

Class Objectives

By the end of today's class you will be able to:



Merge DataFrames together while understanding the differences between inner, outer, left, and right merges.



Slice data using the method and create new values based upon a series of bins.



Feel more confident in the ability to fix Python/Pandas bugs within Jupyter Notebook.



Use Google to explore additional Pandas functionality.



Instructor Demonstration Merging DataFrames

Merging DataFrames

What's merging?



Sometimes an analyst will receive data split across multiple tables and sources



Working across multiple tables is error-prone and confusing



Merging is the process of combining two tables based on shared data



Shared data can be an identical column in both tables, or a shared index



In Pandas, we can merge separate DataFrames using the pd.merge() method

Merging DataFrames: Inner join

Inner joins are the default means through which DataFrames are combined using the pd.merge() method and will only return data whose values match. Rows that do not include matching data will be dropped from the combined DataFrame.

Out[3]:
------	-----

	customer_id	item	cost
0	403	soda	3.0
1	112	chips	4.5
2	543	TV	600.0
3	999	Laptop	900.0
4	654	Cooler	150.0

Merging DataFrames: Outer join

Outer joins will combine the DataFrames regardless of whether any of the rows match and must be declared as a parameter within the pd.merge() method using the syntax how='outer'.

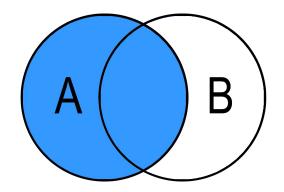
```
In [5]: # Merge two dataframes using an outer join
    merge_df = pd.merge(info_df, items_df, on="customer_id", how="outer")
    merge_df
```

Out[5]:

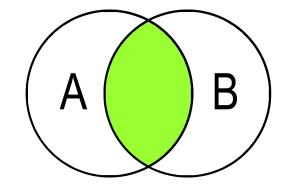
	customer_id	name	email	item	cost
0	112	John	jman@gmail	chips	4.5
1	403	Kelly	kelly@aol.com	soda	3.0
2	999	Sam	sports@school.edu	Laptop	900.0
3	543	April	April@yahoo.com	TV	600.0
4	123	Bobbo	HeylmBobbo@msn.com	NaN	NaN
5	654	NaN	NaN	Cooler	150.0

Joins are VERY important in Data Science!

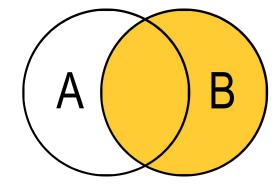
When merging data tables, **joins** tell the program what data to keep. Besides outer joins, there are three other common joins:



Left Outer Join:
All rows from table A,
even if they do not
exist in table B



Inner Join:
Fetch the results that exist in both tables



Right Outer Join: All rows from table B, even if they do not exist in table A

Merging DataFrames: Right and left joins

These joins will protect the data contained within one DataFrame like an outer join does while also dropping the rows with null data from the other DataFrame.

	me	Merge two derge_df = pderge df		ıstome			
t[6]:		customer_id	name	email		item	cost
	0	112	John	jman@gmail		chips	4.5
	1	403	Kelly	kelly@aol.com		soda	3.0
	2	999	Sam	sports@school.edu		Laptop	900.0
	3	543	April	April@yahoo.com	April@yahoo.com		600.
	4	123	Bobbo	HeylmBobbo@msr	n.com	NaN	NaN
		oin on="cu	ıstom				
Out[7]:		customer_id	name	email	item	cost	
	0	112	John	jman@gmail	chips	4.5	
	1	403	Kelly	kelly@aol.com	soda	3.0	
	2	999	Sam	sports@school.edu	Lapto	900.0)
	3	543	April	April@yahoo.com	TV	600.0)
	4	654	NaN	NaN	Coole	er 150.0)



Activity: Cryptocurrency Merging

In this activity, you will merge two datasets on cryptocurrencies, one on Bitcoin and the other on Dash. Once fully merged you will clean it to make it more presentable.



