Intro-to-Python-and-Jupyter-Instructor

June 28, 2022

1 Introduction to Data Science in Python

1.1 WICS - SYP 22

Notebook adapted from UC Berkeley's Data8 and Data6 course materials

1.1.1 Table of Contents

Welcome to Jupyter Notebooks!

- 1. The Python Programming Language
 - a. Expressions and Errors
 - b. Names
 - c. Functions
- 2. Tables
 - a. Attributes
 - b. Transformations
 - c. Sorting Tables

1.2 The Jupyter Notebook

Welcome to the Jupyter Notebook! **Notebooks** are documents that can contain text, code, visualizations, and more. Jupyter notebooks support over 40 different programming languages, including Python, R, Julia, Scala, etc.

Notebooks are used for *literate programming*, a programming paradigm introduced by Donald Knuth in 1984, in which a programming language is accompanied with a documentation language, or a natural language. In other words, the computer program has an explanation in a natural language. This approach to programming effectively treats software as works of literature (Knuth, "Literate Programming"). It supports people to have a strong conceptual map of what is happening in the code and also have clarity on the flow and logic of the code/program. which is helpful for both for the writer and the reader.

Jupyter leverages this idea and enables users to create and share documents that combine code, visualizations, narrative text, equations, and rich media. Notebooks are multipurpose and can be used in any discipline. The notebook is like a laboratory notebook, but for computing. Researchers can write code to work with their data while supplementing their methods with explanations, analysis, or hypotheses. Notebooks are also used in education because they enable students to engage with content presented in different forms, , experience computation with no prior experience, and practice programming in a scaffolded way.

In this session, we'll be using Jupyter Notebooks to introduce you to how data scientists work with data, to learn about issues of justice using real-world data sets, and to also learn how to reason about the human choices embedded in the practice of data science and their significance.

1.2.1 Notebook Structure

A notebook is composed of rectangular sections called **cells**. There are 2 kinds of cells: markdown and code. A **markdown cell**, such as this one, contains text. A **code cell** contains code in Python, a programming language that we will be using for the remainder of this module. You can select any cell by clicking it once. After a cell is selected, you can navigate the notebook using the up and down arrow keys.

To run the code in a code cell, first click on that cell to activate it. It'll be highlighted with a little green or blue rectangle. Next, you can either press the | Run button above or press Shift + Return or Shift + Enter. This will run the current cell and select the next one.

If a code cell is running, you will see an asterisk (*) appear in the square brackets to the left of the cell. Once the cell has finished running, a number will replace the asterisk and any output from the code will appear under the cell.

```
[1]: # run this cell
print("Hello World!")
```

Hello World!

You'll notice that many code cells contain lines of blue text that start with a #. These are *comments*. Comments often contain helpful information about what the code does or what you are supposed to do in the cell. The leading # tells the computer to ignore them.

Editing You can edit a Markdown cell by clicking it twice. Text in Markdown cells is written in **Markdown**, a formatting syntax for plain text, so you may see some funky symbols when you edit a text cell.

Once you've made your changes, you can exit text editing mode by running the cell.

PRACTICE:

Try double-clicking on the text below to edit the cell to fix the misspelling.

Go Michgian Tech Huskies!

Code cells can be edited any time after they are highlighted.

PRACTICE:

Try editing the next code cell to print your name.

```
[2]: # edit the code to print your name print("Hello: my name is NAME")
```

Hello: my name is NAME

Adding Cells You can add a cell by clicking Insert > Insert Cell Below and then choosing the cell type in the drop down menu.

PRACTICE:

Try adding a cell below here and printing your birthday (format: mm/dd/yyyy). Do not forget the quotation marks!

You can add cells by pressing the plus sign icon in the menu bar. This will add (by default) a code cell immediately below your current highlighted cell.

To convert a cell to markdown, you can press 'Cell' in the menu bar, select 'Cell Type', and finally pick the desired option. This works the other way around too! (you can also use the drop down menu in the toolbar above)

Deleting Cells You can delete a cell by clicking the scissors at the top menu or Edit > Cut Cells. Delete the next cell below here.

A common fear is deleting a cell that you needed – but don't worry! This can be undone using 'Edit' > 'Undo Delete Cells'! If you accidentally delete content in a cell, you can use Ctrl + Z to undo.

PRACTICE:

Delete the code cell below.

```
[4]: ## DELETE THIS CELL print("delete me")
```

delete me

The cell above was deleted, do not delete any of the following cells.

