Data 100 Debugging Guide

Yash Dave

Lillian Weng

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About

This text offers pointers for keyboard shortcuts or common mistakes that accompany the coursework in the Spring 2024 Edition of the UC Berkeley course Data 100: Principles and Techniques of Data Science.

Inspiration for this guide was taken from the UC San Diego course DSC 10: Principles of Data Science and their debugging guide.

If you spot any typos or would like to suggest any changes, please fill out the Data 100 Content Feedback Form (Spring 2024). Note that this link will only work if you have an @berkeley.edu email address. If you're not a student at Berkeley and would like to provide feedback, please email us at data100.instructors@berkeley.edu

1 Jupyter 101

Note

If you're using a MacBook, replace ctrl with cmd.

1.1 Shortcuts for Cells

For the following commands, make sure you're in command mode. You can enter this mode by pressing esc.

- a: create a cell above
- b: create a cell below
- dd: delete current cell
- m: convert a cell to markdown (text cell)
- y: convert a cell to code

1.2 Running Cells

For individual cells,

- ctrl + return: run the current cell
- shift + return: run the current cell and move to the next cell

To run all cells in a notebook:

- In the menu bar on the left, click Run. From here, you have several options. The ones we use most commonly are:
 - Run All Above Selected Cell: this runs every cell above the selected cell
 - Run Selected Cell and All Below: this runs the selected cell and all cells below
 - Run All: this runs every cell in the notebook from top-to-bottom

1.3 Saving your notebook

Jupyter autosaves your work, but there can be a delay. As such, it's a good idea to save your work as often as you remember and especially before submitting assignments. To do so, press ctrl + s.

1.4 Restarting Kernel

In the menu bar on the left, click Kernel. From here, you have several options. The ones we use most commonly are:

- Restart Kernel...
- Restart Kernel and Run up to Selected Cell
- Restart Kernel and Run All Cells

1.5 Automatically Closing Brackets

Many IDEs like VSCode have a functionality that automatically closes brackets. For example, pressing (, {, or [would automatically add the second bracket at the other end), }, and], respectively. Datahub does not have this functionality turned on by default, but you can do so by going into Settings -> Auto Close Brackets. If you see a check mark to the left of Auto Close Brackets, then it's enabled.

2 Jupyter / Datahub

2.1 My kernel died, restarted, or is very slow

Jupyterhub connects you to an external container to run your code. That connection could be slow/severed because:

- 1. you haven't made any changes to the notebook for a while
- 2. a cell took too much time to run
- 3. a cell took up too many resources to compute

When you see a message like this:

- 1. Either press the "Ok" button or reload the page
- 2. Restart your kernel
- 3. Rerun your cells

Note that you may loose some recent work if your kernel restarted when you were in the middle of editing a cell. As such, we recommend saving your work as often as possible.

If this does not fix the issue, it could be a problem with your code, usually the last cell that executed before your kernel crashed. Double check your logic, and feel free to make a private post on Ed if you're stuck!

2.2 I can't edit a cell

We set some cells to read-only mode prevent accidental modification. To make the cell writeable,

- 1. Click the cell
- 2. Click setting on the top right corner
- 3. Under "Common Tools", you can toggle between "Editable" (can edit the cell) and "Read-Only" (cannot edit the cell)

2.3 My text cell looks like code

If you double-click on a text (markdown) cell, it'll appear in its raw format. To fix this, simply run the cell. If this doesn't fix the problem, check out the commonly asked question below.

2.4 My text cell changed to a code cell / My code cell changed to a text cell

Sometimes, a text (markdown) cell was changed to a code cell, or a code cell can't be run because it's been changed to a text (markdown) or raw cell. To fix this, toggle the desired cell type in the top bar.

2.5 Why does running a particular cell cause my kernel to die?

If one particular cell seems to cause your kernel to die, this is likely because the computer is trying to use more memory than it has available. For instance: your code is trying to create a gigantic array. To prevent the entire server from crashing, the kernel will "die". This is an indication that there is a mistake in your code that you need to fix.

2.6 I accidentally deleted something in a cell that was provided to me – how do I get it back?

Suppose you're working on Lab 5. One solution is to go directly to DataHub and rename your lab05 folder to something else, like lab05-old. Then, click the Lab 5 link on the course website again, and it'll bring you to a brand-new version of Lab 5. You can then copy your work from your old Lab 5 to this new one, which should have the original version of the assignment.

