```
import numpy as np
import pandas as pd
import seaborn as sns
```

df = sns.load_dataset('titanic')

df.head(5)

→		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	С	First	woman	False	С	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	С	Southampton	yes	False
	4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True

df.columns

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
dtyp	es: bool(2),	category(2), flo	at64(2), int64(4), object(5)
memo	ry usage: 80.	7+ KB	

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

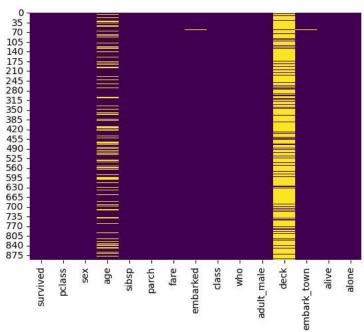
df.isnull().sum()

$\overline{\rightarrow}$	survived	0
	pclass	0
	sex	0

```
177
age
sibsp
                 0
parch
                  0
                 0
fare
embarked
                 2
class
who
                 0
                 0
{\tt adult\_male}
               688
deck
embark_town
                 2
alive
                 0
alone
                 0
dtype: int64
```

sns.heatmap(df.isnull(),cbar = False, cmap = 'viridis')



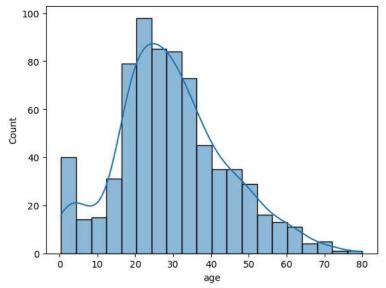


df.head(5)

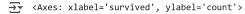
_ →		survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone
	0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	NaN	Southampton	no	False
	1	1	1	female	38.0	1	0	71.2833	С	First	woman	False	С	Cherbourg	yes	False
	2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton	yes	True
	3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	С	Southampton	yes	False
	4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True

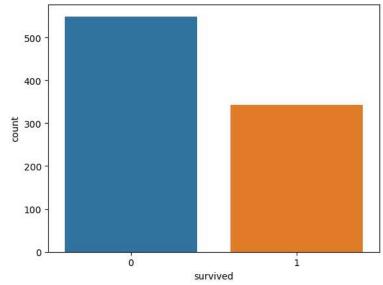
sns.histplot(df['age'],kde = True)

<Axes: xlabel='age', ylabel='Count'>



sns.countplot(data = df, x = 'survived')



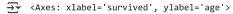


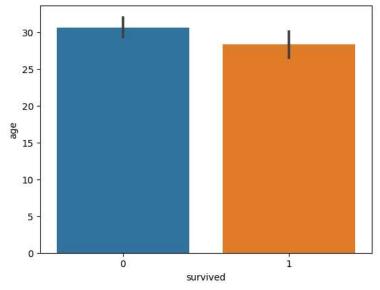
import plotly.express as px

new_df = df['survived'].value_counts().reset_index()

Start coding or generate with AI.

sns.barplot(data = df, x = 'survived', y = 'age')

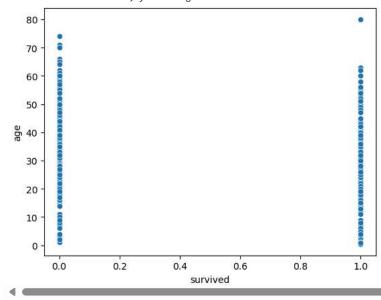




Start coding or generate with AI.

sns.scatterplot(data = df, x = 'survived', y = 'age')





Double-click (or enter) to edit

df.info()