Saving and Loading Your notebook can record all of your text and code edits, as well as any graphs you generate or calculations you make.

You are accessing the Jupyter notebook using "Binder", which allows hosted notebooks (from a url or github repository) to be shared as an interactive environment.

Therefore, if you close this page and relaunch the notebook from the website you will lose your changes.

Alternatively, if you access the notebook from a local computer, you can save the notebook as you make changes by clicking Control-S, clicking the floppy disc icon in the toolbar at the top of the page, or by going to the File menu and selecting "Save and Checkpoint".

Note: after loading a notebook you will see all the outputs (graphs, computations, etc) from your last session, but you won't be able to use any variables you assigned or functions you defined. You can get the functions and variables back by re-running the cells where they were defined- the easiest way is to highlight the cell where you left off work, then go to the Cell menu at the top of the screen and click "Run all above". You can also use this menu to run all cells in the notebook by clicking "Run all".

Downloading as PDF You can download this notebook as a pdf by clicking File > Download as > PDF via LaTeX. Try doing this now.

Downloading the Notebook You can also download the notebook, .ipynb extension, to save a copy with your changes to your local machine with File > Download as > Notebook (.ipynb)

Completing the Notebooks As you navigate the notebooks, you'll see cells with bold, all-capitalized headings that need to be filled in to complete the notebook.

Before we begin, we'll need a few extra tools to conduct our analysis. Run the next cell to load some code packages that we'll use later.

Note: this cell MUST be run in order for most of the rest of the notebook to work.

```
[5]: # dependencies: THIS CELL MUST BE RUN
from datascience import *
import numpy as np
import math
import matplotlib.pyplot as plt
plt.style.use('fivethirtyeight')
import ipywidgets as widgets
%matplotlib inline
```

2 1. Python

Python is programming language- a way for us to communicate with the computer and give it instructions.

Just like any language, Python has a *vocabulary* made up of words it can understand, and a *syntax* giving the rules for how to structure communication.

Errors Python is a language, and like natural human languages, it has rules. It differs from natural language in two important ways: 1. The rules are *simple*. You can learn most of them in a few weeks and gain reasonable proficiency with the language in a semester. 2. The rules are *rigid*. If you're proficient in a natural language, you can understand a non-proficient speaker, glossing over small mistakes. A computer running Python code is not smart enough to do that.

Whenever you write code, you will often accientally break some of these rules. When you run a code cell that doesn't follow every rule exactly, Python will produce an **error message**.

Errors are *normal*; experienced programmers make many errors every day. Errors are also *not dangerous*; you will not break your computer by making an error (in fact, errors are a big part of how you learn a coding language). An error is nothing more than a message from the computer saying it doesn't understand you and asking you to rewrite your command.

We have made an error in the next cell. Run it and see what happens.

```
[6]: print("This line is not missing something."
```

```
File "<ipython-input-6-1a13539de791>", line 1 print("This line is not missing something."
```

SyntaxError: unexpected EOF while parsing

You should see something like this (minus our annotations):

The last line of the error output attempts to tell you what went wrong. The *syntax* of a language is its structure, and this **SyntaxError** tells you that you have created an illegal structure. "EOF" means "end of file," so the message is saying Python expected you to write something more (in this case, a right parenthesis) before finishing the cell.

There's a lot of terminology in programming languages, but you don't need to know it all in order to program effectively. If you see a cryptic message like this, you can often get by without deciphering it. (Of course, if you're frustrated, you can usually find out by searching for the error message online or posting on the Piazza.)

2.0.1 1a. Data

Data is information- the "stuff" we manipulate to make and test hypotheses.

Almost all data you will work with broadly falls into two types: numbers and text. *Numerical data* shows up green in code cells and can be positive, negative, or include a decimal.

```
[7]: # Numerical data
4
87623000983
-667
3.14159
```

[7]: 3.14159

Text data (also called *strings*) shows up red in code cells. Strings are enclosed in double or single quotes. Note that numbers can appear in strings.

```
[8]: # Strings
"a"

"Hi there!"

"We hold these truths to be self-evident that all men are created equal."

# this is a string, NOT numerical data
"3.14159"
```

[8]: '3.14159'

2.0.2 1a. Expressions

A bit of communication in Python is called an **expression**. It tells the computer what to do with the data we give it.