Alternatively, you can access this public repo and navigate to a blank copy of the assignment you were working on. In the case of Lab 5 for example, the notebook would be located at lab/lab05.ipynb. You can then check and copy over the contents of the deleted cell into a new cell in your existing notebook.

2.7 "Click here to download zip file" is not working

When this happens, you can download the zip file through the menu on the left.

Right click on the generated zip file and click "Download".

2.8 I can't export my assignment as a PDF due to a LatexFailed error

Occasionally when running the grader.export(run_tests=True) cell at the end of the note-book, you run into an error where the PDF failed to generate:

Converting a Jupyter notebook to a PDF involves formatting some of the markdown text in LaTeX. However, this process will fail if your free response answers have (unresolved) LaTeX characters like \n, \$, or \$\$. If you're short on time, your best bet is to take screenshots of your free response answers and submit them to Gradescope. If you have more time and would like the Datahub-generated PDF, please remove any special LaTeX characters from your free response answers.

2.9 I can't open Jupyter: HTTP ERROR 431

If this happens, try clearing your browser cache or opening Datahub in an incognito window.

3 Autograder and Gradescope

Citation

Many of these common questions were taken and modified from the UC San Diego course DSC 10: Principles of Data Science and their debugging guide.

3.1 Autograder

3.1.1 Understanding autograder error messages

When you pass a test, you'll see a nice message and a cute emoji!

When you don't, however, the message can be a little confusing.

The best course of action is to find the test case that failed and use that as a starting point to debug your code.

In the example above, we see that the test case in green, max_swing in set(bus['name']), is not passing. The actual output (in blue) is often hard to parse, so the best course of action is to:

- 1. Make a new (temporary) cell after the grader.check(...) cell. Please do not make a new cell in between the given code cell and the grader.check(...) cell, as it could mess with the results.
- 2. Copy and paste the failing test case into your temporary cell and run it.
 - a. If it's giving you an error like in the example above, look at the last line of the error and use the Debugging Guide's search functionality in the top left menu to find the corresponding guide.
 - b. If it's not giving you an error, it'll likely give you an output like False. This means that your code does not cause an error (yay!), but it returns an incorrect output. In these casses, inspect each individual element of the test case. The example above checks if max_swing is in set(bus['name']), so it might be a good idea to display both variables and do a visual check.
 - c. If you're still having issues, post on Ed!
- 3. After your grader.check(...) passes, feel free to delete the temporary cell.

3.1.2 Why do I get an error saying "grader is not defined"?

If it has been a while since you've worked on an assignment, the kernel will shut itself down to preserve memory. When this happens, all of your variables are forgotten, including the grader. That's OK. The easiest way to fix this is by restarting your kernel and rerunning all the cells. To do this, in the top left menu, click Kernel -> Restart and Run All Cells.

3.1.3 I'm positive I have the right answer, but the test fails. Is there a mistake in the test?

While you might see the correct answer displayed as the result of the cell, chances are your solution isn't being stored in the answer variable. Make sure you are assigning the result to the answer variable and that there are no typos in the variable name. Finally, restart your kernel and run all the cells in order: Kernel -> Restart and Run All Cells.

3.1.4 Why does the last grader.export cell fail if all previous tests passed?

This can happen if you "overwrite" a variable that is used in a question. For instance, say Question 1 asks you to store your answer in a variable named stat and, later on in the notebook, you change the value of stat; the test right after Question 1 will pass, but the test at the end of the notebook will fail. It is good programming practice to give your variables informative names and to avoid repeating the same variable name for more than one purpose.

3.1.5 Why does a notebook test fail now when it passed before, and I didn't change my code?

You probably ran your notebook out of order. Re-run all previous cells in order, which is how your code will be graded.

3.2 Gradescope

When submitting to Gradescope, there are often unexpected errors that make students lose more points than expected. Thus, it is imperative that you stay on the submission page until the autograder finishes running, and the results are displayed.

3.2.1 Why did a Gradescope test fail when all the Jupyter notebook's tests passed?

This can happen if you're running your notebook's cells out of order. The autograder runs your notebook from top-to-bottom. If you're defining a variable at the bottom of your notebook and using it at the top, the Gradescope autograder will fail because it doesn't recognize the variable when it encounters it.

