Analytics Jumpstart

Intro to commonly used pandas methods

Nashville Software School



Goals for today

- Learn about the pandas library
- See some common *pandas* methods
- Work on coding tasks



| pandas

Python is a a general purpose programming language and is widely used for web applications, scientific computing, machine learning, and data analysis.

We'll be making use of the <u>pandas</u> library, an open source data analysis and manipulation tool.

In Python, a **library** bundles together functions and objects that have a common functionality (like data analysis).

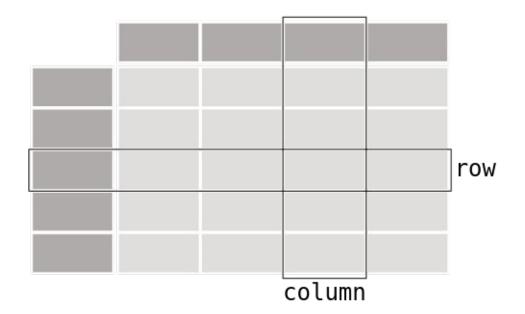
The *pandas* library was installed when you installed Anaconda.





The primary tool for working with data in *pandas* is the **DataFrame**, an object which can hold tabular data.

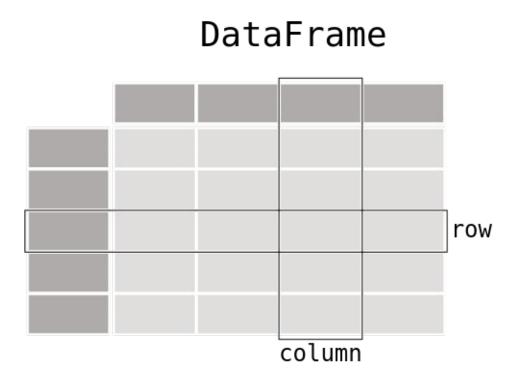
DataFrame





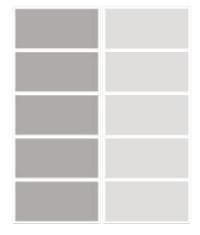


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Each column of a DataFrame is a pandas Series.

Series







DG

	Name	Team	Number	Position	Age
0	Avery Bradley	Boston Celtics	0.0	PG	25.0
1	John Holland	Boston Celtics	30.0	SG	27.0
2	Jonas Jerebko	Boston Celtics	8.0	PF	29.0
3	Jordan Mickey	Boston Celtics	NaN	PF	21.0
4	Terry Rozier	Boston Celtics	12.0	PG	22.0
5	Jared Sullinger	Boston Celtics	7.0	С	NaN
6	Evan Turner	Boston Celtics	11.0	SG	27.0
			L Data		