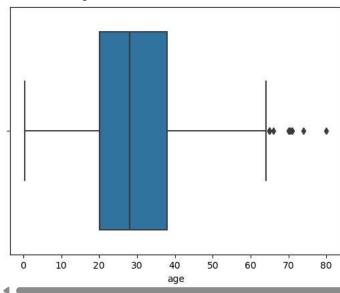
<</pre>
<</pre>

Data	COTUMNIS (COC	ar is corumns).	
#	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
```

sns.boxplot(data = df, x = 'age')

<Axes: xlabel='age'>



df.columns

Data preprocessing

 ${\it from sklearn.preprocessing import Label Encoder}$

Handling the missing values

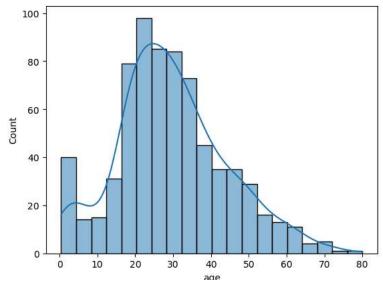
df.isnull().sum()

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

sns.histplot(df['age'],kde = True)

E:\Users\wwwsi\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be remov with pd.option_context('mode.use_inf_as_na', True):

<Axes: xlabel='age', ylabel='Count'>



df['age'].fillna(df['age']).median()

→ 28.0

df['age'].isnull().sum()

→ 177

df['age'] = df['age'].fillna(df['age']).median()

df.isnull().sum()

_	survived	0
	pclass	0
	sex	0
	age	0
	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	

Double-click (or enter) to edit

encoding

df.head(5)

∑ *		survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone
	0	0	3	male	28.0	1	0	7.2500	S	Third	man	True	NaN	Southampton	no	False
	1	1	1	female	28.0	1	0	71.2833	С	First	woman	False	С	Cherbourg	yes	False
	2	1	3	female	28.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton	yes	True
	3	1	1	female	28.0	1	0	53.1000	S	First	woman	False	С	Southampton	yes	False
	4	0	3	male	28.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True

label encoding

df['sex'] = df['sex'].map({"male" : 0, "female": 1})

df['sex'].head(5)

→ 0

1 1 2 1 3 1

4 0

0

Name: sex, dtype: int64

outliers

df['age'].head(5)

→ 0 28.0

1 28.0

2 28

3 28.0

4 28.0

Name: age, dtype: float64

df.head(5)

₹		survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone
	0	0	3	0	28.0	1	0	7.2500	S	Third	man	True	NaN	Southampton	no	False
	1	1	1	1	28.0	1	0	71.2833	С	First	woman	False	С	Cherbourg	yes	False
	2	1	3	1	28.0	0	0	7.9250	S	Third	woman	False	NaN	Southampton	yes	True
	3	1	1	1	28.0	1	0	53.1000	S	First	woman	False	С	Southampton	yes	False
	4	0	3	0	28.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True
	4	0	3	0	28.0	0	0	8.0500	S	Third	man	True	NaN	Southampton	no	True

df.drop('class', axis = 1, inplace = True)

df.head(5)

		survived	pclass	sex	age	sibsp	parch	fare	embarked	who	adult_male	deck	embark_town	alive	alone
	0	0	3	0	28.0	1	0	7.2500	S	man	True	NaN	Southampton	no	False
	1	1	1	1	28.0	1	0	71.2833	С	woman	False	С	Cherbourg	yes	False
	2	1	3	1	28.0	0	0	7.9250	S	woman	False	NaN	Southampton	yes	True
	3	1	1	1	28.0	1	0	53.1000	S	woman	False	С	Southampton	yes	False
	4	0	3	0	28.0	0	0	8.0500	S	man	True	NaN	Southampton	no	True

df.drop('embark_town',axis = 1, inplace = True)

df.head(5)

_		survived	pclass	sex	age	sibsp	parch	fare	embarked	who	adult_male	deck	alive	alone
	0	0	3	0	28.0	1	0	7.2500	S	man	True	NaN	no	False
	1	1	1	1	28.0	1	0	71.2833	С	woman	False	С	yes	False
	2	1	3	1	28.0	0	0	7.9250	S	woman	False	NaN	yes	True
	3	1	1	1	28.0	1	0	53.1000	S	woman	False	С	yes	False
	4	0	3	0	28.0	0	0	8.0500	S	man	True	NaN	no	True

df['embarked'].unique()

⇒ array(['S', 'C', 'Q', nan], dtype=object)

df['who'].unique()

⇒ array(['man', 'woman', 'child'], dtype=object)