Here's an example of an expression. In general when you are writing expressions, you want to have a space between the number and operator. If it is a call expression (these use parentheses), then you do not have a space before the left end of the parentheses

```
[9]: # an expression
14 + 20
```

[9]: 34

When you run the cell, the computer **evaluates** the expression and prints the result. Note that only the last line in a code cell will be printed, unless you explicitly tell the computer you want to print the result.

```
[10]: # more expressions. what gets printed and what doesn't?
    100 / 10

print(4.3 + 10.98)
33 - 9 * (40000 + 1)
884
```

15.28000000000001

[10]: 884

Many basic arithmetic operations are built in to Python, like * (multiplication), + (addition), - (subtraction), and / (division). There are many others, which you can find information about here. Links to open source textbook used in an Introductory Data Science course

The computer evaluates arithmetic according to the PEMDAS order of operations (just like you probably learned in a math class): anything in parentheses is done first, followed by exponents, then multiplication and division, and finally addition and subtraction.

[11]: -2.0

PRACTICE: If you're new to python and coding, one of the best ways to get comfortable is to practice. Try writing and running different expressions in the cell below using numbers and the arithmetic operators * (multiplication), + (addition), - (subtraction), and / (division). See if you can generate different error messages and figure out what they mean.

```
[13]: # Optional: try out different arithmetic operations
# EXAMPLE SOLUTION
2 * ((404 / 2)**0.5)**2 * 5 + 1
```

[13]: 2021.0

2.0.3 1b. Names

Sometimes, the values you work with can get cumbersome- maybe the expression that gives the value is very complicated, or maybe the value itself is long. In these cases it's useful to give the value a **name**.

We can name values using what's called an assignment statement.

```
[14]:  # assigns 442 to x
x = 442
```

The assignment statement has three parts. On the left is the name(x). On the right is the value(442). The $equals\ sign$ in the middle tells the computer to assign the value to the name.

You'll notice that when you run the cell with the assignment, it doesn't print anything. But, if we try to access x again in the future, it will have the value we assigned it.

```
[15]: # print the value of x x
```

[15]: 442

You can also assign names to expressions. The computer will compute the expression and assign the name to the result of the computation.

```
[16]: y = 50 * 2 + 1
```

[16]: 101

We can then use these name as if they were numbers.

```
[17]: x - 42
```

[17]: 400

```
[18]: x + y
```

[18]: 543

PRACTICE: Try rewriting the problem below, so the names make more sense. Do we expect an error below?

[19]: 3000

2.0.4 1c. Functions

We've seen that values can have names (often called **variables**), but operations may also have names. A named operation is called a **function**. Python has some functions built into it.

```
[20]: # a built-in function round
```

[20]: <function round(number, ndigits=None)>

Functions get used in *call expressions*, where a function is named and given values to operate on inside a set of parentheses. The **round** function returns the number it was given, rounded to the nearest whole number.

```
[21]: # a call expression using round round(1988.74699)
```

[21]: 1989

A function may also be called on more than one value (called *arguments*). For instance, the min function takes however many arguments you'd like and returns the smallest. Multiple arguments are separated by commas.

```
[22]: min(9, -34, 0, 99)
```

[22]: -34

PRACTICE

The abs function takes one argument (just like round)

The max function takes one or more arguments (just like min)

Try calling abs and max in the cell below. What does each function do?

Also try calling each function *incorrectly*, such as with the wrong number of arguments. What kinds of error messages do you see?

```
[37]: # replace the ... with calls to abs and max abs(-100)
```

[37]: 100

```
[38]: max(10, -100, 0)
```

[38]: 10

Dot Notation Python has a lot of built-in functions (that is, functions that are already named and defined in Python), but even more functions are stored in collections called *modules*. Earlier, we imported the math module so we could use it later. Once a module is imported, you can use its functions by typing the name of the module, then the name of the function you want from it, separated with a ..

```
[23]: # a call expression with the factorial function from the math module math.factorial(5)
```

[23]: 120

Practice: math also has a function called sqrt that takes one argument and returns the square root. Call sqrt on 16 in the next cell.

```
[25]: # use math.sqrt to get the square root of 16
math.sqrt(16)
```

