



Exploring Pandas

Data Boot Camp
Lesson 4.2



Class Objectives

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



Understand how to navigate through DataFrames using Loc and Iloc.



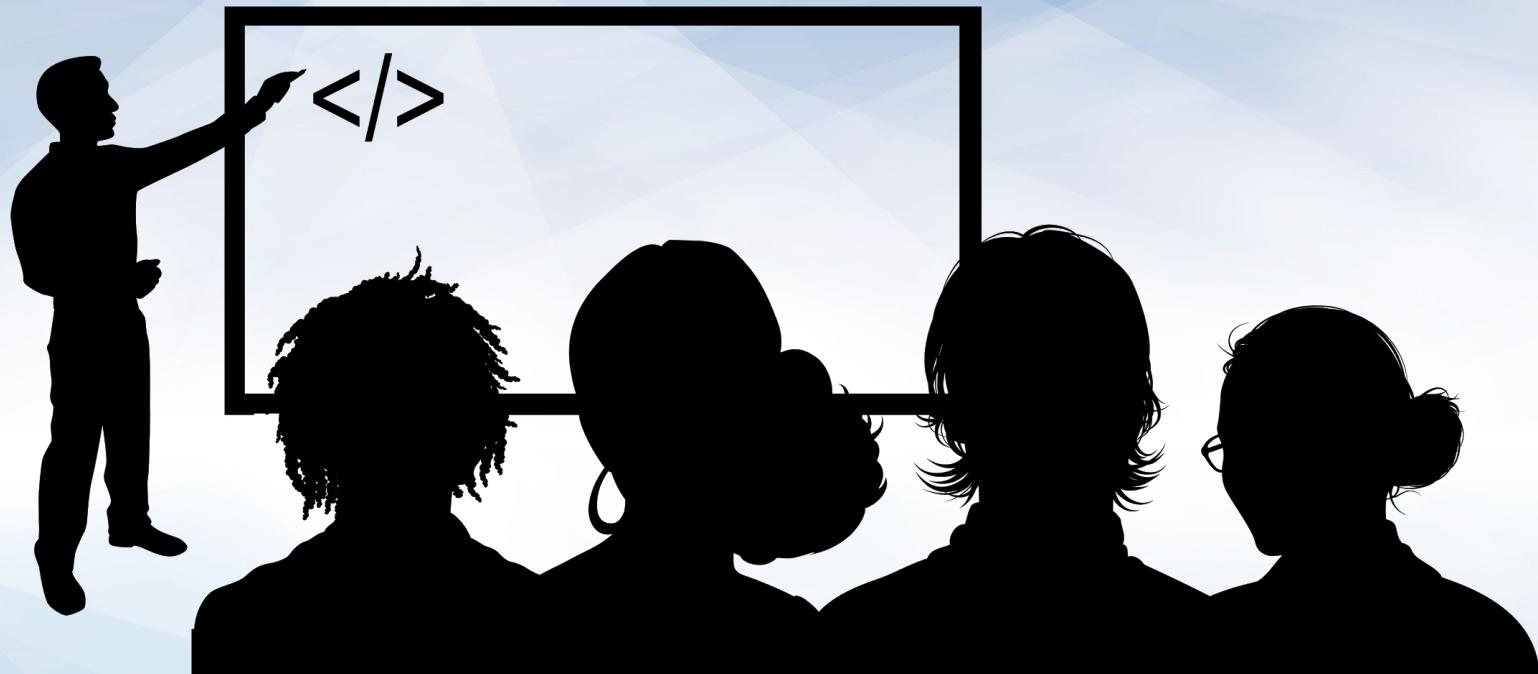
Understand how to filter and slice Pandas DataFrames.



Understand how to create and access Pandas GroupBy objects.



Understand how to sort DataFrames.

