Programming Basics and Data Analytics with Python

simplilearn

Data Manipulation with Pandas

Learning Objectives

By the end of this lesson, you will be able to:

- Explain pandas and its features
- List different data structures of pandas
- Outline the process to create Series and DataFrame with data inputs
- Explain how to view, select, and access elements in a data structure
- Describe the procedure to handle vectorized operations
- Illustrate how to handle missing values
- Analyze data with different data operation methods

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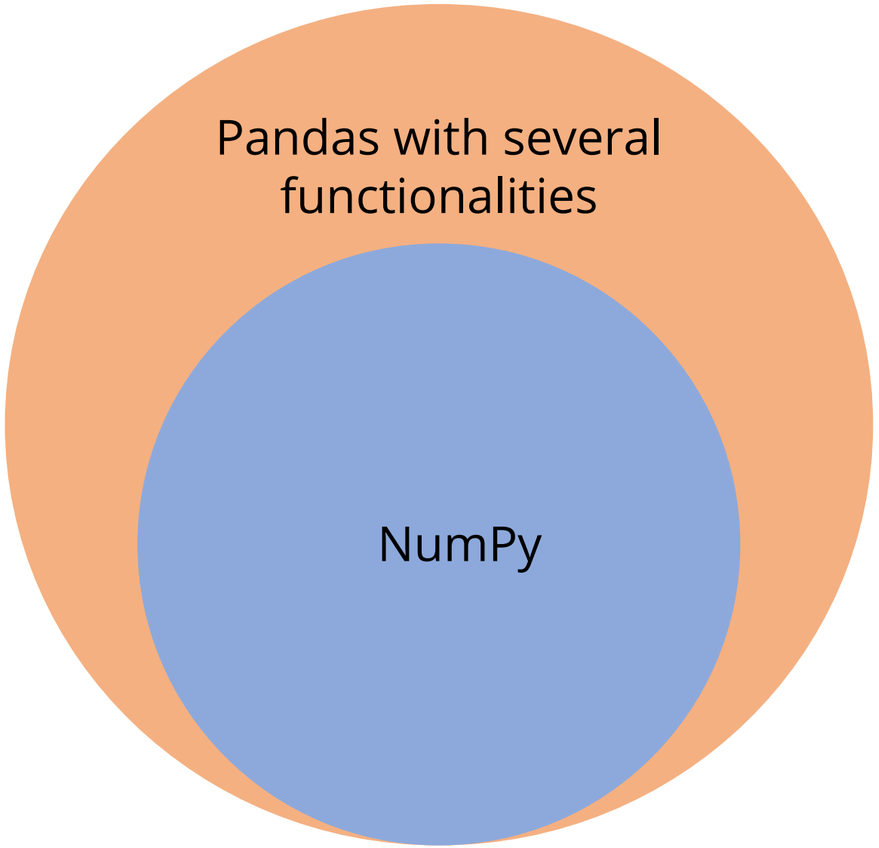
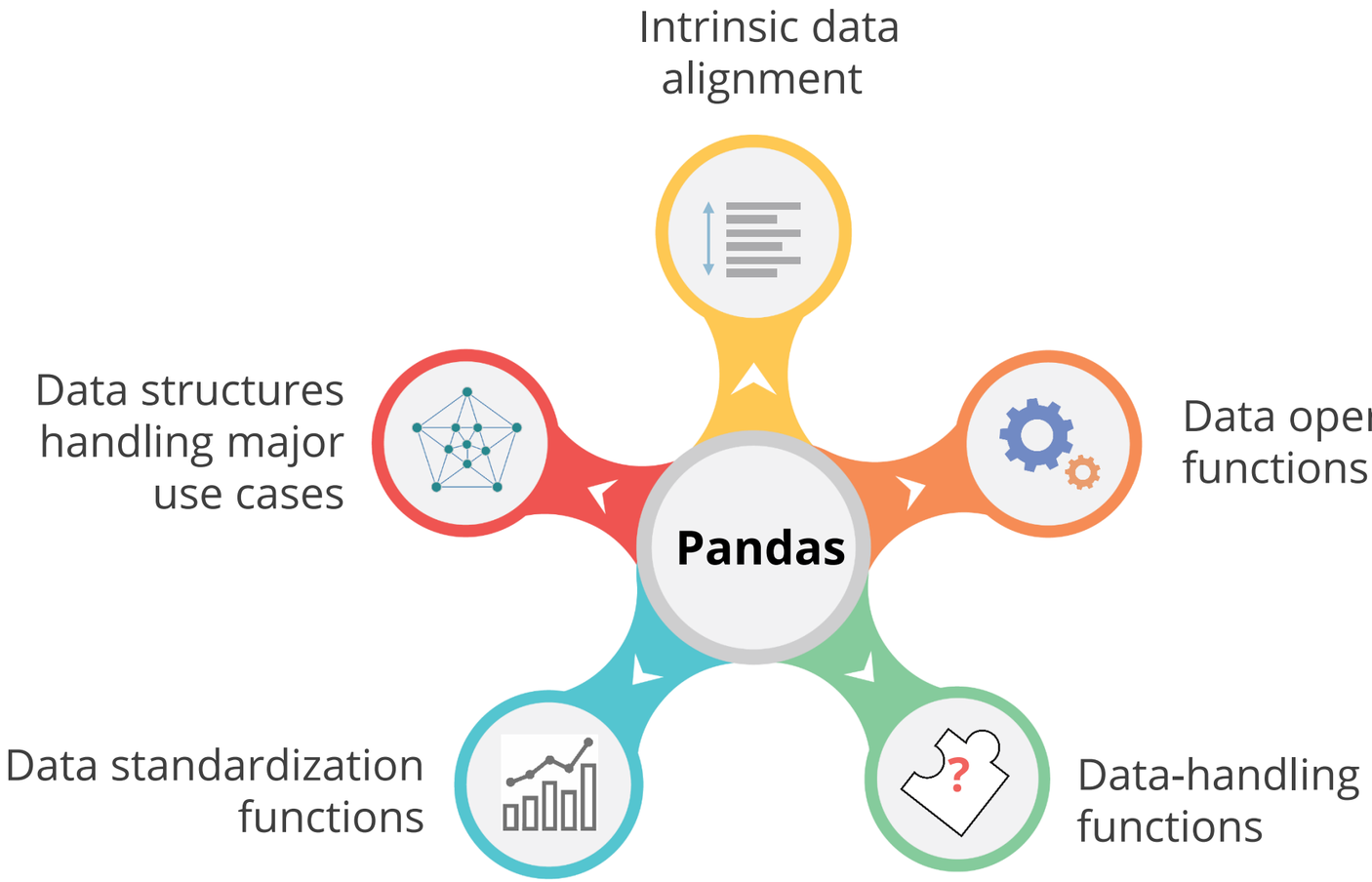
Introduction to Pandas

Why Pandas

NumPy is great for mathematical computing, but, why do we need pandas?

