4.5.1 Find Missing Values

Maria is aware that at least one of the datasets needs to be cleaned before any analysis can be performed. She would like you to use Pandas to do a more thorough inspection of the datasets than you did when you opened them with Excel. Cleaning the data is essential because any missing, malformed, or incorrect data in the rows can jeopardize the analysis.

As a first step in the data-cleaning process, we'll determine if there are missing values in the rows of the CSV files. Let's look at each CSV file separately.

First, open schools_complete.csv. We can see that each row contains a School ID, school name, type of school, student size, and budget. Therefore, there are no missing values in any of the rows, which are also called rows with **null values**. See the following image:

4	Α	В	С	D	E
1	School ID	school_name	type	size	budget
2	0	Huang High School	District	2917	1910635
3	1	Figueroa High School	District	2949	1884411
4	2	Shelton High School	Charter	1761	1056600

5	3	Hernandez High School	District	4635	3022020
6	4	Griffin High School	Charter	1468	917500
7	5	Wilson High School	Charter	2283	1319574
8	6	Cabrera High School	Charter	1858	1081356
9	7	Bailey High School	District	4976	3124928
10	8	Holden High School	Charter	427	248087
11	9	Pena High School	Charter	962	585858
12	10	Wright High School	Charter	1800	1049400
13	11	Rodriguez High School	District	3999	2547363
14	12	Johnson High School	District	4761	3094650
15	13	Ford High School	District	2739	1763916
16	14	Thomas High School	Charter	1635	1043130

While schools_complete.csv has only 15 rows of data and one row for the headers, the student_complete.csv has 39,170 rows. It would be very time-consuming to find missing values in a file so large. Luckily, Pandas has a few methods that can help us determine whether there are missing values in large datasets: the <a href="count("

The count() Method

With the <code>count()</code> method, we can get a count of the rows for each column containing data. By default, "null" values are not counted, so you can often quickly identify which columns have missing data.

Let's use the <u>count()</u> method on the <u>school_data_df</u> DataFrame. Add the following code to a new cell and run the cell:

```
# Determine if there are any missing values in the school data.
school_data_df.count()
```

The output returns the name of the columns and the number of rows that are not null. For the school_data_df DataFrame, there are no missing values, because there are 15 rows that contain data in
schools_complete.csv. In the output, the number 15 is next to each column

header, as shown in the following image:

```
School ID
                 15
school name
                 15
                 15
type
size
                 15
                 15
budget
dtype: int64
```

These results confirm what we observed when we looked at the schools_complete.csv file.

Now let's use the same method on the student_data_df DataFrame. Add the following code to a new cell and run the cell:

```
# Determine if there are any missing values in the student data.
student_data_df.count()
```

Like in the schools_complete.csv file, there are no missing values in any of the columns because the output shows 39,170 rows for the student_complete.csv file:

```
Student ID
                  39170
                  39170
student name
gender
                  39170
```

grade	39170
school_name	39170
reading_score	39170
math_score	39170
dtype: int64	

NOTE

For more information, see the <u>Pandas documentation on the count</u> <u>method (https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.count.html)</u>.

The isnull() Method

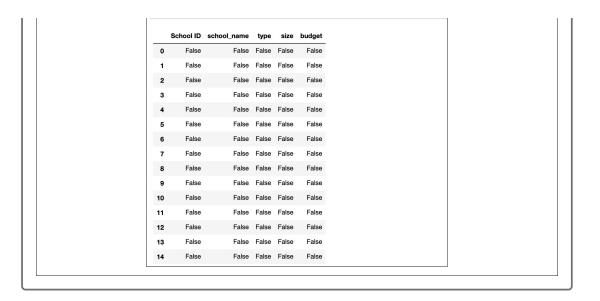
The Pandas library also has the <code>isnull()</code> method for determining empty rows. When you apply the <code>isnull()</code> method to a column, Series, or a DataFrame, a Boolean value will be returned, either "True" for the row or rows that are empty, i.e., null, or "False" for the rows that are not empty.