df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 891 entries, 0 to 890 Data columns (total 13 columns):

#	Column	Non-Null Count	Dtype
0	survived	891 non-null	int64
1	pclass	891 non-null	int64
2	sex	891 non-null	int64
3	age	891 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	who	891 non-null	object
9	adult_male	891 non-null	bool
10	deck	203 non-null	category
11	alive	891 non-null	object
12	alone	891 non-null	bool

dtypes: bool(2), category(1), float64(2), int64(5), object(3)
memory usage: 72.7+ KB

df.drop('deck', axis = 1, inplace = True)

df.head(5)

→		survived	pclass	sex	age	sibsp	parch	fare	embarked	who	adult_male	alive	alone
	0	0	3	0	28.0	1	0	7.2500	S	man	True	no	False
	1	1	1	1	28.0	1	0	71.2833	С	woman	False	yes	False
	2	1	3	1	28.0	0	0	7.9250	S	woman	False	yes	True
	3	1	1	1	28.0	1	0	53.1000	S	woman	False	yes	False
	4	0	3	0	28.0	0	0	8.0500	S	man	True	no	True

df.drop('adult_male',axis = 1, inplace = True)

df.drop('alive',axis = 1, inplace = True)

df.head(5)

→		survived	pclass	sex	age	sibsp	parch	fare	embarked	who	alone
	0	0	3	0	28.0	1	0	7.2500	S	man	False
	1	1	1	1	28.0	1	0	71.2833	С	woman	False
	2	1	3	1	28.0	0	0	7.9250	S	woman	True
	3	1	1	1	28.0	1	0	53.1000	S	woman	False
	4	0	3	0	28.0	0	0	8.0500	S	man	True

df.drop('alone',axis = 1, inplace = True)

df.head(5)

₹		survived	pclass	sex	age	sibsp	parch	fare	embarked	who
	0	0	3	0	28.0	1	0	7.2500	S	man
	1	1	1	1	28.0	1	0	71.2833	С	woman
	2	1	3	1	28.0	0	0	7.9250	S	woman
	3	1	1	1	28.0	1	0	53.1000	S	woman
	4	0	3	0	28.0	0	0	8.0500	S	man
	4									

df.isnull().sum()



df['embarked'].dropna(inplace = True)

df.dropna()

→		survived	pclass	sex	age	sibsp	parch	fare	embarked	who
	0	0	3	0	28.0	1	0	7.2500	S	man
	1	1	1	1	28.0	1	0	71.2833	С	woman
	2	1	3	1	28.0	0	0	7.9250	S	woman
	3	1	1	1	28.0	1	0	53.1000	S	woman
	4	0	3	0	28.0	0	0	8.0500	S	man
	886	0	2	0	28.0	0	0	13.0000	S	man
	887	1	1	1	28.0	0	0	30.0000	S	woman
	888	0	3	1	28.0	1	2	23.4500	S	woman
	889	1	1	0	28.0	0	0	30.0000	С	man
	890	0	3	0	28.0	0	0	7.7500	Q	man
	889 ro	ws × 9 colur	mns							

df['embarked'] = df['embarked'].fillna(df['embarked'].mode()[0])

df.isnull().sum()



df.head(5)

₹		survived	pclass	sex	age	sibsp	parch	fare	embarked	who
	0	0	3	0	28.0	1	0	7.2500	S	man
	1	1	1	1	28.0	1	0	71.2833	С	woman
	2	1	3	1	28.0	0	0	7.9250	S	woman
	3	1	1	1	28.0	1	0	53.1000	S	woman
	4	0	3	0	28.0	0	0	8.0500	S	man

df['embarked'].unique()

```
⇒ array(['S', 'C', 'Q'], dtype=object)
```

```
df['embarked'] = df['embarked'].map({'S':0, 'C' : 1, 'Q' : 2})
```

df['embarked'].unique()

⇒ array([0, 1, 2], dtype=int64)

df.head(5)

₹		survived	pclass	sex	age	sibsp	parch	fare	embarked	who
	0	0	3	0	28.0	1	0	7.2500	0	man
	1	1	1	1	28.0	1	0	71.2833	1	woman
	2	1	3	1	28.0	0	0	7.9250	0	woman
	3	1	1	1	28.0	1	0	53.1000	0	woman
	4	0	3	0	28.0	0	0	8.0500	0	man

Start coding or generate with AI.

df.sample(5)

→		pclass	sex	age	sibsp	parch	fare	embarked	who
	246	3	1	28.0	0	0	7.7750	0	1
	13	3	0	28.0	1	5	31.2750	0	0
	144	2	0	28.0	0	0	11.5000	0	0
	514	3	0	28.0	0	0	7.4958	0	0
	271	3	0	28.0	0	0	0.0000	0	0

x=df.drop("survived",axis=1,inplace=True)

from sklearn.preprocessing import Normalizer