Homework 3: Table Manipulation and Visualization

Reading:

Visualization (https://www.inferentialthinking.com/chapters/07/visualization.html)

Deadline:

This assignment is due Thursday, February 25 at 11:59 P.M. Late work will not be accepted as per the course policies (see the Syllabus and Course policies on Canvas (https://canvas.yale.edu).

Directly sharing answers is not okay, but discussing problems with the course staff or with other students is encouraged. Refer to the policies page to learn more about how to learn cooperatively.

You should start early so that you have time to get help if you're stuck. The drop-in office hours schedule can be found on <u>Canvas (https://canvas.yale.edu)</u>. You can also post questions or start discussions on <u>Ed</u> <u>Discussion (https://edstem.org/us/courses/3558/discussion/)</u>.

Submission:

Submit your assignment as a .pdf on Gradescope. You can access Gradescope through Canvas on the leftside of the class home page. The problems in each homework assignment are numbered. NOTE: When submitting on Gradescope, please select the correct pages of your pdf that correspond to each problem. This will allow graders to find your complete solution to each problem.

To produce the .pdf, please do the following in order to preserve the cell structure of the notebook:

- 1. Go to "File" at the top-left of your Jupyter Notebook
- 2. Under "Download as", select "HTML (.html)"
- 3. After the .html has downloaded, open it and then select "File" and "Print" (note you will not actually be printing)
- 4. From the print window, select the option to save as a .pdf

```
In [ ]:

# Don't change this cell; just run it.

import numpy as np
from datascience import *

# These lines do some fancy plotting magic.\n",
import matplotlib
%matplotlib inline
import matplotlib.pyplot as plots
plots.style.use('fivethirtyeight')
```

1. Differences between Universities

Question 1.1. Suppose you're choosing a university to attend, and you'd like to *quantify* how *dissimilar* any two universities are. You rate each university you're considering on several numerical traits. You decide on a very detailed list of 1000 traits, and you measure all of them! Some examples:

- The cost to attend (per year).
- The average Yelp review of nearby Thai restaurants.
- The USA Today ranking of the Medical school.
- The USA Today ranking of the Engineering school.

You decide that the dissimilarity between two universities is the *total* of the differences in their traits. That is, the dissimilarity is:

- the sum of
- the absolute values of
- the 1000 differences in their trait values.

In the next cell, we've loaded arrays containing the 1000 trait values for Stanford and Berkeley. Compute the dissimilarity (according to the above technique) between Stanford and Berkeley. Call your answer dissimilarity. Use a single line of code to compute the answer.

Note: The data we're using aren't real -- we made them up for this exercise, except for the cost-of-attendance numbers, which were found online.

```
In [ ]:
```

```
stanford = Table.read_table("stanford.csv").column("Trait value")
berkeley = Table.read_table("berkeley.csv").column("Trait value")

dissimilarity = ...
dissimilarity
```

Question 1.2. Why do we sum up the absolute values of the differences in trait values, rather than just summing up the differences?

Write your answer here, replacing this text.

Weighing the traits

After computing dissimilarities between several schools, you notice a problem with your method: the scale of the traits matters a lot.

Since schools cost tens of thousands of dollars to attend, the cost-to-attend trait is always a much bigger *number* than most other traits. That makes it affect the dissimilarity a lot more than other traits. Two schools that differ in cost-to-attend by \$900, but are otherwise identical, get a dissimilarity of 900. But two schools that differ in graduation rate by 0.9 (a huge difference!), but are otherwise identical, get a dissimilarity of only 0.9.

One way to fix this problem is to assign different "weights" to different traits. For example, we could fix the problem above by multiplying the difference in the cost-to-attend traits by .001, so that a difference of \$900 in the attendance cost results in a dissimilarity of $$900 \times .001$, or 0.9.

Here's a revised method that does that for every trait:

- 1. For each trait, subtract the two schools' trait values.
- 2. Then take the absolute value of that difference.
- 3. Now multiply that absolute value by a trait-specific number, like .001 or 2.
- 4. Now, sum the 1000 resulting numbers.

Question 1.3. Suppose you've already decided on a weight for each trait. These are loaded into an array called weights in the cell below. weights.item(0) is the weight for the first trait, weights.item(1) is the weight for the second trait, and so on. Use the revised method to compute a revised dissimilarity between Berkeley and Stanford.

Hint: Using array arithmetic, your answer should be almost as short as in question 1.

```
In [ ]:
weights = Table.read_table("weights.csv").column("Weight")
revised_dissimilarity = ...
revised_dissimilarity
```

2. Unemployment

The Federal Reserve Bank of St. Louis publishes data about jobs in the US. Below, we've loaded data on unemployment in the United States. There are many ways of defining unemployment, and our dataset includes two notions of the unemployment rate:

- 1. Among people who are able to work and are looking for a full-time job, the percentage who can't find a job. This is called the Non-Employment Index, or NEI.
- 2. Among people who are able to work and are looking for a full-time job, the percentage who can't find any job *or* are only working at a part-time job. The latter group is called "Part-Time for Economic Reasons", so the acronym for this index is NEI-PTER. (Economists are great at marketing.)

The source of the data is here (https://fred.stlouisfed.org/categories/33509).

