

pandas tips and tricks

```
import pandas as pd

# load dataset to work with
import seaborn as sns

# titanic dataset ko load karen
kashti = sns.load_dataset('titanic')

kashti.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
```

what every column means in kashti Dataset?

how many datasets are there in
`sns.load_dataset()`? and how to fetch.

```
kashti.head()
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked
class \								
0	0	3	male	22.0	1	0	7.2500	S
Third								
1	1	1	female	38.0	1	0	71.2833	C
First								
2	1	3	female	26.0	0	0	7.9250	S
Third								
3	1	1	female	35.0	1	0	53.1000	S
First								
4	0	3	male	35.0	0	0	8.0500	S
Third								

	who	adult_male	deck	embark_town	alive	alone
0	man	True	NaN	Southampton	no	False
1	woman	False	C	Cherbourg	yes	False
2	woman	False	NaN	Southampton	yes	True
3	woman	False	C	Southampton	yes	False
4	man	True	NaN	Southampton	no	True

kashti.tail()

	survived	pclass	sex	age	sibsp	parch	fare	embarked
class \								
886	0	2	male	27.0	0	0	13.00	S
Second								
887	1	1	female	19.0	0	0	30.00	S
First								
888	0	3	female	NaN	1	2	23.45	S
Third								
889	1	1	male	26.0	0	0	30.00	C
First								
890	0	3	male	32.0	0	0	7.75	Q
Third								

	who	adult_male	deck	embark_town	alive	alone
886	man	True	NaN	Southampton	no	True
887	woman	False	B	Southampton	yes	True
888	woman	False	NaN	Southampton	no	False
889	man	True	C	Cherbourg	yes	True
890	man	True	NaN	Queenstown	no	True

download phool ka data

```
phool = sns.load_dataset('iris')
phool.sample(100)
```

	sepal_length	sepal_width	petal_length	petal_width	species
62	6.0	2.2	4.0	1.0	versicolor
30	4.8	3.1	1.6	0.2	setosa
146	6.3	2.5	5.0	1.9	virginica

81	5.5	2.4	3.7	1.0	versicolor
132	6.4	2.8	5.6	2.2	virginica
...
75	6.6	3.0	4.4	1.4	versicolor
59	5.2	2.7	3.9	1.4	versicolor
126	6.2	2.8	4.8	1.8	virginica
54	6.5	2.8	4.6	1.5	versicolor
31	5.4	3.4	1.5	0.4	setosa

[100 rows x 5 columns]

```
kashti.to_excel('kashti.xlsx')
```

```
phool.to_excel('phool.xlsx')
```

```
phool.to_excel('../day_7/phool.xlsx')
```

```
df = pd.read_excel('kashti.xlsx')
```

```
df.head()
```

	Unnamed: 0	survived	pclass	sex	age	sibsp	parch	fare
embarked \								
0	0	0	3	male	22.0	1	0	7.2500
1	1	1	1	female	38.0	1	0	71.2833
2	2	1	3	female	26.0	0	0	7.9250
3	3	1	1	female	35.0	1	0	53.1000
4	4	0	3	male	35.0	0	0	8.0500

	class	who	adult_male	deck	embark_town	alive	alone
0	Third	man	True	NaN	Southampton	no	False
1	First	woman	False	C	Cherbourg	yes	False
2	Third	woman	False	NaN	Southampton	yes	True
3	First	woman	False	C	Southampton	yes	False
4	Third	man	True	NaN	Southampton	no	True

```
df.columns
```

```
Index(['Unnamed: 0', 'survived', 'pclass', 'sex', 'age', 'sibsp',
      'parch',
      'fare', 'embarked', 'class', 'who', 'adult_male', 'deck',
      'embark_town',
      'alive', 'alone'],
      dtype='object')
```

```
df['embark_town']
```

```

0      Southampton
1      Cherbourg
2      Southampton
3      Southampton
4      Southampton
...
886    Southampton
887    Southampton
888    Southampton
889      Cherbourg
890    Queenstown
Name: embark_town, Length: 891, dtype: object

```

```
df[['embark_town', 'survived']]
```

```

      embark_town  survived
0      Southampton         0
1      Cherbourg         1
2      Southampton         1
3      Southampton         1
4      Southampton         0
..          ...      ...
886    Southampton         0
887    Southampton         1
888    Southampton         0
889      Cherbourg         1
890    Queenstown         0

```

```
[891 rows x 2 columns]
```

```
df.embark_town.unique()
```

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

what is nan, kitna important hy or is ka hona ya na hna kitna zaroori ha?