Activity: Cryptocurrency Merging

Instructions

- Read in both of the CSV files and print out their DataFrames.
- Perform an inner merge that combines both DataFrames on the 'Date' column.
- Rename the columns within the newly merged DataFrame so that the headers are more descriptive.
- Create a summary table that includes the following information: Best Bitcoin Open,
 Best Dash Open,
 Best Bitcoin Close,
 Best Dash Close,
 Total Bitcoin Volume,
 Total Dash Volume.
- Total Bitcoin Volume and Total Dash Volume should be calculated to have units of 'millions' and be rounded to two decimal places.



Time's Up! Let's Review.



Instructor Demonstration
Binning Data



Binning Data: pd.cut()

Use pd.cut() when you need to segment and sort data values into bins. This function is also useful for going from a continuous variable to a categorical variable.

```
In [18]: # Create the bins in which Data will be held
         # Bins are 0, 59.9, 69.9, 79.9, 89.9, 100.
         bins = [0, 59.9, 69.9, 79.9, 89.9, 100]
         # Create the names for the five bins
         group names = ["F", "D", "C", "B", "A"]
In [19]: df["Test Score Summary"] = pd.cut(df["Test Score"], bins, labels=group names, include lowest=True)
Out[19]:
                        Test Score Test Score Summary
            Class Name
         0 Oct
                  Cyndy 90
          1 Oct
                  Logan 59
          2 Jan
                  Laci
          3 Jan
                  Elmer
                        88
          4 Oct
                  Crystle 98
                  Emmie 60
```

Binning Data

What makes binning so powerful is that, after creating and applying these bins, the DataFrame can be grouped according to those values and thus a higher-level analysis can be conducted.

In [19]: df["Test Score Summary"] = pd.cut(df["Test Score"], bins, labels=group names, include lowest=True")

In [19]: df["Test Score Summary"] = pd.cut(df["Test Score"], bins, labels=group names, include lowest=True) Out[19]: Test Score Test Score Summary Class Name 0 Oct Cyndy 90 1 Oct Logan Laci Elmer 4 Oct Crystle 98 Emmie 60 D In [20]: # Creating a group based off of the bins df = df.groupby("Test Score Summary") df.max() Out[20]: Class Name **Test Score Test Score Summary** Oct Logan 59 Emmie 60 72 Jan Laci 88 Jan Elmer Cyndy 98



Activity: Binning TED

In this activity, you will put your binning skills to the test by creating bins for TED Talks based upon their viewership.



Activity: Binning TED

Instructions:

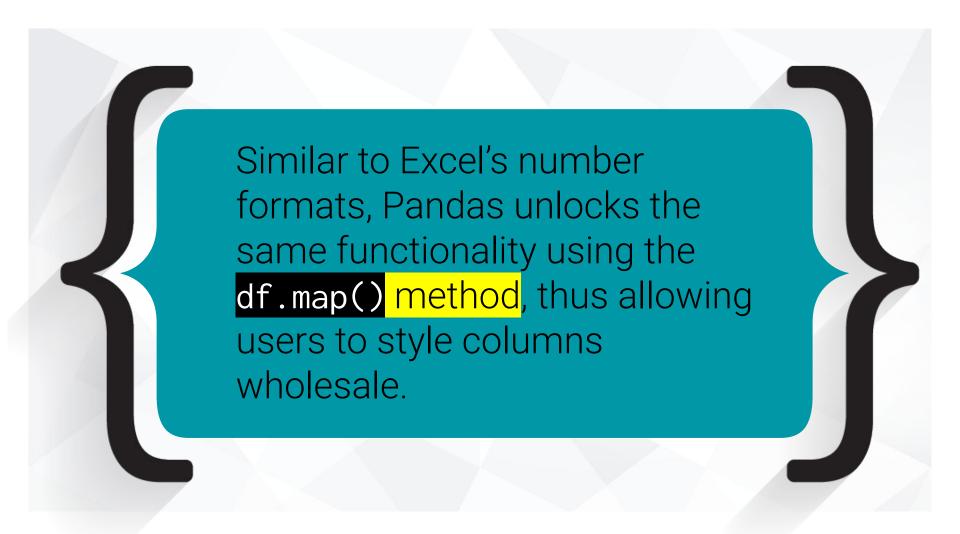
- Read in the CSV file provided and print it to the screen.
- Find the minimum 'views' and maximum 'views'.
- Using the minimum and maximum 'views' as a reference, create 10 bins in which to slice the data.
- Create a new column called 'View Group' and fill it with the values collected through your slicing.
- Group the DataFrame based upon the values within 'View Group'.
- Find out how many rows fall into each group before finding the averages for 'comments', 'duration', and 'languages'.