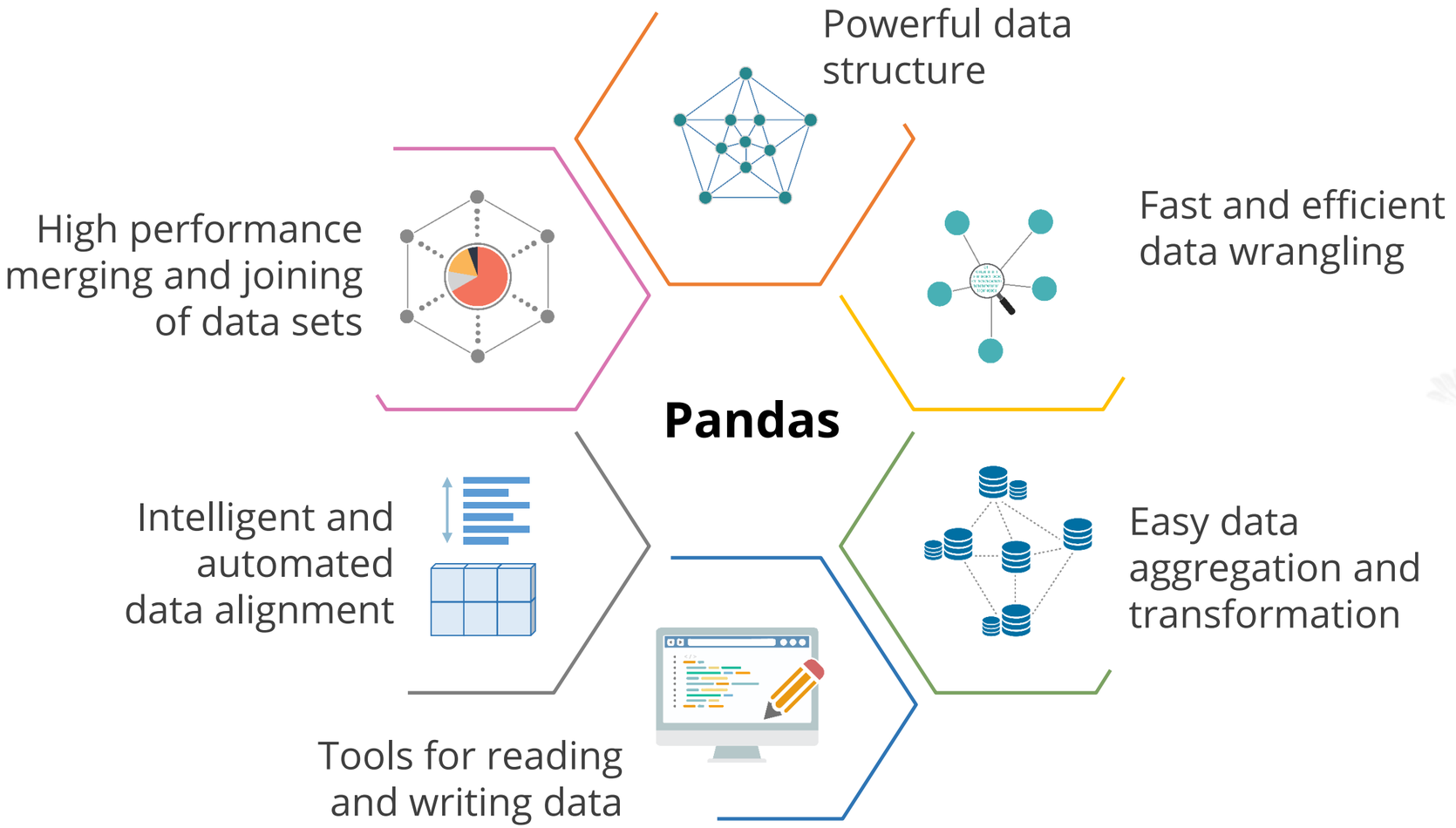


Why Pandas



Features of Pandas

Various features of Pandas make it an efficient library for Data Scientists.



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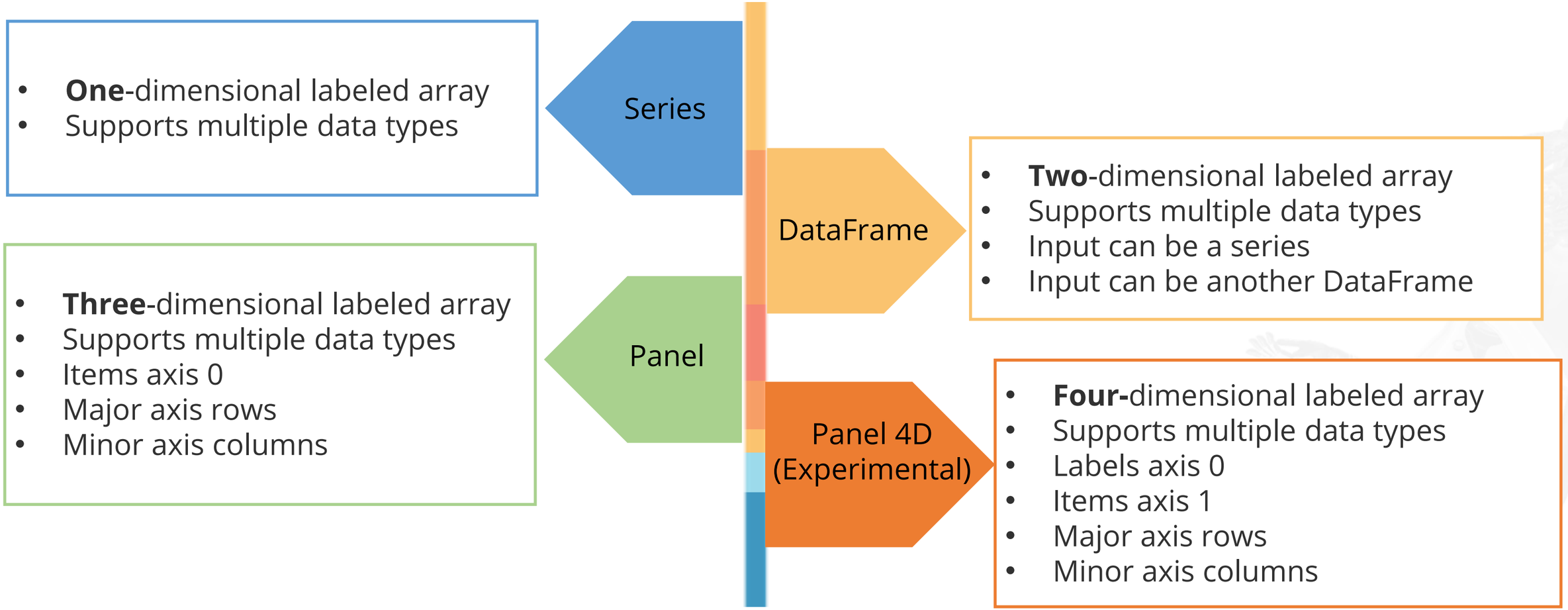
Data Structures

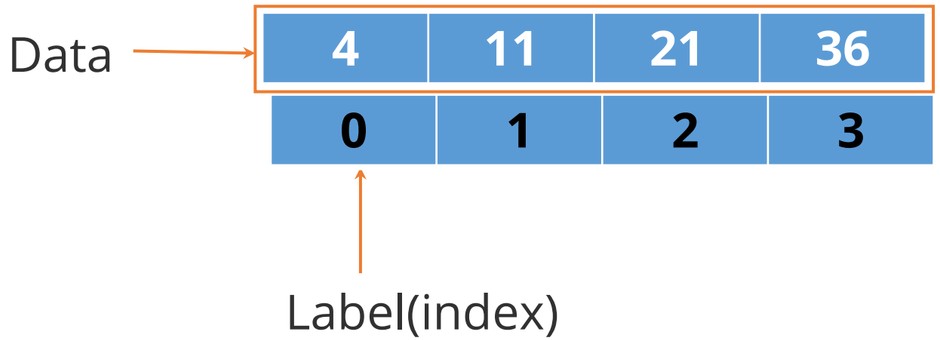
Data Structures

Data structures in Pandas library:



Understanding Series

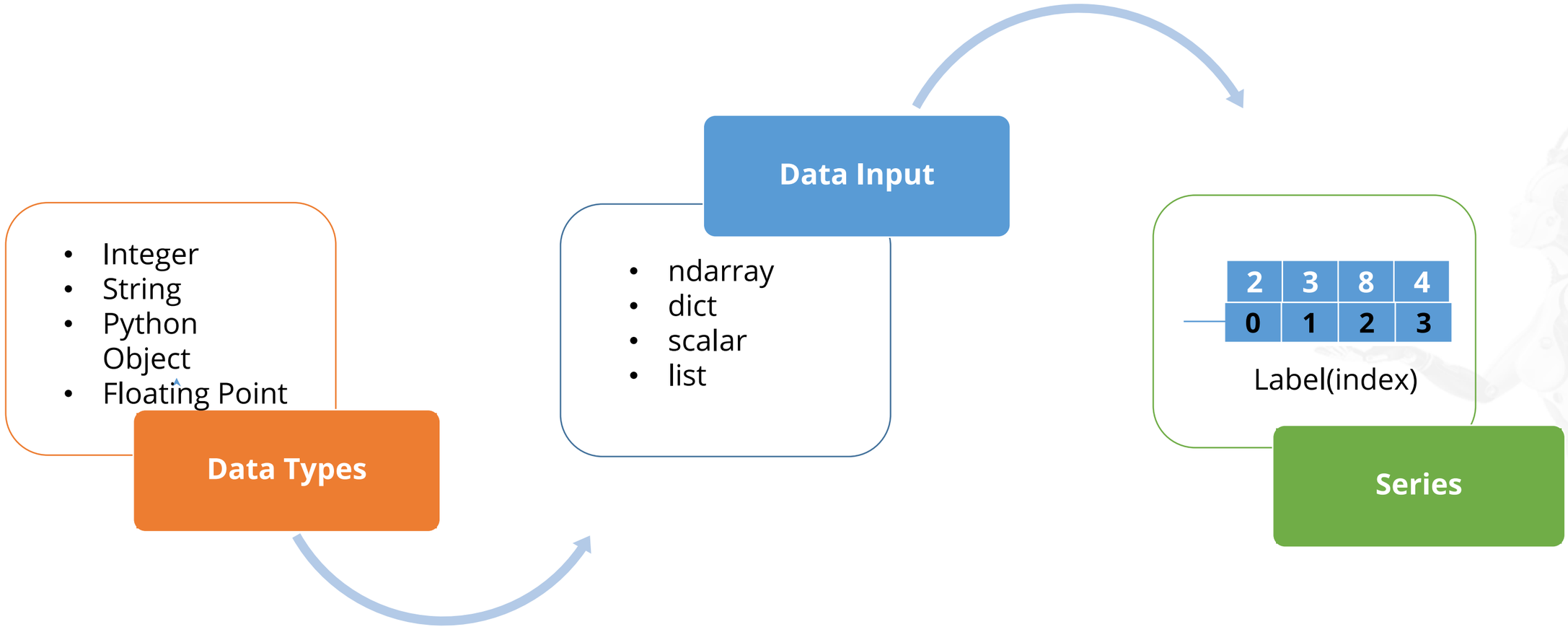
Series is a one-dimensional array-like object containing data and labels (or index).



Data alignment is intrinsic and will not be broken until changed explicitly by program.

Series

Series can be created with different data inputs:



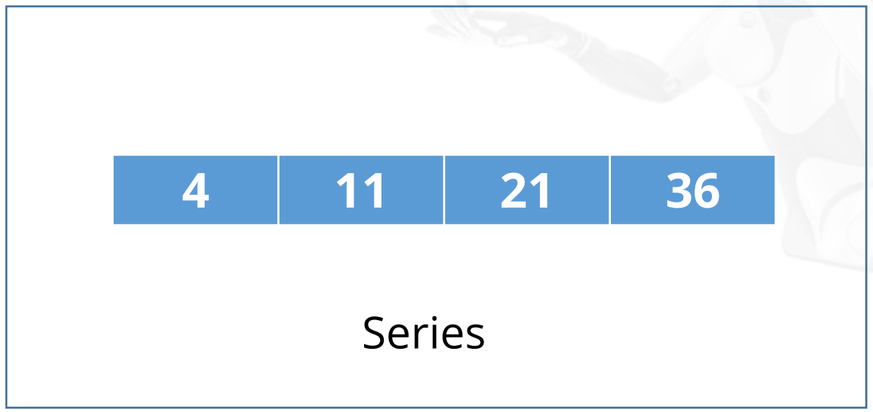
How to Create Series?

Key points to note while creating a series are:

- Import pandas as it is in the main library (Import pandas as pd)
- Import NumPy while working with ndarrays (Import NumPy as np)
- Apply the syntax and pass the data elements as arguments

Basic Method

```
S = pd.Series(data, index = [index])
```

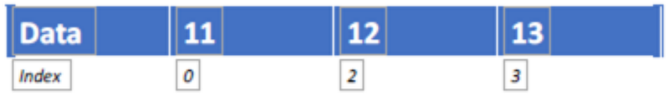


Series: Example

Consider this example of series in pandas:

Code

```
data = [11,12,13]
s = pd.Series(data)
s
0  11
1  23
2  23
dtype: int64
S[1]
23
```

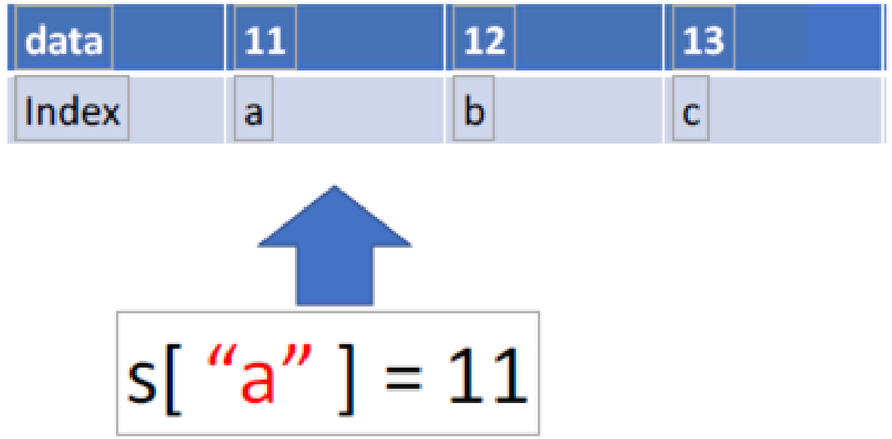


s[1] = 12

Series: Example

Code

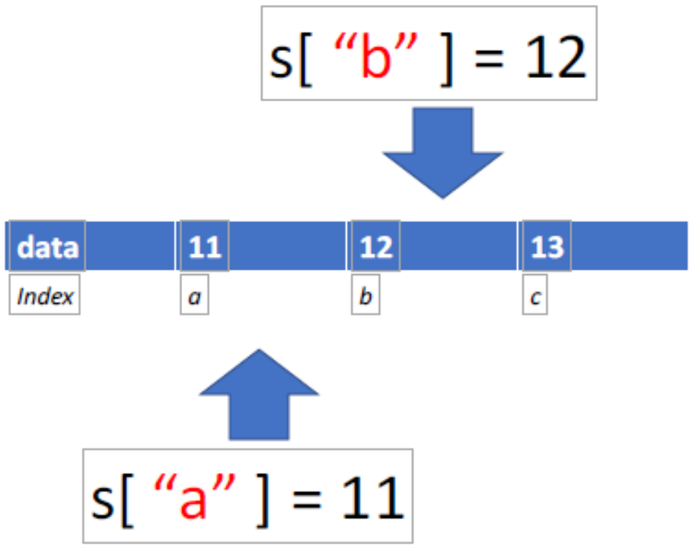
```
data = [11,12,13]
index = [“a”,“b”,“c”]
s = pd.Series(data, index=ind)
s
a  11
b  23
c  23
s[“a”]
11
```



Series: Example

Code

```
import pandas as pd
data = [11,12,13]
ind = [“a”,“b”,“c”]
S = pd.Series(data, index=ind)
S[[“a”, “b”]]
a  11
B  12
dtype: int64
```



Creating Series from a List

```
In [14]: import numpy as np
         import pandas as pd
```

Import libraries

```
In [15]: first_series = pd.Series(list('abcdef'))
```

Pass list as an argument

```
In [16]: print (first_series)
```

```
0    a
1    b
2    c
3    d
4    e
5    f
dtype: object
```

Data value

Index

Data type

Creating Series from an ndarray

ndarray for countries

```
In [17]: np_country = np.array(['Luxembourg','Norway','Japan','Switzerland','United States','Qatar','Iceland','Sweden',
                                'Singapore','Denmark'])

In [18]: s_country = pd.Series(np_country)

In [19]: print (s_country)

0       Luxembourg
1           Norway
2            Japan
3      Switzerland
4    United States
5            Qatar
6          Iceland
7           Sweden
8        Singapore
9          Denmark
dtype: object
```

Pass ndarray as an argument

countries

Data type

Creating Series from dict

A series can also be created with dict data input for faster operations.

dict for countries and their GDP

```
In [10]: #Evaluate countries and their corresponding gdp per capita and print them as series
         dict_country_gdp = pd.Series([52056.01781,40258.80862,40034.85063,39578.07441,39170.41371,37958.23146,37691.02733,
                                       36152.66676,34706.19047,33630.24604,33529.83052,30860.12808],index=['Luxembourg','Macao, China','Norway',
                                     'Japan','Switzerland','Hong Kong, China','United States','Qatar','Iceland','Sweden','Singapore','Denmark'])

In [11]: print (dict_country_gdp)
```

```
Luxembourg          52056.01781
Macao, China        40258.80862
Norway              40034.85063
Japan               39578.07441
Switzerland         39170.41371
Hong Kong, China    37958.23146
United States       37691.02733
Qatar               36152.66676
Iceland             34706.19047
Sweden              33630.24604
Singapore           33529.83052
Denmark             30860.12808
dtype: float64
```

