civ- spouse	Exec- managerial	Husband	Male	0	0	45	United- States	
	7	31	Private	45781	Masters	14	Never- married	Prof- specialty	Not-in- family	Female	14084	0	50	United- States	

(b) read the dataset again and bear in mind that the dataset has no header (use heade
income_data = pd.read_csv("../dataFiles/adult_income_data.csv", header = None)
income_data.head()

Out[3]:		0	1	2	3	4	5	6	7	8	9	10	11	12	
	0	39	State- gov	77516	Bachelors	13	Never- married	Adm- clerical	Not-in- family	Male	2174	0	40	United- States	<=
	1	50	Self- emp- not- inc	83311	Bachelors	13	Married- civ- spouse	Exec- managerial	Husband	Male	0	0	13	United- States	<=
	2	38	Private	215646	HS-grad	9	Divorced	Handlers- cleaners	Not-in- family	Male	0	0	40	United- States	<=
	3	53	Private	234721	11th	7	Married- civ- spouse	Handlers- cleaners	Husband	Male	0	0	40	United- States	<=
	4	28	Private	338409	Bachelors	13	Married- civ- spouse	Prof- specialty	Wife	Female	0	0	40	Cuba	<=

```
In [4]:
# (c) name the columns. Refer to the "adult_income_names.txt"
names=[]
with open("../dataFiles/adult_income_names.txt","r") as f:
    for line in f:
        f.readline()
        var = line.split(":")[0]
        names.append(var)
names
income_data.columns = names
income_data.head()
```

Out[4]:

	age	workclass	fnlwgt	education	education- num	marital- status	occupation	relationship	sex	capital- gain
0	39	State-gov	77516	Bachelors	13	Never- married	Adm- clerical	Not-in- family	Male	2174
1	50	Self-emp- not-inc	83311	Bachelors	13	Married- civ- spouse	Exec- managerial	Husband	Male	0
2	38	Private	215646	HS-grad	9	Divorced	Handlers- cleaners	Not-in- family	Male	0
3	53	Private	234721	11th	7	Married- civ- spouse	Handlers- cleaners	Husband	Male	0

	age	workclass	fnlwgt	education	education- num	marital- status	occupation	relationship	sex	capital- gain
4	28	Private	338409	Bachelors	13	Married- civ- spouse	Prof- specialty	Wife	Female	0

In [5]: # (d) get the statistical summary of the dataset.
income_data.describe()

Out[5]:		age	fnlwgt	education-num	capital-gain	capital-loss	hours-per-week
	count	32561.000000	3.256100e+04	32561.000000	32561.000000	32561.000000	32561.000000
	mean	38.581647	1.897784e+05	10.080679	1077.648844	87.303830	40.437456
	std	13.640433	1.055500e+05	2.572720	7385.292085	402.960219	12.347429
	min	17.000000	1.228500e+04	1.000000	0.000000	0.000000	1.000000
	25%	28.000000	1.178270e+05	9.000000	0.000000	0.000000	40.000000
	50%	37.000000	1.783560e+05	10.000000	0.000000	0.000000	40.000000
	75%	48.000000	2.370510e+05	12.000000	0.000000	0.000000	45.000000
	max	90.000000	1.484705e+06	16.000000	99999.000000	4356.000000	99.000000

(e) Create a DataFrame with only age, education, and occupation
income_data_filtered = income_data[['age','education','occupation']]
income_data_filtered.head()

Out[6]:		age	education	occupation
	0	39	Bachelors	Adm-clerical
	1	50	Bachelors	Exec-managerial
	2	38	HS-grad	Handlers-cleaners
	3	53	11th	Handlers-cleaners
	4	28	Bachelors	Prof-specialty

(g) Find the number of people who are aged between 30 and 50
income_data_filtered = income_data_filtered[income_data['age'].between(30, 50)]
len(income_data_filtered.index) #len(data.index) is fastest counter