Time's Up! Let's Review.



Instructor Demonstration Mapping



Mapping

- df[<COLUMN>].map(<FORMAT STRING>.format) is the method by which, users can modify the styling of an entire column.
- To convert values into a typical dollar format, use "\${:.2f}". This places a dollar sign before the value which has been rounded to two decimal points.
- Using "{:,}" will split a number up so that it uses comma notation.

```
In [3]: # Use Map to format all the columns
    file_df["avg_cost"] = file_df["avg_cost"].map("${:.2f}".format)
    file_df["population"] = file_df["population"].map("{:.}".format)
    file_df["other"] = file_df["other"].map("{:.2f}".format)
    file_df.head()
```

Out[3]:

	id	city	avg_cost	population	other
0	1	Houxiang	\$55.12	609,458	-15.66
1	2	Leribe	\$95.78 601,963 23.79 +40 4d>4d>57.87	589,509	1.31
3	4	Sogcho	\$59.22 948,491 11.38 Kohlu \$23.09	92,206	7.67

Mapping

- Format mapping only really works once and will return errors if the same code is run
 multiple times without restarting the kernel. Because of this, formatting is usually
 applied near the end of an application.
- Format mapping also can change the datatype of a column. As such, all calculations should be handled before modifying the formatting.

```
In [4]: # Mapping has changed the datatypes of the columns to strings
file_df.dtypes

Out[4]: id          int64
          city          object
          avg_cost          object
          population          object
          other                object
          dtype: object
```



Activity: Cleaning Kickstarter

In this activity, you will take a dataset similar to your first homework, clean it up and format it.



Activity: Cleaning Kickstarter

The instructions for this activity are contained within the Jupyter Notebook.

In	[]:	import pandas as pd
In	[]:	# The path to our CSV file
			# Read our Kickstarter data into pandas
In]]:	# Get a list of all of our columns for easy reference
In	[]:	<pre># Extract "name", "goal", "pledged", "state", "country", "staff_pick", # "backers_count", and "spotlight"</pre>
In	[]:	# Remove projects that made no money at all
In]]:	# Collect only those projects that were hosted in the US # Create a list of the columns
			# Create a new df for "US" with the columns above.
In	[]:	# Create a new column that finds the average amount pledged to a project
In]]:	# First convert "average_donation", "goal", and "pledged" columns to float # Then Format to go to two decimal places, include a dollar sign, and use comma notation
In]]:	# Calculate the total number of backers for all US projects
In	[]:	# Calculate the average number of backers for all US projects
In	[]:	# Collect only those US campaigns that have been picked as a "Staff Pick"
In	[2]:	Group by the state of the campaigns and see if staff picks matter (Seems to matter quite a bit)



Time's Up! Let's Review.





Instructor Demonstration Introduction to Bugfixing

Note that an error is being returned as the application attempts to collect the average value within the 'Cocoa Percent' column.

```
In [5]: # Finding the average cocoa percent
      chocolate ratings df["Cocoa Percent"].mean()
      ValueError
                                      Traceback (most recent call last)
      /anaconda3/lib/python3.6/site-packages/pandas/core/nanops.py in ensure numeric(x)
         818
                   try:
      --> 819
                      x = float(x)
         820
                   except Exception:
      0%70%70%70%63%70%63%70%70%60%80%88%72%55%70%70%75%75%75%75%75%75%70%70%70%70%60%60%60%60%60%60%
      60%60%80%60%60%70%70%70%70%70%70%70%70%70%70%870%85%85%85%85%85%85%85%85%85%85%85%85%870%85
      0%70%70%70%70%70%70%70%70%70%80%65%70%65%73%72%80%70%70%90%64%64%64%71%70%70%70%83%78%83%74
      73$72$72$72$55$64$88$72$72$70$74$64$72$76$76$78$86$72$75$70$65$70$78$75$65$75$65$71$75$68$70$70$70$7
      75%100%75%75%77%100%70%70%70%70%68%70%70%72%70%75%85%60%80%70%80%80%60%70%72%70%72%68%70%68%72%72%
      72%72%60%70%75%75%75%75%75%75%72%66%77%75%75%75%74%75%70%74%71%74%72%64%70%69%70%70%72%72%66%65%7
      0%71%71%71%71%71%77%70%70%70%70%70%70%70%63%71%82%91%60%70%75%72%70%100%80%78%70%75%73%70%75%70%77%7
      7%77%55%55%70%70%72%72%72%72%80%65%65%65%65%65%72%72%70%72%72%70%70%70%73%70%70%72%70%70%70%
```