[25]: 4.0

3 2. Tables

Tables are fundamental ways of organizing and displaying data. Run the next cell to load the data. Run the cell below to load a dataset.

```
[27]: # Below we see an assignment statement.
# We are telling the computer to create a Table and read in some data.
```

```
ratings = Table.read_table("data/imdb_ratings.csv")
ratings
```

```
[27]: Votes | Rank | Title
                                                    | Year | Decade
            8.4
                   | M (1931)
                                                    | 1931 | 1930
      88355
      132823 | 8.3
                   | Singin' in the Rain (1952)
                                                    | 1952 | 1950
      74178 | 8.3
                   | All About Eve (1950)
                                                    | 1950 | 1950
                   | Léon (1994)
                                                    | 1994 | 1990
      635139 | 8.6
      145514 | 8.2
                   | The Elephant Man (1980)
                                                    | 1980 | 1980
      425461 | 8.3
                   | Full Metal Jacket (1987)
                                                    | 1987 | 1980
      441174 | 8.1
                   | Gone Girl (2014)
                                                    | 2014 | 2010
      850601 | 8.3
                   | Batman Begins (2005)
                                                    | 2005 | 2000
                   | Judgment at Nuremberg (1961) | 1961 | 1960
            8.2
      46987 | 8
                    | Relatos salvajes (2014)
                                                    | 2014 | 2010
      ... (240 rows omitted)
```

This table is organized into **columns**: one for each *category* of information collected:

You can also think about the table in terms of its **rows**. Each row represents all the information collected about a particular instance, which can be a person, location, action, or other unit.

PRACTICE: What do the rows in this table represent? By default only the first ten rows are shown. Can you see how many rows there are in total?

3.1 2a. Table Attributes

Every table has **attributes** that give information about the table, like the number of rows and the number of columns. Table attributes are accessed using the dot method. But, since an attribute doesn't perform an operation on the table, there are no parentheses (like there would be in a call expression).

Attributes you'll use frequently include num_rows and num_columns, which give the number of rows and columns in the table, respectively.

```
[28]: # get the number of columns ratings.num_columns
```

[28]: 5

PRACTICE: Use num_rows to get the number of rows in our table.

```
[]: # get the number of rows in the table
```

Note on Tables In this notebook, we worked with the datascience library to work with tables. In practice and in industry, you'll most likely use pandas to manipulate tabular data. Using the datascience library is more friendly syntax-wise, but pandas is more powerful overall.

3.2 2b. Table Transformation

Not all of our columns are relevant to every question we want to ask. We can save computational resources and avoid confusion by *transforming* our table before we start work.

Subsetting columns with select and drop

3.2.1 select

The select function is used to get a table containing only particular columns. select is called on a table using dot notation and takes one or more arguments: the name or names of the column or columns you want. Not this does not change the original table. To save your changes, you must assign your change a name.

```
[29]: # make a new table with only selected columns
      ratings.select("Votes", "Title")
[29]: Votes
            | Title
      88355 | M (1931)
      132823 | Singin' in the Rain (1952)
      74178 | All About Eve (1950)
      635139 | Léon (1994)
      145514 | The Elephant Man (1980)
      425461 | Full Metal Jacket (1987)
      441174 | Gone Girl (2014)
      850601 | Batman Begins (2005)
      37664 | Judgment at Nuremberg (1961)
      46987 | Relatos salvajes (2014)
      ... (240 rows omitted)
[30]: # No changes made to the original table
      ratings
```

```
[30]: Votes | Rank | Title
                                                   | Year | Decade
            | 8.4 | M (1931)
                                                    | 1931 | 1930
      132823 | 8.3 | Singin' in the Rain (1952)
                                                   | 1952 | 1950
      74178 | 8.3 | All About Eve (1950)
                                                   | 1950 | 1950
      635139 | 8.6 | Léon (1994)
                                                   | 1994 | 1990
      145514 | 8.2 | The Elephant Man (1980)
                                                   | 1980 | 1980
      425461 | 8.3 | Full Metal Jacket (1987)
                                                   | 1987 | 1980
      441174 | 8.1 | Gone Girl (2014)
                                                   | 2014 | 2010
      850601 | 8.3 | Batman Begins (2005)
                                                   | 2005 | 2000
      37664 | 8.2
                   | Judgment at Nuremberg (1961) | 1961 | 1960
            18
                    | Relatos salvajes (2014)
                                                   | 2014 | 2010
      ... (240 rows omitted)
```

3.2.2 drop

88355 | M (1931)

