STA 130 HW1

September 10, 2024

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STA 130 Week 1 Homework
```

QUESTION 1

 $https://chatgpt.com/share/7f42c95d-01cc-4cbf-a7a3-57edbd7f18e7 \quad (ChatGPT \quad conversation \quad for \quad question \ 1)$

```
Missing values in each column:
row_n
                0
id
                1
                0
name
gender
species
                0
birthday
                0
personality
                0
song
               11
phrase
                0
full_id
                0
url
dtype: int64
Total missing values in the dataset: 12
QUESTION 2
```

 $Chat GPT\ conversation\ link\ for\ question\ 2:\ https://chatgpt.com/share/715a35bc-f635-4c7c-8266-52f10906a4aa$

```
Number of rows: 391
Number of columns: 11
```

Definition of "observation": An observation is an individual data point in a data set. An observation about an individual includes all their characteristics that are being included in the data set.

Definition of "variable": A variable is one single characteristic of an individual, which can be either quantitative (numerical) or qualitative (non-numerical), depending on which one makes sense for measuring that particular piece of data

QUESTION 3

ChatGPT conversation link for question 3: https://chatgpt.com/share/a25cb5e9-8986-49ba-a9bf-f725bf0f9858

First few rows of the dataset:

Summary of each column:

	row_n	id	name	gender	species	${\tt birthday}$	personality	\
count	391.000000	390	391	391	391	391	391	
unique	NaN	390	391	2	35	361	8	
top	NaN	admiral	Admiral	${\tt male}$	cat	1-27	lazy	
freq	NaN	1	1	204	23	2	60	
mean	239.902813	NaN	NaN	NaN	NaN	NaN	NaN	

```
std
              140.702672
                               NaN
                                         NaN
                                                NaN
                                                         NaN
                                                                  NaN
                                                                               NaN
                2.000000
                               NaN
                                                NaN
                                                         NaN
                                                                  NaN
                                                                               NaN
     min
                                         NaN
     25%
              117.500000
                               NaN
                                         NaN
                                                NaN
                                                         NaN
                                                                  NaN
                                                                               NaN
     50%
              240.000000
                               NaN
                                        NaN
                                                NaN
                                                         NaN
                                                                  NaN
                                                                               NaN
              363.500000
                                         NaN
                                                         NaN
     75%
                               NaN
                                                NaN
                                                                  NaN
                                                                               NaN
              483.000000
                               NaN
                                         NaN
                                                         NaN
                                                                  NaN
                                                                               NaN
     max
                                                NaN
                                                full id \
                      song
                              phrase
                       380
                                 391
                                                     391
     count
                        92
                                 388
                                                     391
     unique
     top
              K.K. Country
                             wee one
                                      villager-admiral
                                   2
     freq
                         10
                                                       1
     mean
                       NaN
                                 NaN
                                                    NaN
                                                    NaN
     std
                       NaN
                                 NaN
     min
                       NaN
                                 NaN
                                                    NaN
     25%
                       NaN
                                 NaN
                                                    NaN
     50%
                       NaN
                                 NaN
                                                    NaN
     75%
                       NaN
                                 NaN
                                                    NaN
                       NaN
                                 NaN
                                                    NaN
     max
                                                               url
                                                               391
     count
     unique
                                                               391
     top
              https://villagerdb.com/images/villagers/thumb/...
     freq
                                                                 1
     mean
                                                               NaN
                                                               NaN
     std
     min
                                                               NaN
     25%
                                                               NaN
     50%
                                                               NaN
     75%
                                                               NaN
     max
                                                               NaN
[15]: # New summary after removing column 1 (row number) and including only
       →quantitative data to remove all the NaNs
      import pandas as pd
      url = 'https://raw.githubusercontent.com/rfordatascience/tidytuesday/master/
       ⇔data/2020/2020-05-05/villagers.csv'
      df = pd.read_csv(url)
      df_numeric = df.iloc[:, 1:]
      print("Summary of quantitative columns (excluding the first column):")
      print(df_numeric.describe())
```

Summary of quantitative columns (excluding the first column):

