

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
```

```
warnings.filterwarnings('ignore')
```

```
df = sns.load_dataset('titanic')
```

```
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
4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	NaN	Southampton

```
df.shape
```

(891, 15)

```
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.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
```

```
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['age'].fillna(df['age'].mean(), inplace=True)
```

```
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

```
df['embarked'].value_counts()
```

S	644
C	168
Q	77
Name:	embarked, dtype: int64

```
df['embarked'].fillna('S', inplace=True)
```

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

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

```
df['deck'].value_counts()
```

C	59
B	47
D	33
E	32
A	15
F	13
G	4
Name:	deck, dtype: int64

```
df['deck'].fillna(method='ffill', inplace=True)
```

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

```
survived      0
pclass        0
sex           0
age           0
sibsp         0
parch         0
fare          0
embarked      0
class         0
who           0
adult_male    0
deck          1
embark_town   2
alive         0
alone         0
dtype: int64
```

```
df['deck'].fillna(method='bfill', inplace=True)
```

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

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

```
df['embark_town'].value_counts()
```

```
Southampton    644
Cherbourg      168
Queenstown     77
Name: embark_town, dtype: int64
```

```
df['embark_town'].fillna('Southampton', inplace=True)
```

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

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

▼ EDA (Exploratory Data Analysis)

Use the inbuilt dataset 'titanic'. The dataset contains 891 rows and contains information about the passengers who boarded the unfortunate Titanic ship. Use the Seaborn library to see if we can find any patterns in the 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         891 non-null    float64
 4   sibsp       891 non-null    int64
 5   parch       891 non-null    int64
 6   fare        891 non-null    float64
 7   embarked    891 non-null    object
 8   class       891 non-null    category
 9   who         891 non-null    object
10  adult_male  891 non-null    bool
11  deck        891 non-null    category
12  embark_town 891 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['survived'].value_counts()
```

```
0    549
1    342
Name: survived, dtype: int64
```

```
df['pclass'].value_counts()
```

```
3    491
1    216
2    184
Name: pclass, dtype: int64
```

```
df['sex'].value_counts()
```

```
male      577
female    314
Name: sex, dtype: int64
```

```
df['age'].value_counts()
```

```
29.699118    177
24.000000     30
22.000000     27
18.000000     26
28.000000     25
...
36.500000      1
55.500000      1
0.920000        1
23.500000        1
74.000000        1
Name: age, Length: 89, dtype: int64
```

```
df['sibsp'].value_counts()
```

```
0    608
1    209
2     28
4     18
3     16
8       7
5       5
Name: sibsp, dtype: int64
```

```
df['parch'].value_counts()
```

```
0    678
1    118
2     80
5       5
3       5
4       4
6       1
Name: parch, dtype: int64
```

```
df['fare'].value_counts()

8.0500    43
13.0000    42
7.8958     38
7.7500     34
26.0000     31
..
35.0000     1
28.5000     1
6.2375      1
14.0000     1
10.5167     1
Name: fare, Length: 248, dtype: int64
```

```
df['embarked'].value_counts()

S    646
C    168
Q     77
Name: embarked, dtype: int64
```

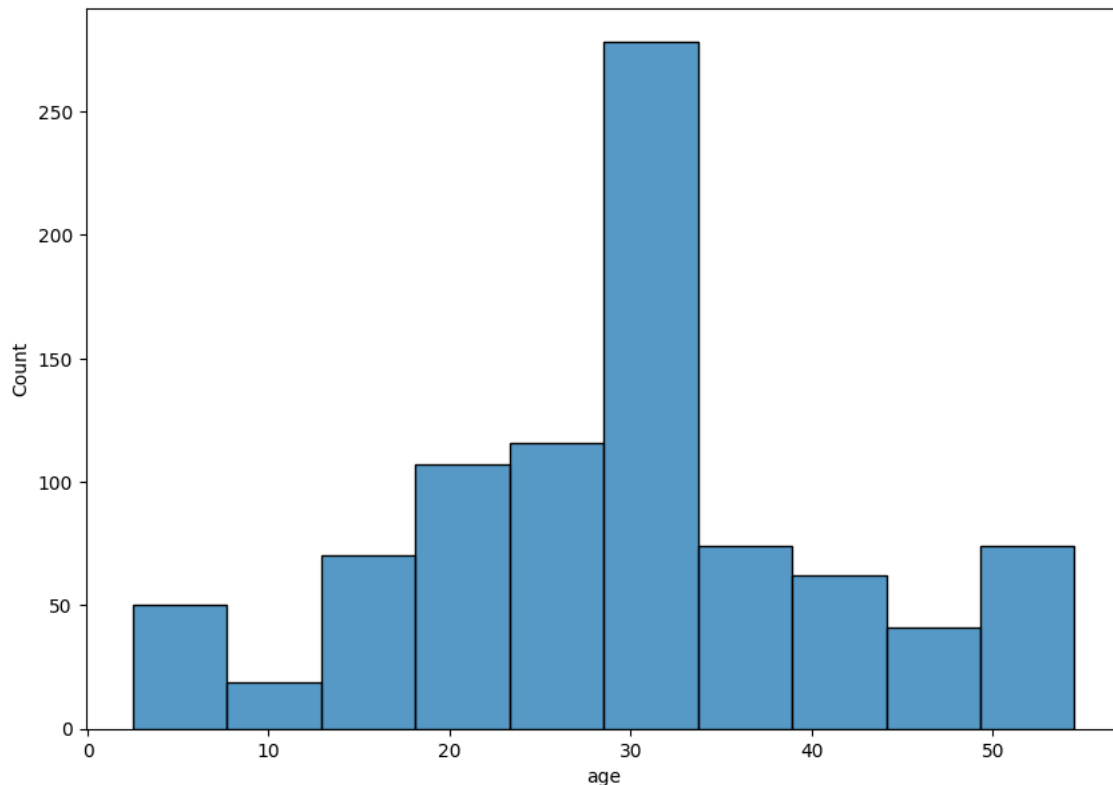
```
df['who'].value_counts()

man    537
woman  271
child   83
Name: who, dtype: int64
```

Write a code to check how the price of the ticket (column name: 'fare') for each passenger is distributed by plotting a histogram.

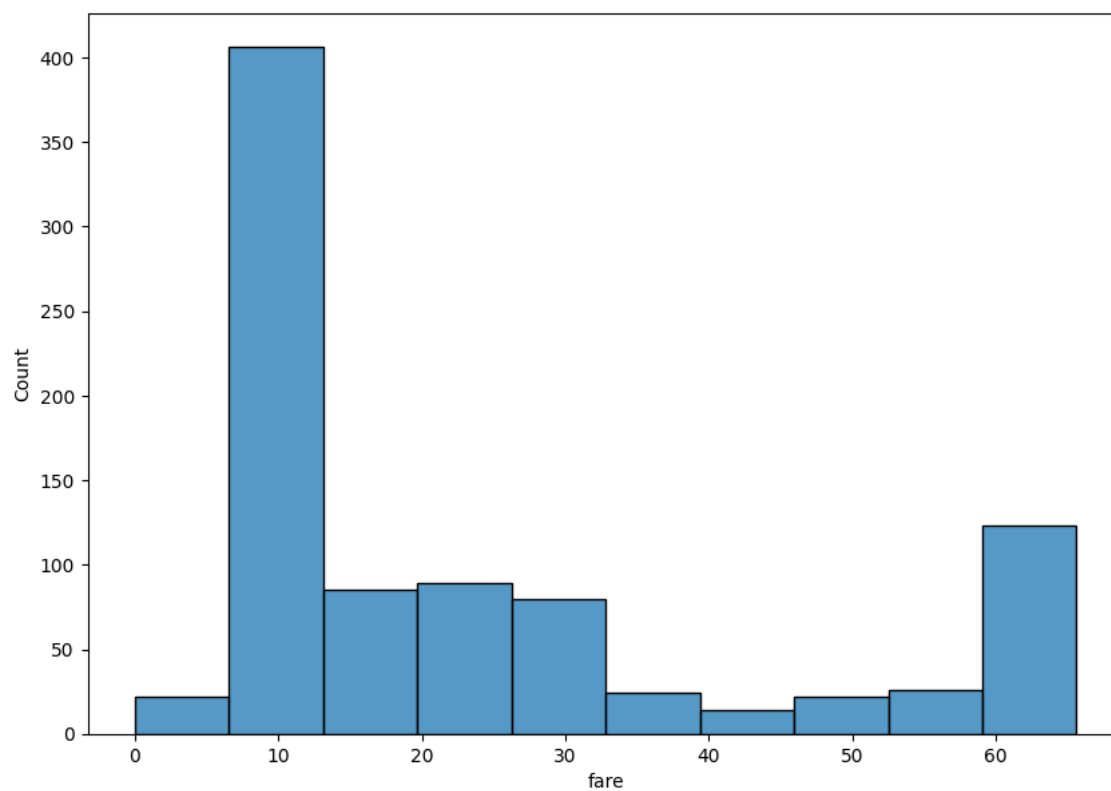
```
plt.figure(figsize = (10,7))
sns.histplot(df['age'],bins=10)
```