This is why we recommend going into the top left menu and clicking Kernel -> Restart -> Run All. The autograder "forgets" all of the variables and runs the notebook from top-to-bottom like the Gradescope autograder does. This will highlight any issues.

Find the first cell that raises an error. Make sure that all of the variables used in that cell have been defined above that cell, and not below.

3.2.2 Why do I get a NameError: name ___ is not defined when I run a grader check?

This happens when you try to access a variable that has not been defined yet. Since the autograder runs all the cells in-order, if you happened to define a variable in a cell further down and accessed it before that cell, the autograder will likely throw this error. Another reason this could occur is because the notebook was not saved before the autograder tests are run. When in doubt, it is good practice to restart your kernel, run all the cells again, and save the notebook before running the cell that causes this error.

3.2.3 My autograder keeps running/timed out

If your Gradescope submission page has been stuck running on this page for a while: $\frac{1}{2}$

or if it times out:

it means that the Gradescope autograder failed to execute in the expected amount of time. This could be due to an inefficiency in your code or a problem on Gradescope's end, so we recommend resubmitting and letting the autograder rerun. It is your responsibility to ensure that the autograder runs properly, and, if it still fails, to follow up by making a private Ed post.

4 Pandas

4.1 Understanding pandas errors

pandas errors can look red, scary, and very long. Fortunately, we don't need to understand the entire thing! The most important parts of an error message are at the **top**, which tells you which line of code is causing the issue, and at the **bottom**, which tells you exactly what the error message is.

This note is (mostly) structured around the error messages that show up at the bottom.

4.2 My code is taking a really long time to run

It is normal for a cell to take a few seconds – sometimes a few minutes – to run. If it's is taking too long, however, you have several options:

- 1. Try restarting the kernel. Sometimes, Datahub glitches or lags, causing the code to run slower than expected. Restarting the kernel should fix this problem, but if the cell is still taking a while to run, it is likely a problem with your code.
- 2. Scrutinize your code. Am I using too many for loops? Is there a repeated operation that I can substitute with a pandas function?

4.3 Why is it generally better avoid using loops or list comprehensions when possible?

In one word: performance. NumPy and pandas functions are optimized to handle large amounts of data in an efficient manner. Even for simple operations, like the elementwise addition of two arrays, NumPy arrays are much faster and scale better (feel free to experiment with this yourself using %%time). This is why we encourage you to vectorize your code (ie. using NumPy arrays, Series, or DataFrames instead of Python lists) and use in-built NumPy/pandas functions wherever possible.

4.4 KeyErrors

4.4.1 KeyError: 'column_name'

This error usually happens when we have a DataFame called df, and we're trying to do an operation on a column 'column_name' that does not exist. If you encounter this error, double check that you're operating on the right column. It might be a good idea to display df and see what it looks like. You could also call df.columns to list all the columns in df.

4.5 TypeErrors

4.5.1 TypeError: '___' object is not callable

This often happens when you use a default keyword (like str, list, range, bool, sum, or max) as a variable name, for instance:

$$sum = 1 + 2 + 3$$

These errors can be tricky because they don't error on their own but cause problems when we try to use the name sum (for example) later on in the notebook.

To fix the issue, identify any such lines of code (Ctrl+F on "sum =" for example), change your variable names to be something more informative, and restart your notebook.

Python keywords like str and list appear in green text, so be on the lookout if any of your variable names appear in green!

4.5.2 TypeError: could not convert string to a float

This error often occurs when we try to do math operations (ie. sum, average, min, max) on a DataFrame column or Series that contains strings instead of numbers (note that we can do math operations with booleans; Python treats True as 1 and False as 0).

Double check that the column you're interested in is a numerical type (int, float, or double). If it looks like a number, but you're still getting this error, you can use .astype(...) (documentation) to change the datatype of a DataFrame or Series.

4.5.3 TypeError: Could not convert <string> to numeric

Related to the above (but distinct), you may run into this error when performing a numeric aggregation function (like mean or sum functions that take integer arguments) after doing a groupby operation on a DataFrame with non-numeric columns.

Working with the elections dataset for example,

```
elections.groupby('Year').agg('mean')
```

would error because pandas cannot compute the mean of the names of presidents. There are three ways to get around this:

- 1. Select only the numeric columns you are interested in before applying the aggregation function. In the above case, both elections.groupby('Year')['Popular Vote'] or elections['Popular vote'].groupby('Year') would work.
- 2. Setting the numeric_only argument to True in the .agg call, thereby applying the aggregation function only to numeric columns. For example, elections.groupby('Year').agg('mean', numeric_only=True).
- 3. Passing in a dictionary to .agg where you specify the column you are applying a particular aggregation function to. Continuing the same example, this looks like elections.groupby('Year').agg({'Popular vote' : 'mean'}).

