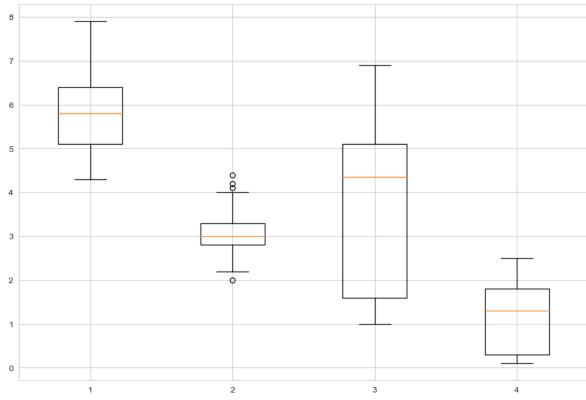
```
In [4]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
 In [5]: col_names = ['Sepal_Length','Sepal_Width','Petal_Length','Petal_Width','Species']
 In [6]: iris = pd.read_csv('C:/Users/Lenovo/Downloads/iris.csv',names=col_names)
In [7]:
          iris
 Out[7]:
               Sepal_Length Sepal_Width Petal_Length Petal_Width
                                                                     Species
            0
                                     3.5
                        5.1
                                                  1.4
                                                              0.2
                                                                    Iris-setosa
            1
                        4.9
                                     3.0
                                                  1.4
                                                              0.2
                                                                    Iris-setosa
            2
                        4.7
                                     3.2
                                                              0.2
                                                  1.3
                                                                   Iris-setosa
            3
                        4.6
                                     3.1
                                                  1.5
                                                              0.2
                                                                    Iris-setosa
            4
                        5.0
                                     3.6
                                                  1.4
                                                              0.2
                                                                   Iris-setosa
          145
                        6.7
                                     3.0
                                                  5.2
                                                              2.3 Iris-virginica
          146
                        6.3
                                     2.5
                                                  5.0
                                                              1.9
                                                                  Iris-virginica
          147
                        6.5
                                     3.0
                                                  5.2
                                                              2.0 Iris-virginica
          148
                        6.2
                                     3.4
                                                  5.4
                                                              2.3 Iris-virginica
          149
                        5.9
                                     3.0
                                                  5.1
                                                              1.8 Iris-virginica
         150 rows × 5 columns
 In [8]: len(list(iris))
Out[8]:
 In [9]: iris.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 150 entries, 0 to 149
          Data columns (total 5 columns):
               Column
                              Non-Null Count Dtype
               Sepal_Length 150 non-null
                                                float64
               Sepal_Width
                               150 non-null
                                                float64
           1
           2
               Petal_Length 150 non-null
                                                float64
               Petal Width
                               150 non-null
                                                float64
           3
                                                object
               Species
                               150 non-null
          dtypes: float64(4), object(1)
          memory usage: 6.0+ KB
          np.unique(iris["Species"])
In
[10]:
          array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
Out[10]:
In
          iris.describe()
[11]:
```

Sepal\_Length Sepal\_Width Petal\_Length Petal\_Width

Out[11]:

fig.add\_subplot(111)



```
In [15]: from scipy import stats
 In [16]: z = np.abs(stats.zscore(iris['Sepal_Length']))
 In [17]: print(z)
          0
                 0.900681
          1
                 1.143017
          2
                 1.385353
          3
                 1.506521
          4
                 1.021849
                   . . .
          145
                 1.038005
          146
                 0.553333
                 0.795669
          147
          148
                 0.432165
          149
                 0.068662
          Name: Sepal_Length, Length: 150, dtype:
                float64
 In [18]: threshold = 0.5
 In [19]: sample_outliers = np.where(z <threshold)</pre>
          sample_outliers
          (array([ 14, 15,
                             18,
                                 33,
                                      36, 53, 55, 61, 62, 63, 64, 66, 67,
Out[19]:
                        69, 70, 71, 73, 78, 79, 80, 81, 82, 83, 85, 88,
                   68,
                        90, 91, 92, 94, 95, 96, 97, 99, 101, 113, 114, 119,
                         121, 126, 127, 134, 138, 142, 148, 149],
                                      dtype=int64),)
 In [20]: sorted_rscore= sorted(iris['Sepal_Length'])
 In [21]: sorted_rscore
```

```
Out[21]: [4.3,
          4.4,
          4.4,
          4.4,
          4.5,
          4.6,
          4.6,
          4.6,
          4.6,
          4.7,
          4.7,
          4.8,
          4.8,
          4.8,
          4.8,
          4.8,
          4.9,
          4.9,
          4.9,
          4.9,
          4.9,
          4.9,
          5.5,
          5.5,
          5.6,
          5.6,
          5.6,
          5.6,
          5.6,
In [22]: q1 = np.percentile(sorted_rscore, 25)
          q3 = np.percentile(sorted_rscore, 75)
         print(q1,q3)
         5.1 6.4
In [23]: IQR = q3-q1
         IQR
In [24]:
         1.30000000000000007
Out[24]:
In [25]:
         lwr_bound = q1-
          (1.5*IQR) upr_bound =
          q3+(1.5*IQR)
         3.149999999999986 8.350000000000001
In [26]: r_outliers = []
          for i in sorted_rscore:
                  if (i<lwr_bound or</pre>
                          i>upr_bound):
                          r_outliers.append(i)
In [27]: print(r_outliers)
          []
In [28]: z = np.abs(stats.zscore(iris['Sepal_Width']))
In [29]: print(z)
```

```
0
                  1.032057
           1
                  0.124958
           2
                  0.337848
           3
                  0.106445
           4
                  1.263460
                   . . .
           145
                  0.124958
           146
                  1.281972
           147
                  0.124958
           148
                  0.800654
           149
                  0.124958
           Name: Sepal_Width, Length: 150, dtype:
                 float64
 In [30]:
           threshold = 0.5
           sample_outliers = np.where(z <threshold)</pre>
 In [31]:
           sample_outliers
                         2,
           (array([ 1,
                              3,
                                   8,
                                         9, 12, 13, 25, 29, 30, 34, 35,
                                                                                37,
Out[31]:
                    38, 42, 45, 47, 50, 51, 52, 58, 61, 63, 64, 65, 66,
                   70, 74, 75, 77, 78, 84, 86, 88, 91, 95, 96, 97, 102,
                   103, 104, 105, 107, 110, 112, 115, 116, 120, 125, 127, 129, 135,
                          137, 138, 139, 140, 141, 143, 145, 147, 149],
                                         dtype=int64),)
 In [32]: sorted_rscore= sorted(iris['Sepal_Width'])
 In [33]: sorted_rscore
 Out[33]: [2.0,
           2.2,
            2.2,
            2.2,
            2.3,
            2.3,
            2.3,
            2.3,
            2.4,
           2.4,
           2.4,
            2.5,
            2.5,
           2.5,
            2.5,
           2.5,
            2.5,
            2.5,
            2.5,
            2.6,
            2.6,
            2.6,
            2.6,
            2.6,
            2.7,
            2.7,
            2.7,
            2.7,
            2.7,
            2.7,
            2.7,
            2.7,
           2.7,
           2.8,
            2.8,
```

```
In [34]: q1 = np.percentile(sorted_rscore, 25)
           q3 = np.percentile(sorted_rscore, 75)
           print(q1,q3)
           2.8 3.3
In [35]: IQR = q3-q1
In [36]: IQR
           0.5
Out[36]:
In [37]: lwr_bound = q1-
            (1.5*IQR) upr_bound =
           q3+(1.5*IQR)
           2.05 4.05
In [38]: r_outliers = []
           for i in sorted_rscore:
                     if (i<lwr_bound or</pre>
                               i>upr_bound):
                               r_outliers.append(i)
           print(r_outliers)
           [2.0, 4.1, 4.2, 4.4]
In [39]: from scipy import stats
In [40]: z = np.abs(stats.zscore(iris['Petal_Length']))
In [41]: print(z)
                               1.341272
                       1
                               1.341272
                       2
                              1.398138
                       3
                               1.284407
                       4
                              1.341272
                       145
                              0.819624
                       146
                               0.705893
                       147
                              0.819624
                       148
                              0.933356
                       149
                              0.762759
                       Name: Petal_Length, Length: 150, dtype:
                             float64
              In [42]: threshold = 0.5
              In [43]: sample_outliers = np.where(z <threshold)</pre>
                        sample_outliers
                       (array([ 51, 53, 54, 55, 57, 58, 59, 60, 61, 62, 64, 65, 67, 68, 69, 71, 74, 75, 78, 79, 80, 81, 82, 84, 87, 88, 89, 90, 91, 92, 93, 94, 95, 96, 97, 98,
                                                                                            66,
             Out[43]:
                                                                                            85,
                               106], dtype=int64),)
              In [44]: sorted_rscore= sorted(iris['Petal_Length'])
              In [45]: q1 = np.percentile(sorted_rscore, 25)
                        q3 = np.percentile(sorted_rscore, 75)
                       print(q1,q3)
                        1.6 5.1
```

```
In [46]: IQR = q3-q1
             lwr_bound = q1-(1.5*IQR)
In [47]:
             upr\_bound = q3+(1.5*IQR)
             print(lwr_bound, upr_bound)
             -3.649999999999999999999999999999
In [60]: r_outliers = []
             for i in sorted_rscore:
                        if (i<lwr_bound or i>upr_bound):
                                  r_outliers.append(i)
             print(r_outliers)
             []
In [61]: print(r_outliers)
             []
In [62]: z = np.abs(stats.zscore(iris['Petal_Width']))
In [63]: print(z)
                            1.312977
                              1.312977
                       2
                              1.312977
                        3
                              1.312977
                             1.312977
                        145
                             1.447956
                       146
                              0.922064
                        147
                              1.053537
                            1.447956
                        148
                        149
                             0.790591
                       Name: Petal_Width, Length: 150, dtype: float64
               In [64]: threshold = 0.5
               In [65]: sample_outliers = np.where(z <threshold)</pre>
                        sample_outliers
                       (array([ 50, 51, 52, 53, 54, 55, 57, 58, 59, 60, 61, 62, 63, 64, 65, 66, 67, 68, 69, 71, 72, 73, 74, 75, 76, 78, 79, 80, 81, 82, 84, 86, 87, 88, 89, 90, 91, 92, 93,
              Out[65]:
                               94, 95, 96, 97, 98, 99, 119, 133, 134], dtype=int64),)
               In [66]: sorted_rscore= sorted(iris['Petal_Width'])
               In [67]: sorted_rscore
```

```
Out[67]: [0.1,
           0.1,
           0.1,
           0.1,
           0.1,
           0.1,
           0.2,
           0.2,
           1.1,
           1.2,
           1.2,
           1.2,
           1.2,
           1.8,
           1.8,
           1.8,
           1.8,
           1.8,
           1.8,
           1.9,
           1.9,
           1.9,
           2.0,
           2.0,
           2.0,
           2.0,
           2.0,
           2.0,
           2.1,
           2.3,
           2.3,
           2.3,
           2.3,
           2.3,
           2.3,
           2.3,
           2.4,
           2.4,
           2.4,
           2.5,
           2.5,
           2.5]
In [68]: q1 = np.percentile(sorted_rscore, 25)
          q3 = np.percentile(sorted_rscore, 75)
          print(q1,q3)
          0.3 1.8
In [69]: IQR = q3-q1
In [70]:
          lwr_bound = q1-
          (1.5*IQR) upr_bound =
          q3+(1.5*IQR)
          print(lwr_bound, upr_bound)
          -1.95 4.05
          r_outliers = []
In [71]:
          for i in sorted_rscore:
                  if (i<lwr_bound or</pre>
                           i>upr_bound):
                           r_outliers.append(i)
          print(r_outliers)
```