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



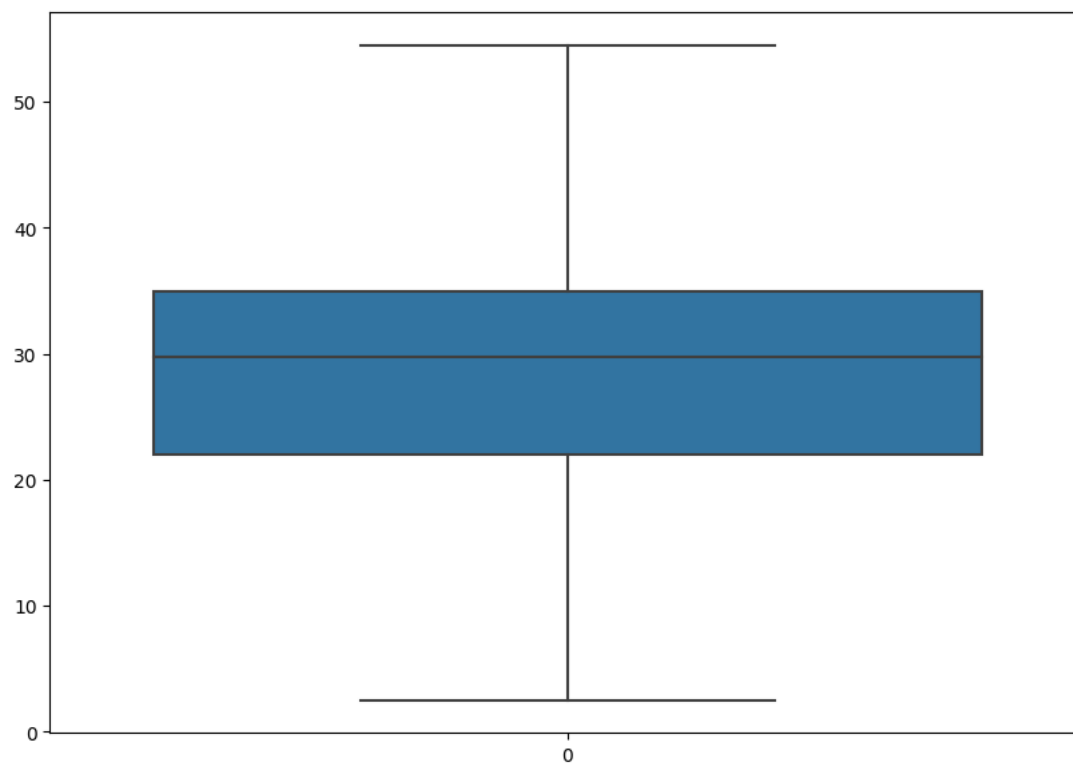
```
plt.figure(figsize = (10,7))  
sns.histplot(df['fare'],bins = 10)
```

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



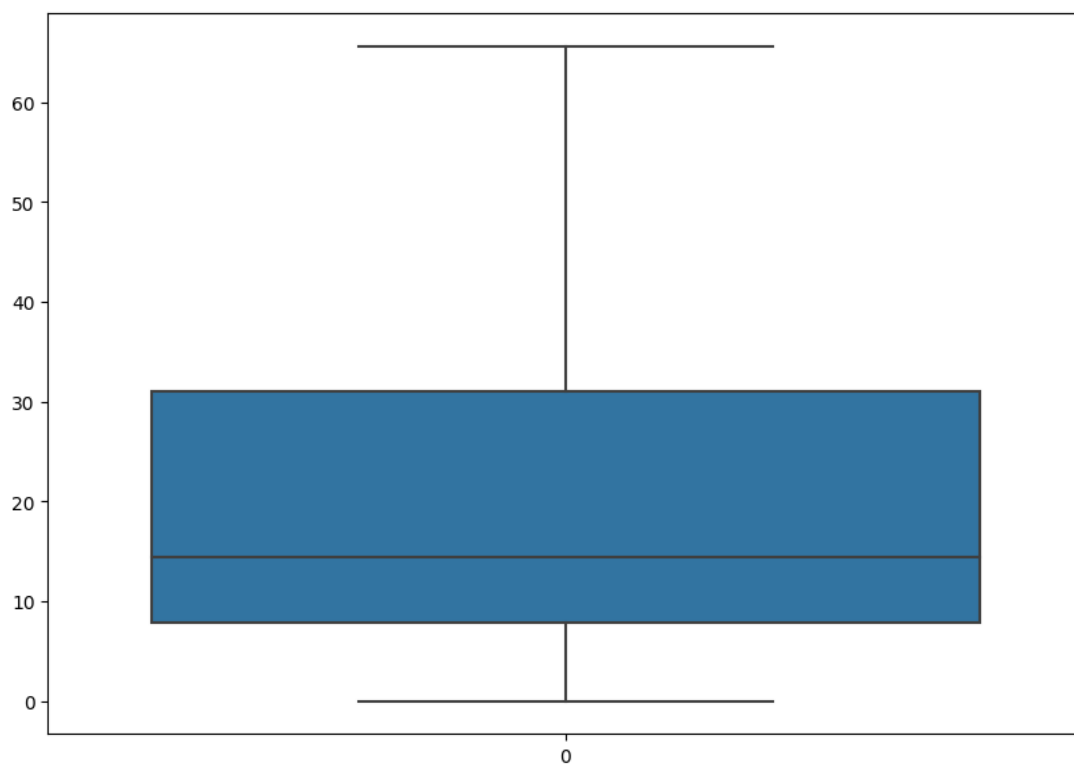
```
plt.figure(figsize=(10,7))  
sns.boxplot(df['age'])
```

