

Getting familiarity with pandas

Series

```
import pandas as pd

ser=pd.Series([3, 7, 14, 18, 21, 14])
ser

0      3
1      7
2     14
3     18
4     21
5     14
dtype: int64

ser=pd.Series([3, 7, 14, 18, 21, 14], index=['a', 'b', 'c', 'd', 'e', 'f'])
ser

a      3
b      7
c     14
d     18
e     21
f     14
dtype: int64

ser.index

Index(['a', 'b', 'c', 'd', 'e', 'f'], dtype='object')

ser.values

array([ 3,  7, 14, 18, 21, 14], dtype=int64)

ser.dtype

dtype('int64')

ser.head(1)

a      3
dtype: int64

ser.tail(1)

f     14
dtype: int64
```

Data Frame

```
data={
    'Name':['Jimmy', 'Kim', 'Gus', 'Mike', 'Victor'],
    'Age':[40, 35, 55, 58, 35],
    'Rank':['A', 'C', 'B', 'C', 'D']
}
```

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

	Name	Age	Rank
0	Jimmy	40	A
1	Kim	35	C
2	Gus	55	B
3	Mike	58	C
4	Victor	35	D

```
df.index
```

```
RangeIndex(start=0, stop=5, step=1)
```

```
df.columns
```

```
Index(['Name', 'Age', 'Rank'], dtype='object')
```

```
df.values
```

```
array([[ 'Jimmy', 40, 'A'],
       [ 'Kim', 35, 'C'],
       [ 'Gus', 55, 'B'],
       [ 'Mike', 58, 'C'],
       [ 'Victor', 35, 'D']], dtype=object)
```

```
df.dtypes
```

```
Name      object
Age        int64
Rank      object
dtype: object
```

```
df.head(1)
```

	Name	Age	Rank
0	Jimmy	40	A

```
df.tail(1)
```

	Name	Age	Rank
4	Victor	35	D

```
df.shape
```

```
(5, 3)
df.size
15
```

Creating Series and DF's from different forms

```
arr=[13, 67, 18, 30, 10]
#This is a normal array

ser=pd.Series(arr)
ser
```

0	13
1	67
2	18
3	30
4	10

```
dtype: int64

data={
    'Name':['Jimmy', 'Kim', 'Gus', 'Mike', 'Victor'],
    'Age':[40, 35, 55, 58, 35],
    'Rank':['A', 'C', 'B', 'C', 'D']
}
df=pd.DataFrame(data)
df
```

	Name	Age	Rank
0	Jimmy	40	A
1	Kim	35	C
2	Gus	55	B
3	Mike	58	C
4	Victor	35	D

```
ser1=pd.Series(df['Name'])
ser1
```

0	Jimmy
1	Kim
2	Gus
3	Mike
4	Victor

```
Name: Name, dtype: object

df1=pd.read_csv('students.csv')
df1
```

	Students	Grades
0	Jimmy	A

```

1  Chandler    C
2   Kerion    B

ser1=pd.Series(df1['Grades'])
ser1

0    A
1    C
2    B
Name: Grades, dtype: object

```

Common operations on data frames

```

df['Age']

0    40
1    35
2    55
3    58
4    35
Name: Age, dtype: int64

```

```
df[['Name', 'Age']]
```

	Name	Age
0	Jimmy	40
1	Kim	35
2	Gus	55
3	Mike	58
4	Victor	35

```
df.loc[0]
```

Name	Jimmy
Age	40
Rank	A

Name: 0, dtype: object

```
df.loc[0:3]
```

	Name	Age	Rank
0	Jimmy	40	A
1	Kim	35	C
2	Gus	55	B
3	Mike	58	C

#This can also be done using iloc

#iloc is used to locate the integer value of index if we dont the what the index is

```
df.iloc[2]
```

```
Name    Gus
Age      55
Rank     B
Name: 2, dtype: object
```

