

▼ making a series

```
a=pd.Series([1,2,3,4,5], index=["A","B","C","D","E"])
a
```

```
A    1
B    2
C    3
D    4
E    5
dtype: int64
```

▼ Making a DataFrame

```
b=pd.DataFrame({"Sher ali":21,"Ahmad":19,"Usman":13}, index=["A","B","C"])
b
```

	Sher ali	Ahmad	Usman
A	21	19	13
B	21	19	13
C	21	19	13

▼ Working on Datasheet from Seaborn Library

```
import pandas as pd
```

Checking first five 5 Entries

▼ Checking information about data

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):
#   Column      Non-Null Count  Dtype
---  -
0   survived    891 non-null    int64
1   pclass      891 non-null    int64
2   sex         891 non-null    object
3   age         714 non-null    float64
4   sibsp       891 non-null    int64
5   parch       891 non-null    int64
6   fare        891 non-null    float64
7   embarked    889 non-null    object
8   class       891 non-null    category
9   who         891 non-null    object
10  adult_male  891 non-null    bool
11  deck        203 non-null    category
12  embark_town 889 non-null    object
13  alive       891 non-null    object
14  alone       891 non-null    bool
dtypes: bool(2), category(2), float64(2), int64(4), object(5)
memory usage: 80.7+ KB
```

```
df.head()
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town
0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	NaN	Southampton
1	1	1	female	38.0	1	0	71.2833	C	First	woman	False	C	Cherbourg
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton
3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	C	Southampton

▼ Checking last 5 Entries

```
df.tail()
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town
886	0	2	male	27.0	0	0	13.00	S	Second	man	True	NaN	Southampton
887	1	1	female	19.0	0	0	30.00	S	First	woman	False	B	Southampton
888	0	3	female	NaN	1	2	23.45	S	Third	woman	False	NaN	Southampton
889	1	1	male	26.0	0	0	30.00	C	First	man	True	C	Cherbourg
890	0	3	male	32.0	0	0	7.75	Q	Third	man	True	NaN	Quebec

▼ Summary Statistics

```
df.describe()
```

	survived	pclass	age	sibsp	parch	fare
count	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
std	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
50%	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

▼ Checking No.of Rows and Columns

```
df.shape[1]
```

15

```
rows="The number of "
```

Double-click (or enter) to edit

Double-click (or enter) to edit

Checking Column name

```
df.columns
```

```
Index(['survived', 'pclass', 'sex', 'age', 'sibsp', 'parch', 'fare',  
      'embarked', 'class', 'who', 'adult_male', 'deck', 'embark_town',  
      'alive', 'alone'],  
      dtype='object')
```

▼ Checking Row Heading

```
df.index
RangeIndex(start=0, stop=244, step=1)
```

▼ Removing Specific Columns

```
df1=df.drop(["deck", "alone"],axis=1)
```

```
-----
KeyError                                Traceback (most recent call last)
<ipython-input-86-f14aa211c575> in <cell line: 1>()
----> 1 df1=df.drop(["deck","alone"],axis=1)

-----
      5 frames -----
/usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in drop(self, labels, errors)
    6932         if mask.any():
    6933             if errors != "ignore":
-> 6934                 raise KeyError(f"{list(labels[mask])} not found in axis")
    6935             indexer = indexer[~mask]
    6936             return self.delete(indexer)

KeyError: "[ 'deck', 'alone'] not found in axis"
```

SEARCH STACK OVERFLOW

▼ Checking Missing Vaules

```
df.isnull().sum()

survived      0
pclass       0
sex           0
age          177
sibsp        0
parch        0
fare         0
embarked      2
class        0
who          0
adult_male   0
deck        688
embark_town   2
alive        0
alone        0
dtype: int64
```

▼ Checking Unique Values

```
df.age.unique()

array([22. , 38. , 26. , 35. , nan, 54. , 2. , 27. , 14. ,
       4. , 58. , 20. , 39. , 55. , 31. , 34. , 15. , 28. ,
       8. , 19. , 40. , 66. , 42. , 21. , 18. , 3. , 7. ,
       49. , 29. , 65. , 28.5 , 5. , 11. , 45. , 17. , 32. ,
       16. , 25. , 0.83, 30. , 33. , 23. , 24. , 46. , 59. ,
       71. , 37. , 47. , 14.5 , 70.5 , 32.5 , 12. , 9. , 36.5 ,
       51. , 55.5 , 40.5 , 44. , 1. , 61. , 56. , 50. , 36. ,
       45.5 , 20.5 , 62. , 41. , 52. , 63. , 23.5 , 0.92, 43. ,
       60. , 10. , 64. , 13. , 48. , 0.75, 53. , 57. , 80. ,
       70. , 24.5 , 6. , 0.67, 30.5 , 0.42, 34.5 , 74. ])
```