<Axes: >



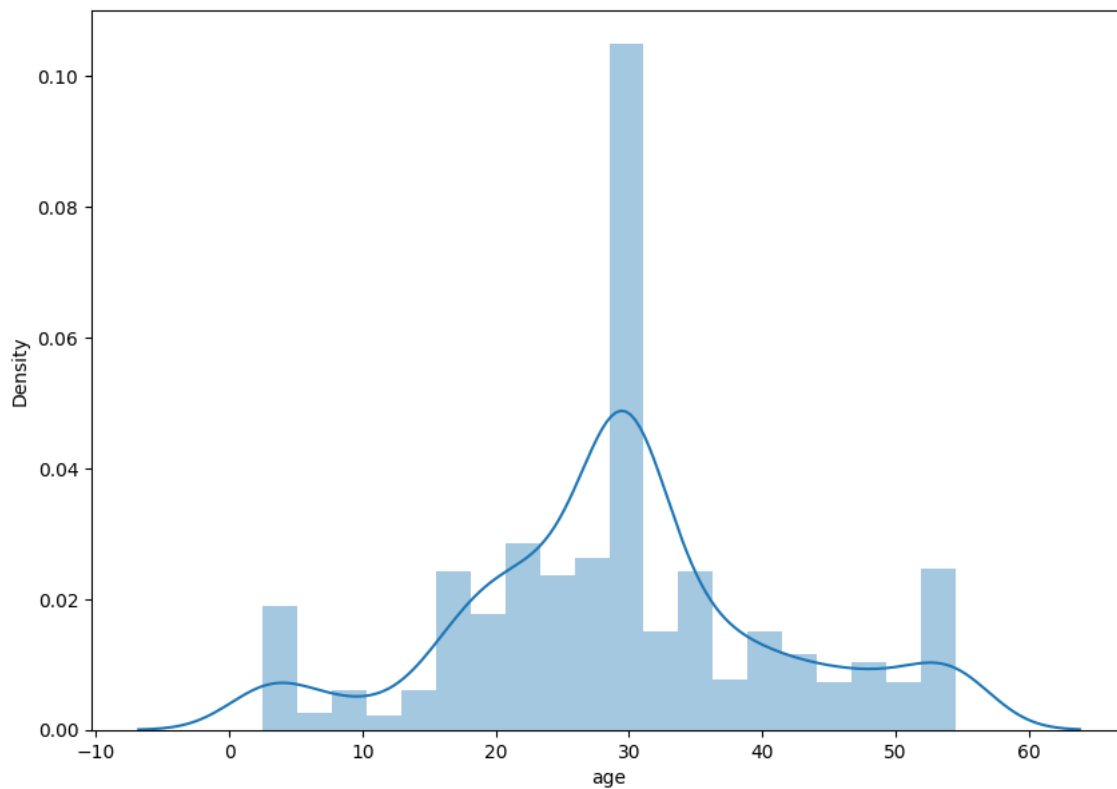
```
plt.figure(figsize=(10,7))  
sns.boxplot(df['fare'])
```

<Axes: >



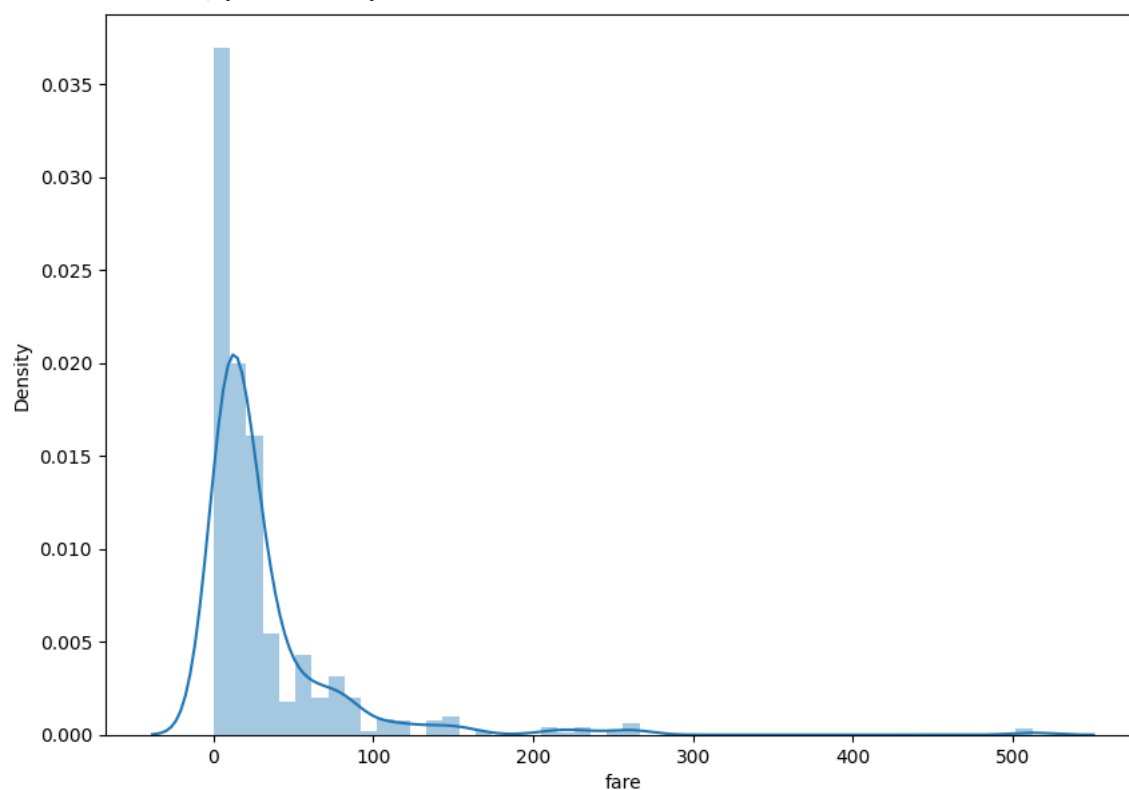
```
plt.figure(figsize=(10,7))  
sns.distplot(df['age'])
```

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



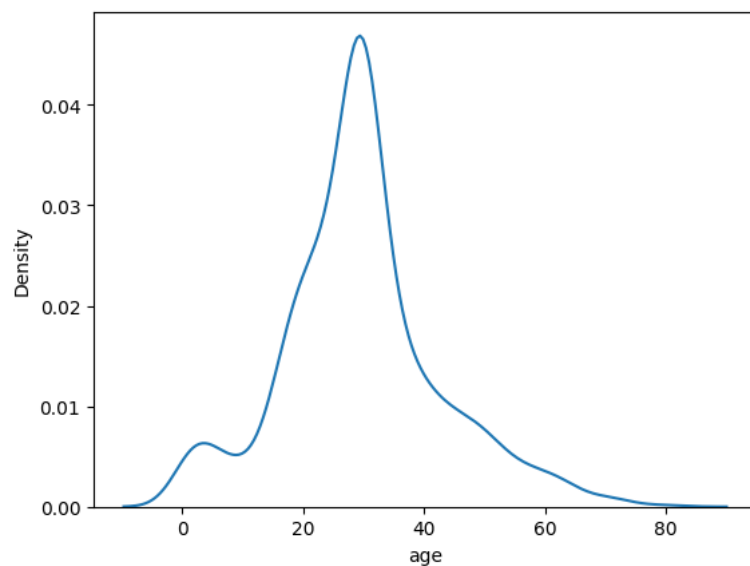
```
plt.figure(figsize=(10,7))  
sns.distplot(df['fare'])
```