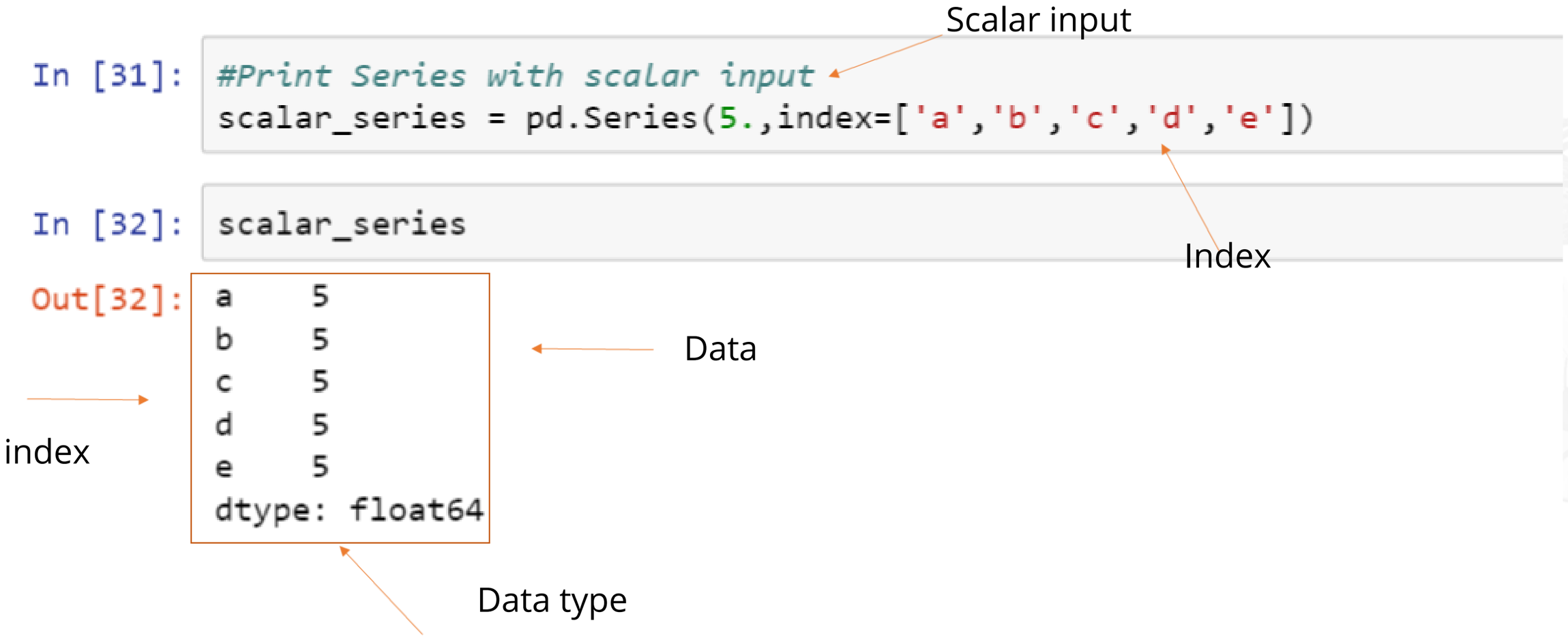
Countries have been passed as an index and GDP as the actual data value

GDP

Country

Data type

Creating Series from Scalar



Accessing Elements in Series

Data can be accessed through different functions like loc and iloc, by passing data element position or index range.

```
In [43]: #access elements in the series
         dict_country_gdp[0]
```

Data element position

```
Out[43]: 52056.017809999998
```

```
In [44]: #access first 5 countries from the series
         dict_country_gdp[0:5]
```

First five data elements and their indices

```
Out[44]: Luxembourg      52056.01781
         Macao, China    40258.80862
         Norway          40034.85063
         Japan           39578.07441
         Switzerland     39170.41371
         dtype: float64
```

```
In [45]: #Look up a country by name or index
         dict_country_gdp.loc['United States']
```

Look up method to access data

```
Out[45]: 37691.027329999997
```

```
In [46]: #Look up by position
         dict_country_gdp.iloc[0]
```

Data elements by looking up the index position

```
Out[46]: 52056.017809999998
```

Vectorized Operations in Series

Vectorized operations are performed by the data element's position.

```
In [52]: first_vector_series = pd.Series([1,2,3,4],index=['a','b','c','d'])
         second_vector_series = pd.Series([10,20,30,40],index=['a','b','c','d'])
```

Add the series

```
In [53]: first_vector_series+second_vector_series
```

```
Out[53]: a    11
         b    22
         c    33
         d    44
         dtype: int64
```

Addition at index level

```
In [54]: second_vector_series = pd.Series([10,20,30,40],index=['a','d','b','c'])
```

```
In [55]: first_vector_series+second_vector_series
```

```
Out[55]: a    11
         b    32
         c    43
         d    24
         dtype: int64
```

Addition after shuffling the indices

Vectorized Operations in Series

```
In [19]: #now replace few indexes with new ones in second vector series
         second_vector_series = pd.Series([10,20,30,40],index=['a','b','e','f'])

In [20]: first_vector_series+second_vector_series

Out[20]: a    11
         b    22
         c   NaN
         d   NaN
         e   NaN
         f   NaN
         dtype: float64
```

Adding two series with a few common and a few different indices

Create Pandas Series

Objective: Create Pandas Series to input employee names and assign employee IDs to each employee. Also, retrieve first five employees from the Pandas Series using their IDs.

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

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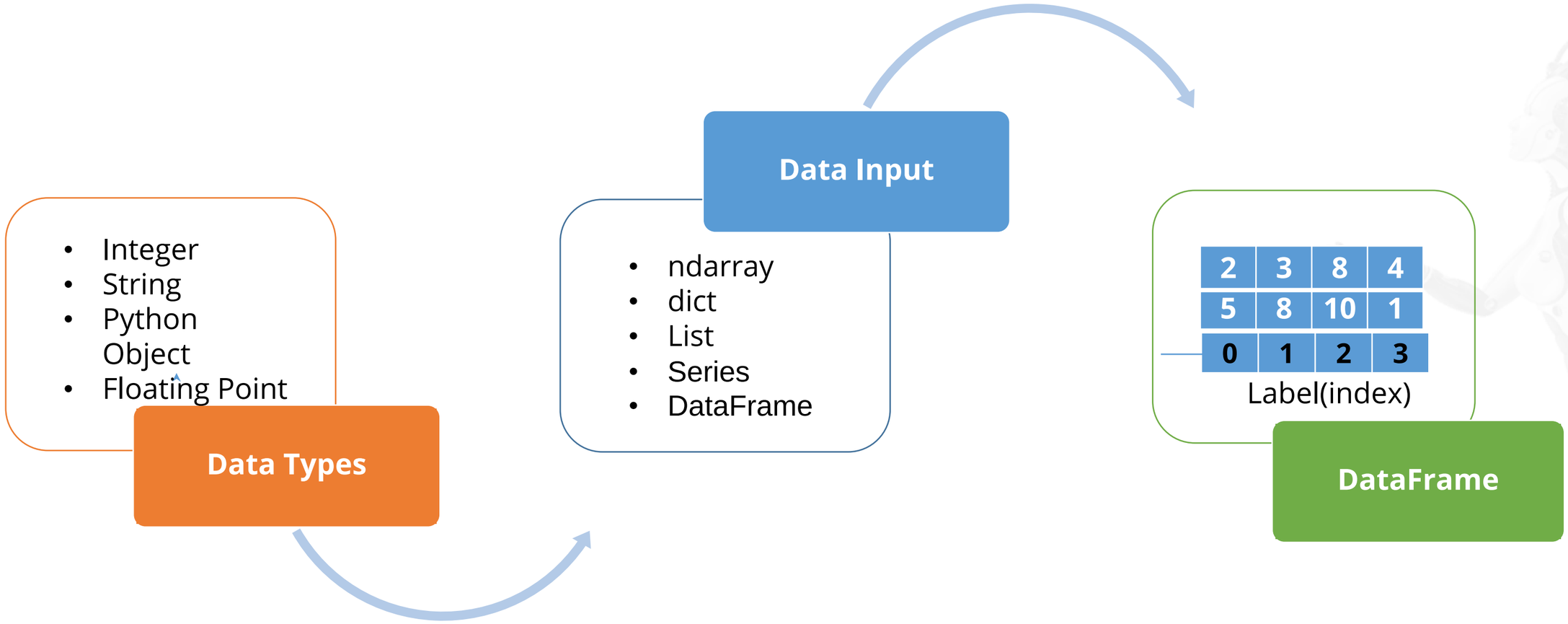
DataFrames

DataFrame

DataFrame is a two-dimensional labeled data structure with columns of potentially different types.