```
390
                           391
                                  391
                                           391
                                                     391
                                                                                 380
    count
                                                                 391
                 390
                           391
                                            35
                                                     361
    unique
                                                                    8
                                                                                 92
    top
             admiral
                      Admiral
                                 male
                                           cat
                                                    1-27
                                                                lazy K.K. Country
                                  204
                                            23
                                                                   60
                                                                                  10
    freq
                   1
                             1
                                                       2
              phrase
                                full id \
                                     391
    count
                 391
                 388
                                    391
    unique
    top
             wee one
                      villager-admiral
                   2
    freq
                                                              url
                                                              391
    count
                                                              391
    unique
             https://villagerdb.com/images/villagers/thumb/...
    top
    freq
                                                                1
    QUESTION 4
[5]: import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv¹
     df = pd.read_csv(url)
     num_rows, num_columns = df.shape
     print(f'Number of rows: {num rows}')
     print(f'Number of columns: {num_columns}')
     print("\nSummary of each column:")
     print(df.describe(include='all'))
    Number of rows: 891
    Number of columns: 15
    Summary of each column:
               survived
                              pclass
                                                               sibsp
                                                                            parch \
                                        sex
                                                     age
    count
             891.000000
                          891.000000
                                        891
                                             714.000000
                                                          891.000000
                                                                       891.000000
    unique
                    NaN
                                 NaN
                                          2
                                                    {\tt NaN}
                                                                 NaN
                                                                              NaN
    top
                    NaN
                                 {\tt NaN}
                                      male
                                                    NaN
                                                                 NaN
                                                                              NaN
                                        577
    freq
                    NaN
                                 NaN
                                                    NaN
                                                                 NaN
                                                                              NaN
    mean
               0.383838
                            2.308642
                                        NaN
                                              29.699118
                                                            0.523008
                                                                         0.381594
    std
               0.486592
                            0.836071
                                       NaN
                                              14.526497
                                                            1.102743
                                                                         0.806057
               0.000000
                            1.000000
                                               0.420000
                                                            0.000000
                                                                         0.000000
    min
                                       NaN
    25%
               0.000000
                            2.000000
                                       {\tt NaN}
                                              20.125000
                                                            0.000000
                                                                         0.000000
    50%
               0.000000
                            3.000000
                                              28.000000
                                                            0.000000
                                                                         0.000000
                                        NaN
```

name gender species birthday personality

song \

id

75%	1.000000	3.0000	00 Na	.N :	38.000000	1.00	0000	0.000	0000	
max	1.000000	3.0000	00 Na	.N 8	80.00000	8.00	0000	6.000	0000	
	fare	${\tt embarked}$	class	who	$adult_male$	deck	embark	_town	alive	\
count	891.000000	889	891	891	891	203		889	891	
unique	NaN	3	3	3	2	7		3	2	
top	NaN	S	Third	man	True	C	Southar	npton	no	
freq	NaN	644	491	537	537	59		644	549	
mean	32.204208	NaN	NaN	NaN	NaN	${\tt NaN}$		NaN	NaN	
std	49.693429	NaN	NaN	NaN	NaN	${\tt NaN}$		NaN	NaN	
min	0.000000	NaN	NaN	NaN	NaN	NaN		NaN	NaN	
25%	7.910400	NaN	NaN	NaN	NaN	NaN		NaN	NaN	
50%	14.454200	NaN	NaN	NaN	NaN	NaN		NaN	NaN	
75%	31.000000	NaN	NaN	NaN	NaN	NaN		NaN	NaN	
max	512.329200	NaN	NaN	NaN	NaN	NaN		NaN	NaN	
	alone									
	001									

count 891 2 unique top True freq 537 mean NaNstd NaN NaN min 25% NaN 50% NaN 75% NaN max NaN

Description for Question 4 (no chatbot used for this question):

The discrepancies between the total number of rows (891) and the count for each column (less that 891 for some columns) exist because of the missing values in the data set. For example, there is a large discrepancy for the column titled "deck," with only 203 values provided out of 891 possible. This is likely because, for most passengers, the deck they were on when the Titanic sank is unknown, and therefore not included in the data set.

QUESTION 5

(methods vs attributes):

In Python, an attribute is a variable which contains data about a particular class/instance/object. A method is a function that belongs to a class, and determines the behavior of its instances. Methods can access and change the attributes of their class or instance.

For example, the attribute "df.shape" provides the number of rows and columns for the data sets above, while the method "df.describe()" creates a table summary of the data, calculating the mean, standard deviation, max, min, and other characteristics of the data sets.

ChatGPT link for question 5: https://chatgpt.com/share/08df9a6d-4613-45a4-b92b-86548321ac46

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ChatGPT summaries (questions 1-5):

Question1 (https://chatgpt.com/share/7f42c95d-01cc-4cbf-a7a3-57edbd7f18e7):

"Here's a summary of our conversation:

1. Python Code for Missing Values:

I provided a Python script using the pandas library to import a CSV dataset and count missing values. The code reads a CSV file, counts missing values in each column, and prints the results.

2. Code for Specific URL:

You requested a specific example using a URL from a GitHub repository. I adapted the previous script to load the CSV file directly from the provided URL and count the missing values in the dataset."

Question2 (https://chatgpt.com/share/715a35bc-f635-4c7c-8266-52f10906a4aa):

"In this conversation, we discussed:

Determining Rows and Columns: To find the number of rows and columns in a dataset, you can use Python and the pandas library. The code reads the CSV file and uses the shape attribute to get these counts.

Observation in Statistics: An observation refers to a single data point or instance within a dataset. For example, in a dataset of test scores, each score is an observation.

Variable in Statistics: A variable is a characteristic or attribute that can take on different values. Variables can be quantitative (e.g., height, weight) or qualitative (e.g., hair color, major) and are essential for data analysis."

Question3 (https://chatgpt.com/share/a25cb5e9-8986-49ba-a9bf-f725bf0f9858):

"Here's a summary of our conversation:

- 1. Request: You asked how to provide summaries of the columns in a dataset from a specified URL using Python code.
- 2. Initial Solution:

I provided Python code using the pandas library to load the dataset and generate a summary of all columns using df.describe(include='all').

3. Refinement:

You requested to exclude the first column and only include quantitative values. I updated the code to exclude the first column and provide a summary of only the quantitative columns using df.iloc[:, 1:] and df_numeric.describe(). If you have any further questions or need more adjustments, feel free to ask!"

Question5 (https://chatgpt.com/share/08df9a6d-4613-45a4-b92b-86548321ac46):

"Sure! Here's a summary:

Attributes vs. Methods in Python:

Attributes are variables that store data or state within a class. They can be instance-specific (unique to each object) or class-wide (shared among all instances). Methods are functions defined within

a class that describe behaviors or actions. They can be instance methods (operate on instance attributes), class methods (operate on class-level data), or static methods (don't access instance or class data).

Class vs. Instance in Python:

A class is a blueprint for creating objects, defining the structure (attributes) and behaviors (methods) of the objects. An instance is a specific object created from a class, with its own unique set of attributes and the ability to use the methods defined by the class.

Mathematical Functions vs. Programming Functions:

Mathematical functions define relationships between inputs and outputs, focusing on abstract properties like continuity and limits. Programming functions are blocks of code designed to perform specific tasks, taking inputs, processing them, and potentially returning outputs. They are practical and concerned with code execution within a program.

Each type of function or concept serves its own purpose within its respective domain—mathematical functions for theoretical analysis and programming functions for practical code execution."

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POST LECTURE QUESTIONS

QUESTION 6

(no ChatGPT used)

- a) The "count" is the total number of values a variable in the data set has
- b) The "mean" is the average of all of a variable's values in the data set (sum of all values / number of values)
- c) The "std" is the standard deviation. This describes how far the values are from the mean. A higher standard deviation means that there is a larger range of values.
- d) The "min" is the lowest value a variable takes in the data set.
- e) "25%" is the 25th percentile, which means that 25% of the values are lower than it, and 75% are greater.
- f) "50%" is the 50th percentile, or median, meaning that half of the values are lower than this, and half are greater.
- g) "75% is the 75th percentile, so 25% of values are greater than it, and 75% are lower.
- h) The "max" is the greatest value that a variable takes in the data set.

QUESTION 7

chatgpt link: https://chatgpt.com/share/7b87e828-5928-4bfb-a570-cf2588184604

- 7.1) df.dropna() is used to remove rows (or columns, but the default is rows) that contain missing values from a dataset. So, it's used primarily in cases when you want to clean/organize the data set better by removing missing or incomplete sections.
- 7.2) del df['col'] is used to delete entire columns from the data set. If you don't need a column and want it completely gon, you would use this instead of df.dropna().

7.3) Applying del df['col'] first will delete the column and all its missing (NaN) values. This will prevent the df.dropna() function from deleting rows which have missing values only in the deleted column, so not as many rows will be deleted, resulting in a larger data set.