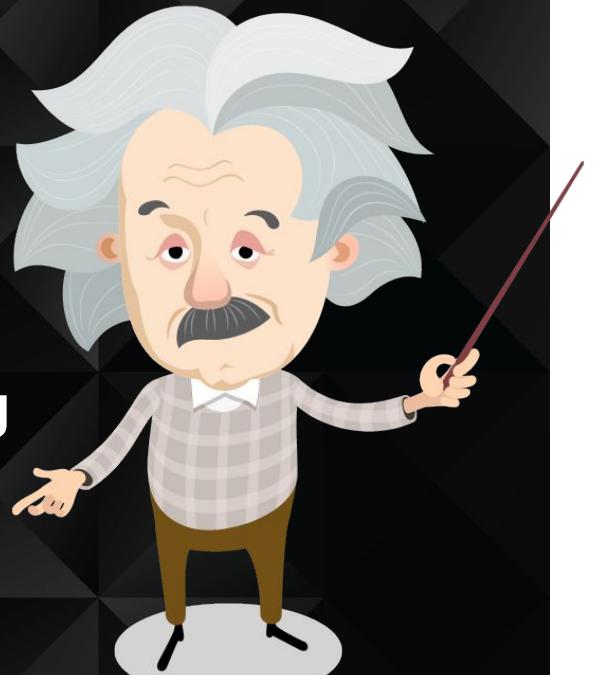


Instructor Demonstration

Exploring Data with Loc and Iloc



Programmers can easily collect specific rows/columns of data from a DataFrame using the loc() and iloc() methods.



One of The Most Powerful Aspects of Pandas... Exploring Data With Loc and Iloc

- The `loc()` method returns data based upon an index of labels/strings
- `loc()` is limited to string types and cannot be used on a numerical index. As an alternative solution you can use the `df.set_index()` function passing in the desired column header for the index.
- On the other hand the `iloc()` method instead of using labels, it uses integer based indexing for selection by position.

```
In [4]: # Set new index to last_name  
df = original_df.set_index("last_name")  
df.head()
```

Out[4]:

	id	first_name	Phone Number	Time zone
last_name				
Richardson	1	Peter	7-(789)867-9023	Europe/Moscow
Berry	2	Janice	86-(614)973-1727	Asia/Harbin
Hudson	3	Andrea	86-(918)527-6371	Asia/Shanghai
Mcdonald	4	Arthur	420-(553)779-7783	Europe/Prague
Morales	5	Kathy	351-(720)541-2124	Europe/Lisbon

Exploring Data With Loc and Iloc

- Both `loc()` and `iloc()` methods use brackets which contain the desired rows, followed by a comma, and then the columns desired.
- For example:

```
loc['Berry', 'Phone Number'] or iloc[1,2]
```

```
In [5]: # Grab the data contained within the "Berry" row and the "Phone Number" column
berry_phone = df.loc["Berry", "Phone Number"]
print("Using Loc: " + berry_phone)

also_berry_phone = df.iloc[1, 2]
print("Using Iloc: " + also_berry_phone)
```

```
Using Loc: 86-(614)973-1727
Using Iloc: 86-(614)973-1727
```

Exploring Data With Loc and Iloc

- Both methods allow us to select a range of columns and rows by providing a list
- We can also use a colon to tell Pandas to look for a range.

```
In [6]: # Grab the first five rows of data and the columns from "id" to "Phone Number"
# The problem with using "last_name" as the index is that the values are not unique so duplicates
# are returned
# If there are duplicates and loc[] is being used, Pandas will return an error
richardson_to_morales = df.loc[['Richardson", "Berry", "Hudson",
                                "McDonald", "Morales"], ["id", "first_name", "Phone Number"]]
print(richardson_to_morales)

print()

# Using iloc[] will not find duplicates since a numeric index is always unique
also_richardson_to_morales = df.iloc[0:4, 0:3]
print(also_richardson_to_morales)
```

	id	first_name	Phone Number
last_name			
Richardson	1	Peter	7-(789)867-9023
Richardson	25	Donald	62-(259)282-5871
Berry	2	Janice	86-(614)973-1727
Hudson	3	Andrea	86-(918)527-6371
Hudson	8	Frances	57-(752)864-4744
Hudson	90	Norma	351-(551)598-1822
McDonald	4	Arthur	420-(553)779-7783
Morales	5	Kathy	351-(720)541-2124

	id	first_name	Phone Number
last_name			
Richardson	1	Peter	7-(789)867-9023
Berry	2	Janice	86-(614)973-1727
Hudson	3	Andrea	86-(918)527-6371
McDonald	4	Arthur	420-(553)779-7783

Exploring Data With Loc and Iloc

- By passing in a colon by itself, `loc()` and `iloc()` will select all rows or columns depending on where it is placed in relation to the comma.

```
In [7]: # The following will select all rows for columns `first_name` and `Phone Number`  
df.loc[:, ["first_name", "Phone Number"]].head()
```

Out[7]:

	first_name	Phone Number
last_name		
Richardson	Peter	7-(789)867-9023
Berry	Janice	86-(614)973-1727
Hudson	Andrea	86-(918)527-6371
Mcdonald	Arthur	420-(553)779-7783
Morales	Kathy	351-(720)541-2124



loc() and iloc() can be used to conditionally filter rows of data based upon the values within a column.