<Axes: xlabel='fare', ylabel='Density'>

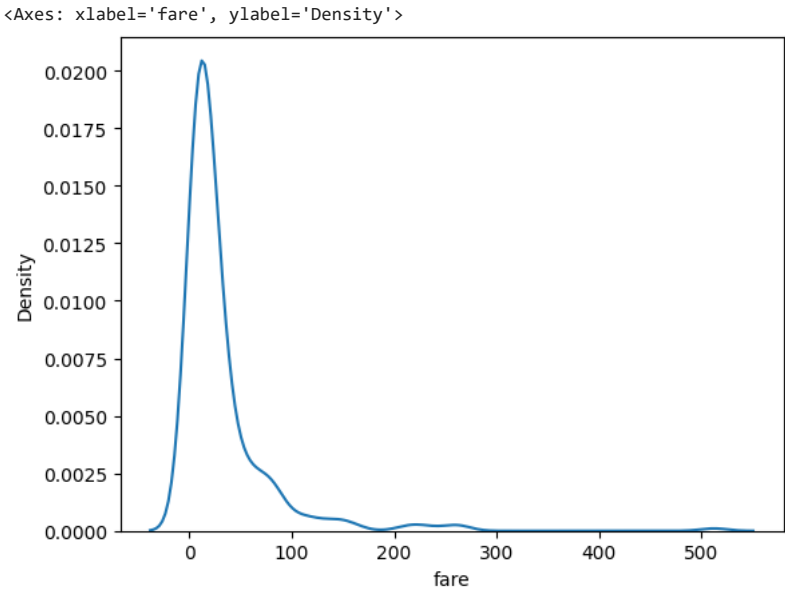


```
sns.kdeplot(df['age'])
```

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




```
sns.kdeplot(df['fare'])
```



```
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         891 non-null    float64
4   sibsp       891 non-null    int64
5   parch       891 non-null    int64
6   fare        891 non-null    float64
7   embarked    891 non-null    object
8   class       891 non-null    category
9   who         891 non-null    object
10  adult_male  891 non-null    bool
11  deck        891 non-null    category
12  embark_town 891 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['age'].skew()
```

0.4344880940129925

```
df['fare'].skew()
```

4.787316519674893

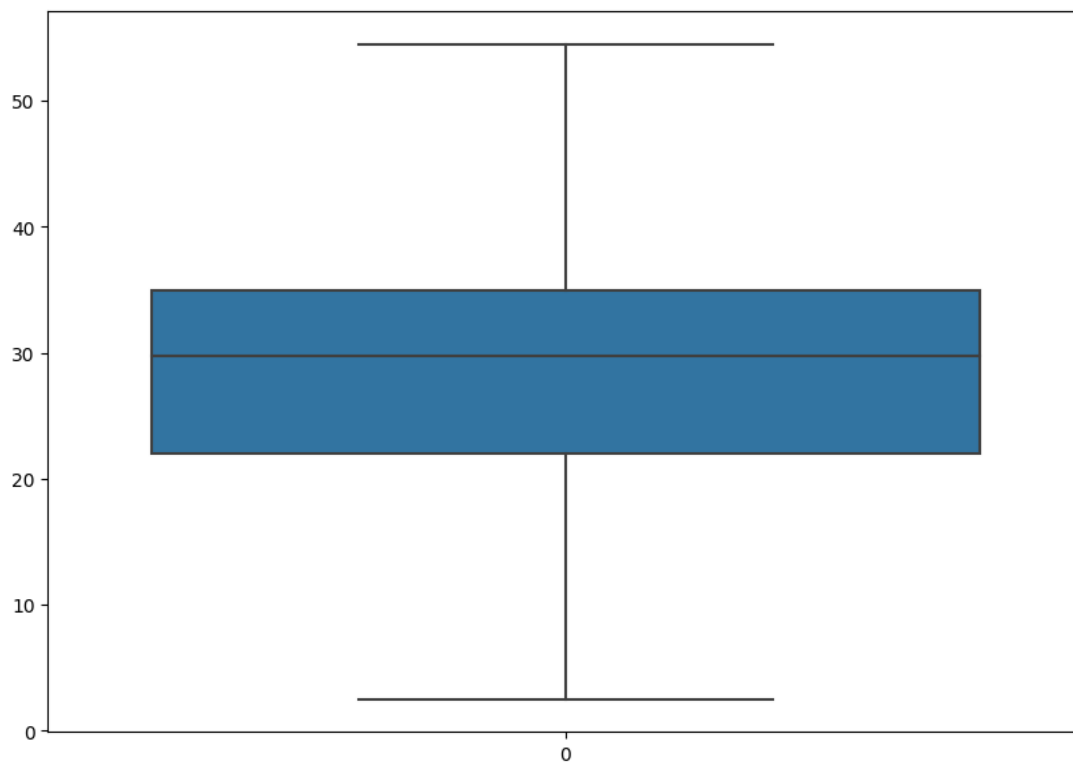
```
df[df['fare']>300]
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone
258	1	1	female	35.0	0	0	512.3292	C	First	woman	False	B	Cherbourg	yes	True
679	1	1	male	36.0	0	1	512.3292	C	First	man	True	B	Cherbourg	yes	False
737	1	1	male	35.0	0	0	512.3292	C	First	man	True	B	Cherbourg	yes	True