```
[1]: # question 7.4 (using the titanic data set from above)
     # before using del df['col'] or df.dropna()
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      GCSV'
     df = pd.read_csv(url)
     num_rows, num_columns = df.shape
     print(f'Number of rows: {num_rows}')
     print(f'Number of columns: {num_columns}')
     print("\nSummary of each column:")
     print(df.describe(include='all'))
    Number of rows: 891
    Number of columns: 15
    Summary of each column:
               survived
                                                                sibsp
                                                                             parch
                              pclass
                                                      age
                                        sex
             891.000000
                          891.000000
                                        891
                                              714.000000
                                                           891.000000
                                                                        891.000000
    count
                                          2
    unique
                     NaN
                                  NaN
                                                     NaN
                                                                  NaN
                                                                                NaN
    top
                     NaN
                                  NaN
                                       male
                                                     NaN
                                                                  NaN
                                                                               NaN
    freq
                     NaN
                                  NaN
                                        577
                                                     NaN
                                                                  NaN
                                                                                NaN
               0.383838
                            2.308642
                                               29.699118
                                                             0.523008
                                                                          0.381594
    mean
                                        NaN
    std
               0.486592
                            0.836071
                                        NaN
                                               14.526497
                                                             1.102743
                                                                          0.806057
               0.000000
                            1.000000
                                        NaN
                                                0.420000
                                                             0.000000
                                                                          0.00000
    min
    25%
               0.000000
                            2.000000
                                               20.125000
                                                             0.000000
                                        NaN
                                                                          0.000000
    50%
               0.000000
                            3.000000
                                        NaN
                                               28.000000
                                                             0.000000
                                                                          0.00000
    75%
               1.000000
                            3.000000
                                        NaN
                                               38.000000
                                                             1.000000
                                                                          0.000000
               1.000000
                            3.000000
                                               80.000000
                                                             8.000000
                                                                          6.000000
    max
                                        NaN
                    fare embarked
                                    class
                                           who adult_male deck
                                                                  embark_town alive
    count
             891.000000
                               889
                                      891
                                            891
                                                        891
                                                             203
                                                                           889
                                                                                  891
                                 3
                                        3
                                              3
                                                          2
                                                               7
                                                                             3
                                                                                    2
    unique
                     NaN
                                 S
                                                               C
                     NaN
                                    Third
                                                                  Southampton
    top
                                           man
                                                      True
                                                                                   no
                                      491
                                                        537
                                                                           644
    freq
                     NaN
                               644
                                            537
                                                              59
                                                                                  549
                                                       {\tt NaN}
    mean
              32.204208
                              NaN
                                      NaN
                                           {\tt NaN}
                                                             {\tt NaN}
                                                                           NaN
                                                                                  NaN
    std
              49.693429
                              NaN
                                      NaN
                                            NaN
                                                       NaN
                                                             NaN
                                                                           NaN
                                                                                  NaN
               0.000000
                              NaN
                                            NaN
                                                        NaN
                                                                           NaN
                                                                                  NaN
    min
                                      NaN
                                                             NaN
    25%
               7.910400
                              NaN
                                      NaN
                                           NaN
                                                        NaN
                                                             NaN
                                                                           NaN
                                                                                  NaN
```