Out[7]: 16390

In [8]: # (h) Find the number of people who are aged between 30 and 50, and made positive capit

```
income_data_filtered = income_data[['age','education','occupation','capital-gain']]
income_data_filtered = income_data_filtered[(income_data['age'] >= 30) & (income_data_fincome_data_filtered.head()
```

```
Out[8]:
                   education
                                   occupation capital-gain
              age
           0
               39
                    Bachelors
                                  Adm-clerical
                                                    2174
           8
               31
                     Masters
                                  Prof-specialty
                                                    14084
           9
               42
                    Bachelors
                               Exec-managerial
                                                    5178
          59
               30
                     HS-grad Machine-op-inspct
                                                    5013
          60
               30
                    Bachelors
                                        Sales
                                                    2407
 In [9]:
           # (i) group by "occupation", show the median age of each group.
           grouped = income_data.groupby('occupation')['age'].median()
           grouped
 Out[9]: occupation
                                 35
           Adm-clerical
                                 35
           Armed-Forces
                                 29
           Craft-repair
                                 38
           Exec-managerial
                                 41
           Farming-fishing
                                 39
           Handlers-cleaners
                                 29
           Machine-op-inspct
                                 36
           Other-service
                                 32
           Priv-house-serv
                                 40
           Prof-specialty
                                 40
           Protective-serv
                                 36
                                 35
           Sales
           Tech-support
                                 36
           Transport-moving
                                 39
          Name: age, dtype: int64
In [10]:
           # (j) use the following codes to create two dataframes
           df1 = income_data[['age', 'workclass', 'occupation']].sample(5,random_state=101)
           df2 = income_data[['education','occupation']].sample(5,random_state=101)
```

```
# (j) use the following codes to create two dataframes
df1 = income_data[['age', 'workclass', 'occupation']].sample(5,random_state=101)
df2 = income_data[['education','occupation']].sample(5,random_state=101)
# merge df1 with df2. The key is 'occupation'
occupation_data = pd.merge(df1,df2,on='occupation')
display(occupation_data)
```

education	occupation	workclass	age	
HS-grad	Machine-op-inspct	Private	51	0
11th	Sales	Private	19	1
HS-grad	Exec-managerial	Private	40	2
10th	Handlers-cleaners	Private	17	3
7th-8th	Craft-repair	Private	61	4

```
In [11]: #Q.2.a Read in dataset 'Energy Indicators.xls'. Exclude the footer & header info, the f
    header_names = ['toDelete','Country Name','Energy Supply','Energy Supply per Capita','%
    energy_data = pd.read_excel("../dataFiles/Energy Indicators.xls", skipfooter=246, skipr
    energy_data.drop(columns='toDelete', inplace=True)
    display(energy_data)
```

Country Name	Energy Supply	Energy Supply per Capita	% Renewable
Afghanistan	321	10	78.669280
Albania	102	35	100.000000
Algeria	1959	51	0.551010
American Samoa			0.641026
Andorra	9	121	88.695650
Angola	642	27	70.909090
Anguilla	2	136	0.000000
Antigua and Barbuda	8	84	0.000000
Argentina	3378	79	24.064520
Armenia	143	48	28.236060
Aruba	12	120	14.870690
Australia1	5386	231	11.810810
Austria	1391	164	72.452820
Azerbaijan	567	60	6.384345
Bahamas	45	118	0.000000
Bahrain	574	425	0.000000
Bangladesh	1625	10	1.966329
Barbados	19	69	0.000000
Belarus	1142	120	0.463389

```
In [12]:
#Q.2.a.continue Rename the following countries ["Republic of Korea":"South Korea", "Unit
replacement_names = {
    "Republic of Korea":"South Korea",
    "Australia1":"Australia",
    "United States of America":"United States",
    "United Kingdom of Great Britain and Northern Ireland":"United Kingdom",
    "China, Hong Kong Special Administrative Region":"Hong Kong",
    "Bolivia (Plurinational State of)":"Bolivia",
    "Switzerland17":"Switzerland"
    }
    energy_data["Country Name"].replace(replacement_names, inplace=True)
display(energy_data)
```

```
Country NameEnergy SupplyEnergy Supply per Capita% RenewableAfghanistan3211078.669280
```