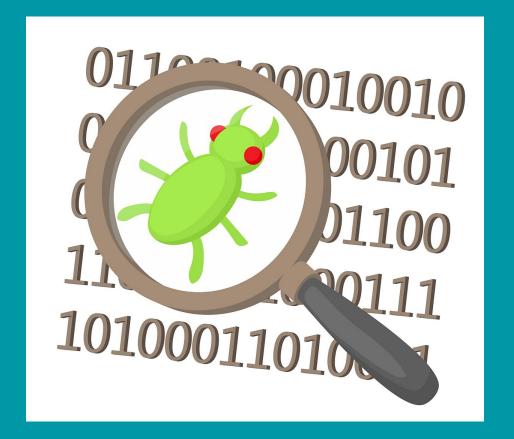
The first step: Keep calm

Bugs happen all the time and they are rarely the end of the world. In fact, most bugs that programmers run across are simple enough to solve so long as they know how and where to look for the solution.



The second step: Figure out what the bug is and where it is located

Bugs happen all the time and they are rarely the end of the world. In fact, most bugs that programmers run across are simple enough to solve so long as they know how and where to look for the solution.



The second step: Figure out what the bug is and where it is located

Unfortunately, Pandas is not known for returning clearly understandable error text. In fact, it often returns large blocks of text that is complex and confusing to those who do not know the library's underlying code. Looking for the line following KeyError: is generally a good starting point.



The second step: Figure out what the bug is and where it is located

For example, the text following TypeError: within the current code lets the programmer know that Pandas cannot convert the string values in the 'Cocoa Percent' column to floats.

ValueError: could not convert string to float:

The second step: Figure out what the bug is and where it is located

If the error text is not entirely clear, it is oftentimes helpful to print out variables/columns to the console in order to uncover where the bug is. For example, printing out the 'Cocoa Percent' series lets the programmer know that the dtype of this series is an object and not a float.

ValueError: could not convert string to float:

The third step: To look up the error online and source for solutions

The key part to this step is coming up with an accurate way to describe the bug. This may take multiple tries and is a skill that will develop over time.

Google is the programmer's best friend as typing in a description of the bug being faced will often bring up links to some possible solutions. If not, simply change the search up a little bit until a solution is discovered.

Pandas cannot string to float





The third step: To look up the error online and source for solutions

This particular problem requires the code to drop the percentages within the 'Cocoa Percent' column, so the search should be a bit more specific.

Pandas cannot convert string to float percentages







Activity: Bugfixing Bonanza

In this activity, you will be provided with a Pandas project containing TONS of bugs inside of it. Your job will be to take the application and fix it up so that it works properly.



Activity: Bugfixing Bonanza

Instructions Hints Dig through the Jupyter Notebook provided After fixing the bugs in each block of code, be sure to and attempt to fix as many bugs as run the cell below for an updated error. possible. There are a lot of them and the bugs get harder to deal with as the code progresses. Once you have finished bugfixing, perform There are a few new concepts being covered within some additional analysis on the dataset this Jupyter Notebook. The most complex of these provided. See what interesting trends are concepts is that of multi-indexing and it is very likely buried deep within these bug logs for the that this is where many will get held up. Do not worry Eclipse IDE. So long as you challenge though, multi-indexing is not in the homework and is yourself, bugs will pop up and you will get not required outside of this activity. It is simply an even more bugfixing practice. interesting/powerful feature of Pandas.



Time's Up! Let's Review.

last
but
not least..

