import pandas as pd

Making 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 dataframe

```
b=pd.DataFrame({"arooj":19,"farwa":20,"neha":19}, index=["A","B","C"]) b
```

		arooj	farwa	neha	1	ıl.
-	4	19	20	19		
E	3	19	20	19		
(2	19	20	19		

Working on dataset from seaborn library

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

urvived 0	pclass	sex	age				ombonkod	class	who	adult mala	dock	embark town	211110	alono
0			3-	этогр	parch	таге	embarked	Class	WIIO	auuit_maie	ueck	eliibai k_town	alive	атопе
U	3	male	22.0	1	0	7.2500	S	Third	man	True	NaN	Southampton	no	False
1	1	female	38.0	1	0	71.2833	С	First	woman	False	С	Cherbourg	yes	False
1	3	female	26.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton	yes	True
1	1	female	35.0	1	0	53.1000	S	First	woman	False	С	Southampton	yes	False
0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True
0	2	male	27.0	0	0	13.0000	S	Second	man	True	NaN	Southampton	no	True
1	1	female	19.0	0	0	30.0000	S	First	woman	False	В	Southampton	yes	True
0	3	female	NaN	1	2	23.4500	S	Third	woman	False	NaN	Southampton	no	False
1	1	male	26.0	0	0	30.0000	С	First	man	True	С	Cherbourg	yes	True
0	3	male	32.0	0	0	7.7500	Q	Third	man	True	NaN	Queenstown	no	True
	1 1 0 0 1 0	1 3 1 1 0 3 0 2 1 1 0 3 1 1 0 3	1 3 female 1 1 female 0 3 male 0 2 male 1 1 female 0 3 female 1 1 male 0 3 male	1 3 female 26.0 1 1 female 35.0 0 3 male 35.0 0 2 male 27.0 1 female 19.0 0 3 female NaN 1 male 26.0	1 3 female 26.0 0 1 1 female 35.0 1 0 3 male 35.0 0 0 2 male 27.0 0 1 1 female 19.0 0 0 3 female NaN 1 1 1 male 26.0 0 0 3 male 32.0 0	1 3 female 26.0 0 0 1 1 female 35.0 1 0 0 3 male 35.0 0 0 0 2 male 27.0 0 0 1 1 female 19.0 0 0 0 3 female NaN 1 2 1 1 male 26.0 0 0 0 3 male 32.0 0 0	1 3 female 26.0 0 0 7.9250 1 1 female 35.0 1 0 53.1000 0 3 male 35.0 0 0 8.0500 0 2 male 27.0 0 0 13.0000 1 1 female 19.0 0 0 30.0000 0 3 female NaN 1 2 23.4500 1 1 male 26.0 0 0 30.0000 0 3 male 32.0 0 0 7.7500	1 3 female 26.0 0 0 7.9250 S 1 1 female 35.0 1 0 53.1000 S 0 3 male 35.0 0 0 8.0500 S 0 2 male 27.0 0 0 13.0000 S 1 1 female 19.0 0 0 30.0000 S 0 3 female NaN 1 2 23.4500 S 1 1 male 26.0 0 0 30.0000 C 0 3 male 32.0 0 7.7500 Q	1 3 female 26.0 0 0 7.9250 S Third 1 1 female 35.0 1 0 53.1000 S First 0 3 male 35.0 0 0 8.0500 S Third 0 2 male 27.0 0 0 13.0000 S Second 1 1 female 19.0 0 0 30.0000 S First 0 3 female NaN 1 2 23.4500 S Third 1 1 male 26.0 0 0 30.0000 C First 0 3 male 32.0 0 7.7500 Q Third	1 3 female 26.0 0 0 7.9250 S Third woman 1 1 female 35.0 1 0 53.1000 S First woman 0 3 male 35.0 0 0 8.0500 S Third man 0 2 male 27.0 0 0 13.0000 S Second man 1 1 female 19.0 0 0 30.0000 S First woman 0 3 female NaN 1 2 23.4500 S Third woman 1 1 male 26.0 0 0 30.0000 C First man 0 3 male 32.0 0 7.7500 Q Third man	1 3 female 26.0 0 0 7.9250 S Third woman False 1 1 female 35.0 1 0 53.1000 S First woman False 0 3 male 35.0 0 0 8.0500 S Third man True	1 3 female 26.0 0 0 7.9250 S Third woman False NaN 1 1 female 35.0 1 0 53.1000 S First woman False C 0 3 male 35.0 0 0 8.0500 S Third man True NaN </td <td>1 3 female 26.0 0 0 7.9250 S Third woman False NaN Southampton 1 1 female 35.0 1 0 53.1000 S First woman False C Southampton 0 3 male 35.0 0 0 8.0500 S Third man True NaN Southampton 0 2 male 27.0 0 0 13.0000 S Second man True NaN Southampton 1 1 female 19.0 0 0 30.0000 S First woman False B Southampton 0 3 female NaN 1 2 23.4500 S Third woman False NaN Southampton 1 1 male 26.0 0 0 30.0000 C First man True C Cherbourg 0 3 male 32.0 0 0 7.7500 Q Third man True NaN<td>1 3 female 26.0 0 0 7.9250 S Third woman False NaN Southampton yes 1 1 female 35.0 1 0 53.1000 S First woman False C Southampton yes 0 3 male 35.0 0 0 8.0500 S Third man True NaN Southampton no 0 2 male 27.0 0 0 13.0000 S Second man True NaN Southampton no 1 1 female 19.0 0 0 30.0000 S First woman False B Southampton no 0 3 female NaN 1 2 23.4500 S Third woman False NaN Southampton no 1 1 male 26.0 0 0 30.0000 C First man True C Cherbourg yes 0 3 male 32.0 0</td></td>	1 3 female 26.0 0 0 7.9250 S Third woman False NaN Southampton 1 1 female 35.0 1 0 53.1000 S First woman False C Southampton 0 3 male 35.0 0 0 8.0500 S Third man True NaN Southampton 0 2 male 27.0 0 0 13.0000 S Second man True NaN Southampton 1 1 female 19.0 0 0 30.0000 S First woman False B Southampton 0 3 female NaN 1 2 23.4500 S Third woman False NaN Southampton 1 1 male 26.0 0 0 30.0000 C First man True C Cherbourg 0 3 male 32.0 0 0 7.7500 Q Third man True NaN <td>1 3 female 26.0 0 0 7.9250 S Third woman False NaN Southampton yes 1 1 female 35.0 1 0 53.1000 S First woman False C Southampton yes 0 3 male 35.0 0 0 8.0500 S Third man True NaN Southampton no 0 2 male 27.0 0 0 13.0000 S Second man True NaN Southampton no 1 1 female 19.0 0 0 30.0000 S First woman False B Southampton no 0 3 female NaN 1 2 23.4500 S Third woman False NaN Southampton no 1 1 male 26.0 0 0 30.0000 C First man True C Cherbourg yes 0 3 male 32.0 0</td>	1 3 female 26.0 0 0 7.9250 S Third woman False NaN Southampton yes 1 1 female 35.0 1 0 53.1000 S First woman False C Southampton yes 0 3 male 35.0 0 0 8.0500 S Third man True NaN Southampton no 0 2 male 27.0 0 0 13.0000 S Second man True NaN Southampton no 1 1 female 19.0 0 0 30.0000 S First woman False B Southampton no 0 3 female NaN 1 2 23.4500 S Third woman False NaN Southampton no 1 1 male 26.0 0 0 30.0000 C First man True C Cherbourg yes 0 3 male 32.0 0