```
df.deck.unique()
```

```
array([nan, 'C', 'E', 'G', 'D', 'A', 'B', 'F'], dtype=object)
```

```
df.describe()
```

	Unnamed: 0	survived	pclass	age	sibsp
parch \					
count	891.000000	891.000000	891.000000	714.000000	891.000000
891.000000					
mean	445.000000	0.383838	2.308642	29.699118	0.523008

```

0.381594
std      257.353842      0.486592      0.836071      14.526497      1.102743
0.806057
min       0.000000      0.000000      1.000000      0.420000      0.000000
0.000000
25%      222.500000      0.000000      2.000000      20.125000      0.000000
0.000000
50%      445.000000      0.000000      3.000000      28.000000      0.000000
0.000000
75%      667.500000      1.000000      3.000000      38.000000      1.000000
0.000000
max      890.000000      1.000000      3.000000      80.000000      8.000000
6.000000

```

```

                fare
count  891.000000
mean    32.204208
std     49.693429
min      0.000000
25%      7.910400
50%     14.454200
75%     31.000000
max     512.329200

```

```
df = df.drop(columns='Unnamed: 0')
```

```
df
```

```

   survived  pclass    sex  age  sibsp  parch    fare embarked
class \
0          0      3   male  22.0     1     0    7.2500         S
Third
1          1      1  female  38.0     1     0   71.2833         C
First
2          1      3  female  26.0     0     0    7.9250         S
Third
3          1      1  female  35.0     1     0   53.1000         S
First
4          0      3   male  35.0     0     0    8.0500         S
Third
...      ...     ...     ...     ...     ...     ...     ...
...
886        0      2   male  27.0     0     0   13.0000         S
Second
887        1      1  female  19.0     0     0   30.0000         S
First
888        0      3  female   NaN     1     2   23.4500         S
Third
889        1      1   male  26.0     0     0   30.0000         C
First

```

```
890      0      3    male  32.0      0      0  7.7500      Q
Third
```

	who	adult_male	deck	embark_town	alive	alone
0	man	True	NaN	Southampton	no	False
1	woman	False	C	Cherbourg	yes	False
2	woman	False	NaN	Southampton	yes	True
3	woman	False	C	Southampton	yes	False
4	man	True	NaN	Southampton	no	True
...
886	man	True	NaN	Southampton	no	True
887	woman	False	B	Southampton	yes	True
888	woman	False	NaN	Southampton	no	False
889	man	True	C	Cherbourg	yes	True
890	man	True	NaN	Queenstown	no	True

```
[891 rows x 15 columns]
```

```
df.groupby(['sex']).describe().T
```

sex		female	male
survived	count	314.000000	577.000000
	mean	0.742038	0.188908
	std	0.438211	0.391775
	min	0.000000	0.000000
	25%	0.000000	0.000000
	50%	1.000000	0.000000
	75%	1.000000	0.000000
pclass	max	1.000000	1.000000
	count	314.000000	577.000000
	mean	2.159236	2.389948
	std	0.857290	0.813580
	min	1.000000	1.000000
	25%	1.000000	2.000000
	50%	2.000000	3.000000
age	75%	3.000000	3.000000
	max	3.000000	3.000000
	count	261.000000	453.000000
	mean	27.915709	30.726645
	std	14.110146	14.678201
	min	0.750000	0.420000
	25%	18.000000	21.000000
sibsp	50%	27.000000	29.000000
	75%	37.000000	39.000000
	max	63.000000	80.000000
	count	314.000000	577.000000
	mean	0.694268	0.429809
	std	1.156520	1.061811
	min	0.000000	0.000000
	25%	0.000000	0.000000

	50%	0.000000	0.000000
	75%	1.000000	0.000000
	max	8.000000	8.000000
parch	count	314.000000	577.000000
	mean	0.649682	0.235702
	std	1.022846	0.612294
	min	0.000000	0.000000
	25%	0.000000	0.000000
	50%	0.000000	0.000000
	75%	1.000000	0.000000
	max	6.000000	5.000000
fare	count	314.000000	577.000000
	mean	44.479818	25.523893
	std	57.997698	43.138263
	min	6.750000	0.000000
	25%	12.071875	7.895800
	50%	23.000000	10.500000
	75%	55.000000	26.550000
	max	512.329200	512.329200

```
df.count()
```

```
survived      891
pclass        891
sex           891
age           714
sibsp         891
parch         891
fare          891
embarked      889
class         891
who           891
adult_male    891
deck         203
embark_town   889
alive        891
alone        891
dtype: int64
```