Question 2.1. The data are in a CSV file called unemployment.csv. Load that file into a table called unemployment.

```
In [ ]:
unemployment = ...
unemployment
```

Question 2.2. Sort the data in descending order by NEI, naming the sorted table by_nei. Create another table called by nei pter that's sorted in descending order by NEI-PTER instead.

```
In [ ]:
by_nei = ...
by_nei_pter = ...
```

Question 2.3. Use take to make a table containing the data for the 10 quarters when NEI was greatest. Call that table <code>greatest_nei</code>.

```
In [ ]:
greatest_nei = ...
greatest_nei
```

Question 2.4. It's believed that many people became PTER (recall: "Part-Time for Economic Reasons") in the "Great Recession" of 2008-2009. NEI-PTER is the percentage of people who are unemployed (and counted in the NEI) plus the percentage of people who are PTER. Compute an array containing the percentage of people who were PTER in each quarter. (The first element of the array should correspond to the first row of unemployment, and so on.)

Note: Use the original unemployment table for this.

```
In [ ]:

pter = ...
pter
```

Question 2.5. Add pter as a column to unemployment (named "PTER") and sort the resulting table by that column in descending order. Call the table by _pter .

Try to do this with a single line of code, if you can.

```
In [ ]:
by_pter = ...
by_pter
```

Question 2.6. Create a line plot of the PTER over time. To do this, first add the year array and the pter array to the unemployment table; label these columns "Year" and "PTER", respectively. Then, generate a line plot using one of the table methods you've learned in class. Assign this new table to pter over time.

```
In [ ]:

year = 1994 + np.arange(by_pter.num_rows)/4

pter_over_time = ...
...
```

Question 2.7. Were PTER rates high during or directly after the Great Recession (that is to say, were PTER rates particularly high in the years 2008 through 2011)? Assign highPTER to True if you think PTER rates were high in this period, and False if you think they weren't.

```
In [ ]:
highPTER = ...
```

3. Birth Rates

The following table gives census-based population estimates for each state on both July 1, 2015 and July 1, 2016. The last four columns describe the components of the estimated change in population during this time interval. For all questions below, assume that the word "states" refers to all 52 rows including Puerto Rico & the District of Columbia.

The data was taken from http://www2.census.gov/programs-surveys/popest/datasets/2010-2016/national/totals/nst-est2016-alldata.csv).

If you want to read more about the different column descriptions, go http://www2.census.gov/programs-surveys/popest/datasets/2010-2015/national/totals/nst-est2015-alldata.pdf)! As of February 2017, no descriptions were posted for 2010 - 2016.

```
In [ ]:
```

```
# Don't change this cell; just run it.
pop = Table.read_table('nst-est2016-alldata.csv').where('SUMLEV', 40).select([1, 4, 12, 13, 27, 34, 62, 69])
pop = pop.relabeled('POPESTIMATE2015', '2015').relabeled('POPESTIMATE2016', '2016')
pop = pop.relabeled('BIRTHS2016', 'BIRTHS').relabeled('DEATHS2016', 'DEATHS')
pop = pop.relabeled('NETMIG2016', 'MIGRATION').relabeled('RESIDUAL2016', 'OTHER')
pop.set_format([2, 3, 4, 5, 6, 7], NumberFormatter(decimals=0)).show(5)
```

Question 3.1. Assign us_birth_rate to the total US annual birth rate during this time interval. The annual birth rate for a year-long period is the total number of births in that period as a proportion of the population size at the start of the time period.

Hint: What year corresponds to the start of the time period?

```
In [ ]:

us_birth_rate = ...
us_birth_rate
```

Question 3.2. Assign fastest_growth to an array of the names of the five states with the fastest population growth rates in *descending order of growth rate*. We have first created a new version of the pop table, called growth_rates, which includes a column with the growth rate of each state. Making intermediate tables can improve the readability of the code and make it easier to follow when revisting at a later time.

```
In [ ]:
```

```
growth_rates = pop.with_column('Growth Rate', (pop.column(3) / pop.column(2)) -
1)
fastest_growth = ...
fastest_growth
```

Question 3.3. Assign movers to the number of states for which the **absolute value** of the **annual rate of migration** was higher than 1%. The annual rate of migration for a year-long period is the net number of migrations (in and out) as a proportion of the population size at the start of the period. The MIGRATION column contains estimated annual net migration counts by state.

```
In [ ]:
```

```
migration_rates = ...
movers = ...
movers
```

Question 3.4. Assign west_births to the total number of births that occurred in region 4 (the Western US).

Hint: Make sure you double check the type of the values in the region column.

```
In [ ]:
```

```
west_births =
west_births
```

Question 3.5. Assign less_than_west_births to the number of states that had a total population in 2016 that was smaller than the *total number of births in region 4 (the Western US)* during this time interval.

```
In [ ]:
```

```
less_than_west_births = ...
less_than_west_births
```

Question 3.6.

In the code cell below, create a visualization that will help us determine if there is an association between birth rate and death rate during this time interval. It may be helpful to create an intermediate table here.

```
In [ ]:
```

```
# Generate your chart in this cell ...
```

Question 3.7. True or False: There appears to be a negative association between birth rate and death rate during this time interval.

Assign assoc to True or False in the cell below.

```
In [ ]:

assoc = ...
```

4. Marginal Histograms

Consider the following scatter plot:

The axes of the plot represent values of two variables: x and y.

Suppose we have a table called t that has two columns in it:

- x: a column containing the x-values of the points in the scatter plot
- y: a column containing the y-values of the points in the scatter plot

Question 4.1: Match each of the following histograms to the code that produced them. Explain your reasoning.

Histogram A:

Histogram B:

```
Line 1: t.hist('x')
```

Histogram for Line 1: ...

Explanation:...

```
Line 2: t.hist('y')
```

Histogram for Line 2:...

Explanation:...

5. Submission

then print to save as a .pdf) on the class Canvas site.	
<pre>In []:</pre>	

Once you're finished, submit your assignment as a .ipynb (Jupyter Notebook) and .pdf (download as .html,