```
50%
              14.454200
                              {\tt NaN}
                                                       NaN NaN
                                                                                  NaN
                                      \mathtt{NaN}
                                           NaN
                                                                           NaN
    75%
              31.000000
                                                                                  NaN
                              NaN
                                      {\tt NaN}
                                           NaN
                                                       NaN NaN
                                                                           NaN
             512.329200
                              NaN
                                           NaN
                                                       NaN NaN
                                                                           NaN
                                                                                 NaN
    max
                                      {\tt NaN}
            alone
              891
    count
    unique
    top
             True
              537
    freq
    mean
              NaN
    std
              NaN
    min
              NaN
    25%
              NaN
    50%
              NaN
    75%
              NaN
    max
              NaN
[2]: # same data set, after applying first del df['col'] to the "deck" column, then
      \hookrightarrow df.dropna()
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
     ⇔csv'
     df = pd.read_csv(url)
     num_rows, num_columns = df.shape
     print(f'Number of rows: {num_rows}')
     print(f'Number of columns: {num_columns}')
     del df['deck']
     print(df.dropna())
     # this removed the "deck column," and removed all other rows with missing values
     # so, the data set now has 14 columns and 712 rows, rather than the original 15_{\sqcup}
      ⇔columns and 891 rows
    Number of rows: 891
    Number of columns: 15
```

survived pclass age sibsp parch fare embarked class \ sex Third 0 0 3 male 22.0 1 0 7.2500 S 1 1 female 38.0 1 0 71.2833 First 1 2 1 3 female 26.0 0 0 7.9250 Third 3 1 1 female 35.0 1 0 53.1000 S First 0 male 35.0 0 0 8.0500 Third

```
. .
                                   39.0
885
             0
                       3
                          female
                                               0
                                                       5
                                                          29.1250
                                                                            Q
                                                                                Third
                      2
                                   27.0
             0
                            male
                                               0
                                                       0
                                                          13.0000
                                                                            S
                                                                               Second
886
              1
                       1
                                   19.0
                                               0
                                                       0
                                                          30.0000
                                                                            S
                                                                                First
887
                          female
              1
                       1
                                               0
                                                                            C
889
                            male
                                   26.0
                                                       0
                                                           30.0000
                                                                                First
              0
                       3
                                   32.0
                                               0
890
                            male
                                                       0
                                                            7.7500
                                                                            Q
                                                                                Third
        who
              adult_male
                           embark town alive
                                                 alone
0
                                                 False
       man
                    True
                           Southampton
                                            no
                   False
1
     woman
                              Cherbourg
                                            yes
                                                 False
2
                   False
                           Southampton
     woman
                                            yes
                                                  True
3
                           Southampton
     woman
                   False
                                           yes
                                                 False
4
        man
                    True
                           Southampton
                                            no
                                                  True
. .
885
     woman
                   False
                            Queenstown
                                            no
                                                 False
                           Southampton
                                                  True
886
                    True
        man
                                            no
887
                   False
                           Southampton
                                                  True
     woman
                                            yes
                    True
889
        man
                              Cherbourg
                                            yes
                                                  True
                            Queenstown
890
                    True
                                                  True
        man
                                            no
```

[712 rows x 14 columns]

QUESTION 8