Exploring Data With Loc and Iloc

- Instead of passing a list of indices, we can use a logic statement!
- In case of multiple conditions that should be checked for, `&` and `|` may be added into logic test as representations of `and` and `or`.

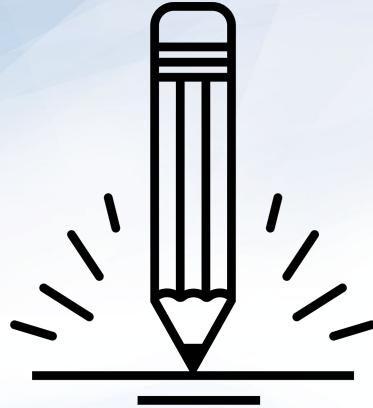
```
In [9]: # Loc and Iloc also allow for conditional statements to filter rows of data
# using Loc on the logic test above only returns rows where the result is True
only_billys = df.loc[df["first_name"] == "Billy", :]
print(only_billys)

print()

# Multiple conditions can be set to narrow down or widen the filter
only_billy_and_peter = df.loc[(df["first_name"] == "Billy") | (
    df["first_name"] == "Peter"), :]
print(only_billy_and_peter)
```

	id	first_name	Phone Number	Time zone
last_name				
Clark	20	Billy	62-(213)345-2549	Asia/Makassar
Andrews	23	Billy	86-(859)746-5367	Asia/Chongqing
Price	59	Billy	86-(878)547-7739	Asia/Shanghai

	id	first_name	Phone Number	Time zone
last_name				
Richardson	1	Peter	7-(789)867-9023	Europe/Moscow
Clark	20	Billy	62-(213)345-2549	Asia/Makassar
Andrews	23	Billy	86-(859)746-5367	Asia/Chongqing
Price	59	Billy	86-(878)547-7739	Asia/Shanghai



Activity: Good Movies

In this activity, you will create an application that looks through IMDB data in order to find only the best movies out there.