4.5.4 TypeError: 'NoneType' object is not subscriptable / AttributeError: 'NoneType' object has no attribute 'shape'

This usually occurs as you assign a None value to a variable, then try to either index into or access some attribute of that variable. For Python functions like append and extend, you do not need to do any variable assignment because they mutate the variable directly and return None. Assigning None tends to happen as a result of code like:

```
some_list = some_list.append(element)
```

In contrast, an operation like np.append does not mutate the variable in place and, instead, returns a copy. In these cases, (re)assignment is necessary:

```
some_array = np.append(some_array, element)
```

4.5.5 TypeError: 'int'/'float' object is not subscriptable

This occurs when you try and index into an integer or other numeric Python data type. It can be confusing to debug amidst a muddle of code, but you can use the error message to identify which variable is causing this error. Using type(var_name) to check the data type of the variable in question can be a good starting point.

4.6 IndexErrors

4.6.1 IndexError: invalid index to scalar variable.

This error is similar to the last TypeError in the previous section. However, it is slightly different in that scalar variables come up in the context of NumPy data types which have slightly different attributes.

For a concrete example, if you defined

```
numpy_arr = np.array([1])
```

and indexed into it twice (numpy_arr[0][0]), you would get the above error. Unlike a Python integer whose type is int, type(numpy_arr[0]) returns the NumPy version of an integer, numpy.int64. Additionally, you can check the data type by accessing the .dtype attribute of NumPy array (numpy_arr.dtype) or scalar variable (numpy_arr[0].dtype).

4.6.2 IndexError: index _ is out of bounds for axis _ with size _

This error usually happens when you try to index a value that's greater than the size of the array/list/DataFrame/Series. For example,

```
some_list = [2, 4, 6, 8]
```

some_list has a length of 4. Trying some_list[6] will error because index 6 is greater than the length of the array. Note that some_list[4] will also cause an IndexError because Python and pandas uses zero indexing, which means that the first element has index 0, the second element has index 1, etc.; some_list[4] would grab the fifth element, which is impossible when the list only has 4 elements.

4.7 ValueErrors

4.7.1 ValueError: Truth value of a Series is ambiguous

This error could occur when you apply Python logical operators (or, and, not), which only operate on a single boolean values, to NumPy arrays or Series objects, which can contain multiple values. The fix is to use bitwise operators |, &, ~, respectively, to allow for elementwise comparisons between values in arrays or Series.

Alternatively, these errors could emerge due to overwriting Python keywords like bool and sum that may be used in the autograder tests, similar to what's described here. You should follow a similar procedure of identifying the line of code erroring, checking if you've overwritten any Python keywords using Ctrl+F, and renaming those variables to something more informative before restarting your kernel and running the erroring tests again.

4.7.2 ValueError: Can only compare identically-labeled Series objects

As the message would suggest, this error occurs when comparing two Series objects that have different lengths. You can double check the lengths of the Series using len(series_name) or series_name.size.

4.7.3 ValueError: -1 is not in range / KeyError: -1

This error occurs when you try and index into a Series or DataFrame as you would a Python list. Unlike a list where passing an index of -1 gives the last element, pandas interprets df[-1] as an attempt to find the row corresponding to index -1 (that is, df.loc[-1]). If your intention is to pick out the last row in df, consider using integer-position based indexing by doing df.iloc[-1]. In general, to avoid ambiguity in these cases, it is also good practice to write out both the row and column indices you want with df.iloc[-1, :].

5 RegEx

RegEx syntax can be incredibly confusing, so we highly encourage using sources like the Data 100 Exam reference sheet (you can find this under the "Exam Resources" section on our Resources page) or websites like regex101.com to help build your understanding.

5.1 How to Interpret regex101

Regex101 is a great tool that helps you visually interact with RegEx patterns. Let's take a look at its components with a simple example.

