

### This Week: Pandas

#### By the end of this week, you'll know how to:



Read an external CSV file into a DataFrame.



Determine data types of row values in a DataFrame.



Format and retrieve data from columns of a DataFrame.



Merge, filter, slice, and sort a DataFrame.



Apply the groupby() function to a DataFrame.



Use multiple methods to perform a function on a DataFrame.

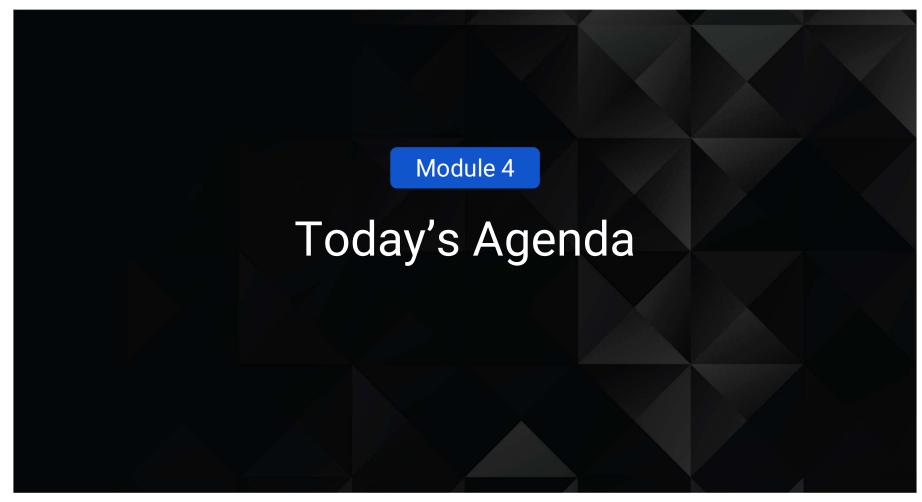


Perform mathematical calculations on columns of a DataFrame or Series.



## This Week's Challenge

Using the skills learned throughout the week, help a mock school board with their investigation by adjusting specific data.



## Today's Agenda

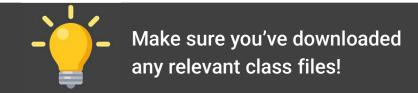
By completing today's activities, you'll learn the following skills:

01

Working with DataFrames

02

Segment and sort values into bins









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

### 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

#### count()

### <DataFrame>.dropna(how='any')

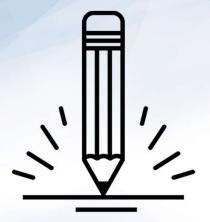
```
In [6]: # Identify incomplete rows
        df.count()
Out[6]: LastName
                     1776
        FirstName
                     1776
        Employer
                     1743
        City
                     1776
                     1776
        State
        Zip
                     1776
        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
                     1743
        State
                     1743
        Zip
        Amount
                     1743
        dtype: int64
```

