**DATE**: 20 june 2024

DAY: Thursday

TOPICS: Library: pandas

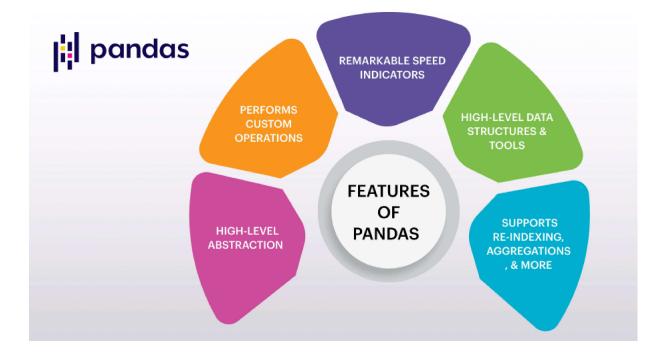


Pandas is a powerful and widely-used open-source library in Python for data manipulation and analysis. It provides data structures and functions needed to work with structured data seamlessly and efficiently. Here are some key features and components of the Pandas library:

#### Key Features of Pandas

#### 1. Data Structures:

- Series: A one-dimensional array-like object containing a sequence of values. It is similar to a list or a one-dimensional NumPy array but with labeled axes (index).
- **DataFrame**: A two-dimensional table of data with labeled axes (rows and columns). It can be thought of as a collection of Series objects sharing the same index.
- 2. **Data Alignment**: Automatic alignment of data for arithmetic operations on Series and DataFrame objects, making it easy to manage missing data.
- 3. Data Cleaning: Functions for handling missing data, such as filling, replacing, and dropping missing values.
- 4. **Data Transformation**: Tools for reshaping, pivoting, and transforming data, including functions for merging, joining, and concatenating DataFrame objects.
- 5. **Data Aggregation and Grouping**: Powerful group-by functionality to split data into groups, apply functions to each group, and combine the results.
- 6. Time Series Handling: Specialized tools for working with time series data, including date range generation and frequency conversion.
- 7. Input and Output: Functions for reading and writing data to and from various file formats, including CSV, Excel, SQL databases, and more.
- 8. **Visualization**: Integration with plotting libraries like Matplotlib to provide basic data visualization capabilities directly from DataFrame and Series objects.



import numpy as np
import pandas as pd

### Series

```
11 = [1, 2, 3, 4, 5, 6]
labels = ['a', 'b', 'c', 'd', 'e', 'f']
d1 = {"A":10, "B":20, "C":30, "D":40, "E":50}
s1 = pd.Series(l1)
s1
 → 0
         1
           2
      2
           3
      3
           4
      dtype: int64
s1[4]
→ 5
s2 = pd.Series(labels)
s2
 → 0
           b
      2
           С
      3
           d
           е
      dtype: object
s2[4]
<u>→</u> 'e'
s3 = pd.Series(data=11, index=labels)
s3
```

```
3
     С
     d
           4
           5
           6
     dtype: int64
s3['a']
<u>→</u> 1
s3[0]
<u>→</u> 1
pd.Series(d1)
<u>→</u> A
           10
           20
           30
     D
           40
           50
     dtype: int64
```

### DataFrame

```
arr = np.random.randint(low=1, high=100, size=(5, 6))
array([[55, 53, 65, 2, 56, 33], [50, 26, 87, 5, 79, 44], [7, 73, 51, 32, 6, 19],
           [21, 28, 26, 5, 43, 64],
[22, 58, 71, 18, 67, 47]])
type(arr)
→ numpy.ndarray
pd.DataFrame(arr)
         0 1 2 3 4 5
     0 55 53 65
                    2 56 33
     1 50 26 87
                   5 79 44
        7 73 51 32 6 19
     3 21 28 26 5 43 64
     4 22 58 71 18 67 47
df = pd.DataFrame(arr, index=["A", "B", "C", "D", "E"], columns=["U", "V", "W", "X", "Y", "Z"])
df
₹
         A 55 53 65 2 56 33
     B 50 26 87 5 79 44
     C 7 73 51 32 6 19
     D 21 28 26 5 43 64
     E 22 58 71 18 67 47
type(df)
pandas.core.frame.DataFrame
```