```
# Defining function for Outliers Treatment
def Outlier_Treatment(col):
    Q1 = df[col].quantile(0.25)
    Q3 = df[col].quantile(0.75)
    IQR = Q3 - Q1
    upper = Q3 + (1.5 * IQR)
    lower = Q1 - (1.5 * IQR)
    np.clip(df[col], lower, upper, inplace = True)
```

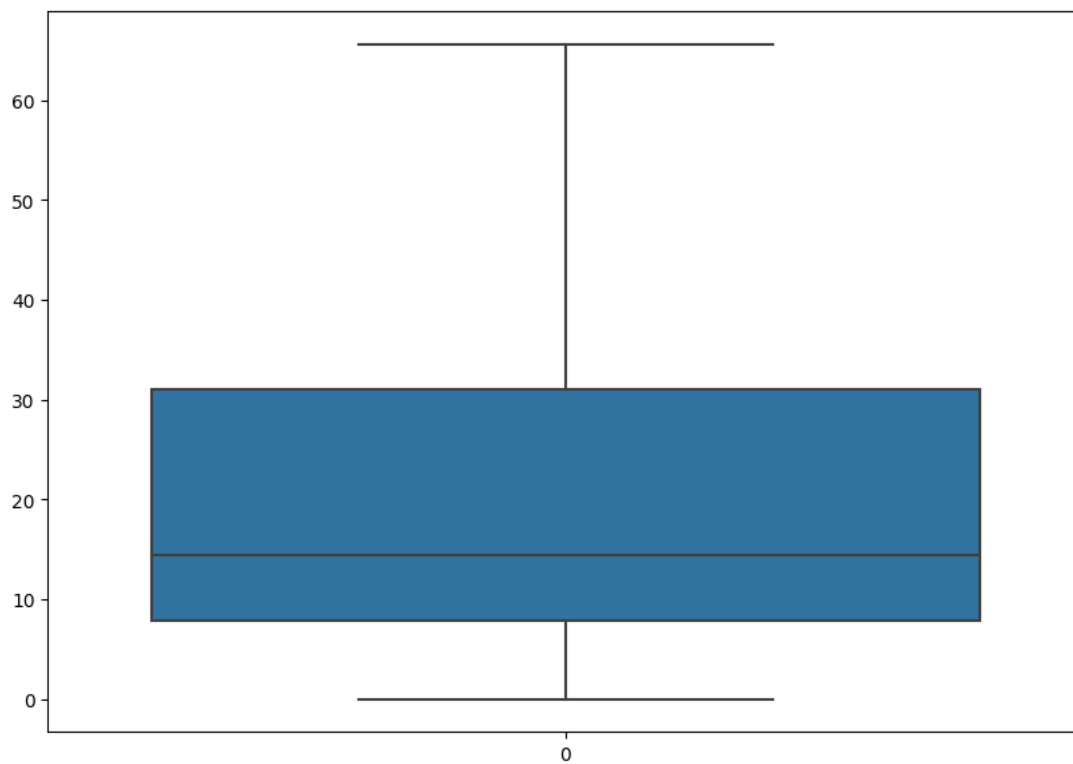
```
Outlier_Treatment('age')  
plt.figure(figsize = (10,7))  
sns.boxplot(df['age'])
```

<Axes: >



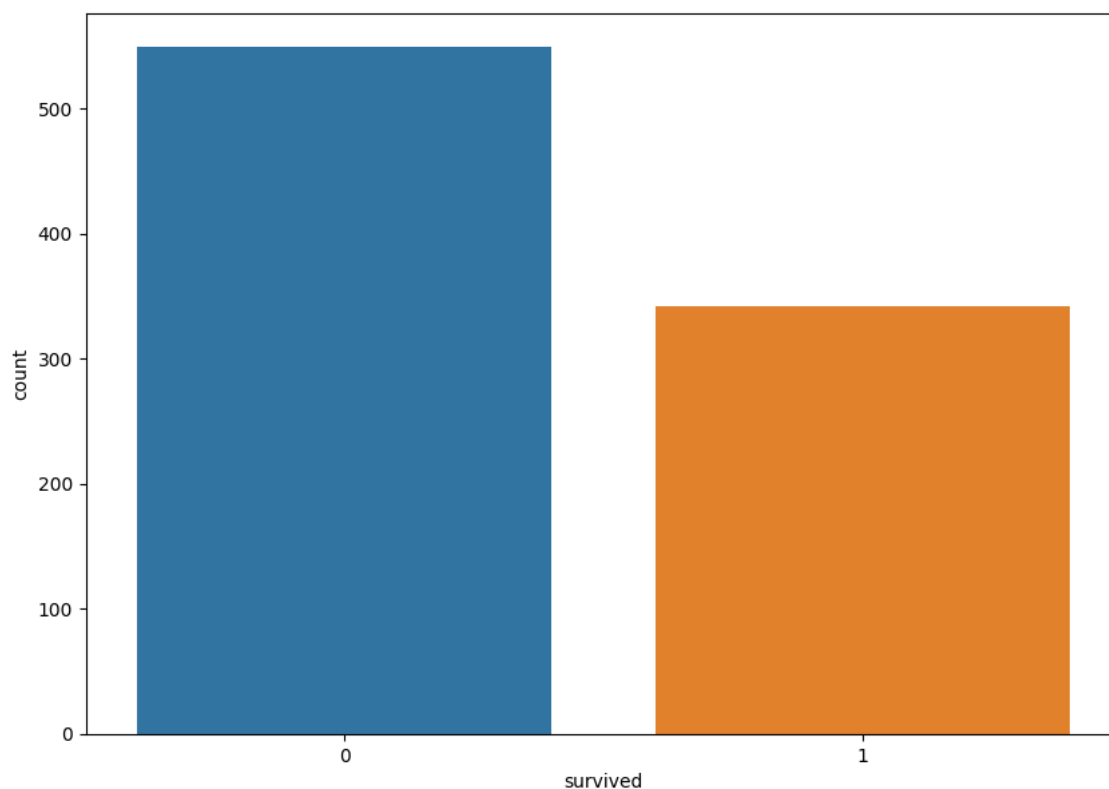
```
Outlier_Treatment('fare')  
plt.figure(figsize = (10,7))  
sns.boxplot(df['fare'])
```

<Axes: >



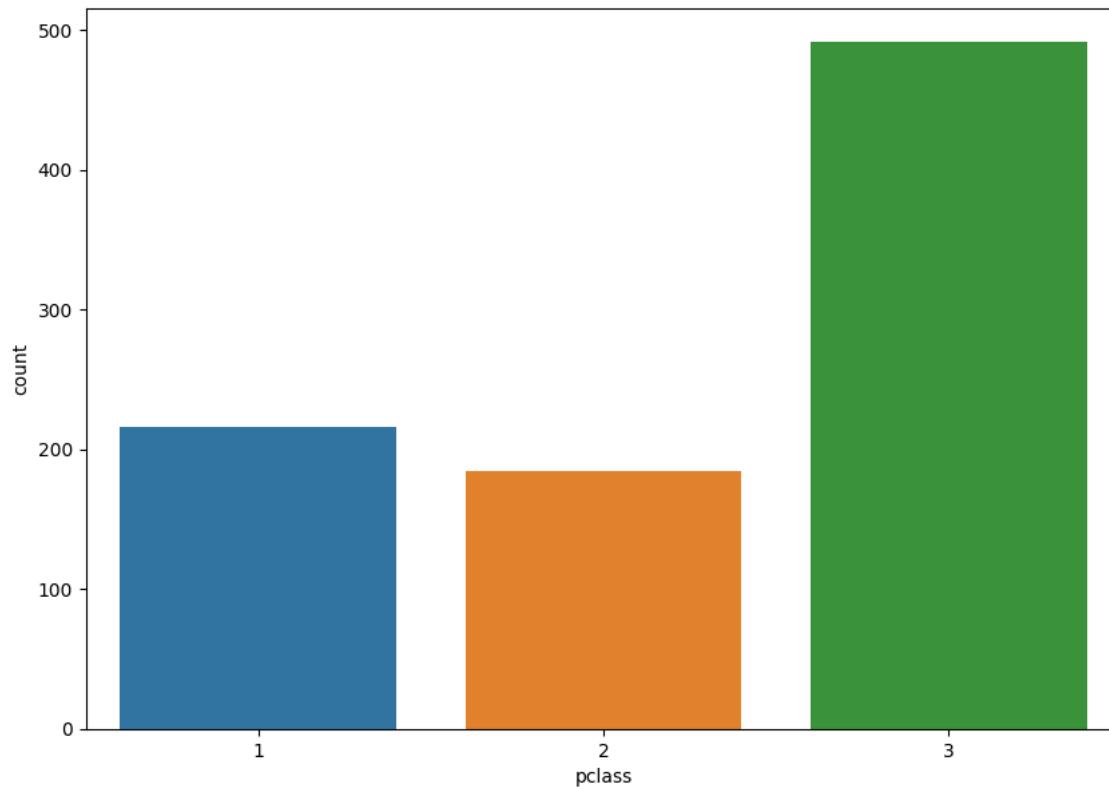
```
plt.figure(figsize = (10,7))  
sns.countplot(data = df, x = 'survived')
```

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



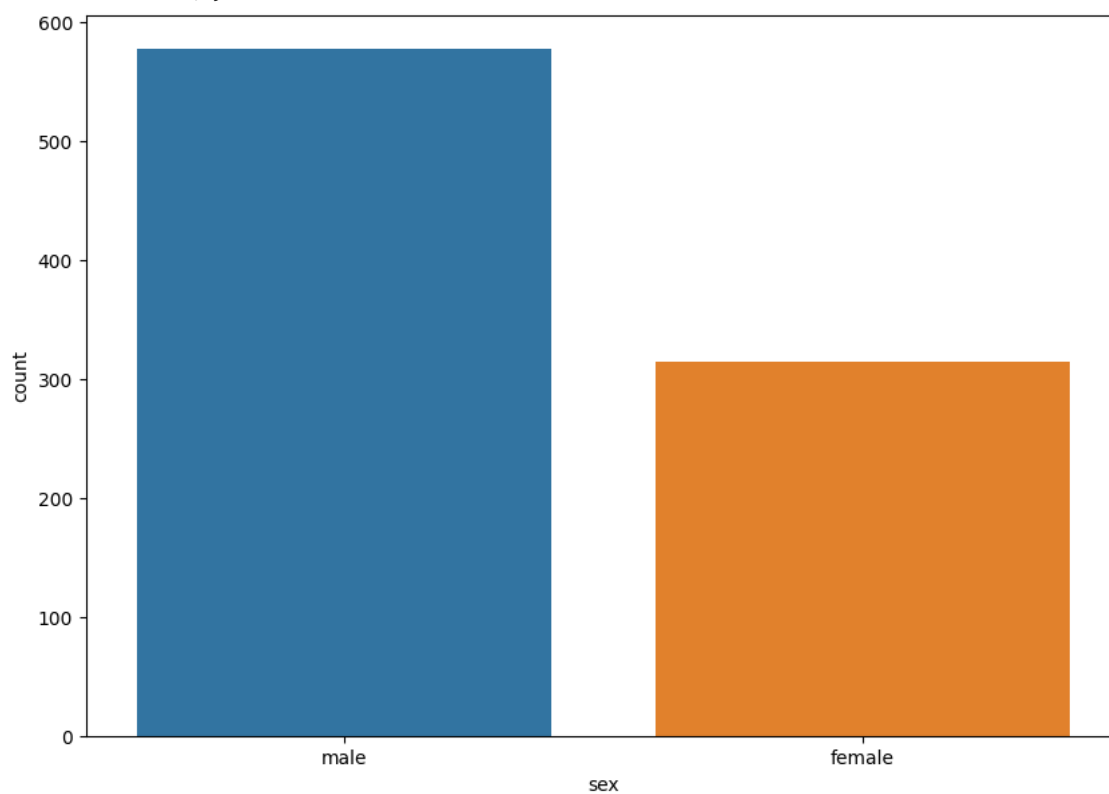
```
plt.figure(figsize = (10,7))  
sns.countplot(data = df, x = 'pclass')
```

<Axes: xlabel='pclass', ylabel='count'>



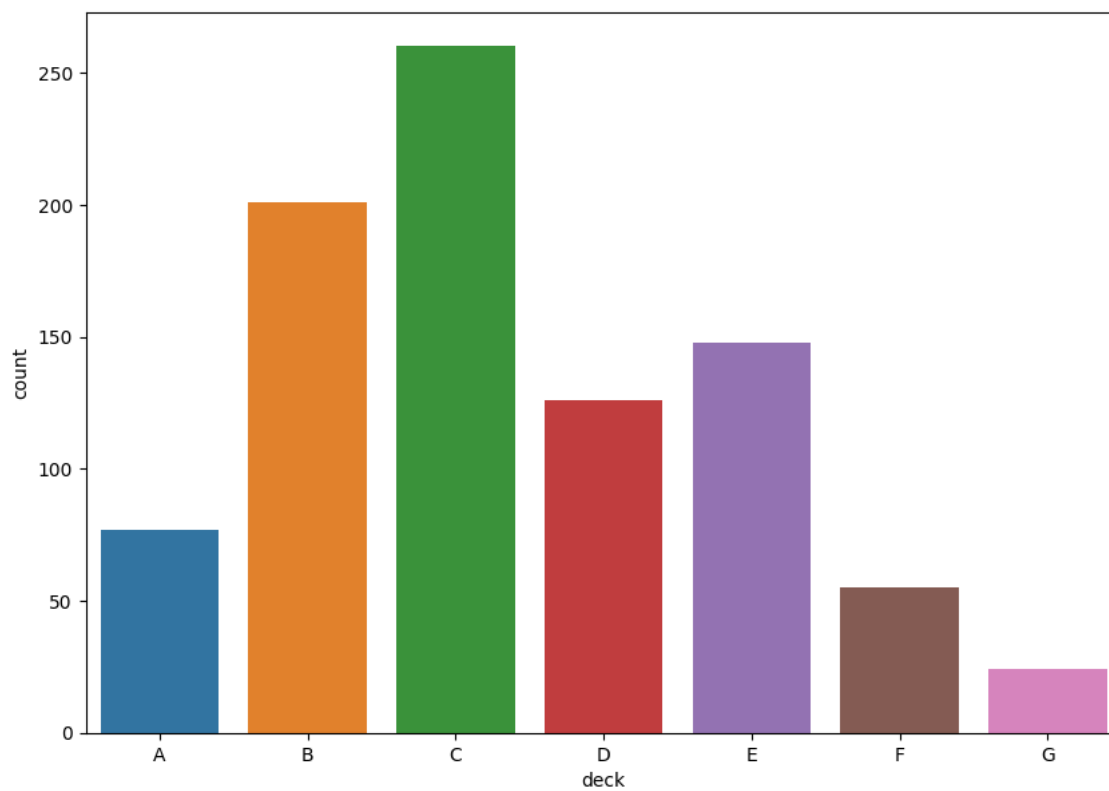
```
plt.figure(figsize = (10,7))  
sns.countplot(data = df, x = 'sex')
```

<Axes: xlabel='sex', ylabel='count'>



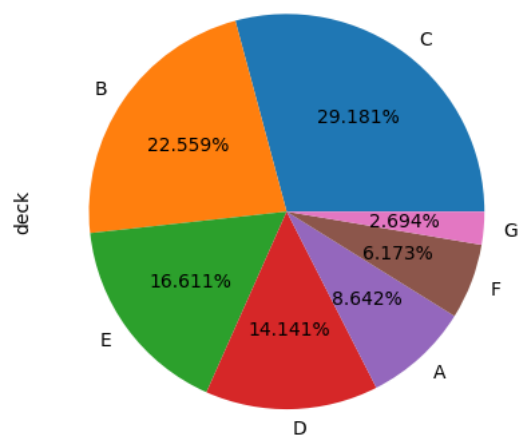
```
plt.figure(figsize = (10,7))  
sns.countplot(data = df, x = 'deck')
```

<Axes: xlabel='deck', ylabel='count'>



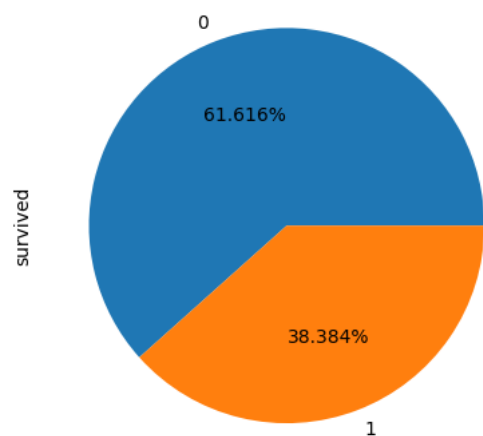
```
df['deck'].value_counts().plot(kind='pie', autopct='%0.3f%%')
```

<Axes: ylabel='deck'>



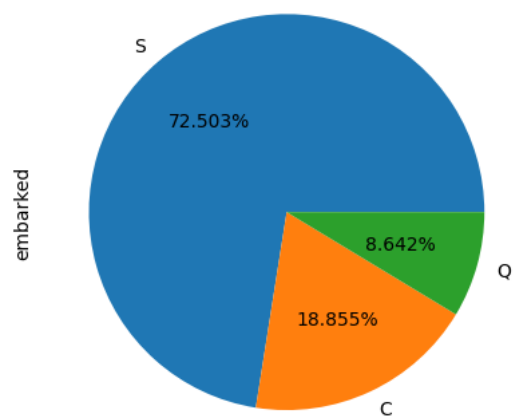
```
df['survived'].value_counts().plot(kind='pie', autopct='%0.3f%%')
```

<Axes: ylabel='survived'>

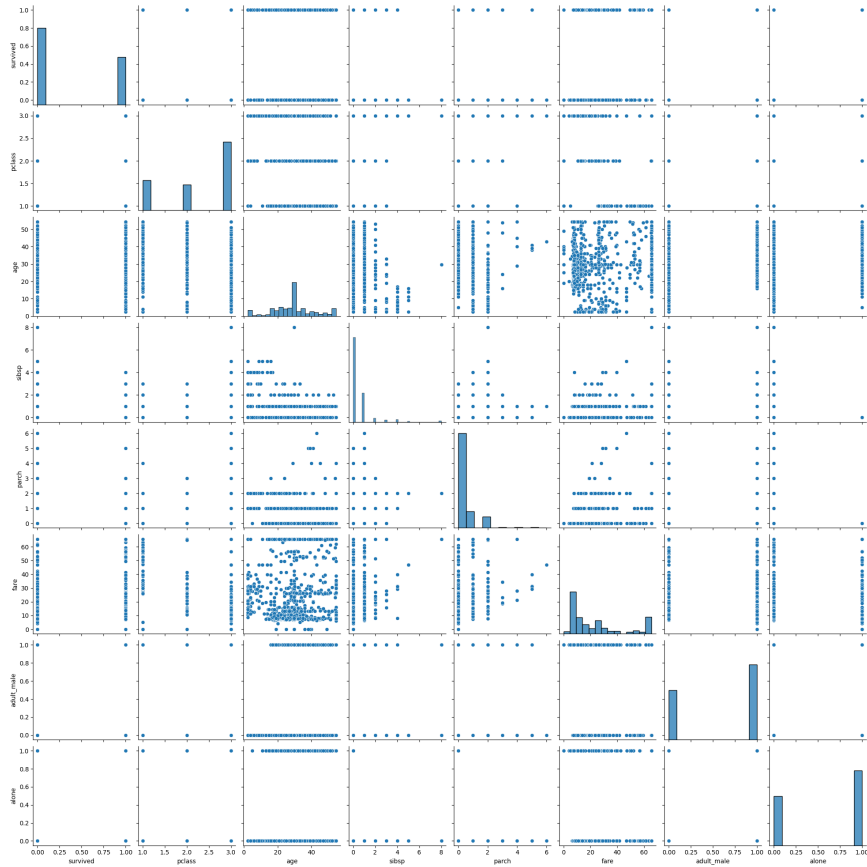