```
[3]:
                                                          25%
                                                                 50%
                                                                        75%
                                            std
                                                   min
                 count
                              mean
                                                                              max
     survived
     0
                 424.0
                         30.626179
                                     14.172110
                                                  1.00
                                                         21.0
                                                               28.0
                                                                      39.0
                                                                             74.0
     1
                 290.0
                         28.343690
                                     14.950952
                                                  0.42
                                                         19.0
                                                               28.0
                                                                      36.0
                                                                             80.0
```

- 8.1) explanation: This code took all the individuals from the "survived" column and divided them into groups of survivors, and non-survivors. It then calculated statistics about the ages of the people in each group, such as their average age, median age, oldest, and youngest individuals.
- 8.2) df.describe does not remove missing values, so it can't calculate the statistics for any column with any missing values, instead giving the result NaN. Therefore, if I did df.describe, no rows would be removed, and so there would be a total count of 891. However, df.groupby DOES remove rows with missing values in the relevant columns ("survived" and "age") automatically, so the count in the table from question 8.1 only totals to 714.

```
[7]: \# 8.3.A)
     # did not import pandas
    url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
     ⇔csv'
    df = pd.read_csv(url)
    df.groupby("survived")["age"].describe()
     NameError
                                               Traceback (most recent call last)
     Cell In[7], line 4
           1 # 8.3.A)
           3 url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/
      →titanic.csv¹
      ----> 4 df = pd.read_csv(url)
           6 df.groupby("survived")["age"].describe()
     NameError: name 'pd' is not defined
[8]: # 8.3.A) fixed with ChatGPT and Google search help
    import pandas as pd
    url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
     ⇔csv'
    df = pd.read_csv(url)
    df.groupby("survived")["age"].describe()
[8]:
              count
                          mean
                                      std
                                            min
                                                  25%
                                                        50%
                                                              75%
    survived
              424.0 30.626179 14.172110 1.00 21.0 28.0 39.0 74.0
    1
              290.0 28.343690 14.950952 0.42 19.0 28.0 36.0 80.0
[9]: # 8.3.B)
    import pandas as pd
    url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanics.
     ⇔csv' # added an s to titanic in url
    df = pd.read_csv(url)
    df.groupby("survived")["age"].describe()
```

```
HTTPError
                                                        Traceback (most recent call last)
Cell In[9], line 6
        3 import pandas as pd
        5 url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/
 ⇔titanics.csv'
----> 6 df = pd.read csv(url)
        8 df.groupby("survived")["age"].describe()
File /opt/conda/lib/python3.11/site-packages/pandas/io/parsers/readers.py:948,
 in read_csv(filepath_or_buffer, sep, delimiter, header, names, index_col,_u 
usecols, dtype, engine, converters, true_values, false_values,_u
skipinitialspace, skiprows, skipfooter, nrows, na_values, keep_default_na,_u
una_filter, verbose, skip_blank_lines, parse_dates, infer_datetime_format,_u
keep_date_col, date_parser, date_format, dayfirst, cache_dates, iterator,_u
uchunksize, compression, thousands, decimal, lineterminator, quotechar,_u
undergoon, doublequote, escapechar, comment, encoding, encoding_errors, dialect
undergoon, datume_datume_baskerd)
  ⇔storage_options, dtype_backend)
     935 kwds defaults = refine defaults read(
     936
                dialect,
     937
                delimiter,
    (...)
     944
                dtype_backend=dtype_backend,
     945 )
     946 kwds.update(kwds_defaults)
--> 948 return _read(filepath_or_buffer, kwds)
File /opt/conda/lib/python3.11/site-packages/pandas/io/parsers/readers.py:611,u
  →in read(filepath or buffer, kwds)
     608 _validate_names(kwds.get("names", None))
     610 # Create the parser.
--> 611 parser = TextFileReader(filepath_or_buffer, **kwds)
     613 if chunksize or iterator:
     614
                return parser
File /opt/conda/lib/python3.11/site-packages/pandas/io/parsers/readers.py:1448,
  →in TextFileReader.__init__(self, f, engine, **kwds)
                self.options["has_index_names"] = kwds["has_index_names"]
    1447 self.handles: IOHandles | None = None
-> 1448 self._engine = self._make_engine(f, self.engine)
File /opt/conda/lib/python3.11/site-packages/pandas/io/parsers/readers.py:1705,

    in TextFileReader. make engine(self, f, engine)

                if "b" not in mode:
    1703
                     mode += "b"
    1704
-> 1705 self.handles = get_handle(
    1706
                f,
    1707
                mode,
                encoding=self.options.get("encoding", None),
    1708
                compression=self.options.get("compression", None),
    1709
```