```
df.groupby(["sex"]).count()
```

	survived	pclass	age	sibsp	parch	fare	embarked	class
who \ sex								
female	314	314	261	314	314	314	312	314
male	577	577	453	577	577	577	577	577

	adult_male	deck	embark_town	alive	alone
sex					
female	314	97	312	314	314
male	577	106	577	577	577

```
df.groupby(["sex", ("survived"))].count()
```

		pclass	age	sibsp	parch	fare	embarked	class	who
\	sex	survived							
female	0		81	64	81	81	81	81	81
	1		233	197	233	233	231	233	233
male	0		468	360	468	468	468	468	468
	1		109	93	109	109	109	109	109

		adult_male	deck	embark_town	alive	alone
sex	survived					
female	0		81	6	81	81
	1		233	91	231	233
male	0		468	61	468	468
	1		109	45	109	109

```
df.groupby(["survived", "sex"])["survived"].count()
```

survived	sex	
0	female	81
	male	468
1	female	233
	male	109

Name: survived, dtype: int64

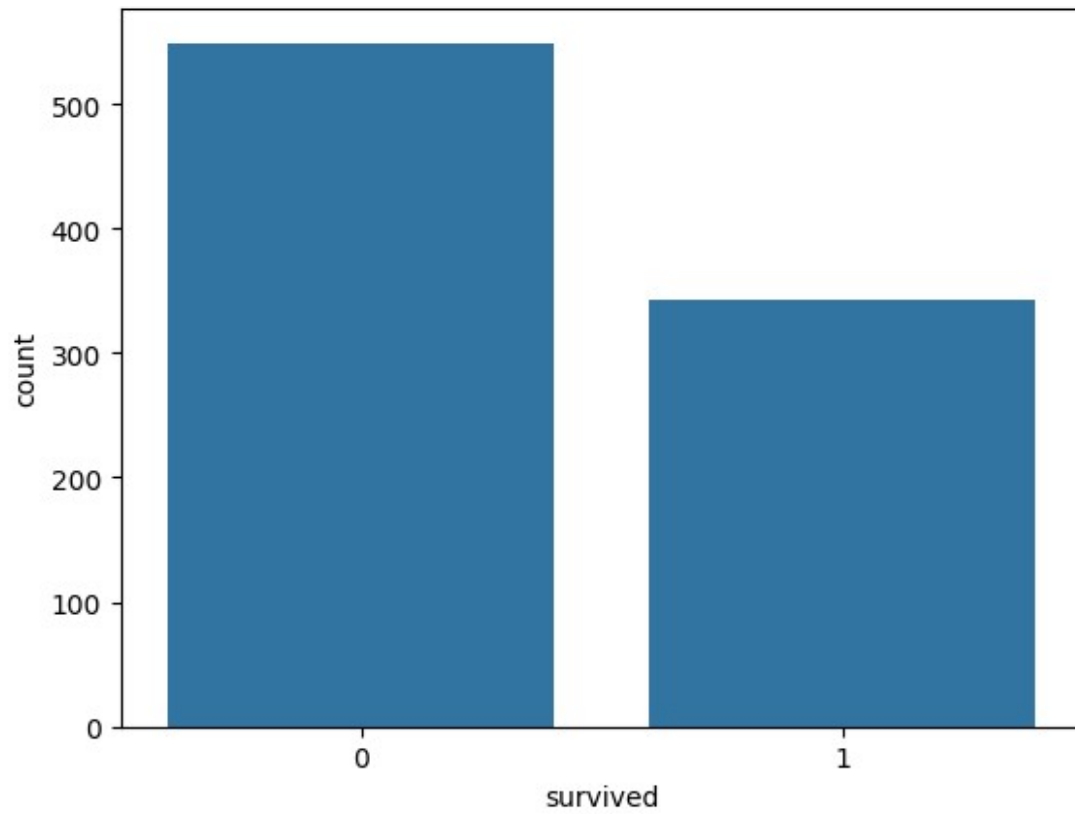
```
df.groupby(["survived", "sex", "who"])["who"].count()
```

survived	sex	who	
0	female	child	15
		woman	66
	male	child	19
		man	449
1	female	child	28
		woman	205
	male	child	21
		man	88

Name: who, dtype: int64

```
sns.countplot(x="survived", data=df)
```

```
<Axes: xlabel='survived', ylabel='count'>
```

```
sns.countplot(x="survived", hue="sex", data=df)  
<Axes: xlabel='survived', ylabel='count'>
```