5.1.1 Example 1: Basic

- 0. **Flavor**: Regular expressions work slightly differently depending on the programming language you use. In Data 100, we only use the Python flavor. By default, regex101 opens on the PCRE2 flavor, so make sure to change to Python before experimenting.
- 1. **Regular Expression**: This is where the RegEx expression goes. For this example, our pattern is Data 100. In Python, we denote it as a string r"Data 100" with the prefix r to indicate that this is a RegEx expression, not a normal Python string. In regex101, because we changed to the Python flavor, we don't need to type out the r" at the start or the " at the end, as that's already set up for us.
- Explanation: This portion of the website explains each component of the pattern above. Since it does not contain any special characters, Data 100 will match any portion of a string containing Data 100.
- 3. **Test String**: This is where you can try out different inputs and see if they match the RegEx pattern. Of the 4 example sentences, we see that only the first sentence contains characters that match the pattern, highlighted in blue. (Note that while sentence 3 does contain data 100, RegEx is sensitive to capitalization. d and D are different characters)
- 4. **Match Information**: Each match between the RegEx expression and test strings is shown here.

5.1.2 Example 2: Greedy

For this example, let's replace the 100 in our original expression with $\d+$ so that our pattern is Data $\d+$

\d and + are both special operators, and the explanation on the top right (boxed in red) tells us what they do:

- \d matches digits, or any number between 0 and 9. It's equivalent to [0-9].
- + matches the previous token ≥ 1 times. It is a *greedy operation*, meaning it will match as many characters as possible.

Altogether, the expression \d+ will match any digit one or more times. Look at each match under "Match Information". Can you see why they align with Data \d+?

5.1.3 Example 3: Capturing Groups

Let's say we're given a body of text with dates formatted as DD/Month/YYYY (ie. 04/Jan/2014), and we're interested in extracting the dates. An expression like r"\d+\/\w+\/\d+" would match any string with the DD/Month/YYYY format:

- the first \d+ matches DD patterns (ie. 04)
- \/ matches the / separator. Since / is a special operator in RegEx, we need to escape it with \ to get the literal character.
- \w+ in the middle matches Month patterns we're interested in (ie. Jan, January)
- lastly, \d+ matches YYYY patterns (ie. 2014)

That's great! This pattern will match the entirety of DD/Month/YYYY, but what if we want to access DD individually? What about YYYY? This is where **capturing groups** comes in handy. Capturing groups are RegEx expressions surrounded by parenthesis () that are used to remember the text they match so that it can be referenced later. Putting capturing groups around $\d+$ and $\d+$ to get $\d+$ ($\d+$)/($\d+$)/($\d+$) "gives us the following:

- The "Explanation" section now shows an explanation for each of the 3 capturing groups.
- In our test strings, the portion matching the RegEx expression is highlighted in blue per usual. Additionally, each capturing group is highlighted with a particular color: green, orange, and purple.
- These colored highlights correspond to their match/group under "Match Information". "Match #" (light blue) shows the entire portion that matches the expression while "Group #" shows the match per group.

5.1.3.1 How do I access captured groups?

To access each group, we use the following syntax:

```
target_string = "Today's date is 01/March/2024."
result = re.search(r"(\d+)\/(\w+)\/(\d+)", target_string)

result # re.Match object
result.groups() # all captured groups: ('01', 'March', '2024')
result.group(0) # '01/March/2024', the full match
result.group(1) # '01', the first captured group
result.group(2) # 'March', the second captured group
result.group(3) # '2024', the third captured group
```

5.2 RegEx Misconceptions & General Errors

5.2.1 I'm certain my RegEx pattern in .str.replace is correct, but I'm not passing the grader check.

Here's the skeleton from the exam reference sheet:

```
s.str.replace(pat, repl, regex=False)
```

Notice how the regex= argument has a default value of False, causing pandas to treat pat like a normal Python string. Make sure to set regex=True if you're using RegEx!

5.2.2 My RegEx pattern matches the test cases on regex101, but is not working in pandas with .str.findall / .str.extractall / .str.extract.

The most likely reason for this is forgetting to include the r before the string with your regular expression. Without including the r in .str.findall(r".*"), pandas will not interpret your pattern as a regex expression and will only match it literally.

5.2.3 Value Error: pattern contains no capture groups

These errors usually occur when using s.str.extract or s.str.extractall. Read more about it in the RegEx course notes. This error means that your RegEx pattern does not match anything in the given Series of strings. To debug this, try putting your pattern into regex101.com and use example strings from the Series as test cases.