Creating DataFrame from Lists

In [1]:
```
import pandas as pd
```

Create DataFrame from dict of equal length lists

In [2]:
```
#last five olymnics data: place, year and number of countries participated
olympic_data_list = {'HostCity':['London','Beijing','Athens','Sydney','Atlanta'],
                     'Year':[2012,2008,2004,2000,1996],
                      'No. of Participating Countries':[205,204,201,200,197]
                     }
```

In [3]:
```
df_olympic_data = pd.DataFrame(olympic_data_list)
```

Pass the list to the DataFrame

In [4]:
```
df_olympic_data
```

Out[4]:

	HostCity	No. of Participating Countries	Year
0	London	205	2012
1	Beijing	204	2008
2	Athens	201	2004
3	Sydney	200	2000
4	Atlanta	197	1996

Creating DataFrame from dict

This example shows you how to create a DataFrame from a series of dicts:

Create DataFrame from dict of dicts

dict one

dict two

```
In [5]: olympic_data_dict = {'London':{2012:205},'Beijing':{2008:204}}

In [6]: df_olympic_data_dict = pd.DataFrame(olympic_data_dict)

In [7]: df_olympic_data_dict
```

Entire dict

Out[7]:

	Beijing	London
2008	204	NaN
2012	NaN	205

Viewing DataFrame

You can view a DataFrame by referring to the column name or the describe function.

```
In [8]: #select by City name
        df_olympic_data.HostCity
```

Viewing a DataFrame

```
Out[8]: 0     London
        1    Beijing
        2     Athens
        3     Sydney
        4    Atlanta
        Name: HostCity, dtype: object
```

```
In [9]: #use describe function to display the content
        df_olympic_data.describe
```

Viewing the entire contents of the dataset

```
Out[9]: <bound method DataFrame.describe of   HostCity  No. of Participating Countries  Year
        0   London                          205  2012
        1  Beijing                          204  2008
        2   Athens                          201  2004
        3   Sydney                          200  2000
        4  Atlanta                          197  1996>
```

Creating DataFrame from dict of Series

Create DataFrame from dict of series

In [10]:
```
olympic_series_participation = pd.Series([205,204,201,200,197],index=[2012,2008,2004,2000,1996])
olympic_series_country = pd.Series(['London','Beijing','Athens','Sydney','Atlanta'],
                                   index=[2012,2008,2004,2000,1996])
```

In [11]:
```
df_olympic_series = pd.DataFrame({'No. of Participating Countries':olympic_series_participation,
                                  'Host Cities':olympic_series_country})
```

In [12]:
```
df_olympic_series
```

Out[12]:

	Host Cities	No. of Participating Countries
2012	London	205
2008	Beijing	204
2004	Athens	201
2000	Sydney	200
1996	Atlanta	197

Creating DataFrame from ndarray

Create DataFrame from dict of ndarray

```
In [13]: import numpy as np

In [14]: np_array = np.array([2012,2008,2004,2006])
         dict_ndarray = {'year':np_array}

In [15]: df_ndarray = pd.DataFrame(dict_ndarray)

In [16]: df_ndarray
```

Create an ndarray with years

Create a dict with the ndarray

Pass this dict to a new DataFrame

Out[16]:

	year
0	2012
1	2008
2	2004
3	2006

Creating DataFrame from DataFrame Object

In [17]:
```
df_from_df = pd.DataFrame(df_olympic_series)
```
Create a DataFrame from a DataFrame object

In [18]:
```
df_from_df
```

Out[18]:

	Host Cities	No. of Participating Countries
2012	London	205
2008	Beijing	204
2004	Athens	201
2000	Sydney	200
1996	Atlanta	197

Create Pandas DataFrames

Objective: Create DataFrames in the following conditions:

1. Input the marks of two subjects for five students and create a DataFrame using the two series
2. Extract data from the given SalaryGender CSV file and store the data from each column in a separate NumPy array
3. Create a DataFrame using dictionary of names and age of five people as input
4. Create a DataFrame from the dictionary of series

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

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Create Pandas DataFrames

Objective: Create the following dataframes:

1. A DataFrame of weather data using dictionary with the keys like day, temperature, and weather conditions
2. Create a DataFrame with a list of dictionaries, rows, and columns

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

UNASSISTED PRACTICE

Unassisted Practices: Create Pandas DataFrame

```
[10]: import pandas as pd
weather_data = {
    'day': ['01-09-2019','02-09-2019','03-09-2019','04-09-2019','05-09-2019','06-09-2019',
            '07-09-2019','08-09-2019','09-09-2019','10-09-2019'],
    'temperature': [32,35,28,24,32,31,33,34,30,29],
    'Weather': ['Rainy', 'Sunny', 'Overcast','Windy','Sunny','Sunny','Rainy','Cloudy','Overcast','Windy']
}
df = pd.DataFrame(weather_data)
print(df)
#df = pd.read_csv("weather_data.csv")
```

Dictionary of list of weather data

DataFrame of dictionary

	day	temperature	Weather
0	01-09-2019	32	Rainy
1	02-09-2019	35	Sunny
2	03-09-2019	28	Overcast
3	04-09-2019	24	Windy
4	05-09-2019	32	Sunny
5	06-09-2019	31	Sunny
6	07-09-2019	33	Rainy
7	08-09-2019	34	Cloudy
8	09-09-2019	30	Overcast
9	10-09-2019	29	Windy

Output

Unassisted Practices: Create Pandas DataFrame

```
[14]: import pandas as pd

people_data = [{'John': 45, 'Daisy': 55}, {'John': 65, 'Daisy': 60, 'Alice': 20}]

dataframe = pd.DataFrame(people_data, index =['Maths','Science'], columns =['John','Daisy'])

dataframe2 = pd.DataFrame(people_data, index =['Maths','Science'], columns =['John', 'Alice'])

print (dataframe,  "\n")
print (dataframe2)
```

Initializing the list data

Two column indices of values same as the dictionary keys

Print the first dataframe

Print the second dataframe

```
         John  Daisy
Maths      45     55
Science    65     60

         John  Alice
Maths      45    NaN
Science    65   20.0
```

Output

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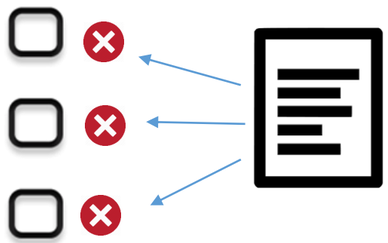
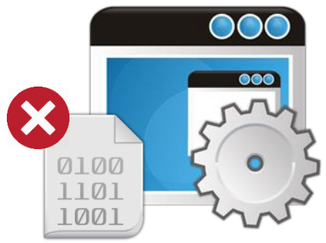
Missing Values

Missing Values

Various factors may lead to missing data values:

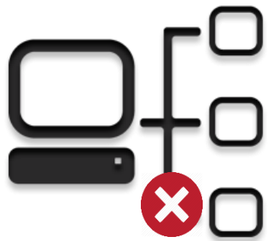


Data not provided by the source



Software issue

Data integration issue



Network issue

Handling Missing Values

It’s difficult to operate a dataset when it has missing values or uncommon indices.

```
In [3]: import pandas as pd

In [4]: #declare first series
        first_series = pd.Series([1,2,3,4,5],index=['a','b','c','d','e'])

In [5]: #declare second series
        second_series=pd.Series([10,20,30,40,50],index=['c','e','f','g','h'])

In [6]: sum_of_series = first_series+second_series

In [7]: sum_of_series

Out[7]: a    NaN
        b    NaN
        c     13
        d    NaN
        e     25
        f    NaN
        g    NaN
        h    NaN
        dtype: float64
```