Country Name	Energy Supply	Energy Supply per Capita	% Renewable
Albania	102	35	100.000000
Algeria	1959	51	0.551010
American Samoa			0.641026
Andorra	9	121	88.695650
Angola	642	27	70.909090
Anguilla	2	136	0.000000
Antigua and Barbuda	8	84	0.000000
Argentina	3378	79	24.064520
Armenia	143	48	28.236060
Aruba	12	120	14.870690
Australia	5386	231	11.810810
Austria	1391	164	72.452820
Azerbaijan	567	60	6.384345
Bahamas	45	118	0.000000
Bahrain	574	425	0.000000
Bangladesh	1625	10	1.966329
Barbados	19	69	0.000000
Belarus	1142	120	0.463389

```
In [13]:
#Q.2.b Load the GDP Data from "world_bank.xlsx". Exclude the header, rename the followi
replacement_names = {
    "Korea, Rep.":"South Korea",
    "Iran, Islamic Rep.":"Iran",
    "Hong Kong SAR, China":"Hong Kong"
}
gdp_data = pd.read_excel("../dataFiles/world_bank.xlsx", skiprows=2, header=2, keep_def
gdp_data.rename
gdp_data["Country Name"].replace(replacement_names, inplace=True)
display(gdp_data)
```

	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	,
0	Aruba	ABW	GDP (current US\$)	NY.GDP.MKTP.CD			
1	Afghanistan	AFG	GDP (current US\$)	NY.GDP.MKTP.CD	537777811111111	54888889555556	54666667777
2	Angola	AGO	GDP (current US\$)	NY.GDP.MKTP.CD			

	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	
3	Albania	ALB	GDP (current US\$)	NY.GDP.MKTP.CD			
4	Andorra	AND	GDP (current US\$)	NY.GDP.MKTP.CD			
•••							
259	Kosovo	XKX	GDP (current US\$)	NY.GDP.MKTP.CD			
260	Yemen, Rep.	YEM	GDP (current US\$)	NY.GDP.MKTP.CD			
261	South Africa	ZAF	GDP (current US\$)	NY.GDP.MKTP.CD	75752484950301	797284054318914	84978300433
262	Zambia	ZMB	GDP (current US\$)	NY.GDP.MKTP.CD	713000000	696285714285714	69314285714
263	Zimbabwe	ZWE	GDP (current US\$)	NY.GDP.MKTP.CD	1052990400	1096646600	111760

264 rows × 62 columns

In [14]:

```
#Q.2.c Load the dataset scimagojr.xlsx
header_names=["Rank", "Country Name", "Documents", "Citable Documents", "Citations", "S
scimagojr = pd.read_excel("../dataFiles/scimagojr.xlsx", names=header_names)
display(scimagojr)
```

	Rank	Country Name	Documents	Citable Documents	Citations	Self- Citations	Citations per document	H index
0	1	China	147887	147512	856806	583858	5.79	162
1	2	United States	113579	111426	1085684	370574	9.56	259
2	3	Japan	34294	34054	275980	73491	8.05	145
3	4	United Kingdom	24328	23671	278694	52119	11.46	159
4	5	India	21450	21183	179494	54929	8.37	132
•••	•••							•••
187	188	Reunion	1	1	4	1	4.00	1
188	189	Saint Helena	1	1	36	0	36.00	1

	Rank	Country Name	Documents	Citable Documents	Citations	Self- Citations	Citations per document	H index
189	190	American Samoa	1	1	0	0	0.00	0
190	191	Belize	1	1	9	0	9.00	1
191	192	British Indian Ocean Territory	1	1	45	0	45.00	1

192 rows × 8 columns

In [15]:

```
#Q.2.d Join the three datasets, retain only the last 10 years of GDP data and only the
```

```
data_final = pd.merge(scimagojr, energy_data, on="Country Name")
data_final = pd.merge(data_final, gdp_data, on="Country Name")
years = list(range(1960,2006,1))
data_final.drop(labels=years,axis=1,inplace=True)
display(data_final)
```

	Rank	Country Name	Documents	Citable Documents	Citations	Self- Citations	Citations per document	H index	Energy Supply	Energy Supply per Capita
0	14	Australia	10616	10496	129788	22759	12.23	123	5386	231
1	38	Austria	2133	2102	19946	2234	9.35	61	1391	164
2	40	Algeria	1645	1626	11257	2286	6.84	46	1959	51
3	43	Argentina	1389	1379	14060	2050	10.12	52	3378	79
4	63	Bangladesh	623	617	2895	309	4.65	26	1625	10
5	70	Azerbaijan	392	385	655	145	1.67	11	567	60
6	72	Belarus	340	334	1297	251	3.81	17	1142	120
7	83	Bahrain	154	150	1279	44	8.31	17	574	425
8	88	Armenia	105	105	648	250	6.17	13	143	48
9	104	Angola	56	56	140	2	2.50	7	642	27
10	120	Albania	27	27	217	14	8.04	7	102	35
11	148	Barbados	7	7	16	3	2.29	3	19	69
12	159	Afghanistan	5	5	5	0	1.00	1	321	10
13	169	Andorra	2	2	15	0	7.50	1	9	121
14	171	Antigua and Barbuda	2	2	0	0	0.00	0	8	84
15	190	American Samoa	1	1	0	0	0.00	0		