Suggested Time:
20 Minutes



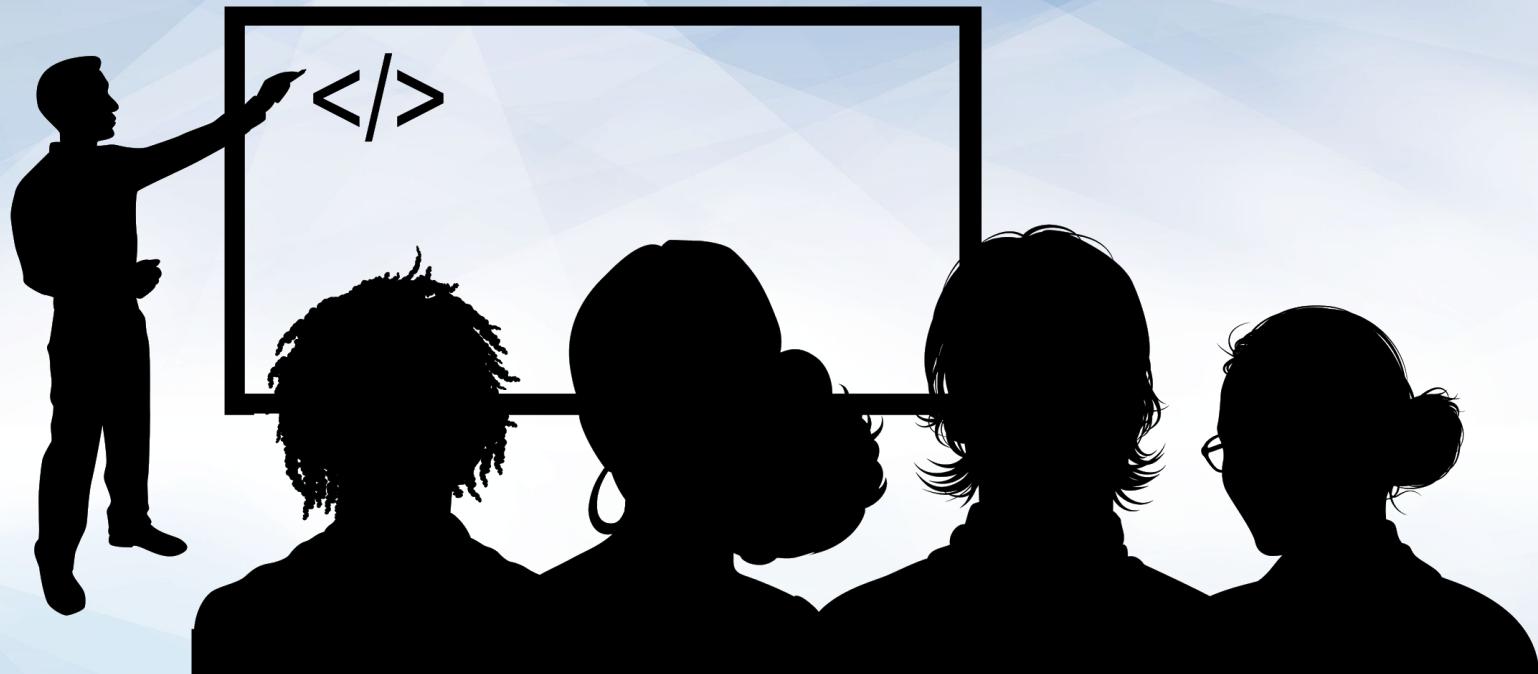
Activity: Good Movies

Instructions:

- Use Pandas to load and display the CSV provided in Resources.
- List all the columns in the data set.
- We're only interested in IMDb data, so create a new table that takes the Film and all the columns relating to IMDB.
- Filter out only the good movies—i.e., any film with an IMDb score greater than or equal to 7 and remove the norm ratings.
- Find less popular movies that you may not have heard about - i.e., anything with under 20K votes.
- Finally, export this file to a spreadsheet, excluding the index, so we can keep track of our future watchlist.



Time's Up! Let's Review.



Instructor Demonstration
Cleaning Data



When dealing with massive datasets it is almost inevitable that duplicate rows, inconsistent spelling, and missing values will crop up.



Cleaning Data

- `del <DataFrame>[<columns>]`

```
In [4]: # Preview of the DataFrame  
# Note that FIELD8 is likely a meaningless column  
df.head()
```

Out[4]:

	LastName	FirstName	Employer	City	State	Zip	Amount	FIELD8
0	Aaron	Eugene	State Department	Dulles	VA	20189	500.0	NaN
1	Abadi	Barbara	Abadi & Co.	New York	NY	10021	200.0	NaN
2	Adamany	Anthony	Retired	Rockford	IL	61103	500.0	NaN
3	Adams	Lorraine	Self	New York	NY	10026	200.0	NaN
4	Adams	Marion	None	Exeter	NH	03833	100.0	NaN

```
In [5]: # Delete extraneous column  
del df['FIELD8']  
df.head()
```

Out[5]:

	LastName	FirstName	Employer	City	State	Zip	Amount
0	Aaron	Eugene	State Department	Dulles	VA	20189	500.0
1	Abadi	Barbara	Abadi & Co.	New York	NY	10021	200.0
2	Adamany	Anthony	Retired	Rockford	IL	61103	500.0
3	Adams	Lorraine	Self	New York	NY	10026	200.0
4	Adams	Marion	None	Exeter	NH	03833	100.0

Cleaning Data

- `count()`
- `<DataFrame>.dropna(how='any')`

```
In [6]: # Identify incomplete rows  
df.count()
```

```
Out[6]: LastName      1776  
FirstName     1776  
Employer      1743  
City          1776  
State          1776  
Zip            1776  
Amount         1776  
dtype: int64
```

```
In [7]: # Drop all rows with missing information  
df = df.dropna(how='any')
```

```
In [8]: # Verify dropped rows  
df.count()
```

```
Out[8]: LastName      1743  
FirstName     1743  
Employer      1743  
City          1743  
State          1743  
Zip            1743  
Amount         1743  
dtype: int64
```

Cleaning Data

- `value_counts()`
- `replace()`

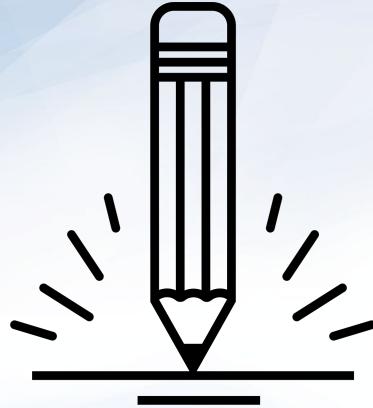
```
In [12]: # Display an overview of the Employers column  
df['Employer'].value_counts()
```

```
Out[12]: None 249  
Self 241  
Retired 126  
Self Employed 39  
Self-Employed 34
```

```
In [13]: # Clean up Employer category. Replace 'Self Employed' and 'Self' with 'Self-Employed'  
df['Employer'] = df['Employer'].replace(  
    {'Self Employed': 'Self-Employed', 'Self': 'Self-Employed'})
```

```
In [14]: # Verify clean-up.  
df['Employer'].value_counts()
```

```
Out[14]: Self-Employed 314  
None 249  
Retired 126  
Google 6
```



Activity: Portland Crime

In this activity, you will take a crime dataset from Portland and do your best to clean it up so that the DataFrame is consistent and no rows with missing data are present.