#multiple rows can be accessed using iloc also
`df.iloc[0:3]`

```
   Name  Age Rank
0  Jimmy  40   A
1   Kim   35   C
2   Gus   55   B
```

```
df[df['Age']>50]
```

```
   Name  Age Rank
2   Gus   55   B
3  Mike   58   C
```

```
df[df['Rank']=='C']
```

```
   Name  Age Rank
1   Kim   35   C
3  Mike   58   C
```

```
df[(df['Age']>52) | (df['Rank']=='B')]
```

```
   Name  Age Rank
2   Gus   55   B
3  Mike   58   C
```

#we can add a new column to our data frame

```
df['DOB']=2024-df['Age']
df
```

```
   Name  Age Rank  DOB
0  Jimmy  40   A  1984
1   Kim   35   C  1989
2   Gus   55   B  1969
3  Mike   58   C  1966
4  Victor  35   D  1989
```

```
df['Name']=df['Name'].str.upper()
df
```

```
   Name  Age Rank  DOB
0  JIMMY  40   A  1984
1   KIM   35   C  1989
2   GUS   55   B  1969
3  MIKE   58   C  1966
4  VICTOR  35   D  1989
```

```
df=df.drop(columns=['DOB'])
df
```

	Name	Age	Rank
0	JIMMY	40	A
1	KIM	35	C
2	GUS	55	B
3	MIKE	58	C
4	VICTOR	35	D

```
#To modify the excisting data
df.loc[2, 'Rank']='A'
df
```

	Name	Age	Rank
0	JIMMY	40	A
1	KIM	35	C
2	GUS	55	A
3	MIKE	58	C
4	VICTOR	35	D

Data Handling

```
import pandas as pd
import numpy as np
```

```
# Sample DataFrame with missing values
```

```
data = {
    'Name': ['Ram', 'Bheem', 'Laxman', np.nan],
    'Age': [25, np.nan, 35, 40],
    'City': ['New York', 'Los Angeles', np.nan, 'Chicago']
}
df = pd.DataFrame(data)
```

```
# Check for missing values
```

```
missing_values = df.isna()
```

```
print(missing_values)
```

	Name	Age	City
0	False	False	False
1	False	True	False
2	False	False	True
3	True	False	False

```
count_missing_val=df.isna().sum()
count_missing_val
```

Name	1
Age	1

```

City      1
dtype: int64

# Drop rows with any missing values
df_dropped_rows = df.dropna()

# Drop columns with any missing values
df_dropped_cols = df.dropna(axis=1)

print(df_dropped_rows)
print(df_dropped_cols)

   Name  Age  City
0  Ram  25.0 New York
Empty DataFrame
Columns: []
Index: [0, 1, 2, 3]

df_filled_value = df.fillna({'Name': 'Ravan', 'Age': df['Age'].mean(),
                             'City': 'Colomba'})

df_filled_value

   Name  Age  City
0  Ram  25.000000  New York
1  Bheem  33.333333  Los Angeles
2  Laxman  35.000000  Colomba
3  Ravan  40.000000  Chicago

```

Data transformation

```

# Sample DataFrame
df = pd.DataFrame({
    'Value': ['1', '2', '3', '4']
})

# Convert 'Value' to integer
df['Value'] = df['Value'].astype(int)
df

   Value
0      1
1      2
2      3
3      4

# Sample DataFrames
df1 = pd.DataFrame({'A': [1, 2], 'B': [3, 4]})
df2 = pd.DataFrame({'A': [5, 6], 'B': [7, 8]})

# Concatenate DataFrames vertically

```

```
df_concat = pd.concat([df1, df2])
df_concat
```

	A	B
0	1	3
1	2	4
0	5	7
1	6	8

```
# Sample DataFrames
```

```
df1 = pd.DataFrame({'ID': [1, 2, 3], 'Value': ['A', 'B', 'C']})
```

```
df2 = pd.DataFrame({'ID': [1, 2, 4], 'Description': ['X', 'Y', 'Z']})
```

```
# Merge DataFrames on 'ID'
```

```
df_merged = pd.merge(df1, df2, on='ID', how='inner')
```

```
df_merged
```

	ID	Value	Description
0	1	A	X
1	2	B	Y

Conclusion

Applications