5.2.4 When do I need to escape characters?

If you want to match exactly those characters in a RegEx expression, you need to "escape" them by preceding them with a backslash \. However, the rules around this can change in the context of character classes.

For example, the pattern r"[.]" matches '.', the literal period. In this context, it is not treated as a special character. The hyphen, while not included in the list of special characters, also changes its behavior depending on its position in a character class. It can be used to specify a range of characters (e.g. r"[0-9]") based on their Unicode values, or match a literal '-' if it does not have two adjacent characters (e.g. r"[-09]" matches -, 0, 9). To be on the safer side, you could escape - like in $r"[0\-9]"$ to achieve the same result.

Finally, it's generally good practice to escape both single and double quotes for greater readability. Technically, patterns like r"'(.)*)" and r'"(.)*)" do work as you'd expect, but you can already see how confusing it is to decipher what's going on. Escaping the quotes inside the pattern does not affect what matches you get, but makes it easier to figure out what the intended match was.

5.2.5 The three uses of ^

The ^ character can be tricky to wrap your head around given how its function changes depending on the context:

- 1. If used at the start of a pattern, like in $r"^\w"$, it means that a lowercase letter must begin the string in order for a match to occur.
- 2. If included at the start of a character class, like in r"[^abc]", it negates all the characters in that class and will match with any other character in the above example, any character that is not a, b, c.
- 3. Finally, if escaped as in r"" it is treated as a literal and will match any instance of $\^$.

5.2.6 What's the difference between all the re functions?

The exam reference sheets give a few **re** functions, but how can you determine which one to use?

re.match and re.search only return *one* instance of a match between string and pattern (or None if there's no match) - re.match only considers characters at the beginning of a string - re.search considers characters anywhere in the string - For example: "' pattern = r"Data 100" example1 = "Data 100 is the best!" example2 = "I love Data 100!"

re.match(pattern, example1).group(0) # matches "Data 100" re.match(pattern, example2) # does not match "Data 100" because it's not at the beginning of a string; returns None

re.search(pattern, example1).group(0) # matches "Data 100" re.search(pattern, example2).group(0) # matches "Data 100" "'

If, instead, you're interested in finding *all* matches between the given string and pattern, re.findall will find them all, returning the matches in a list.

```
re.findall(r'\d+', 'Data 100, Data 8, Data 101')
# returns a list of strings: ['100', '8', '101']

re.findall(r'\d+', 'Data science is great')
# no matches found, returns empty list: []

re.sub will find them all and replace it with a string of your choice.

re.sub(r'\d+', 'panda', 'Data 100, Data 8, Data 101')
# returns 'Data panda, Data panda, Data panda'

re.sub(r'\d+', 'panda', 'Data science is great')
# no matches found, returns the original string "Data science is great"
```

5.2.7 What's the difference between re functions and pd.Series.str functions?

Generally, all the pd.Series.str functions are used when you want to apply a Python or RegEx string function to a *Series of strings*. In contrast, re functions are applied to string objects. The reference sheet gives a great overview of the different use cases of each of the pd.Series.str functions.

6 Visualizations

Visualizations are how data scientists use to communicate their insights to the world. In the process of making a good visualization, be it while adding a legend or setting the x-axis label, we may run into some errors. Let's take a look at how to resolve them below!

6.0.1 My legend's labels don't match up / my legend isn't displaying properly

If you simply add plt.legend() after your plotting line of code, you should see a legend; seaborn will sometimes automatically populate the legend. However, if you're plotting multiple lines or sets of points on a single plot, the labels in the legend may not correctly line up with what's shown.

Make sure to pass in the label argument into the sns plotting function with the label you want associated with that individual plot. For example,

```
sns.histplot(means_arr, label = 'simulated values') # informative label name
plt.title('Simulated values')
plt.plot(original, 10, 'bo', label = 'original test statistic') # informative label name
plt.legend(loc = 'upper left') # can specify location of legend
```

6.1 The y-axis of my histplot shows the count, not the density

Look into the sns.histplot documentation and see what arguments the stat parameter takes in. By default, stat=count, and the number of elements in each histogram bin is the y axis. But if you wanted to normalize the distribution such that the total area is 1, consider passing stat=density into the plot function.

6.2 I'm having trouble labeling the axes/title of my graph

To label the axes and title of a graph, we use the following syntax:

```
plt.xlabel("x axis name")
plt.ylabel("y axis name")
plt.title("graph title")
```

Where plt.xlabel, plt.ylabel, and plt.title are matplotlib functions that we call.