### Grabbing Columns

### Grabbing Rows

**E** 18 47 58

```
df.loc["C"]
→ U
        73
       19
    Name: C, dtype: int64
df.loc[["A", "B", "E"]]
₹
        U V W X Y Z
    A 55 53 65 2 56 33
    B 50 26 87 5 79 44
    E 22 58 71 18 67 47
df.iloc[2]
        51
       32
        6
    Z 19
    Name: C, dtype: int64
```

## → Adding a New Column

dҒ



- **A** 55 53 65 2 56 33
- **B** 50 26 87 5 79 44
- **C** 7 73 51 32 6 19
- **D** 21 28 26 5 43 64
- E 22 58 71 18 67 47
- df['New'] = [10, 20, 30, 40, 50]

df



- A 55 53 65 2 56 33 10
- **B** 50 26 87 5 79 44 20
- **C** 7 73 51 32 6 19 30
- **D** 21 28 26 5 43 64 40
- **E** 22 58 71 18 67 47 5
- df['New'] = [100, 200, 300, 400, 500]

df



- A 55 53 65 2 56 33 100
- **B** 50 26 87 5 79 44 200
- **C** 7 73 51 32 6 19 300
- **D** 21 28 26 5 43 64 400
- E 22 58 71 18 67 47 500
- Deleting a Column

df



- A 55 53 65 2 56 33 100
  - **B** 50 26 87 5 79 44 200
  - **C** 7 73 51 32 6 19 300
  - **D** 21 28 26 5 43 64 400
  - E 22 58 71 18 67 47 500

df.drop('New', axis=1)



- U V W X Y Z
  A 55 53 65 2 56 33
- **B** 50 26 87 5 79 44
- **C** 7 73 51 32 6 19
- **D** 21 28 26 5 43 64
- E 22 58 71 18 67 47

df

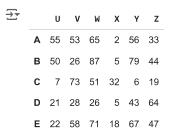
df.drop('New', axis=1, inplace=True)

df

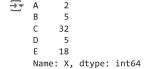


### Conditional Selection

df



df['X']



df['X'] % 2 == 0

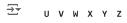
df[df['X'] % 2 == 0]

df[df['X'] % 2 == 0]['Y']

A False
B False
C False
D False

E False Name: X, dtype: bool

$$df[(df['X'] \% 2 == 0) \& (df['X'] > 50)]$$



# Setting an Index

df

<b>→</b>		U	V	W	Х	Υ	z
	Α	55	53	65	2	56	33
	В	50	26	87	5	79	44
	С	7	73	51	32	6	19
	D	21	28	26	5	43	64
	Е	22	58	71	18	67	47

df.reset\_index()

₹		index	U	٧	W	Х	Υ	Z
	0	Α	55	53	65	2	56	33
	1	В	50	26	87	5	79	44
	2	С	7	73	51	32	6	19
	3	D	21	28	26	5	43	64
	4	Е	22	58	71	18	67	47

df.reset\_index(inplace=True)

df

₹		index	U	v	W	х	Υ	z
	0	Α	55	53	65	2	56	33
	1	В	50	26	87	5	79	44
	2	С	7	73	51	32	6	19
	3	D	21	28	26	5	43	64
	4	Е	22	58	71	18	67	47

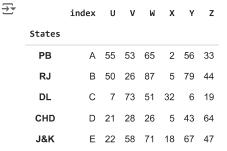
df['States'] = "PB RJ DL CHD J&K".split()

"PB RJ DL CHD J&K".split()

df

₹		index	U	V	W	Х	Υ	Z	States
	0	А	55	53	65	2	56	33	РВ
	1	В	50	26	87	5	79	44	RJ
	2	С	7	73	51	32	6	19	DL
	3	D	21	28	26	5	43	64	CHD
	4	Е	22	58	71	18	67	47	J&K