Checking information about data

```
survived
                   891 non-null
                                   int64
     pclass
                   891 non-null
                                   int64
 2
                   891 non-null
                                   object
     sex
 3
                   714 non-null
                                   float64
     age
 4
     sibsp
                   891 non-null
                                   int64
                   891 non-null
                                   int64
     parch
                  891 non-null
 6
                                   float64
     fare
     embarked
                  889 non-null
                                   object
                                   category
     class
                   891 non-null
     who
                   891 non-null
                                   object
 10
     adult_male
                  891 non-null
                                   bool
     deck
                   203 non-null
                                   category
     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
```

Checking first five entries

df.head()



→ Checking last five entries

df.tail()

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone	1
886	0	2	male	27.0	0	0	13.00	S	Second	man	True	NaN	Southampton	no	True	
887	1	1	female	19.0	0	0	30.00	S	First	woman	False	В	Southampton	yes	True	
888	0	3	female	NaN	1	2	23.45	S	Third	woman	False	NaN	Southampton	no	False	
889	1	1	male	26.0	0	0	30.00	С	First	man	True	С	Cherbourg	yes	True	
890	0	3	male	32.0	0	0	7.75	Q	Third	man	True	NaN	Queenstown	no	True	

▼ Summary statics

df.describe()

	survived	pclass	age	sibsp	parch	fare	1	ıl.
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		

th

Checking number of rows and columns

→ checking columns name

Checking row headings

```
df.index
    RangeIndex(start=0, stop=891, step=1)
```

▼ Removing specific columns

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

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	embark_town	alive	10.	th
0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	Southampton	no		
1	1	1	female	38.0	1	0	71.2833	С	First	woman	False	Cherbourg	yes		
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	Southampton	yes		
3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	Southampton	yes		
4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	Southampton	no		
886	0	2	male	27.0	0	0	13.0000	S	Second	man	True	Southampton	no		
887	1	1	female	19.0	0	0	30.0000	S	First	woman	False	Southampton	yes		
888	0	3	female	NaN	1	2	23.4500	S	Third	woman	False	Southampton	no		
889	1	1	male	26.0	0	0	30.0000	С	First	man	True	Cherbourg	yes		
890	0	3	male	32.0	0	0	7.7500	Q	Third	man	True	Queenstown	no		
891 ro	ws × 13 colu	ımns													

Checking missing values

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

survived 0
pclass 0
sex 0
age 177
sibsp 0
parch 0
```

```
fare
embarked
                 2
class
                a
who
adult_male
                0
deck
               688
embark town
                2
alive
                 0
alone
dtype: int64
```

Checking unique values

→ Groupby