Suggested Time:
20 Minutes



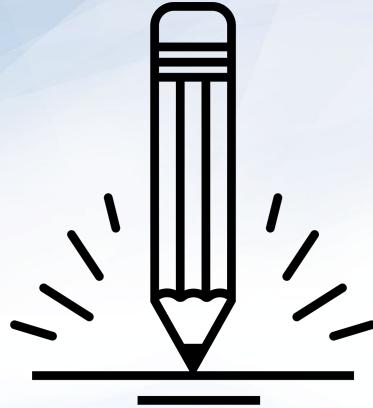
Activity: Portland Crime

Instructions:

- Read in the csv using Pandas and print out the DataFrame that is returned.
- Get a count of rows within the DataFrame in order to determine if there are any null values
- Drop the rows which contain null values.
- Search through the "Offense Type" column and replace any similar values with one consistent value.
- Create a couple DataFrames that look into one Neighborhood only and print them to the screen.



Time's Up! Let's Review.



Activity: Pandas Recap and Data Types

In this activity, we will recap what has been covered in Pandas up to this point.

Suggested Time:
15 Minutes



Activity: Pandas Recap and Data Types

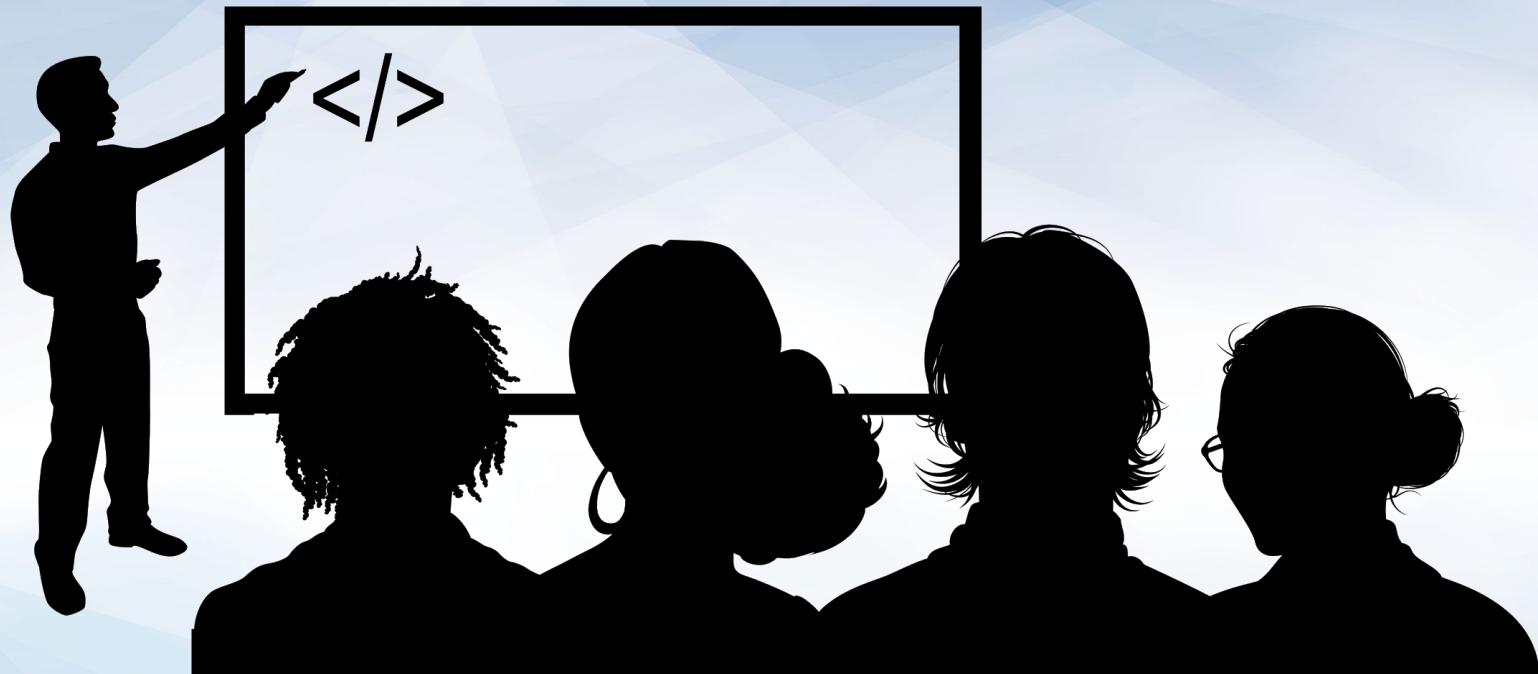
- Open up ‘PandasRecap.ipynb’ under the ‘unsolved’ folder in your Jupyter Notebook.
- Go through the cells following the notes.
- Hint:
 - A list of a DataFrame's data types can be seen by accessing its `dtypes` property.
 - In order to change a non-numeric column to a numeric column, use the `df.astype(<datatype>)` method and pass in the desired datatype as the parameter.