# 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
                                                                                          249
         None
         Retired
                                                                                          126
         Google
```

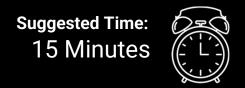
count()	To look for missing values, we use the count() method on the DataFrame.
dropna(how="any")	To drop rows with null values, we use dropna(how="any"), then verify the counts.
value_counts()	To look for any misspelled offenses and to find if similar offenses can be combined, we use value_counts() on the Offense Type column.
replace()	We combine similar offenses using the replace() method on the column in question and pass a dictionary into it, with the keys being those values to replace and the value being a common offense in the column.



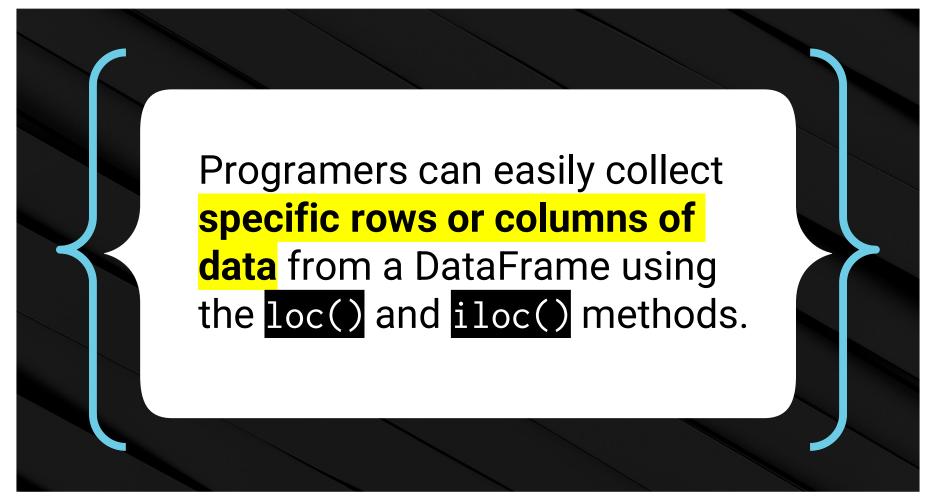


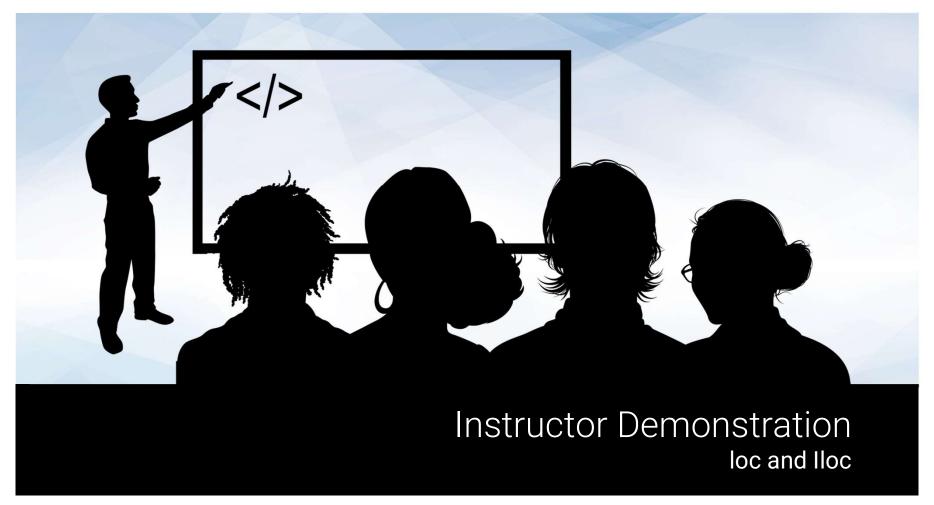