However, we often see students use the following incorrect syntax to try and label their plot:

```
plt.xlabel = "x name"
plt.ylabel = "y name"
plt.title = "graph title"
```

Now, instead of plt.xlabel, plt.ylabel, and plt.title being functions, they are strings. Trying to call one of the labelling function using the correct syntax in the next few cells (ie.plt.xlabel("x name")) will result in a TypeError: str object is not callable. If this happens to you, comb through your notebook and look for places when you used the incorrect syntax. After fixing it, restart your kernel and rerun your cells.

6.3 My sns.lineplot has an unwanted shaded region around the solid lines.

Note: the following examples are taken from sns.lineplot's documentation.

sns.lineplot gives us a clean line when each x value has one y value. For example, the table

Year	May
1948	120
1949	122
1950	123
i :	:

will give us a clean plot because each year corresponds to a single "Number of Flights in May" value.

When each x value has multiple y values, sns.lineplot will automatically plot a shaded region around the solid line, where the solid line is the mean of the y values for that x value (think of a groupby on x aggregated by .mean() on y) and the shaded region is the 95% confidence interval (read more about confidence intervals in the Data 8 textbook). For example, the table

Year	May
1948	115
1948	120
1948	125
1949	118
1949	122
1949	126
<u>:</u>	:

will plot a lineplot with a shaded region.

If you do not want the shaded region, aggregate the data such that there is only one y-value for a given x-value; then, make the plot.

7 Project A1 Common Questions

7.1 Question 6

7.1.1 TypeError: could not convert string to float: 'SF'

Type errors like these usually stem from applying a numeric aggregation function to a non-numeric column as described in the pandas section of the debugging guide.

Aggregation functions like np.median and np.mean are only well-defined for columns with numeric types like int and float. Your code is likely trying to aggregate across all columns in training_data, including those of type str. Instead of aggregating across the entire DataFrame, try just selecting the relevant columns.

7.1.2 TypeError: unhashable type: 'Series'

This error can occur if you try and use Python's in to check whether values in a Series are contained in a list. If you're trying to perform boolean filtering in this manner, you should look into the .isin (documentation) function as introduced in HW 2.

7.2 Question 7

7.2.1 I'm not sure how to use sklearn to do One Hot Encoding

A good starting point is to revisit the One Hot Encoding question in Lab 7. It's recommended you look through this portion of the walkthrough, so you have a good understanding of how to use the OneHotEncoder object. Pay attention to what each variable represents and the expected outputs of the functions used. Can you map the logic from the lab to this project? A nice way to start is to make a new cell and experiment with examples from the documentation.

7.2.2 My OHE columns contain a lot of NaN values

This may happen if you try and merge the OHE columns with the training_data table without making sure both DataFrame have the *same index values*. Look into the pd.merge documentation for ways to resolve this.

8 Project A2 Common Questions

8.1 Questions 5d and 5f

8.1.1 General Debugging Tips

Question 5 is a challenging question that mirrors a lot of data science work in the real world: cleaning, exploring, and transforming data; fitting a model, working with a pre-defined pipeline and evaluating your model's performance. Here are some general debugging tips to make the process easier:

- Separate small tasks into helper functions, especially if you will execute them multiple times. For example, a helper function that one-hot encodes a categorical variable may be helpful as you could perform it on multiple such columns. If you're parsing a column with RegEx, it also might be a good idea to separate it to a helper function. This allows you to verify that you're not making errors in these small tasks and prevents unknown bugs from appearing.
- Feel free to make new cells to play with the data! As long as you delete them afterward, it will not affect the autograder.
- The feature_engine_final function looks daunting at first, but start small. First, try and implement a model with a single feature to get familiar with how the pipeline works, then slowly experiment with adding one feature at a time and see how that affects your training RMSE.

8.1.2 My training RMSE is low, but my validation/test RMSE is high

Your model is likely overfitting to the training data and does not generalize to the test set. Recall the bias-variance tradeoff discussed in lecture. As you add more features and make your model more complex, it is expected that your training error will decrease. Your validation and test error may also decrease initially, but if your model is too complex, you end up with high validation and test RMSE.