Break





Instructor Demonstration Pandas Grouping



**.groupby() is a simpler method to
filter data.**

Pandas Grouping

- In order to split the DataFrame into multiple groups and group by state the `df.groupby([<Columns>])` is used.
- The `.groupby()` method returns a GroupBy object that can only be accessed by using a data function on it.

```
In [9]: # Using GroupBy in order to separate the data into fields according to "state" values
grouped_usa_df = usa_ufo_df.groupby(['state'])

# The object returned is a "GroupBy" object and cannot be viewed normally...
print(grouped_usa_df)

# In order to be visualized, a data function must be used...
grouped_usa_df.count().head(10)
```

```
<pandas.core.groupby.groupby.DataFrameGroupBy object at 0x10cde6278>
```

Out[9]:

	datetime	city	country	shape	duration (seconds)	duration (hours/min)	comments	date posted	latitude	longitude
state										
ak	311	311	311	311	311	311	311	311	311	311
al	629	629	629	629	629	629	629	629	629	629
ar	578	578	578	578	578	578	578	578	578	578
az	2362	2362	2362	2362	2362	2362	2362	2362	2362	2362
ca	8683	8683	8683	8683	8683	8683	8683	8683	8683	8683
co	1385	1385	1385	1385	1385	1385	1385	1385	1385	1385
ct	865	865	865	865	865	865	865	865	865	865
dc	7	7	7	7	7	7	7	7	7	7
de	165	165	165	165	165	165	165	165	165	165
fl	3754	3754	3754	3754	3754	3754	3754	3754	3754	3754

Pandas Grouping

- The `pd.DataFrame()` method makes possible to create new DataFrames using solely GroupBy data.
- A DataFrame can also be created by selecting a single series from a GroupBy object and passing it in as the values for a specified column.

```
In [11]: # Since "duration (seconds)" was converted to a numeric time, it can now be summed up per state  
state_duration = grouped_usa_df["duration (seconds)"].sum()  
state_duration.head()
```

```
Out[11]: state  
ak    1455863.00  
al    900453.50  
ar    66986144.50  
az    15453494.60  
ca    24865571.47  
Name: duration (seconds), dtype: float64
```

```
In [12]: # Creating a new DataFrame using both duration and count  
state_summary_df = pd.DataFrame({"Number of Sightings": state_counts,  
                                 "Total Visit Time": state_duration})  
state_summary_df.head()
```

	Number of Sightings	Total Visit Time
ak	311	1455863.00
al	629	900453.50
ar	578	66986144.50
az	2362	15453494.60
ca	8683	24865571.47

Pandas Grouping

- It is possible to perform a `df.groupby()` method on multiple columns as well. This can be done by simply passing two or more column references into the list parameter.

```
In [13]: # It is also possible to group a DataFrame by multiple columns  
# This returns an object with multiple indexes, however, which can be harder to deal with  
grouped_international_data = converted_ufo.groupby(['country', 'state'])
```

```
grouped_international_data.count().head(20)
```

```
Out[13]:
```

		datetime	city	shape	duration (seconds)	duration (hours/min)	comments	date posted	latitude	longitude
country	state									
au	al	1	1	1	1	1	1	1	1	1
	dc	1	1	1	1	1	1	1	1	1
	nt	2	2	2	2	2	2	2	2	2
	oh	1	1	1	1	1	1	1	1	1
	sa	2	2	2	2	2	2	2	2	2
	wa	2	2	2	2	2	2	2	2	2
	yt	1	1	1	1	1	1	1	1	1
ca	ab	284	284	284	284	284	284	284	284	284
	bc	677	677	677	677	677	677	677	677	677
	mb	124	124	124	124	124	124	124	124	124
	nb	86	86	86	86	86	86	86	86	86
	nf	15	15	15	15	15	15	15	15	15
	ns	101	101	101	101	101	101	101	101	101
	nt	13	13	13	13	13	13	13	13	13
	on	1335	1335	1335	1335	1335	1335	1335	1335	1335
	pe	10	10	10	10	10	10	10	10	10
	pq	62	62	62	62	62	62	62	62	62
	qc	124	124	124	124	124	124	124	124	124
	sa	27	27	27	27	27	27	27	27	27
	sk	77	77	77	77	77	77	77	77	77

