

# pandas

January 28, 2026

```
[1]: import pandas as pd
import numpy as np

[2]: # creating series
# creating series from list
s1 = pd.Series([10, 20, 30, 40])
print(s1)

# creating series with custom index
s2 = pd.Series([100, 200, 300], index=["a", "b", "c"])
print(s2)

# creating series from numpy array
arr = np.array([1, 2, 3])
s3 = pd.Series(arr)
print(s3)

# creating series from dictionary
data = {"Math": 85, "Science": 90, "English": 88}
s4 = pd.Series(data)
print(s4)
```

```
0    10
1    20
2    30
3    40
dtype: int64
a    100
b    200
c    300
dtype: int64
0    1
1    2
2    3
dtype: int32
Math      85
Science   90
English   88
```

```
dtype: int64
```

```
[3]: # series attributes
s = pd.Series([10, 20, 30])

print(s.index)
print(s.values)
print(s.dtype)
print(s.size)
print(s.ndim)
```

```
RangeIndex(start=0, stop=3, step=1)
[10 20 30]
int64
3
1
```

```
[ ]: # Accessing series elements
s = pd.Series([100, 200, 300], index=["x", "y", "z"])

print(s[0])    # by position
print(s["y"])  # by label
print(s[:2])   # slicing
print(s[["x", "z"]])
print(s[s > 150])  # conditional filtering
```

```
# updating values
s = pd.Series([10, 20, 30])

s[1] = 200
print(s)

s[s > 20] = 999
print(s)
```

```
[6]: # Arithmetic operations
s1 = pd.Series([10, 20, 30], index=["a", "b", "c"])
s2 = pd.Series([1, 2, 3], index=["a", "b", "d"])

print(s1 + s2)  # alignment by index

# Aggregation functions
s = pd.Series([10, 20, 30, 40])

print(s.sum())
print(s.mean())
print(s.median())
```

```
print(s.min())
print(s.max())
print(s.std())
print(s.var())
print(s.count())
```

```
a    11.0
b    22.0
c    NaN
d    NaN
dtype: float64
100
25.0
25.0
10
40
12.909944487358056
166.66666666666666
4
```

```
[5]: # value counts
s = pd.Series(["apple", "banana", "apple", "orange", "banana"])

print(s.value_counts())

# unique & nunique
s = pd.Series([1, 2, 2, 3, 3, 3])

print(s.unique())
print(s.nunique())
```

```
apple    2
banana   2
orange   1
Name: count, dtype: int64
[1 2 3]
3
```

```
[7]: # Handling missing data
s = pd.Series([1, 2, np.nan, 4])

print(s.isna())
print(s.notna())

print(s.fillna(0))
print(s.dropna())
```

```
# combining series
s1 = pd.Series([1, 2, 3])
s2 = pd.Series([4, 5, 6])

print(pd.concat([s1, s2]))
```

```
0    False
1    False
2     True
3    False
dtype: bool
0    True
1    True
2   False
3   True
dtype: bool
0    1.0
1    2.0
2    0.0
3    4.0
dtype: float64
0    1.0
1    2.0
3    4.0
dtype: float64
0    1
1    2
2    3
0    4
1    5
2    6
dtype: int64
```

### 0.0.1 DATAFRAME

```
[8]: # creating dataframe from dictionary
data = {"Name": ["A", "B", "C"], "Age": [20, 21, 22], "Marks": [80, 85, 90]}

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

	Name	Age	Marks
0	A	20	80
1	B	21	85
2	C	22	90

```
[9]: # dataframe from a list of dictionaries
# dataframe from list of dicts
```

```
data = [{"Name": "A", "Age": 20}, {"Name": "B", "Age": 21}, {"Name": "C", "Age":  
    ↪ 22}]
```

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

```
   Name  Age  
0     A    20  
1     B    21  
2     C    22
```

```
[10]: # Dataframe attributes  
print(df.shape) # rows, columns  
print(df.columns)  
print(df.index)  
print(df.dtypes)  
print(df.ndim)  
print(df.size)  
  
# head & tail  
print(df.head())  
print(df.tail(2))
```

```
(3, 2)  
Index(['Name', 'Age'], dtype='object')  
RangeIndex(start=0, stop=3, step=1)  
Name      object  
Age       int64  
dtype: object  
2  
6  
   Name  Age  
0     A    20  
1     B    21  
2     C    22  
   Name  Age  
1     B    21  
2     C    22
```

```
[13]: # Selecting columns  
# selecting single column  
print(df["Name"])  
  