If instead you need all columns except a few, the drop function can get rid of specified columns. drop works very similarly to select: call it on the table using dot notation, then give it the name or names of what you want to drop.

```
[31]: # drop a column
      ratings.drop("Decade")
[31]: Votes | Rank | Title
                                                   | Year
      88355 | 8.4 | M (1931)
                                                   l 1931
      132823 | 8.3 | Singin' in the Rain (1952)
                                                   I 1952
      74178 | 8.3 | All About Eve (1950)
                                                   | 1950
      635139 | 8.6 | Léon (1994)
                                                   l 1994
      145514 | 8.2 | The Elephant Man (1980)
                                                   I 1980
      425461 | 8.3 | Full Metal Jacket (1987)
                                                   I 1987
      441174 | 8.1 | Gone Girl (2014)
                                                   | 2014
      850601 | 8.3 | Batman Begins (2005)
                                                   1 2005
      37664 | 8.2 | Judgment at Nuremberg (1961) | 1961
                    | Relatos salvajes (2014)
      46987 | 8
                                                   | 2014
      ... (240 rows omitted)
```

PRACTICE: Pick two columns from our table. Create a new table containing only those two columns two different ways: once using select and once using drop.

```
[33]: # use select
      ratingsOne = ratings.select("Votes", "Title")
      ratingsOne
[33]: Votes | Title
      88355 | M (1931)
      132823 | Singin' in the Rain (1952)
      74178 | All About Eve (1950)
      635139 | Léon (1994)
      145514 | The Elephant Man (1980)
      425461 | Full Metal Jacket (1987)
      441174 | Gone Girl (2014)
      850601 | Batman Begins (2005)
      37664 | Judgment at Nuremberg (1961)
      46987 | Relatos salvajes (2014)
      ... (240 rows omitted)
[34]: # use drop
      ratings2 = ratings.drop("Rank", "Year", "Decade")
      ratings2
[34]: Votes
            | Title
```

```
132823 | Singin' in the Rain (1952)
74178 | All About Eve (1950)
635139 | Léon (1994)
145514 | The Elephant Man (1980)
425461 | Full Metal Jacket (1987)
441174 | Gone Girl (2014)
850601 | Batman Begins (2005)
37664 | Judgment at Nuremberg (1961)
46987 | Relatos salvajes (2014)
... (240 rows omitted)
```

3.3 2c. Sorting Tables

The last section covered select and drop. In this next section, we'll test our knowledge and use what we have learned and understand how we can sort and manipulate data that are placed in tables.

3.3.1 show

In a table, we can display a specific amount of rows using the **show** operation. The **show** operations allows you to enter the amount of rows you want displayed from a table.

```
[35]: # use show to display 20 rows ratings.show(20)
```

<IPython.core.display.HTML object>

3.3.2 sort

Some details about sort from the UC Berkeley's Data 8 lab 2 (Spring 2020):

- 1. The first argument to **sort** is the name of a column to sort by.
- 2. If the column has text in it, sort will sort alphabetically; if the column has numbers, it will sort numerically.
- 3. The descending=False bit is called an *optional argument*. It has a default value of False, so when you explicitly tell the function descending=True, then the function will sort in descending order.
- 4. The distinct=True bit is also an *optional argument*. When the function distinct=True is used, the function will delete any duplicate values based on the column_or_label value that was also passed in the sort function.
- 5. Rows always stick together when a table is sorted. It wouldn't make sense to sort just one column and leave the other columns alone. For example, in this case, if we sorted just the Year column, the movies would all end up with the wrong year.

The format for sorting code is written as....