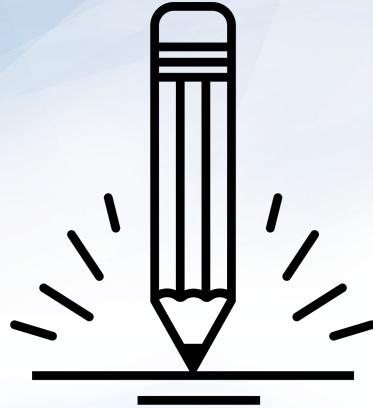
Pandas Grouping

- A new DataFrame can be created from a GroupBy object.

```
In [14]: # Converting a GroupBy object into a DataFrame
international_duration_df = pd.DataFrame(
    grouped_international_data[ "duration (seconds)" ].sum())
international_duration_df.head(10)
```

Out[14]:

		duration (seconds)
country	state	
au	al	900.00
	dc	300.00
	nt	360.00
	oh	180.00
	sa	305.00
	wa	450.00
	yt	30.00
ca	ab	530994.00
	bc	641955.82
	mb	160132.00



Activity: Building a PokeDex

In this activity, you will create a DataFrame that visualizes the average stats for each type of Pokemon from the popular video game series.

Suggested Time:
25 Minutes

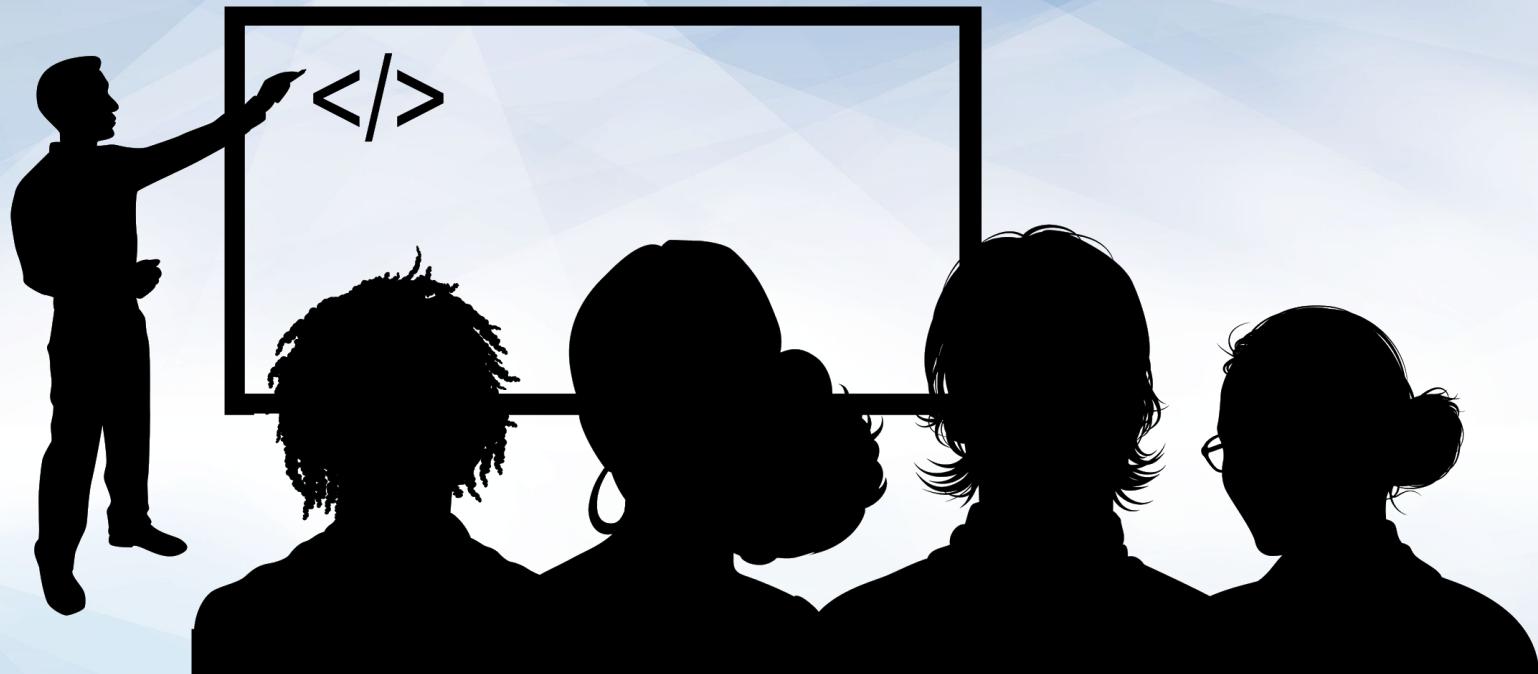


Activity: Pandas Recap and Data Types

- Read the Pokemon CSV file with Pandas.
- Create a new table by extracting the following columns: "Type 1", "HP", "Attack", "Sp. Atk", "Sp. Def", and "Speed".
- Find the average stats for each type of Pokemon.
- Create a new DataFrame out of the averages.
- Calculate the total power level of each type of Pokemon by summing all of the previous stats together and place the results into a new column.
- **Bonus:**
 - Sort the table by strongest type and export the resulting table to a new CSV.



Time's Up! Let's Review.



Instructor Demonstration
Sorting Made Easy

Sorting Made Easy

- In order to sort a DataFrame based upon the values within a column, simply use the `df.sort_values()` method and pass the column name to sort by in as a parameter.
- The parameter of "ascending" is always marked as True by default. This means that the `sort_values()` method will always sort from lowest to highest unless the parameter of `ascending=False` is passed into the `sort_values()` method as well.

```
In [3]: # Sorting the DataFrame based on "Freedom" column  
# Will sort from lowest to highest if no other parameter is passed  
freedom_df = happiness_df.sort_values("Freedom")  