```
1710
                                        memory_map=self.options.get("memory_map", False),
          1711
                                        is_text=is_text,
                                        errors=self.options.get("encoding_errors", "strict"),
          1712
                                        storage_options=self.options.get("storage_options", None),
          1713
          1714
          1715 assert self.handles is not None
          1716 f = self.handles.handle
File /opt/conda/lib/python3.11/site-packages/pandas/io/common.py:718, in in in the condition of the conditio
    aget handle(path or buf, mode, encoding, compression, memory map, is text,
    ⇔errors, storage_options)
             715
                                       codecs.lookup_error(errors)
             717 # open URLs
--> 718 ioargs = _get_filepath_or_buffer(
             719
                                       path or buf,
             720
                                        encoding=encoding,
                                        compression=compression,
             721
             722
                                        mode=mode,
             723
                                        storage options=storage options,
             724
             726 handle = ioargs.filepath_or_buffer
             727 handles: list[BaseBuffer]
File /opt/conda/lib/python3.11/site-packages/pandas/io/common.py:372, in packages/pandas/io/common.py:372, in packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/packages/
    م_get_filepath_or_buffer(filepath_or_buffer, encoding, compression, mode, و
    ⇔storage_options)
             370 # assuming storage options is to be interpreted as headers
             371 req_info = urllib.request.Request(filepath_or_buffer,_
    ⇔headers=storage options)
 --> 372 with urlopen(req_info) as req:
                                        content_encoding = req.headers.get("Content-Encoding", None)
             373
             374
                                        if content_encoding == "gzip":
             375
                                                     # Override compression based on Content-Encoding header
File /opt/conda/lib/python3.11/site-packages/pandas/io/common.py:274, in_
    →urlopen(*args, **kwargs)
             268 """
             269 Lazy-import wrapper for stdlib urlopen, as that imports a big chunk of
             270 the stdlib.
             271 """
             272 import urllib.request
--> 274 return urllib.request.urlopen(*args, **kwargs)
File /opt/conda/lib/python3.11/urllib/request.py:216, in urlopen(url, data, u
    →timeout, cafile, capath, cadefault, context)
             214 else:
                                        opener = _opener
             215
--> 216 return opener open(url, data, timeout)
```

```
File /opt/conda/lib/python3.11/urllib/request.py:525, in OpenerDirector.
        ⇔open(self, fullurl, data, timeout)
           523 for processor in self.process_response.get(protocol, []):
                   meth = getattr(processor, meth name)
           524
                   response = meth(req, response)
       --> 525
           527 return response
      File /opt/conda/lib/python3.11/urllib/request.py:634, in HTTPErrorProcessor.
        http_response(self, request, response)
           631 # According to RFC 2616, "2xx" code indicates that the client's
           632 # request was successfully received, understood, and accepted.
           633 if not (200 <= code < 300):
       --> 634
                   response = self.parent.error(
                       'http', request, response, code, msg, hdrs)
           637 return response
      File /opt/conda/lib/python3.11/urllib/request.py:563, in OpenerDirector.
        ⇔error(self, proto, *args)
           561 if http err:
                   args = (dict, 'default', 'http_error_default') + orig_args
           562
       --> 563
                  return self. call chain(*args)
      File /opt/conda/lib/python3.11/urllib/request.py:496, in OpenerDirector.
        →_call_chain(self, chain, kind, meth_name, *args)
           494 for handler in handlers:
                   func = getattr(handler, meth_name)
           495
                  result = func(*args)
       --> 496
                   if result is not None:
           497
           498
                       return result
      File /opt/conda/lib/python3.11/urllib/request.py:643, in HTTPDefaultErrorHandle.
        ⇔http_error_default(self, req, fp, code, msg, hdrs)
           642 def http_error_default(self, req, fp, code, msg, hdrs):
       --> 643
                   raise HTTPError(req.full url, code, msg, hdrs, fp)
      HTTPError: HTTP Error 404: Not Found
[11]: # 8.3.B) fixed with ChatGPT help
      import pandas as pd
      url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
       ⇔csv'
      df = pd.read_csv(url)
      df.groupby("survived")["age"].describe()
```