Handling Missing Values with Functions

The dropna function drops all the values with uncommon indices.

```
In [5]: sum_of_series

Out[5]: a     NaN
        b     NaN
        c    13.0
        d     NaN
        e    25.0
        f     NaN
        g     NaN
        h     NaN
        dtype: float64

In [6]: # drop NaN( Not a Number) values from dataset
        dropna_s = sum_of_series.dropna()

In [7]: dropna_s

Out[7]: c    13.0
        e    25.0
        dtype: float64
```

Handling Missing Values: Example

Code

```
import pandas as pd
import numpy      as np

Raw_data = {'name':['Joe', np.nan, 'Tina', 'Mike', 'Amy'],
              'last_name':['Miller', np.nan, 36,24,73]}
                    'AGE':[42, np.nan, 36, 24,73]}
df =pd.DataFrame(raw_data)
df
```

	age	last_name	name
0	42.0	Miller	Joe
1	NaN	NaN	NaN
2	36.0	Ali	Tina
3	24.0	Bob	Mike
4	73.0	Das	Amy

Handling Missing Values: Example

Code

```
df.dropna()
```

	age	last_name	name
0	42.0	Miller	Joe
2	36.0	Ali	Tina
3	24.0	Bob	Mike
4	73.0	Das	Amy

Handling Missing Values with Functions

The fillna function fills all the uncommon indices with a number instead of dropping them.

```
In [8]: dropna_s.fillna(0)
```

Fill the missing values with zero

```
Out[8]: c    13.0
        e    25.0
        dtype: float64
```

```
In [9]: # Fill NaN( Not a Number) values with Zeroes (0)
        fillna_s = sum_of_series.fillna(0)
```

```
In [10]: fillna_s
```

```
Out[10]: a     0.0
         b     0.0
         c    13.0
         d     0.0
         e    25.0
         f     0.0
         g     0.0
         h     0.0
         dtype: float64
```

Handle Missing Values with Functions: Example

```
In [10]: #fill values with zeroes before performing addition operation for missing indices
         fill_NaN_with_zeros_before_sum =first_series.add(second_series,fill_value=0)

In [11]: fill_NaN_with_zeros_before_sum

Out[11]: a     1
         b     2
         c    13
         d     4
         e    25
         f    30
         g    40
         h    50
         dtype: float64
```

Handle Missing Values

Objective: Create a dataframe with a list of dictionaries, rows indices, and column indices with one index having a different name. Handle the missing values by:

1. Removing the NaN values
2. Filling all the uncommon or NaN values with a number, instead of dropping them

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

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Data Operation

Data Operation

Data operation can be performed through various built-in methods for faster data processing.

In [1]:
```
import pandas as pd
```

In [2]:
```
#declare movie rating dataframe: ratings from 1 to 5 (star * rating)
df_movie_rating = pd.DataFrame(
                  {'movie 1': [5,4,3,3,2,1],
                  'movie 2': [4,5,2,3,4,2]},
                  index=['Tom','Jeff','Peter','Ram','Ted','Paul']
)
```

In [3]:
```
df_movie_rating
```

Out[3]:

	movie 1	movie 2
Tom	5	4
Jeff	4	5
Peter	3	2
Ram	3	3
Ted	2	4
Paul	1	2

Data Operation with Functions

While performing data operation, custom functions can be applied using the applymap method.

```
In [4]: def movie_grade(rating):
            if rating==5:
                return 'A'
            if rating==4:
                return 'B'
            if rating==3:
                return 'C'
            else:
                return 'F'
```

Declare a custom function

```
In [5]: print (movie_grade(5))
A
```

Test the function

```
In [6]: df_movie_rating.applymap(movie_grade)
```

Apply the function to the DataFrame

Out[6]:

	movie 1	movie 2
Tom	A	B
Jeff	B	A
Peter	C	F
Ram	C	C
Ted	F	B
Paul	F	F

Data Operation with Statistical Functions

```
In [7]: df_test_scores = pd.DataFrame(
                    {'Test1': [95,84,73,88,82,61],
                    'Test2': [74,85,82,73,77,79]},
                    index=['Jack','Lewis','Patrick','Rich','Kelly','Paula']
)
```

Create a DataFrame with two tests

```
In [8]: df_test_scores.max()
```

Apply the max function to find the maximum score

```
Out[8]: Test1    95
        Test2    85
        dtype: int64
```

```
In [9]: df_test_scores.mean()
```

Apply the mean function to find the average score

```
Out[9]: Test1    80.500000
        Test2    78.333333
        dtype: float64
```

```
In [10]: df_test_scores.std()
```

Apply the std function to find the standard deviation for both the tests

```
Out[10]: Test1    11.979149
         Test2     4.633213
         dtype: float64
```

Data Operation Using Groupby

In [16]:
```
df_president_name = pd.DataFrame({'first':['George','Bill', 'Ronald','Jimmy','George'],
                                  'last':['Bush','Clinton', 'Regan', 'Carter', 'Washington']})
```

In [17]:
```
df_president_name
```

Out[17]:

	first	last
0	George	Bush
1	Bill	Clinton
2	Ronald	Regan
3	Jimmy	Carter
4	George	Washington

Create a DataFrame with first and last name as former presidents

In [18]:
```
grouped = df_president_name.groupby('first')
```

Group the DataFrame with the first name

In [19]:
```
grp_data = grouped.get_group('George')
grp_data
```

Out[19]:

	first	last
0	George	Bush
4	George	Washington

Data Operation: Sorting

```
In [20]: df_president_name.sort_values('first')
```

Sort values by first name

Out[20]:

	first	last
1	Bill	Clinton
0	George	Bush
4	George	Washington
3	Jimmy	Carter
2	Ronald	Regan

Data Operations in Pandas DataFrame

Objective: Consider the **SalaryGender** dataset to perform the following operations on pandas dataframe:

1. Replace the values 0 and 1 of the gender column with female and male respectively
2. Find the maximum salary and the minimum salary
3. Find the number of men and women with PhD
4. Store age and Phd columns in the dataframe and remove all the people without PhD
5. Calculate the total number of PhD holders
6. Sort the dataframe on the basis of salary

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

ASSISTED PRACTICE

Data Operations in Pandas DataFrame

Objective: Consider the **Pandas-results.csv** dataset to perform the following operations on pandas dataframe:

1. Create a dataframe of home_team and home_score
2. Find the home team with maximum home scores
3. Find the teams with home score greater than zero and get the basic statistical details of the dataframe
4. Create a dataframe in the date range 2019-07-10 to 2019-07-20
5. Compare the home score and away score of the teams and add the winning team to a column named winner

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

UNASSISTED PRACTICE

Unassisted Practice: Data Operations in Pandas DataFrame

```
import pandas as pd, numpy as np
df=pd.read_csv('Pandas-results.csv',delimiter=',')
df[['home_team','home_score']]
```

Reading the Pandas-results.csv file

Creating a dataframe of home team and home score

	home_team	home_score
0	Scotland	0
1	England	4
2	Scotland	2
3	England	2
4	Scotland	3
5	Scotland	4
6	England	1
7	Wales	0
:	:	:
:	:	:
;	;	;

Output

Unassisted Practice: Data Operations in Pandas DataFrame

```
import pandas as pd
sort_df=df.sort_values('home_score')
df
print("Maximum home_goals:",df['home_score'].max())
#sort_df.iloc[23781]
df.loc[df['home_score'] == 31]
```

Using the max() function to find the maximum goals

Extracting the team details with maximum home score

```
Maximum home_goals: 31
```

	date	home_team	away_team	home_score	away_score	tournament	city	country	neutral
23781	2001-04-11	Australia	American Samoa	31	0	FIFA World Cup qualification	Coffs Harbour	Australia	False