16 rows × 26 columns

```
In [16]:
```

```
#Q.2.e What is the average GDP over the Last 10 years for each country
columns = [2007,2008,2009,2010,2011,2012,2013,2014,2015,2016,2017]
data_final['Average GDP over 10 years'] = round(data_final.mean(axis=1),1)
display(data_final)
```

	Rank	Country Name	Documents	Citable Documents	Citations	Self- Citations	Citations per document	H index	Energy Supply	Energy Supply per Capita
0	14	Australia	10616	10496	129788	22759	12.23	123	5386	231
1	38	Austria	2133	2102	19946	2234	9.35	61	1391	164
2	40	Algeria	1645	1626	11257	2286	6.84	46	1959	51
3	43	Argentina	1389	1379	14060	2050	10.12	52	3378	79
4	63	Bangladesh	623	617	2895	309	4.65	26	1625	10
5	70	Azerbaijan	392	385	655	145	1.67	11	567	60
6	72	Belarus	340	334	1297	251	3.81	17	1142	120
7	83	Bahrain	154	150	1279	44	8.31	17	574	425
8	88	Armenia	105	105	648	250	6.17	13	143	48
9	104	Angola	56	56	140	2	2.50	7	642	27
10	120	Albania	27	27	217	14	8.04	7	102	35
11	148	Barbados	7	7	16	3	2.29	3	19	69
12	159	Afghanistan	5	5	5	0	1.00	1	321	10
13	169	Andorra	2	2	15	0	7.50	1	9	121
14	171	Antigua and Barbuda	2	2	0	0	0.00	0	8	84
15	190	American Samoa	1	1	0	0	0.00	0		

16 rows × 27 columns

```
In [17]:
#Q.2.f By how much had the GDP changed over the 10-year span for the country with the 6
gdp_difference = str(data_final.iloc[5][2016] - data_final.iloc[5][2006])
name = str(data_final.iloc[5]['Country Name'])
print("Country: " + name + " GDP Difference over 10 years: " + gdp_difference)
```

Country: Azerbaijan GDP Difference over 10 years: 168646958121543

```
In [18]:
#Q.2.g Create a new column that is the ration of Self-Citations to Total Citations. Wha
#self citations = data final["Self-Citations"]
```

#total_citations = data_final["Citations"] + self_citations
#zipper = zip(self_citations, total_citations)
#data_final["Citation Ratio"] = [str(Fraction(x,y)).replace('/',':') for x, y in zipper
data_final["Citation Ratio"] = data_final["Self-Citations"]/data_final["Citations"]
data_final.sort_values(by="Citation Ratio", ascending=False, inplace=True)
display(data_final)

	Rank	Country Name	Documents	Citable Documents	Citations	Self- Citations	Citations per document	H index	Energy Supply	Energy Supply per Capita
8	88	Armenia	105	105	648	250	6.17	13	143	48
5	70	Azerbaijan	392	385	655	145	1.67	11	567	60
2	40	Algeria	1645	1626	11257	2286	6.84	46	1959	51
6	72	Belarus	340	334	1297	251	3.81	17	1142	120
11	148	Barbados	7	7	16	3	2.29	3	19	69
0	14	Australia	10616	10496	129788	22759	12.23	123	5386	231
3	43	Argentina	1389	1379	14060	2050	10.12	52	3378	79
1	38	Austria	2133	2102	19946	2234	9.35	61	1391	164
4	63	Bangladesh	623	617	2895	309	4.65	26	1625	10
10	120	Albania	27	27	217	14	8.04	7	102	35
7	83	Bahrain	154	150	1279	44	8.31	17	574	425
9	104	Angola	56	56	140	2	2.50	7	642	27
12	159	Afghanistan	5	5	5	0	1.00	1	321	10
13	169	Andorra	2	2	15	0	7.50	1	9	121
14	171	Antigua and Barbuda	2	2	0	0	0.00	0	8	84
15	190	American Samoa	1	1	0	0	0.00	0		

16 rows × 28 columns