```
survived
               424.0 30.626179 14.172110 1.00 21.0 28.0 39.0 74.0
               290.0 28.343690 14.950952 0.42 19.0 28.0 36.0 80.0
[12]: # 8.3.C)
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv'
     df = pd.read_csv(url)
     DF.groupby("survived")["age"].describe() # capitalized DF; python is case_
       \hookrightarrow sensitive
                                                Traceback (most recent call last)
      NameError
      Cell In[12], line 8
            5 url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/
       ⇔titanic.csv'
            6 df = pd.read_csv(url)
       ----> 8 DF.groupby("survived")["age"].describe()
      NameError: name 'DF' is not defined
[13]: # 8.3.C) fixed with ChatGPT
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv¹
     df = pd.read_csv(url)
     df.groupby("survived")["age"].describe()
[13]:
                                                   25%
                                                         50%
                                       std
                                             min
                                                               75%
                count
                           mean
                                                                     max
     survived
               424.0 30.626179 14.172110 1.00 21.0 28.0 39.0 74.0
               290.0 28.343690 14.950952 0.42 19.0 28.0 36.0 80.0
[14]: # 8.3.D)
     import pandas as pd
```

std min 25%

50%

75%

[11]:

count

mean

```
url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
       GCSV'
     df = pd.read_csv(url # removed ()
     df.groupby("survived"["age"].describe() # removed ()
        Cell In[14], line 8
          df.groupby("survived"["age"].describe() # removed ()
      SyntaxError: '(' was never closed
[15]: # 8.3.D) fixed with ChatGPT and Google search
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv'
     df = pd.read_csv(url) # added )
     df.groupby("survived")["age"].describe() # added )
[15]:
                                             min 25%
                                                         50%
                                                               75%
               count
                           mean
                                       std
                                                                     max
     survived
               424.0 30.626179 14.172110 1.00 21.0 28.0 39.0 74.0
               290.0 28.343690 14.950952 0.42 19.0 28.0 36.0 80.0
[16]: # 8.3.E)
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
       ocsv'
     df = pd.read_csv(url)
     df.group_by("survived")["age"].deescrib() # added _ and miss-spelled describe
      AttributeError
                                                Traceback (most recent call last)
      /tmp/ipykernel_86/2399913520.py in ?()
            5 url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/
       ⇔titanic.csv¹
            6 df = pd.read_csv(url)
```

```
---> 8 df.group_by("survived")["age"].deescrib() # added _ and miss-spelled_
        \hookrightarrowdescribe
      /opt/conda/lib/python3.11/site-packages/pandas/core/generic.py in ?(self, name)
         6200
                          and name not in self. accessors
         6201
                          and self._info_axis.
        6202
         6203
                          return self[name]
      -> 6204
                      return object.__getattribute__(self, name)
      AttributeError: 'DataFrame' object has no attribute 'group_by'
[17]: # 8.3.E) fixed with ChatGPT
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv'
     df = pd.read_csv(url)
      # Correct method is groupby, not group_by
     df.groupby("survived")["age"].describe()
      # even though the error message above only listed the \_ error, ChatGPT also_{\sqcup}
       ⇒caught and fixed the spelling mistake!
[17]:
               count
                           mean
                                       std
                                            min
                                                   25%
                                                        50%
                                                              75%
                                                                    max
     survived
               424.0 30.626179 14.172110 1.00 21.0 28.0 39.0 74.0
               290.0 28.343690 14.950952 0.42 19.0 28.0 36.0 80.0
[18]: # 8.3.F)
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv'
     df = pd.read_csv(url)
     df.groupby("Sex")["Pclass"].describe()
      KeyError
                                                Traceback (most recent call last)
      Cell In[18], line 8
            5 url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/
       ⇔titanic.csv¹
            6 df = pd.read_csv(url)
```