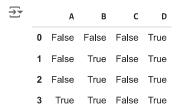
df.set\_index('States')



### Missing Values

```
d = {"A":[1, 2, 3, np.nan],
   "B":[5, np.nan, np.nan, np.nan],
   "C":[10, 20, 30, 40],
   "D":[np.nan, np.nan, np.nan, np.nan]}
df = pd.DataFrame(d)
df
→
                 С
                        D
          Α
               В
     0 1.0
             5.0 10 NaN
        2.0 NaN 20 NaN
        3.0 NaN 30 NaN
     3 NaN NaN 40 NaN
```

#### df.isnull()



### df.isnull().sum()



#### df.dropna(axis=1)



df.dropna(axis=1, thresh=2)

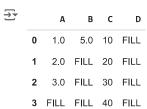
```
A C
0 1.0 10
1 2.0 20
2 3.0 30
```

3 NaN 40

df

<del></del>		Α	В	С	D
	0	1.0	5.0	10	NaN
	1	2.0	NaN	20	NaN
	2	3.0	NaN	30	NaN
	3	NaN	NaN	40	NaN

#### df.fillna("FILL")



#### df.fillna(df.mean())

₹		Α	В	c	D
	0	1.0	5.0	10	NaN
	1	2.0	5.0	20	NaN
	2	3.0	5.0	30	NaN
	3	2.0	5.0	40	NaN

#### df.fillna(0)

<b>→</b> *		Α	В	C	D
	0	1.0	5.0	10	0.0
	1	2.0	0.0	20	0.0
	2	3.0	0.0	30	0.0
	3	0.0	0.0	40	0.0

### Grouping

_				
<del></del>		Company	Employee	Sales
	0	FB	Sam	1000
	1	GOOGLE	Rachel	500
	2	MICROSOFT	Maddy	550
	3	FB	Joe	2000
	4	GOOGLE	Srishti	890
	5	FB	Shivay	500
	6	MICROSOFT	Pushpa	350
	7	FB	Kirti	350

df.min()

Company FB
Employee Joe
Sales 350
dtype: object

df.max()

Company MICROSOFT
Employee Srishti
Sales 2000
dtype: object

grouped\_df= df.groupby('Company')
grouped\_df

grouped\_df.min()

<del>_</del> ₹		Employee	Sales
	Company		
	FB	Joe	350
	GOOGLE	Rachel	500
	MICROSOFT	Maddy	350

grouped\_df.max()

Employee Sales

Company

FB Shivay 2000

GOOGLE Srishti 890

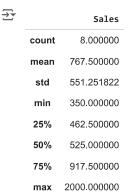
Pushpa

grouped\_df.describe()

**MICROSOFT** 

₹ Sales 25% 50% 75% count mean std min max Company FΒ 4.0 962.5 745.402576 350.0 462.5 750.0 1250.0 2000.0 GOOGLE 2.0 695.0 275.771645 500.0 597.5 695.0 792.5 890.0 MICROSOFT 2.0 450.0 141.421356 350.0 400.0 450.0 500.0 550.0

df.describe()



# v pandas Exercise :

#### Import pandas

import pandas as pd

#### Read the data from Salaries.csv and store it in a dataframe

df = pd.read\_csv("/content/Salaries.csv")