Let's use the <code>isnull()</code> method on the <code>school_data_df</code> DataFrame.

```
# Determine if there are any missing values in the school data.
school_data_df.isnull()
```

When we execute this code, we see that every row that is not empty is given the Boolean value "False," which tells us that there are no missing values. See the following screenshot:

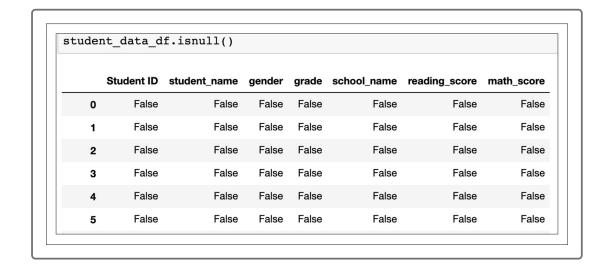
```
# Determine if there are any missing values in the school data.
school_data_df.isnull()
```



Now we'll use the same method on the student_data_df DataFrame. Add the following code to a new cell and run the cell:

```
# Determine if there are any missing values in the student data.
student_data_df.isnull()
```

The output shows that there are no rows that contain missing values, as they are all labeled "False." See the following image:



To get the total number of empty rows, or rows that are "True," we can use

the Pandas sum() method after the isnull() method, like this:

```
# Determine if there are any missing values in the student data.
student_data_df.isnull().sum()
```

REWIND

The process of joining two or more methods or functions together that are separated by a period is called **chaining**.

The output after running this code shows the total number of rows that are empty is zero for each column:

This output allows us to more easily determine at a glance how many rows are empty in the student_data_df DataFrame (zero). This is more

straightforward than output that shows "False" in thousands of rows.

NOTE

For more information, see the **Pandas documentation on the isnull()** method (https://pandas.pydata.org/pandas-docs/stable/reference /api/pandas.DataFrame.isnull.html).

The notnull() Method

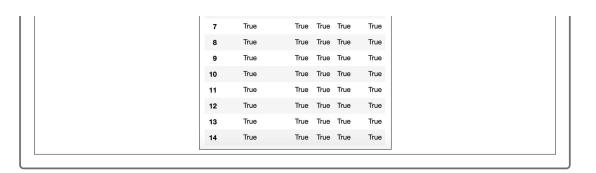
Another method that we can use to find missing values is the notnull() method. When you apply the notnull() method to a column, Series, or a DataFrame, a Boolean will be returned: "True" for the row or rows that are not empty, or "False" for the row or rows that are empty. This method returns the opposite output of the <code>isnull()</code> method.

Let's use the notnull() method on the school data df DataFrame. Run the following code:

```
# Determine if there are not any missing values in the school data.
school_data_df.notnull()
```

When we run this code, the output returns a copy of our school data df DataFrame, where all the rows that do not have any missing values are labeled "True." See the following image.

school_data_df.notnull()							
School ID	Schoo	ol ID	school_n	name	type	size	budget
) True	0	True		True	True	True	True
T rue	1	True		True	True	True	True
2 True	2	True		True	True	True	True
3 True	3	True		True	True	True	True
f True	4	True		True	True	True	True
5 True	5	True		True	True	True	True
3 True	6	True		True	True	True	True



Like we did with the student_data_df DataFrame, we can chain the notnull() method and the sum() method to get the sum of all the columns that are "True."

```
# Determine if there are not any missing values in the student data.
student_data_df.notnull().sum()
```

When we execute this code, we get the number of rows that are not null, which is 39,170 for each column.

```
student_data_df.notnull().sum()
Student ID
                 39170
student name
                 39170
                 39170
gender
grade
                 39170
school name
                 39170
reading_score
                 39170
math score
                 39170
dtype: int64
```

NOTE

For more information, see the **Pandas documentation on the notnull()**

method (http://pandas.pydata.org/pandas-docs/stable/reference /api/pandas.notnull.html)_.

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