```
----> 8 df.groupby("Sex")["Pclass"].describe()
 File /opt/conda/lib/python3.11/site-packages/pandas/core/frame.py:8869, in_
  DataFrame.groupby(self, by, axis, level, as index, sort, group keys, observed u
  ⇔dropna)
    8866 if level is None and by is None:
             raise TypeError("You have to supply one of 'by' and 'level'")
    8867
 -> 8869 return DataFrameGroupBy(
    8870
             obj=self,
    8871
             keys=by,
    8872
             axis=axis,
    8873
             level=level,
    8874
             as index=as index,
    8875
             sort=sort,
    8876
             group keys=group keys,
    8877
             observed=observed,
    8878
             dropna=dropna,
    8879
 File /opt/conda/lib/python3.11/site-packages/pandas/core/groupby/groupby.py:
  →1278, in GroupBy.__init__(self, obj, keys, axis, level, grouper, exclusions,
  ⇒selection, as_index, sort, group_keys, observed, dropna)
    1275 self.dropna = dropna
    1277 if grouper is None:
 -> 1278
             grouper, exclusions, obj = get grouper(
    1279
                 obj,
    1280
                 keys,
    1281
                 axis=axis,
    1282
                 level=level,
    1283
                 sort=sort,
                 observed=False if observed is lib.no_default else observed,
    1284
    1285
                 dropna=self.dropna,
    1286
    1288 if observed is lib.no_default:
    1289
             if any(ping._passed_categorical for ping in grouper.groupings):
 File /opt/conda/lib/python3.11/site-packages/pandas/core/groupby/grouper.py:
  41009, in get grouper(obj, key, axis, level, sort, observed, validate, dropna)
    1007
                 in_axis, level, gpr = False, gpr, None
    1008
             else:
 -> 1009
                 raise KeyError(gpr)
    1010 elif isinstance(gpr, Grouper) and gpr.key is not None:
             # Add key to exclusions
    1012
             exclusions.add(gpr.key)
 KeyError: 'Sex'
```

```
[19]: # 8.3.F) fixed with ChatGPT
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
     df = pd.read_csv(url)
     df.groupby("survived")["age"].describe()
       # chatGPT didn't catch the capitalization error, but told me to check spelling
Γ197:
                                       std min
                                                   25%
                                                         50%
                                                               75%
               count
                           mean
     survived
               424.0 30.626179 14.172110 1.00 21.0 28.0 39.0 74.0
               290.0 28.343690 14.950952 0.42 19.0 28.0 36.0 80.0
[20]: # 8.3.G)
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
      ⇔csv'
     df = pd.read_csv(url)
     df.groupby(survived)[age].describe() # removed "" for the stuff in parentheses/
       ⇒brackets
      NameError
                                                Traceback (most recent call last)
      Cell In[20], line 8
            5 url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/
        ⇔titanic.csv'
            6 df = pd.read_csv(url)
       ----> 8 df.groupby(survived)[age].describe() # removed "" for the stuff in
        →parentheses/brackets
      NameError: name 'survived' is not defined
[21]: # 8.3.G) fixed with ChatGPT
     import pandas as pd
     url = 'https://raw.githubusercontent.com/mwaskom/seaborn-data/master/titanic.
       GCSV'
     df = pd.read_csv(url)
```

```
result = df.groupby('survived')['age'].describe()
#Chat GPT had trouble catching this until I got very specific
print(result)
```

	count	mean	std	min	25%	50%	75%	max
survived								
0	424.0	30.626179	14.172110	1.00	21.0	28.0	39.0	74.0
1	290.0	28.343690	14.950952	0.42	19.0	28.0	36.0	80.0

Question 8.3 overall impression

For all the errors above, it was much faster to fix them with ChatGPT, and for all but A and D, a Google Search was almost completely useless for solving the problem, as it gave very generic, technical answers, not simply worded, specific ones like the chatbot.

QUESTION 9

Somewhat

ChatGPT links and summaries for Questions 6-8:

Question 6: no chat gpt used

Question 7: https://chatgpt.com/share/7b87e828-5928-4bfb-a570-cf2588184604

Summary: "In this conversation, we discussed the difference between using df.dropna() and del df['col'] in Python.

df.dropna() is used to remove rows or columns with missing values (NaN), while del df['col'] is used to delete a specific column by name, regardless of whether it contains missing values. The order of these operations can matter depending on your goals, especially if missing values are present. Additionally, there was an issue with using print(del df['col']) in the user's code, as del is a statement, not a function, and should not be used inside print(). The solution was to remove the print() around del and ensure the column name exists in the DataFrame. The corrected code was provided."

Question 8: https://chatgpt.com/share/c18b554a-671a-452c-b026-5a8090357f38 (part 1) and https://chatgpt.com/share/f39dbf07-b7d4-47a0-8d4c-f1462baa5946 (part 2)

Summary: "In this chat, you encountered two issues with your Python code using the Titanic dataset:

KeyError with groupby and describe:

Issue: The error was due to trying to access a column that didn't exist or was incorrectly referenced. Solution: Verify that the columns you are referencing are correct and exist in the DataFrame. Use print(df.columns) to check column names and ensure they are correctly spelled in your code. NameError with Column References:

Issue: You tried to reference column names without quotes in the groupby method, causing a NameError. Solution: Column names should be enclosed in quotes. For example, use df.groupby('survived')['age'].describe() instead of df.groupby(survived)[age].describe(). To avoid these errors:

Always check that column names are correct and exist in the DataFrame. Use strings (quotes) to reference column names in DataFrame methods."