df

	Id	EmployeeName	JobTitle	BasePay	OvertimePay	OtherPay	Benefits	TotalPay	TotalPayBenefits	Year	Notes	
0	1	NATHANIEL FORD	GENERAL MANAGER- METROPOLITAN TRANSIT AUTHORITY	167411.18	0.00	400184.25	NaN	567595.43	567595.43	2011	NaN	F
1	2	GARY JIMENEZ	CAPTAIN III (POLICE DEPARTMENT)	155966.02	245131.88	137811.38	NaN	538909.28	538909.28	2011	NaN	F
2	3	ALBERT PARDINI	CAPTAIN III (POLICE DEPARTMENT)	212739.13	106088.18	16452.60	NaN	335279.91	335279.91	2011	NaN	F
3	4	CHRISTOPHER CHONG	WIRE ROPE CABLE MAINTENANCE MECHANIC	77916.00	56120.71	198306.90	NaN	332343.61	332343.61	2011	NaN	F
4	5	PATRICK GARDNER	DEPUTY CHIEF OF DEPARTMENT, (FIRE DEPARTMENT)	134401.60	9737.00	182234.59	NaN	326373.19	326373.19	2011	NaN	F
148649	148650	Roy I Tillery	Custodian	0.00	0.00	0.00	0.0	0.00	0.00	2014	NaN	F
148650	148651	Not provided	Not provided	NaN	NaN	NaN	NaN	0.00	0.00	2014	NaN	F
4												

#### Check if the dataframe is properly read or not using the head function

df.head()

₹		Id	EmployeeName	JobTitle	BasePay	OvertimePay	OtherPay	Benefits	Total
	0	1	NATHANIEL FORD	GENERAL MANAGER- METROPOLITAN TRANSIT AUTHORITY	167411.18	0.00	400184.25	NaN	56759
	1	2	GARY JIMENEZ	CAPTAIN III (POLICE DEPARTMENT)	155966.02	245131.88	137811.38	NaN	538909
	2	3	ALBERT PARDINI	CAPTAIN III (POLICE DEPARTMENT)	212739.13	106088.18	16452.60	NaN	335279
	4								•

#### What columns exist in this dataframe?

```
df.columns
```

#### How many rows does this dataframe have?

```
df.index
```

```
RangeIndex(start=0, stop=148654, step=1)
```

#### Display the information about the dataframe using the info function. Which of these columns have missing values in them?

```
df.info()
```

```
<<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 148654 entries, 0 to 148653
    Data columns (total 13 columns):
    # Column
                       Non-Null Count
                                        Dtype
    --- -----
    0 Id
                        148654 non-null int64
       EmployeeName 148654 non-null object
        JobTitle
                         148654 non-null object
     3
        BasePay
                         148045 non-null float64
     4 OvertimePay
                       148650 non-null float64
        OtherPay
                        148650 non-null float64
     6
        Benefits
                         112491 non-null float64
        TotalPay
                        148654 non-null float64
        TotalPayBenefits 148654 non-null float64
     8
        Year
                         148654 non-null int64
    10 Notes
                         0 non-null
                                         float64
                         148654 non-null object
     11 Agency
    12 Status
                         0 non-null
                                         float64
    dtypes: float64(8), int64(2), object(3)
    memory usage: 14.7+ MB
```

#### What is the total BasePay?

```
df["BasePay"].mean()
```

<del>→</del>▼ 66325.4488404877

#### What is the highest amount of overtime pay?

```
df["OvertimePay"].max()
```

→ 245131.88

What is the job title of JOSEPH DRISCOLL? Note: Use all caps, otherwise you may get an answer that doesn't match up (there is also a lowercase Joseph Driscoll).

How much does JOSEPH DRISCOLL make (including benefits)?

```
df[df["EmployeeName"] == "JOSEPH DRISCOLL"]["TotalPayBenefits"]

24    270324.91
    Name: TotalPayBenefits, dtype: float64
```

What is the name of highest paid person (including benefits)?

```
df[df['TotalPayBenefits']== df['TotalPayBenefits'].max()]

Id EmployeeName JobTitle BasePay OvertimePay OtherPay Benefits TotalP

GENERAL
MANAGER-
```

What was the average (mean) BasePay of all employees per year? (2011-2014)?

```
df.groupby('Year')['BasePay'].mean()

→ Year
2011 63595.956517
2012 65436.406857
2013 69630.030216
2014 66564.421924
Name: BasePay, dtype: float64
```

How many unique job titles exist in the dataframe?

```
df["JobTitle"].nunique()

→ 2159
```

What is the name of lowest paid person (including benefits)? Do you notice something strange about how much he or she is paid?