```
df['embarked'].value_counts().plot(kind='pie', autopct='%0.3f%%')
```

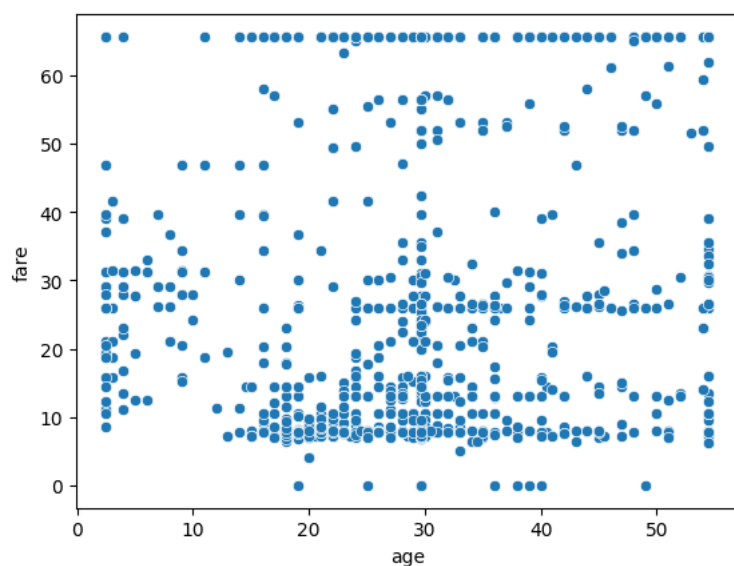
<Axes: ylabel='embarked'>