```
###
```

```
df.embark_town.unique()

array(['Southampton', 'Cherbourg', 'Queenstown', nan], dtype=object)
```

Grouping

```
df1.groupby(["sex"]).mean()

<ipython-input-89-753766b12dbc>:1: FutureWarning: The default value of numeric_only in l
df1.groupby(["sex"]).mean()
      total_bill      tip      size
sex
Male      20.744076  3.089618  2.630573
Female    18.056897  2.833448  2.459770
```

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```
df1.groupby([
df1.groupby([
```

```
import seaborn as sns
df=sns.load_dataset("tips")
df
```

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4
...
239	29.03	5.92	Male	No	Sat	Dinner	3
240	27.18	2.00	Female	Yes	Sat	Dinner	2
241	22.67	2.00	Male	Yes	Sat	Dinner	2
242	17.82	1.75	Male	No	Sat	Dinner	2
243	18.78	3.00	Female	No	Thur	Dinner	2

244 rows × 7 columns

Double-click (or enter) to edit

▼ checking information about data

```
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 244 entries, 0 to 243
Data columns (total 7 columns):
#   Column      Non-Null Count  Dtype
---  -
0   total_bill  244 non-null   float64
1   tip         244 non-null   float64
2   sex         244 non-null   category
3   smoker      244 non-null   category
4   day         244 non-null   category
5   time        244 non-null   category
6   size        244 non-null   int64
dtypes: category(4), float64(2), int64(1)
memory usage: 7.4 KB
```

Checking first five 5 Entries

```
df.head()
```

	total_bill	tip	sex	smoker	day	time	size	
0	16.99	1.01	Female	No	Sun	Dinner	2	
1	10.34	1.66	Male	No	Sun	Dinner	3	
2	21.01	3.50	Male	No	Sun	Dinner	3	
3	23.68	3.31	Male	No	Sun	Dinner	2	
4	24.59	3.61	Female	No	Sun	Dinner	4	

Checking last 5 Entries

```
df.tail()
```

	total_bill	tip	sex	smoker	day	time	size	
239	29.03	5.92	Male	No	Sat	Dinner	3	
240	27.18	2.00	Female	Yes	Sat	Dinner	2	
241	22.67	2.00	Male	Yes	Sat	Dinner	2	
242	17.82	1.75	Male	No	Sat	Dinner	2	
243	18.78	3.00	Female	No	Thur	Dinner	2	

Summary Satatistics

```
df.describe()
```

	total_bill	tip	size	
count	244.000000	244.000000	244.000000	
mean	19.785943	2.998279	2.569672	
std	8.902412	1.383638	0.951100	
min	3.070000	1.000000	1.000000	
25%	13.347500	2.000000	2.000000	
50%	17.795000	2.900000	2.000000	
75%	24.127500	3.562500	3.000000	
max	50.810000	10.000000	6.000000	

Checking No.of Rows and Columns

```
df.shape
```

```
(244, 7)
```

```
df.shape[0]
```

```
244
```

```
df.shape[1]
```

```
7
```

Checking Column name

```
df.columns
```

```
Index(['total_bill', 'tip', 'sex', 'smoker', 'day', 'time', 'size'], dtype='object')
```

Checking Row Heading

```
df.index

RangeIndex(start=0, stop=244, step=1)
```

Checking Missing Vaules

```
df.isnull().sum()

total_bill    0
tip           0
sex           0
smoker        0
day           0
time          0
size          0
dtype: int64
```

Checking Unique Values

```
df.day.unique()

['Sun', 'Sat', 'Thur', 'Fri']
Categories (4, object): ['Thur', 'Fri', 'Sat', 'Sun']
```

```
import seaborn as sns
df=sns.load_dataset("titanic")
df
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone
0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	NaN	Southampton	no	False
1	1	1	female	38.0	1	0	71.2833	C	First	woman	False	C	Cherbourg	yes	False
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton	yes	True
3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	C	Southampton	yes	False
4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True
...
886	0	2	male	27.0	0	0	13.0000	S	Second	man	True	NaN	Southampton	no	True
887	1	1	female	19.0	0	0	30.0000	S	First	woman	False	B	Southampton	yes	True
888	0	3	female	NaN	1	2	23.4500	S	Third	woman	False	NaN	Southampton	no	False
889	1	1	male	26.0	0	0	30.0000	C	First	man	True	C	Cherbourg	yes	True
890	0	3	male	32.0	0	0	7.7500	Q	Third	man	True	NaN	Queenstown	no	True

891 rows × 15 columns