## **Activity:** Portland Crime

In this activity, we will take a crime dataset from Portland and do our best to clean it up so the DataFrame is consistent and has no rows with missing data.

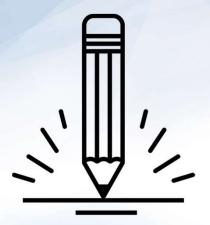






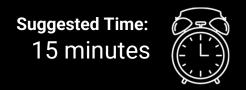






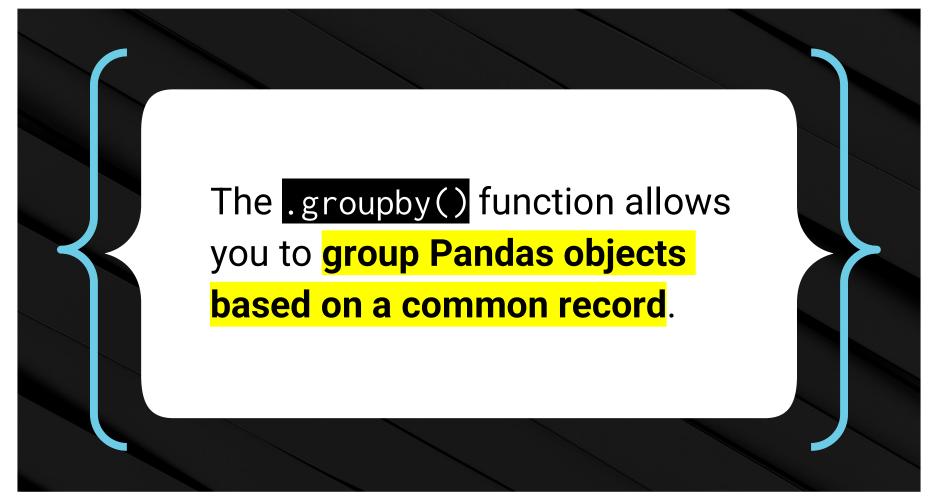
## **Activity:** Good Movies

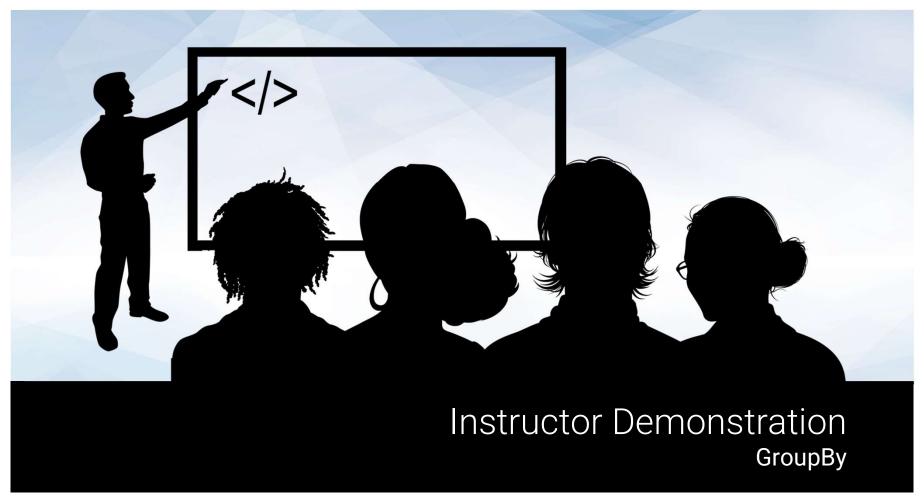
In this activity, you will create an application that looks through IMDB data in order to find only the best movies out there.











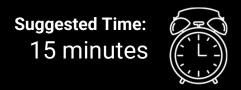




### **Pair Programming Activity:**

## **Training Groupby**

In this exercise, you will work in pairs and use groupby() to get the average weight and length membership of the gym members for each trainer.



### **Pair Programming**

There are 2 main roles in pair programming:

01

#### Driver

The first is the Driver, whose role is to focus on resolving the current task while talking through their thought process out loud.





### **Navigator**

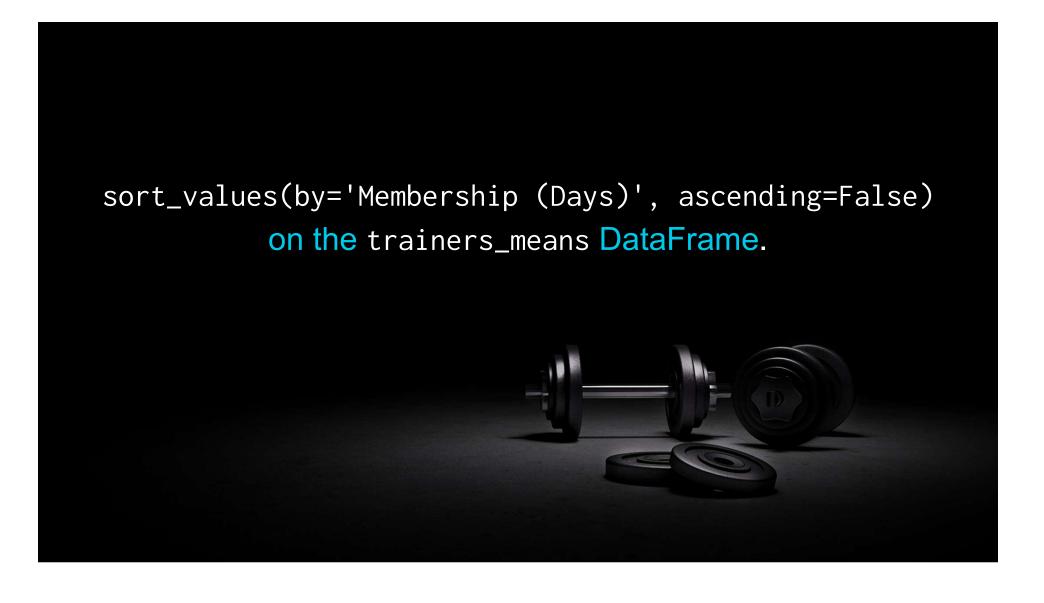