```
sns.pairplot(data=df)  
plt.show()
```

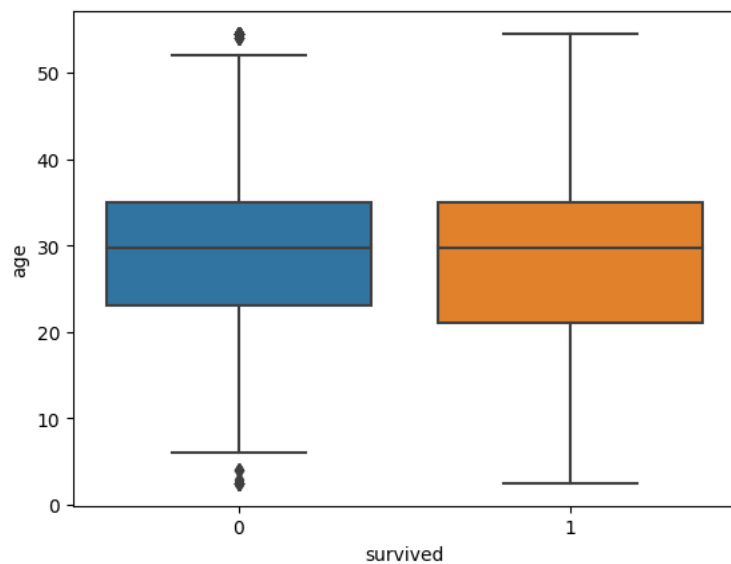