# selecting multiple columns  
print(df[["Name", "Age"]])  
  
# selecting rows(loc & iloc)  
# loc -> label based
```

```

print(df.loc[0])
print(df.loc[0, "Name"])

# iloc -> index based
print(df.iloc[1])
print(df.iloc[0,1])

```

```

0    A
1    B
2    C
Name: Name, dtype: object
   Name  Age
0     A   20
1     B   21
2     C   22
Name      A
Age      20
Name: 0, dtype: object
A
Name      B
Age      21
Name: 1, dtype: object
20

```

```

[ ]: # slicing rows
print(df[0:2])
print(df.iloc[:2])

# adding new column
df["Passed"] = df["Marks"] >= 85
print(df)

# update values
df.loc[0, "Marks"] = 95
df["Age"] = df["Age"] + 1
print(df)

# delete column
df.drop(columns=["Passed"], inplace=True)

# delete row
df.drop(index=2, inplace=True)
print(df)

# filtering rows
print(df[df["Marks"] > 80])

```

```
[ ]: # sorting & searching
# sorting
print(df.sort_values("Marks"))
print(df.sort_values("Marks", ascending=False))

print(df.sort_index())

# searching
print(df["Marks"].idxmax())
print(df["Marks"].idxmin())

# aggregation functions
print(df["Marks"].sum())
print(df["Marks"].mean())
print(df["Marks"].max())
print(df["Marks"].min())
print(df["Marks"].count())
print(df["Marks"].std())
print(df["Marks"].var())
```

```
[17]: # describe() --> statistical summary
print(df.describe())
```

```
Age
count    3.0
mean     21.0
std      1.0
min      20.0
25%     20.5
50%     21.0
75%     21.5
max     22.0
```

```
[18]: # value counts
print(df["Age"].value_counts())
```

```
Age
20     1
21     1
22     1
Name: count, dtype: int64
```

```
[19]: # handling missing data
df2 = pd.DataFrame({"A": [1, 2, None], "B": [4, None, 6]})

print(df2.isna())
print(df2.fillna(0))
print(df2.dropna())
```

```

          A      B
0  False  False
1  False   True
2   True  False
          A      B
0    1.0   4.0
1    2.0   0.0
2    0.0   6.0
          A      B
0    1.0   4.0

```

```
[ ]: # apply function
df["Grade"] = df["Marks"].apply(lambda x: "A" if x >= 90 else "B")
print(df)
```

```
[21]: # rename columns
df.rename(columns={"Marks": "Score"}, inplace=True)

# rename index
df.rename(index={0: "Row1", 1: "Row2"}, inplace=True)
print(df)

# String operations
df["Name"] = df["Name"].str.lower()
print(df)
```

	Name	Age
Row1	A	20
Row2	B	21
2	C	22

  

	Name	Age
Row1	a	20
Row2	b	21
2	c	22

```
[22]: # Group by (IMP)
data = {"Dept": ["IT", "IT", "HR", "HR"], "Salary": [50000, 60000, 40000, ↴45000]}
df = pd.DataFrame(data)

print(df.groupby("Dept").mean())
print(df.groupby("Dept").sum())
```

Dept	Salary
HR	42500.0
IT	55000.0

```
Dept
HR      85000
IT      110000
```

```
[23]: # Merge/Join
df1 = pd.DataFrame({"id": [1, 2, 3], "Name": ["A", "B", "C"]})

df2 = pd.DataFrame({"id": [1, 2, 4], "Salary": [50000, 60000, 70000]})

print(pd.merge(df1, df2, on="id", how="inner"))
print(pd.merge(df1, df2, on="id", how="left"))
```

```
   id Name  Salary
0   1     A    50000
1   2     B    60000
   id Name  Salary
0   1     A    50000.0
1   2     B    60000.0
2   3     C      NaN
```

```
[24]: # concatenation
df1 = pd.DataFrame({"A": [1, 2]})
df2 = pd.DataFrame({"A": [3, 4]})

print(pd.concat([df1, df2]))
```

```
   A
0   1
1   2
0   3
1   4
```

```
[25]: # reading & writing files

# reading csv
# df = pd.read_csv("data.csv")

# writing csv
df.to_csv("output.csv", index=False)
```

```
[26]: # example
# student dataframe
students = pd.DataFrame({"Name": ["A", "B", "C", "D"], "Marks": [78, 85, 62, 90]})

print(students[students["Marks"] >= 75])
print("Topper:", students.loc[students["Marks"].idxmax()])
```

```
Name  Marks
```

```
0      A      78
1      B      85
3      D      90
Topper: Name      D
Marks      90
Name: 3, dtype: object
```