



ENGR 1330

**Computational Thinking with
Data Science**

Pandas



Topic Outline



- Pandas library
 - ✓ Data representation: Dataframes
 - ✓ Data operations: Indexing, summarizing statistics, aggregation, grouping, filling and dropping values, and read/write files



Objectives



- To be able to represent data in the form of dataframes via the Pandas library
- To be able to access and manipulate data within a dataframe
- To be able to obtain basic statistical measures of data within a dataframe



Computational Thinking Concepts

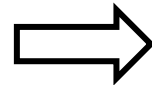


Pandas dataframes



Data representation

Data interpretation,
manipulation, and analysis
of Pandas dataframes



Decomposition

Algorithm



Pandas



- Pandas: Derived from the term 'Panel Data'
- Dataframe: 2-dimensional mutable and heterogenous tabular data structure
 - ✓ Provides rich data structures and functions designed to make working with data fast, easy, and expressive



Pandas Dataframes

- Creating a dataframe:

In [1]: `import pandas as pd` → Importing Pandas library

In [6]: `df = pd.DataFrame(np.random.randint(1,100,(5,4)), ['A','B','C','D','E'], ['W','X','Y','Z'])`
`df`

rows → columns → index → columns

Function to create a Pandas dataframe

- What will be the shape of the above 2D Pandas dataframe?

(Demo)



Pandas Dataframes

- Creating a dataframe: from dictionary

```
data = {  
    "name": ["Bob", "Mary", "Tom"],  
    "section": ["009", "011", "012"]  
}  
  
df = pd.DataFrame(data)  
df
```



	name	section
0	Bob	009
1	Mary	011
2	Tom	012

Dictionary's keys



Auto indexing





Dataframes: Indexing, slicing



Selecting rows and all columns: same as indexing in list.

```
df = pd.DataFrame(np.random.randint(1, 10, (4, 3)),  
                  columns=['col1', 'col2', 'col3'])  
df
```

	col1	col2	col3
0	6	9	1
1	7	5	3
2	2	8	1
3	5	2	2

df[start:end:step]



```
df[0:1]
```

	col1	col2	col3
0	6	9	1

Selecting rows and some columns: include in a list of names of selecting columns.



```
df[0:1][["col1"]]
```

	col1
0	6



Dataframes: Indexing, slicing



Select columns first then rows later.

```
df
```

	col1	col2	col3
0	6	9	1
1	7	5	3
2	2	8	1
3	5	2	2

```
df[["col1"]][0::2]
```

	col1
0	6
2	2



Indexing with conditions



Operator	Meaning
&	Both must true
	Either condition true

```
df
```

	col1	col2	col3
0	6	9	1
1	7	5	3
2	2	8	1
3	5	2	2

or

```
df[(df["col1"] > 2) | (df["col2"] > 1)]
```

	col1	col2	col3
0	6	9	1
1	7	5	3
2	2	8	1
3	5	2	2

and

```
df[(df["col1"] > 2) & (df["col2"] > 2)]
```

	col1	col2	col3
0	6	9	1
1	7	5	3



Dataframes: Basic Operations



- Functions to do basic operations on Pandas dataframes
 - ✓ `df.head()`: Returns first 5 rows of a dataframe
 - ✓ `df.info()`: Returns information such as number of rows and columns about a dataframe
 - ✓ `df.describe()`: Returns basic statistical measures of a dataframe

(Demo)