```
sns.scatterplot(x='age', y='fare', data=df)
plt.show()
```



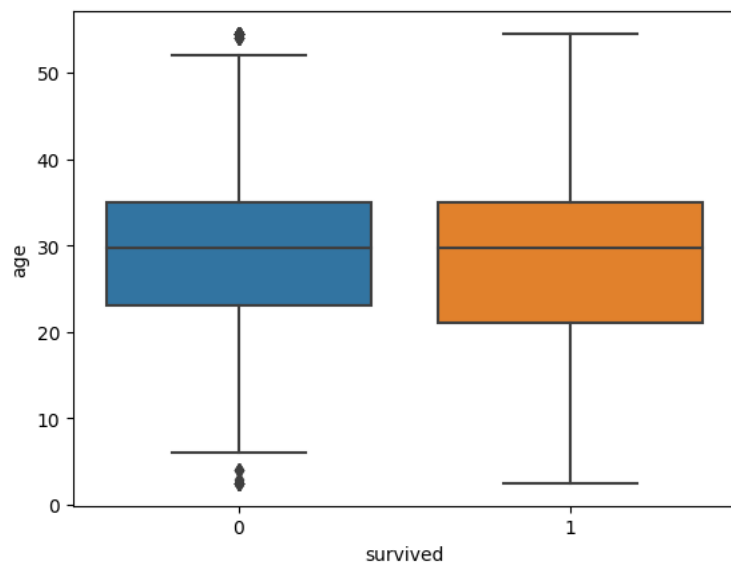
```
sns.boxplot(x='survived', y='age', data=df)
```

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



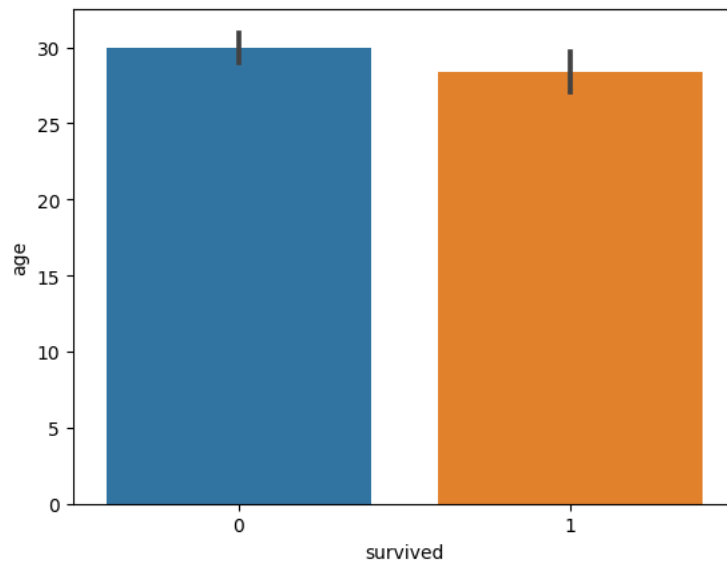
```
sns.boxplot(x='survived', y='age', data=df)
```

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



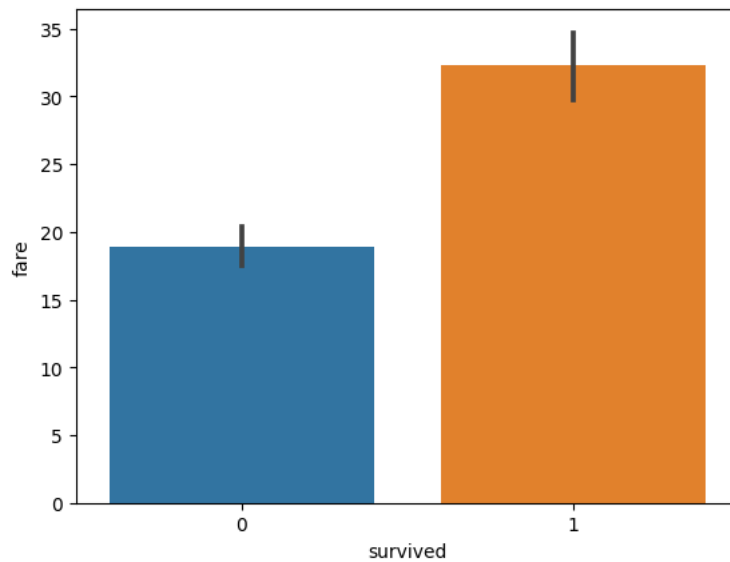
```
sns.barplot(x='survived', y='age',data=df)
```

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



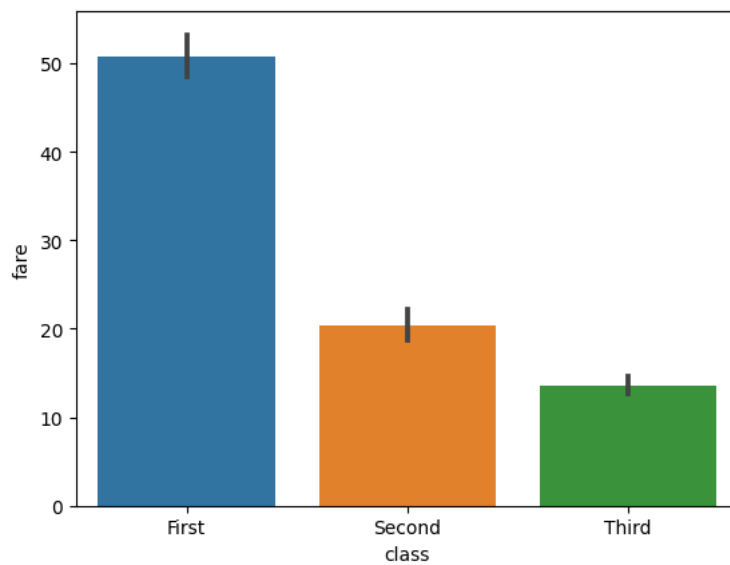
```
sns.barplot(x='survived', y='fare',data=df)
```

<Axes: xlabel='survived', ylabel='fare'>

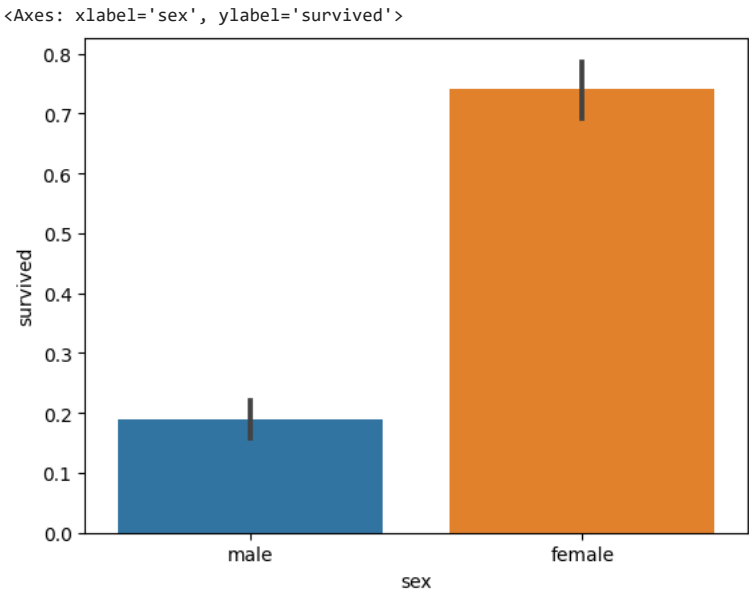


```
sns.barplot(x='class', y='fare',data=df)
```

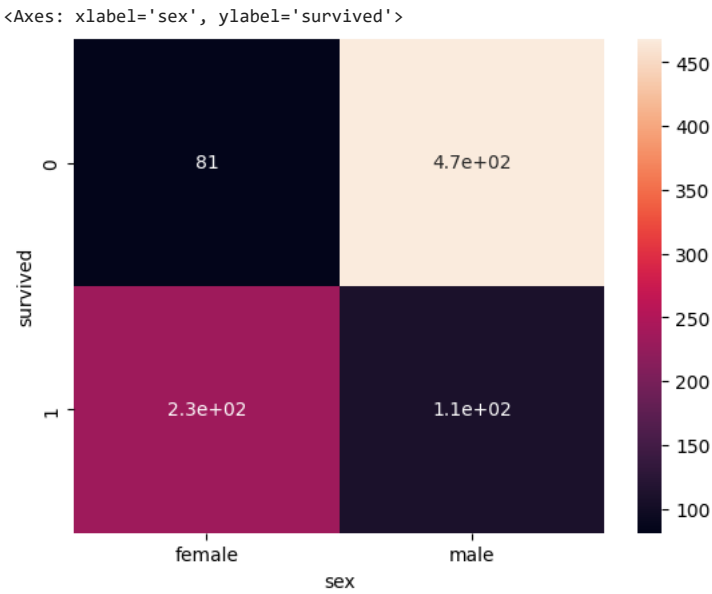
<Axes: xlabel='class', ylabel='fare'>



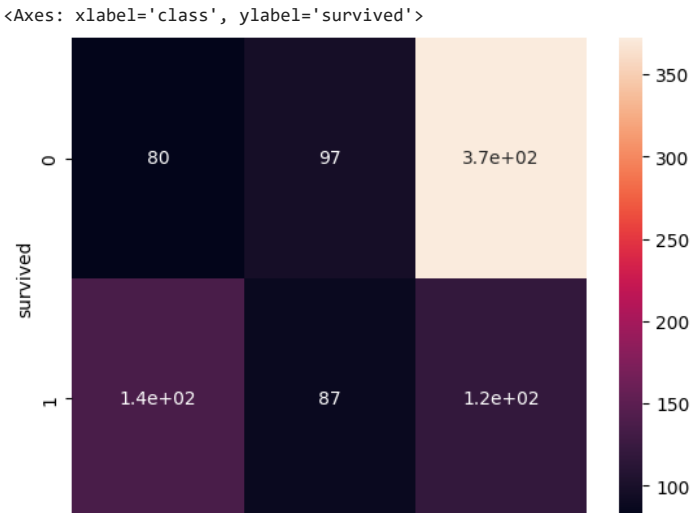

```
sns.barplot(x='sex', y='survived',data=df)
```



```
sns.heatmap(pd.crosstab(df['survived'],df['sex']), annot=True)
```



```
sns.heatmap(pd.crosstab(df['survived'],df['class']), annot=True)
```



```
sns.clustermap(pd.crosstab(df['survived'],df['class']), annot=True)
```