Output

Unassisted Practice: Data Operations in Pandas DataFrame

```
print(df[df['home_score']>0]) #find the teams with home_score greater than zero
#df.describe() #get the basic statistical details of the dataframe
```

```
           date        home_team        away_team  home_score  away_score  \
1    1873-03-08          England         Scotland           4           2
2    1874-03-07         Scotland          England           2           1
3    1875-03-06          England         Scotland           2           2
4    1876-03-04         Scotland          England           3           0
5    1876-03-25         Scotland            Wales           4           0
6    1877-03-03          England         Scotland           1           3
8    1878-03-02         Scotland          England           7           2
9    1878-03-23         Scotland            Wales           9           0
10   1879-01-18          England            Wales           2           1
11   1879-04-05          England         Scotland           5           4
13   1880-03-13         Scotland          England           5           4
14   1880-03-15            Wales          England           2           3
15   1880-03-27         Scotland            Wales           5           1
17   1881-03-12          England         Scotland           1           6
18   1881-03-14            Wales         Scotland           1           5
20   1882-02-25            Wales  Northern Ireland           7           1

                                  ::::
                                  ::::
                                  ;;;;
```

Output

Unassisted Practice: Data Operations in Pandas DataFrame

```
df.loc[40809:40838]
```

Creating a dataframe in the date range 2019-07-10 to 2019-07-20

	date	home_team	away_team	home_score	away_score	tournament	city	country	neutral
40809	2019-07-10	Senegal	Benin	1	0	African Cup of Nations	Cairo	Egypt	True
40810	2019-07-10	Papua New Guinea	Vanuatu	2	0	Pacific Games	Apia	Samoa	True
40811	2019-07-10	Tuvalu	Tahiti	0	7	Pacific Games	Apia	Samoa	True
40812	2019-07-10	American Samoa	Fiji	0	9	Pacific Games	Apia	Samoa	True
40813	2019-07-10	Solomon Islands	New Caledonia	0	2	Pacific Games	Apia	Samoa	True
40814	2019-07-10	Tajikistan	Syria	2	0	Intercontinental Cup	Ahmedabad	India	True
40815	2019-07-11	Ivory Coast	Algeria	1	1	African Cup of Nations	Suez	Egypt	True
40816	2019-07-11	Madagascar	Tunisia	0	3	African Cup of Nations	Cairo	Egypt	True
40817	2019-07-12	Samoa	Tonga	2	0	Pacific Games	Apia	Samoa	False
40818	2019-07-12	American Samoa	Tuvalu	1	1	Pacific Games	Apia	Samoa	True
40819	2019-07-12	Solomon Islands	Tahiti	0	3	Pacific Games	Apia	Samoa	True
40820	2019-07-12	New Caledonia	Fiji	1	0	Pacific Games	Apia	Samoa	True
40821	2019-07-13	India	North Korea	2	5	Intercontinental Cup	Ahmedabad	India	False
40822	2019-07-14	Algeria	Nigeria	2	1	African Cup of Nations	Cairo	Egypt	True

Output

Unassisted Practice: Data Operations in Pandas DataFrame

Adding a column named winner to the filtered dataframe

```
dfx=df[df['home_score']>0]
dfx['winner'] = np.where((dfx['home_score'] > dfx['away_score']), dfx['home_team'], dfx['away_team'])
dfx
```

Filtering the data with home score greater than zero

	date	home_team	away_team	home_score	away_score	tournament	city	country	neutral	win	winner
1	1873-03-08	England	Scotland	4	2	Friendly	London	England	False	England	England
2	1874-03-07	Scotland	England	2	1	Friendly	Glasgow	Scotland	False	Scotland	Scotland
3	1875-03-06	England	Scotland	2	2	Friendly	London	England	False	Scotland	Scotland
4	1876-03-04	Scotland	England	3	0	Friendly	Glasgow	Scotland	False	Scotland	Scotland
5	1876-03-25	Scotland	Wales	4	0	Friendly	Glasgow	Scotland	False	Scotland	Scotland

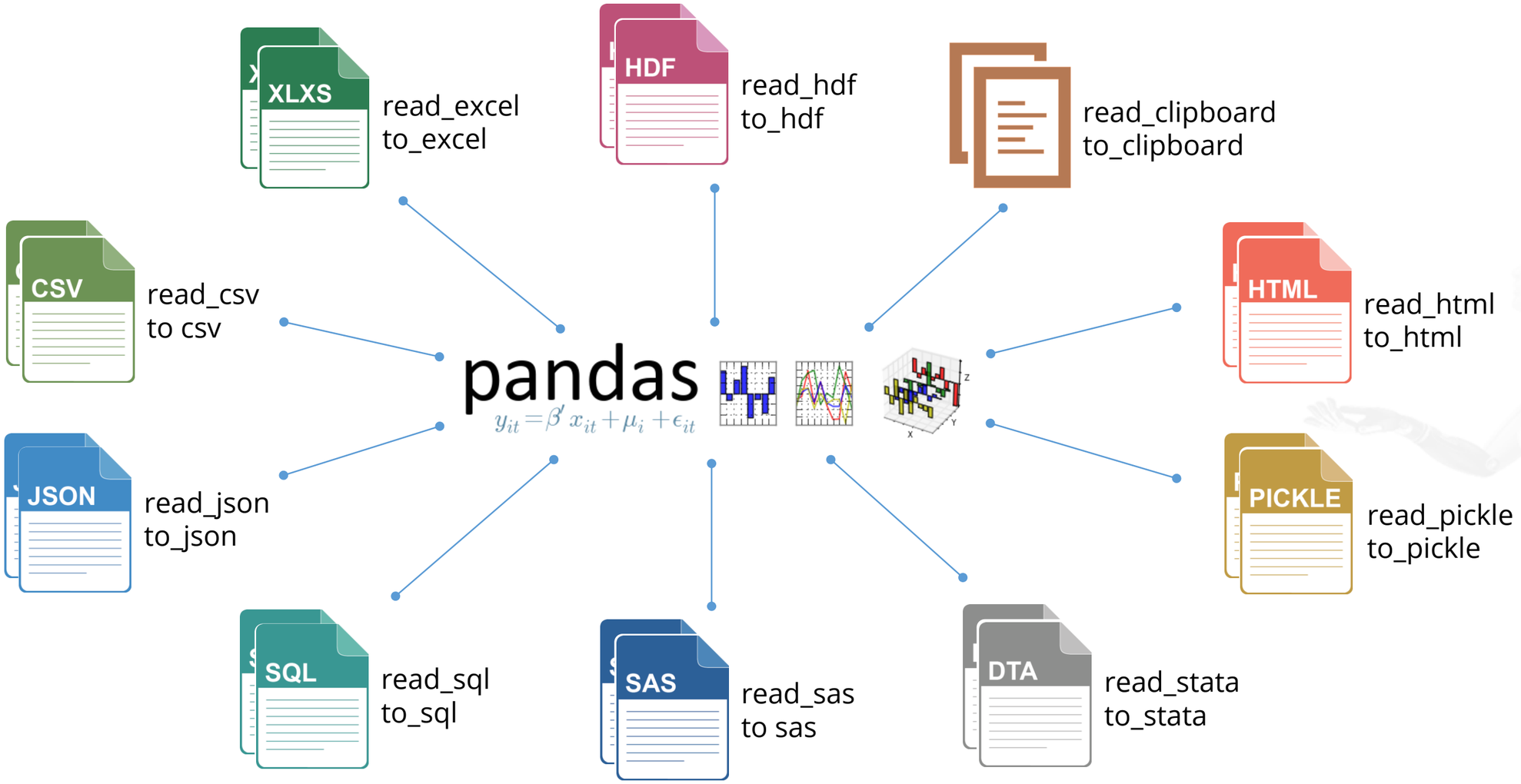
: : : :
: : : :
; ; ; ;

Output

DATA AND
ARTIFICIAL INTELLIGENCE

Data Standardization

File Read and Write Support



Pandas SQL Operation

In [1]:
```
#import pandas library
import pandas as pd
```

In [2]:
```
#import sqllite
import sqlite3
```

In [3]:
```
#Create SQL table
create_table = """
CREATE TABLE student_score
(Id INTEGER, Name VARCHAR(20), Math REAL,
Science REAL
);"""
```

In [4]:
```
#execute the SQL statement
executeSQL = sqlite3.connect(':memory:')
executeSQL.execute(create_table)
executeSQL.commit()
```

In [5]:
```
#prepare a SQL query
SQL_query = executeSQL.execute('select * from student_score')
```

In [7]:
```
#fetch result from the SQLlite database
resulset = SQL_query.fetchall()
```

In [8]:
```
#view result (empty data)
resulset
```

Out[8]: []