To decrease model complexity, consider visualizing the relationship between the features you've chosen with the (Log) Sale Price and removing features that are not highly correlated. Removing outliers can also help your model generalize better and prevent it from fitting to noise in the data. Methods like cross-validation allow you to get a better sense of where you

lie along the validation error curve. Feel free to take a look at the code used in Lecture 16 if you're confused on how to implement cross-validation.

8.1.3 ValueError: Per-column arrays must each be 1-dimensional

If you're passing the tests for question 5d but getting this error in question 5f, then your Y variable is likely a DataFrame, not a Series. sklearn models like LinearRegression expect X to be a 2D datatype (ie. DataFrame, 2D NumPy array) and Y to be a 1D datatype (ie. Series, 1D NumPy array).

8.1.4 KeyError: 'Sale Price'/KeyError: 'Log Sale Price'

KeyErrors are raised when a column name does not exist in your DataFrame. You could be getting this error because:

- The test set does not contain a "(Log) Sale Price" as that's what we're trying to predict. Make sure you only reference the "(Log) Sale Price" column when working with training data (is_test_set=False).
- You dropped the "Sale Price" column twice in your preprocessing code.

8.1.5 Value Error: could not convert string to float

This error usually occurs when your final design matrix has non-numeric columns. The fit function of a lm.LinearRegression object can take a pandas DataFrame as the X argument, but requires that the DataFrame is only composed of numeric values.

8.1.6 ValueError: Input X contains infinity or a value too large for dtype('float64')

The reason why your X data contains infinity is likely because you are taking the logarithm of 0 somewhere in your code. To prevent this, try:

- Adding a small number to the features that you want to perform the log transformation on so that all values are positive and greater than 0. Note that whatever value you add to your train data should also be added to your test data.
- Removing zeroes before taking the logarithm. Note that this is only possible on the training data as you cannot drop rows from the test set.

8.1.7 ValueError: Input X contains NaN

The reason why your design matrix X contains NaN values is likely because you take the log of a negative number somewhere in your code. To prevent this, try:

- Shifting the range of values for features that you want to perform the logging operation on to positive values greater than 0. Note that whatever value you add to your train data should also be added to your test data.
- Removing negative values before taking the log. Note that this is only possible on the training data as you cannot drop rows from the test set.

8.1.8 ValueError: The feature names should match those that were passed during fit

This error is followed by one or both of the following:

```
Feature names unseen at fit time:
- FEATURE NAME 1
- FEATURE NAME 2
...

Feature names seen at fit time, yet now missing
- FEATURE NAME 1
- FEATURE NAME 2
```

This error occurs if the columns/features you're passing in for the test dataset aren't the same as the features you used to train the model. sklearn's models expect the testing data's column names to match the training data's. The features listed under Feature names unseen at fit time are columns that were present in the training data but not the testing data, and features listed under Feature names seen at fit time, yet now missing were present in the testing data but not the training data.

Potential causes for this error:

- Your preprocessing for X is different for training and testing. Double-check your code in feature_engine_final! Besides removing any references to 'Sale Price' and code that would remove rows from the test set, your preprocessing should be the same.
- Some one-hot-encoded categories are present in training but not in testing (or vice versa). For example, let's say that the feature "Data100" has categories "A", "B", "C", and "D". If "A", "B", and "C" are present in the training data, but "B", "C", and "D" are present in the testing data, you will get this error:

The feature names should match those that were passed during fit. Feature names unseen a - Data100_D ...

Feature names seen at fit time, yet now missing - Data100 A

8.2 Gradescope

8.2.1 I don't have many Gradescope submissions left

If you're almost out of Gradescope submissions, try using k-fold cross-validation to check the accuracy of your model. Results from cross-validation will be closer to the test set accuracy than results from the training data. Feel free to take a look at the code used in Lecture 16 if you're confused on how to implement cross-validation.

8.2.2 "Wrong number of lines (___ instead of ___)"

This occurs when you remove outliers when preprocessing the testing data. *Please do not remove any outliers from your test set.* You may only remove outliers in training data.

8.2.3 Numerical Overflow

This error is caused by overly large predictions that create an extremely large RMSE. The cell before you generate your submission runs submission_df["Value"].describe(), which returns some summary statistics of your predictions. Your maximum value for Log Sale Price should not be over 25.

For your reference, a log sale price of 25 corresponds to a sale price of $e^{25} \approx 70$ billion, which is far bigger than anything found in the dataset. If you see such large predictions, you can try removing outliers from the *training* data or experimenting with new features so that your model generalizes better.