freedom_df.head()
```

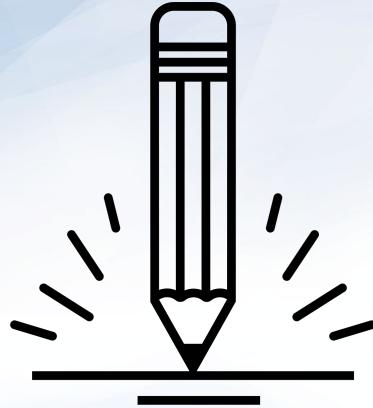
Out[3]:

	Country	Happiness.Rank	Happiness.Score	Whisker.high	Whisker.low	Economy..GDP.per.Capita.	Family	Health..L
139	Angola	140	3.795	3.951642	3.638358	0.858428	1.104412	0.049869
129	Sudan	130	4.139	4.345747	3.932253	0.659517	1.214009	0.290921
144	Haiti	145	3.603	3.734715	3.471285	0.368610	0.640450	0.277321
153	Burundi	154	2.905	3.074690	2.735310	0.091623	0.629794	0.151611
151	Syria	152	3.462	3.663669	3.260331	0.777153	0.396103	0.500533

```
In [4]: # To sort from highest to lowest, ascending=False must be passed in  
freedom_df = happiness_df.sort_values("Freedom", ascending=False)  
freedom_df.head()
```

Out[4]:

	Country	Happiness.Rank	Happiness.Score	Whisker.high	Whisker.low	Economy..GDP.per.Capita.	Family	Health..L
46	Uzbekistan	47	5.971	6.065538	5.876463	0.786441	1.548969	0.4982
0	Norway	1	7.537	7.594445	7.479556	1.616463	1.533524	0.7966
128	Cambodia	129	4.168	4.278518	4.057483	0.601765	1.006238	0.4297
2	Iceland	3	7.504	7.622030	7.385970	1.480633	1.610574	0.8335
1	Denmark	2	7.522	7.581728	7.462272	1.482383	1.551122	0.7925



Activity: Search For the Worst

In this activity, you will take a dataset composed of soccer player statics and will attempt to determine which players are the worst in the world at their particular position.

Suggested Time:
25 Minutes



Activity: Search For the Worst

- Read in the CSV file provided and print it to the screen.
- Print out a list of all of the values within the "Preferred Position" column.
- Select a value from this list and create a new DataFrame that only includes players who prefer that position.
- Sort the DataFrame based upon a player's skill in that position.
- Reset the index for the DataFrame so that the index is in order.
- Print out the statistics for the worst player in a position to the screen.



Time's Up! Let's Review.

*The
End*