Pandas SQL Operation

In [9]:
```
#prepare records to be inserted into SQL table through SQL statement
insertSQL = [(10,'Jack',85,92),
                         (29,'Tom',73,89),
                          (65,'Ram',65.5,77),
                          (5,'Steve',55,91)
                         ]
```

In [10]:
```
#insert records into SQL table through SQL statement
insert_statement = "Insert into student_score values(?,?,?,?)"
executeSQL.executemany(insert_statement,insertSQL)
executeSQL.commit()
```

In [11]:
```
#prepare SQL query
SQL_query = executeSQL.execute("select * from student_score")
```

In [12]:
```
#fetch the resultset for the query
resulset = SQL_query.fetchall()
```

In [13]:
```
#view the resultset
resulset
```

Out[13]:
```
[(10, u'Jack', 85.0, 92.0),
 (29, u'Tom', 73.0, 89.0),
 (65, u'Ram', 65.5, 77.0),
 (5, u'Steve', 55.0, 91.0)]
```

Pandas SQL Operation

In [14]:
```
#put the records together in dataframe
df_student_recors = pd.DataFrame(resulset,columns=zip(*SQL_query.description)[0])
```

In [15]:
```
#view the records in pandas dataframe
df_student_recors
```

Out[15]:

	Id	Name	Math	Science
0	10	Jack	85.0	92.0
1	29	Tom	73.0	89.0
2	65	Ram	65.5	77.0
3	5	Steve	55.0	91.0

Pandas SQL Operations

Objective: Perform the following SQL operations with pandas:

1. Read the **Pandas-results.csv** dataset and create a dataframe
2. Write the SQL query to retrieve the first 10 items of the dataframe
3. Create a SQL table named movie rating with attributes like movie name, genre, and rating
4. Insert records into SQL table through SQL statement
5. Fetch and view the result
6. Create a dataframe of the result

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

ASSISTED PRACTICE

Pandas SQL Operations

Objective: Perform the following SQL operations:

1. Create a dataframe of customer details with columns like ID, customer, billing address, and shipping address
2. Write a SQL query to retrieve customers based on a particular ID

   Example: For ID=1233, output is customer A
3. Create a small dataframe and write the records in dataframe to a SQL database

Access: To execute the practice, follow these steps:

- Go to the **PRACTICE LABS** tab on your LMS
- Click the **START LAB** button
- Click the **LAUNCH LAB** button to start the lab

UNASSISTED PRACTICE

Unassisted Practice: Pandas SQL Operations

```
import pandas as pd, numpy as np
import pandasql as ps
```

Import pandasql to query pandas dataframes using SQL syntax

```
df = pd.DataFrame([[1234, 'Customer A', '123 Street', np.nan],
                   [1234, 'Customer A', np.nan, '333 Street'],
                   [1233, 'Customer B', '444 Street', '333 Street'],
                  [1233, 'Customer B', '444 Street', '666 Street']], columns=
['ID', 'Customer', 'Billing Address', 'Shipping Address'])
```

Create a dataframe of customer details

```
q1 = """SELECT Customer FROM df WHERE ID=1233 """
print(ps.sqldf(q1, locals()))
```

Select the customer based on the ID number

```
     Customer
0  Customer B
1  Customer B
```

Output

Unassisted Practice: Pandas SQL Operations

```
 from sqlalchemy import create_engine
engine = create_engine('sqlite://', echo=False)
df1=  pd.DataFrame({'name' : ['User 1', 'User 2', 'User 3']})
df1
```

Use sqlalchemy to create engine

Create an engine

Create a simple dataframe

	name
0	User 1
1	User 2
2	User 3

```
df.to_sql('users', con=engine)
```

Create a database

```
df.to_sql('users', con=engine)
>>> engine.execute("SELECT * FROM users").fetchall()
[(0, 'User 1'), (1, 'User 2'), (2, 'User 3')]
```

Write the records in the dataframe to an SQL database

```
[(0, 'User 1'), (1, 'User 2'), (2, 'User 3')]
```

Output

Key Takeaways

You are now able to:

- Explain Pandas and its features
- List different data structures of Pandas
- Outline the process to create series and DataFrame with data inputs
- Explain how to view, select, and access elements in a data structure
- Describe the procedure to handle vectorized operations
- Illustrate how to handle missing values
- Analyze data with different data operation methods

Knowledge Check

Knowledge Check

1

How is an index for data elements assigned while creating a Pandas series ? Select all that apply?

a. Created automatically

b. Needs to be assigned

c. Once created can not be changed or altered

d. Index is not applicable as series is one-dimensional

Knowledge Check

1

How is an index for data elements assigned while creating a Pandas series ? Select all that apply?

a. Created automatically

b. Needs to be assigned

c. Once created can not be changed or altered

d. Index is not applicable as series is one-dimensional

The correct answer is **a, b**

Data alignment is intrinsic in Pandas data structure and happens automatically. One can also assign index to data elements.

Knowledge Check

2

What will the result be in vector addition if label is not found in a series?

a. Marked as zeros for missing labels

b. Labels will be skipped

c. Marked as NaN for missing labels

d. Will prompt an exception, index not found

Knowledge Check

2

What will the result be in vector addition if label is not found in a series?

a. Marked as zeros for missing labels

b. Labels will be skipped

c. Marked as NaN for missing labels

d. Will prompt an exception, index not found

The correct answer is **c**

The result will be marked as NaN (Not a Number) for missing labels.

Knowledge Check

3

What is the result of DataFrame[3:9]?

a. Series with sliced index from 3 to 9

b. dict of index positions 3 and 9

c. DataFrame of sliced rows index from 3 to 9

d. DataFrame with data elements at index 3 to 9

Knowledge Check

3

What is the result of DataFrame[3:9]?

a. Series with sliced index from 3 to 9

b. dict of index positions 3 and 9

c. DataFrame of sliced rows index from 3 to 9

d. DataFrame with data elements at index 3 to 9

The correct answer is **C**

This is DataFrame slicing technique with indexing or selection on data elements. When a user passes the range 3:9, the entire range from 3 to 9 gets sliced and displayed as output.

Knowledge Check

4

What does the fillna() method do?

a. Fills all NaN values with zeros

b. Fills all NaN values with one

c. Fills all NaN values with values mentioned in the parenthesis

d. Drops NaN values from the dataset

Knowledge Check

4

What does the fillna() method do?

a. Fills all NaN values with zeros

b. Fills all NaN values with one

c. Fills all NaN values with values mentioned in the parenthesis

d. Drops NaN values from the dataset

The correct answer is **c**

fillna is one of the basic methods to fill NaN values in a dataset with a desired value by passing that in parenthesis.

Knowledge Check

5

Which of the following data structures is used to store three-dimensional data?

a. Series

b. DataFrame

c. Panel

d. PanelND

Knowledge Check

5

Which of the following data structures is used to store three-dimensional data?

a. Series

b. DataFrame

c. Panel

d. PanelND

The correct answer is **C**

Panel is a data structure used to store three-dimensional data.

Analyze the Federal Aviation Authority (FAA) Dataset using Pandas

Problem Statement:

Analyze the Federal Aviation Authority (FAA) dataset using Pandas to do the following:

1. View
    a. Aircraft manufacturer name
    b. State name
    c. Aircraft model name
    d. Text information
    e. Flight phase
    f. Event description type
    g. Fatal flag
2. Clean the dataset and replace the fatal flag NaN with “No”
3. Find the aircraft types and their occurrences in the dataset
4. Remove all the observations where aircraft names are not available
5. Display the observations where fatal flag is “Yes”

LESSON-END PROJECT

Analyze the Federal Aviation Authority (FAA) Dataset using Pandas

Instructions to perform:

Download the FAA dataset from the “Resource” tab. Upload the dataset to your Jupyter notebook to view and evaluate it.

LESSON-END PROJECT

Analyzing the Dataset

Problem Statement:

A dataset in CSV format is given for the Fire Department of the New York City.

Analyze the dataset to determine:

1. The total number of fire department facilities in the New York city
2. The number of fire department facilities in each borough
3. The facility names in Manhattan

Instructions to perform:

Download the FDNY dataset from the “Resource” tab. You can upload the dataset to your Jupyter notebook to use it.

LESSON-END PROJECT

DATA AND
ARTIFICIAL INTELLIGENCE

Thank You