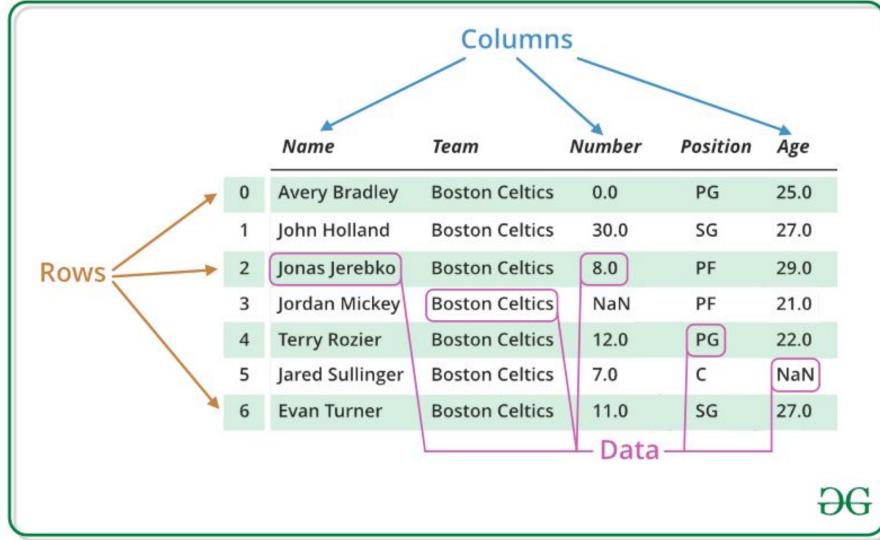






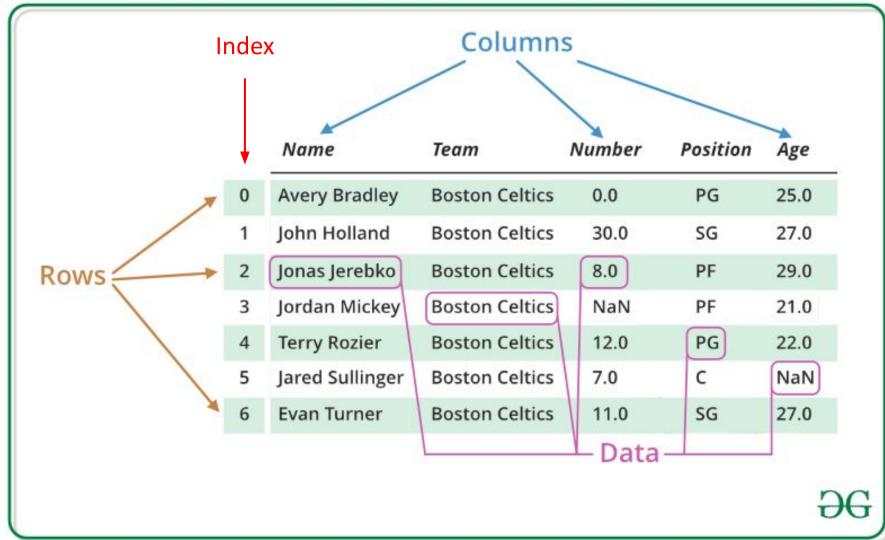














To make use of pandas, we'll import it using the command

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import pandas as pd
```



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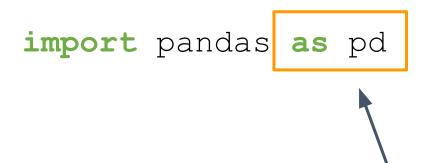
```
import pandas as pd
```



Allows us to use the components of *pandas*.



To make use of pandas, we'll import it using the command



This is an **alias**, meaning that when referring to pandas objects, we will use the abbreviation pd instead of the full word.



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pd.read_csv('../data/public_art.csv')
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This function reads in data from a csv (comma separated values) file into a pandas DataFrame.



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We need parentheses since we're calling a function.



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Here, we need an argument, which tells the function where to find our data.



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```
art = pd.read_csv('../data/public_art.csv')
```

We can **assign** the result of this function to a **variable** named art which will allow us to reuse it.



Function can be part of a library (like *pd.read_csv()*) or they can be built in to objects. In this case, we call them **methods**.

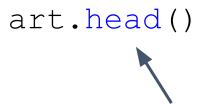


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```
art.head()
```



Function can be part of a library (like pd.read_csv()) or they can be built in to objects. In this case, we call them methods.



The name of the method.

This function returns the first five rows of a DataFrame.



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It is a built-in function of *pandas*DataFrames. This is the name of the

DataFrame we are working with.



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art.head()

It does not require any arguments, but still needs parentheses since it is a function.



Dataframes also have additional characteristics called **attributes** or **properties** which are also accessed with a period but do not require parentheses.



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For example, the *shape* attribute contains the dimensions (number of rows and number of columns) of a DataFrame.

art.shape



Now, let's look at some commonly-used *pandas* tools which we'll practice in our first notebook.



Importing Data

• **pd.read_csv()** – read a comma delimited file; good practice is to look at the raw file in a text editor (like Visual Studio Code, not Excel); additional arguments may be needed to handle extra rows at the top and extra data (footnotes) at the bottom.



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Inspecting

- **df.head()** looks at the top of the DataFrame; 5 rows by default
- df.tail() looks at the bottom of the DataFrame; 5 rows by default
- **df.shape** returns the dimensions of a DataFrame: (number of rows, number of columns)
- **df.info()** method to get information about the DataFrame



Modifying

- **df.columns** column labels attribute
- **df.rename()** rename values (can pass in a dictionary with existing columns as the key and new ones as the values)
- **df.drop()** drop the specified labels (either rows or columns) from the DataFrame



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- **series.unique()** returns the unique values in a column
- series.nunique() returns the number of unique elements in a column
- series.value_counts() returns the unique elements in a column and the number of appearances of each



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Slicing/Filtering

- df.loc[] pass in row name and column name to access data at that location
- df[[]] creates a slice (subset) of the DataFrame including just the columns passed in

Notice that these last accessors use square brackets instead of parentheses.



Let's open our first shared notebook so we can see these in action:

notebook_01_public_art_part_1.ipynb