Dataframes: Basic Aggregation

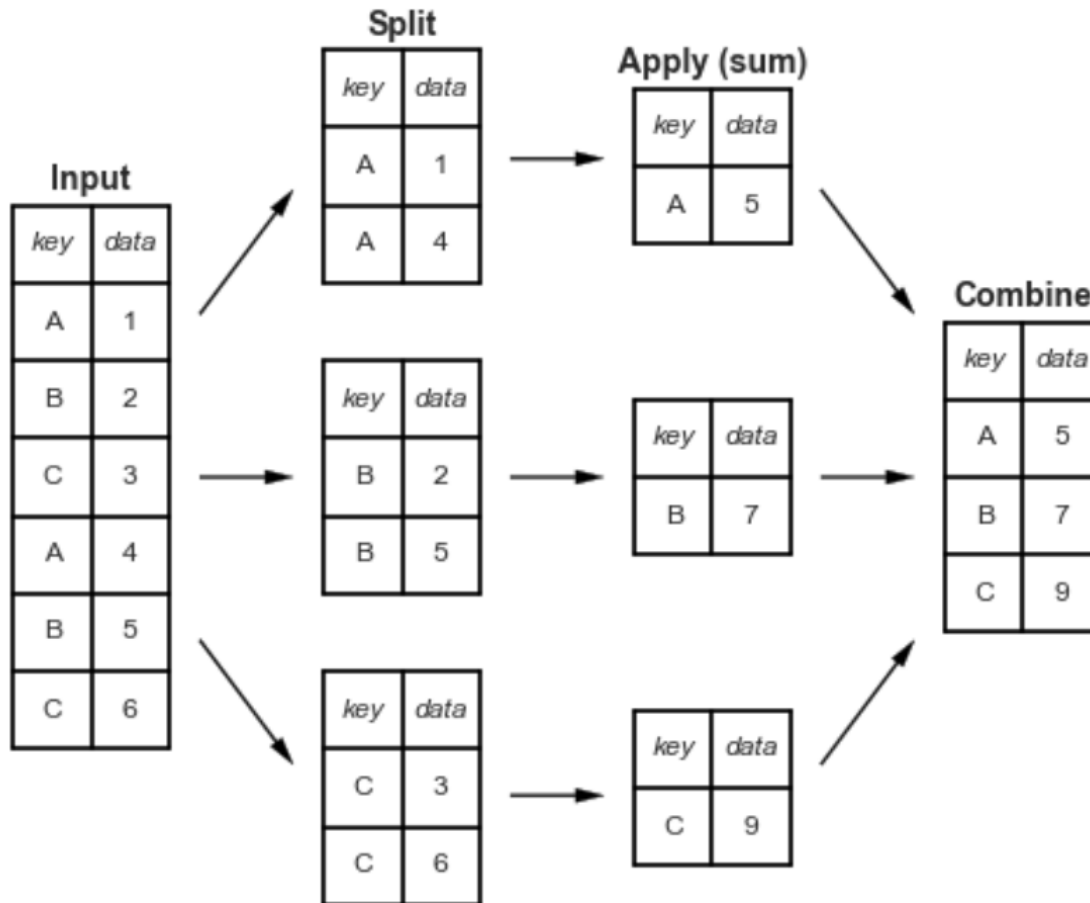


- Functions to do basic operations on Pandas dataframes
 - ✓ `df.mean(.)`: mean of rows or columns
 - ✓ `df.min(.)`: min element of rows or columns
 - ✓ `df.max(.)`: max element of rows or columns
 - ✓ `df.sum(.)`: Returns the sum of rows or columns

(Demo)



Aggregation with Grouping



Grouping all columns

```
df.groupby('key').sum()
```

key	data
A	5
B	7
C	9



Aggregation with Grouping



```
data = {  
    "key": ['A', 'B', 'C', 'A', 'B', 'C'],  
    "data1": [1, 2, 3, 4, 5, 6],  
    "data2": [10, 11, 12, 13, 14, 15],  
    "data3": [20, 21, 22, 13, 24, 25]  
}
```

```
df = pd.DataFrame(data)  
df
```

	key	data1	data2	data3
0	A	1	10	20
1	B	2	11	21
2	C	3	12	22
3	A	4	13	13
4	B	5	14	24
5	C	6	15	25

Grouping all columns

```
df.groupby('key').sum()
```

	data1	data2	data3
key			
A	5	23	33
B	7	25	45
C	9	27	47

Grouping for selected columns

```
df.groupby('key')[["data1", "data2"]].sum()
```

	data1	data2
key		
A	5	23
B	7	25
C	9	27



Dataframes: Dropping values

- Often, the data will consist of missing values 'NaN'

df =

	col1	col2	col3
0	1.0	444.0	orange
1	2.0	555.0	apple
2	3.0	NaN	grape
3	4.0	444.0	mango
4	NaN	666.0	jackfruit
5	6.0	111.0	watermelon
6	7.0	NaN	banana
7	NaN	222.0	peach

- Missing values lead to problems in the data analysis process



Dataframes: Dropping values



- You can use the `dropna()` function to drop all the rows consisting of the missing values

In [27]:

```
df_dropped = df.dropna()
```

↑
New dataframe
after filling values

↑
Function to
drop values

(Demo)



Dataframes: Filling values

- You can also use the `fillna()` function to fill values (e.g. a value of '0' in the place of 'NaN') in the place of missing values

In [29]:

```
df_filled = df.fillna(0)
```

↑
New dataframe
after filling values

↑
Function to
drop values

↙
Fill value

(Demo)



Dataframes: Filling values

- You can also use the `fillna()` function to fill values (e.g. mean value of each column in the place of 'NaN') in the place of missing values

In [31]:

```
df_filled = df.fillna(df.mean())
```

↑
New dataframe
after filling values

↑
Function to
drop values

↙
Fill value

(Demo)

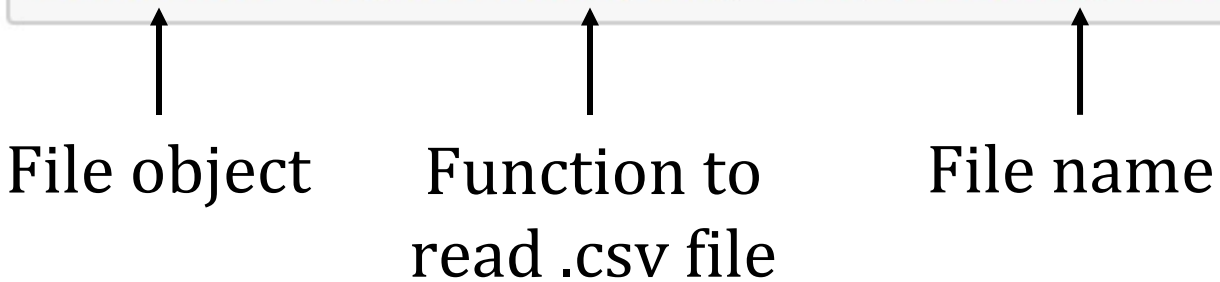


Dataframes: Reading a File



- Objective is to read the data in a '.csv' (comma separated values) file and print it as a dataframe

```
In [4]: readfile = pd.read_csv('CSV_ReadingFile.csv')
```



- Printing the contents of the .csv file to the output screen

```
In [7]: readfile
```

(Demo)



Dataframes: Writing a File

- Objective is to write the data in a new '.csv' (comma separated values) file

```
In [11]: readfile.to_csv('CSV_WritingFile.csv', index=False)
```

Function to
read .csv file

File name

Excludes
row labels

- Note: File name that you give will first be created in the same folder where the Jupyter notebook is present

(Demo)



Summary



- Concepts of representing data in the form of Pandas dataframes are covered
- Concepts of interpreting, manipulating, and analyzing data within Pandas dataframes are covered