Table.sort(column_or_label, descending=False, distinct=False)

```
[36]: # sort a column
     ratings.sort("Year")
[36]: Votes | Rank | Title
                                                         | Year | Decade
     55784 | 8.3 | The Kid (1921)
                                                         | 1921 | 1920
     58506 | 8.2 | The Gold Rush (1925)
                                                         | 1925 | 1920
     46332 | 8.2 | The General (1926)
                                                         | 1926 | 1920
     98794 | 8.3 | Metropolis (1927)
                                                         | 1927 | 1920
     88355 | 8.4 | M (1931)
                                                         | 1931 | 1930
     92375 | 8.5 | City Lights (1931)
                                                         | 1931 | 1930
     56842 | 8.1 | It Happened One Night (1934)
                                                         | 1934 | 1930
     121668 | 8.5 | Modern Times (1936)
                                                         | 1936 | 1930
     69510 | 8.2 | Mr. Smith Goes to Washington (1939) | 1939 | 1930
     259235 | 8.1 | The Wizard of Oz (1939)
                                                         | 1939 | 1930
     ... (240 rows omitted)
[37]: # Sorted from most recent year to oldest year
     ratings.sort("Year", descending=True)
[37]: Votes | Rank | Title
                                                       | Year | Decade
     79615 | 8.5 | Inside Out (2015/I)
                                                       | 2015 | 2010
     262425 | 8.3 | Mad Max: Fury Road (2015)
                                                       | 2015 | 2010
     441174 | 8.1 | Gone Girl (2014)
                                                       | 2014 | 2010
     46987 | 8 | Relatos salvajes (2014)
                                                       | 2014 | 2010
     427099 | 8
                   | X-Men: Days of Future Past (2014) | 2014 | 2010
     321834 | 8
                   | The Imitation Game (2014)
                                                       | 2014 | 2010
     689541 | 8.6 | Interstellar (2014)
                                                       | 2014 | 2010
     527349 | 8 | Guardians of the Galaxy (2014)
                                                      | 2014 | 2010
     369141 | 8.1 | The Grand Budapest Hotel (2014)
                                                       | 2014 | 2010
     264333 | 8.5 | Whiplash (2014)
                                                       | 2014 | 2010
     ... (240 rows omitted)
     PRACTICE: Sort the table from highest ranked movie to lowest ranked.
[38]: # Sorted from highest ranked movie to lowest ranked
     ratings.sort("Rank", descending=True)
[38]: Votes
              | Rank | Title
                                                                           | Year |
     Decade
     1027398 | 9.2 | The Godfather (1972)
                                                                           l 1972 l
     1970
     1498733 | 9.2 | The Shawshank Redemption (1994)
                                                                           l 1994 l
     692753 | 9 | The Godfather: Part II (1974)
                                                                           l 1974 l
     1970
     447875 | 8.9 | Il buono, il brutto, il cattivo (1966)
                                                                           l 1966 l
```

```
1960
1473049 | 8.9
               | The Dark Knight (2008)
                                                                         | 2008 |
2000
               | 12 Angry Men (1957)
384187 | 8.9
                                                                         | 1957 |
1950
               | The Lord of the Rings: The Return of the King (2003) | 2003 |
1074146 | 8.9
2000
               | Schindler's List (1993)
761224 | 8.9
                                                                         | 1993 |
1990
1166532 | 8.9
               | Pulp Fiction (1994)
                                                                         | 1994 |
1990
1177098 | 8.8
               | Fight Club (1999)
                                                                         | 1999 |
... (240 rows omitted)
```

PRACTICE: Sort the table by year in ascending order, making sure each year is distinct.

```
[39]: ratings.sort("Year", distinct=True)
[39]: Votes
            | Rank | Title
                                                          | Year | Decade
      55784
             8.3
                   | The Kid (1921)
                                                          | 1921 | 1920
            | 8.2 | The Gold Rush (1925)
                                                          | 1925 | 1920
      58506
             8.2
                   | The General (1926)
      46332
                                                          | 1926 | 1920
                   | Metropolis (1927)
      98794
            8.3
                                                            1927 | 1920
      88355
            8.4
                   | M (1931)
                                                            1931 | 1930
            8.1
                   | It Happened One Night (1934)
                                                          | 1934 | 1930
      56842
      121668 | 8.5
                   | Modern Times (1936)
                                                            1936 | 1930
      69510 | 8.2
                   | Mr. Smith Goes to Washington (1939) | 1939 | 1930
      55793 | 8.1
                                                          | 1940 | 1940
                   | The Grapes of Wrath (1940)
      101754 | 8.1 | The Maltese Falcon (1941)
                                                          | 1941 | 1940
      ... (70 rows omitted)
 []:
```

References

- Sections of "Intro to Jupyter", "Table Transformation" adapted from materials by Kelly Chen and Ashley Chien in UC Berkeley Data Science Modules core resources
- "A Note on Errors" subsection and "error" image adapted from materials by Chris Hench and Mariah Rogers for the Medieval Studies 250: Text Analysis for Graduate Medievalists data science module.
- Rocket Fuel data and discussion questions adapted from materials by Zsolt Katona and Brian Bell, BerkeleyHaas Case Series

Initially authored by Keeley Takimoto, adapted by BUDS team, further adapted by Laura Brown.

```
[]:
```