The second is the Navigator, which is equally as important. They will help catch bugs and typos, think about issues to address for efficiency, and use documentation to find resources to help the driver get past a hurdle.



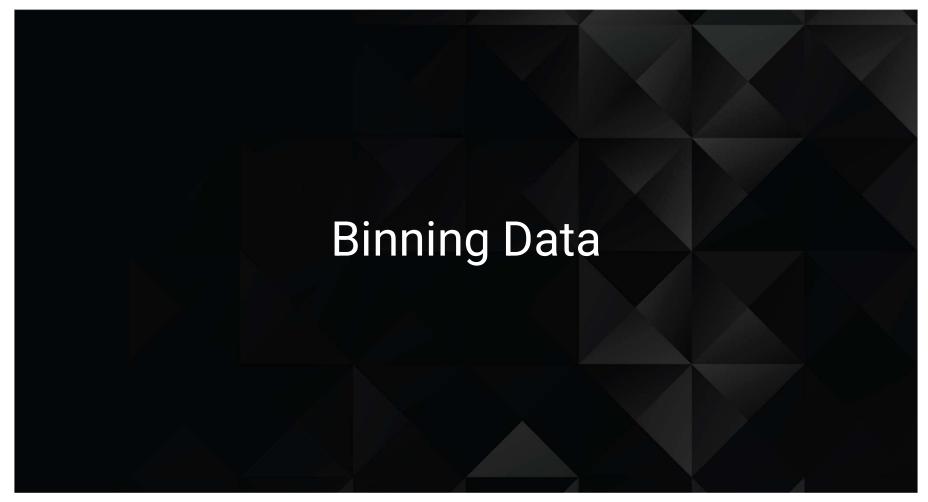


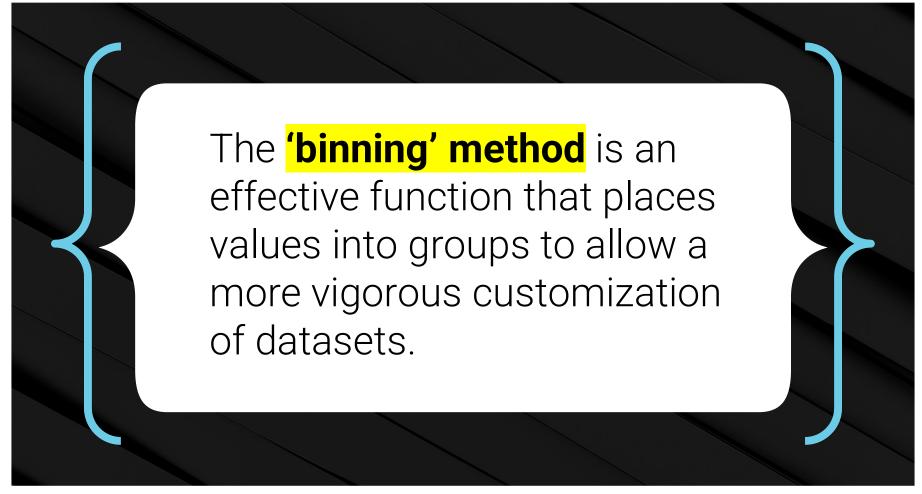


How would we sort the DataFrame from greatest to fewest length of membership in days?









## **Binning Data**

### Understand

Not everyone is a numbers person, and sometimes there are so many values within a DataFrame that it becomes difficult to comprehend what exactly is going on.

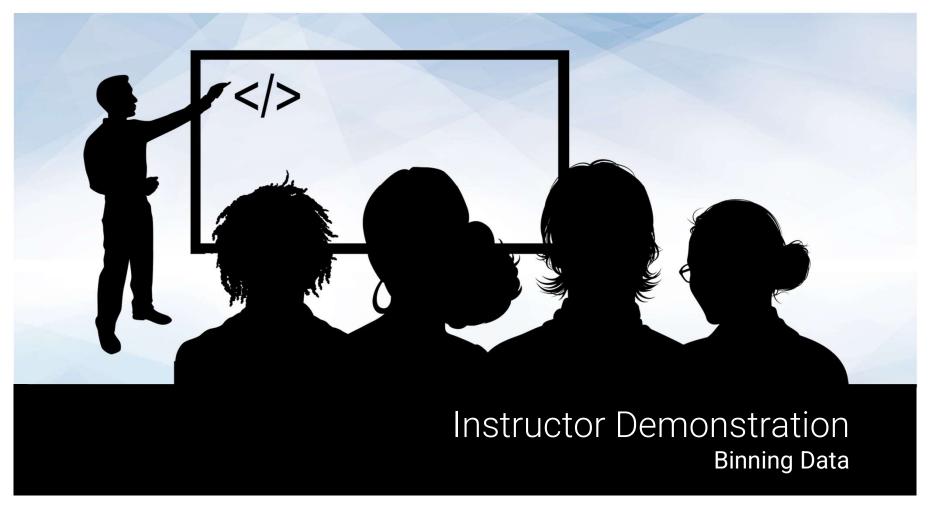
### Visualize

Grouping these values in bins can make it easier to visualize large datasets.

### Function

Using the Pandas

pd.cut() function will
allow us to "bin" values
into groups, which
enables more vigorous
customization of
datasets.







## **Activity:** Binning TED

In this activity, you will create bins for TED Talks based on their viewership. After creating the bins, you'll group the DataFrame based on those bins, and then perform some analysis on them.