[]

```
import pandas as pd
import numpy as np

import matplotlib.pyplot as plt
import seaborn as sns

dataset = sns.load_dataset('titanic')

dataset.head()
```

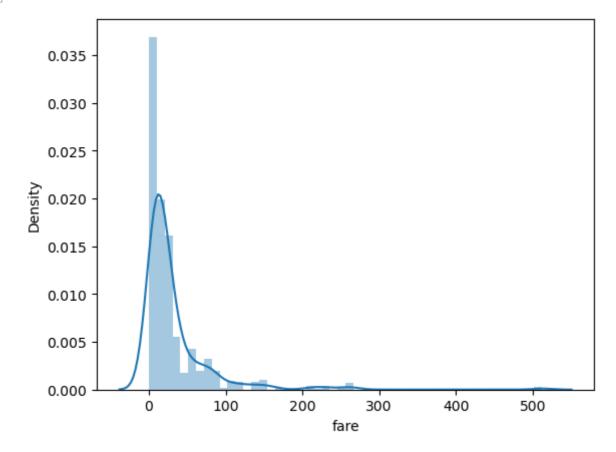
Out[3]:		survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	de
	0	0	3	male	22.0	1	0	7.2500	S	Third	man	True	Ni
	1	1	1	female	38.0	1	0	71.2833	С	First	woman	False	
	2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False	Ni
	3	1	1	female	35.0	1	0	53.1000	S	First	woman	False	
	4	0	3	male	35.0	0	0	8.0500	S	Third	man	True	Ni

In [4]: sns.distplot(dataset['fare'])

C:\Users\Lenovo\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureW arning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

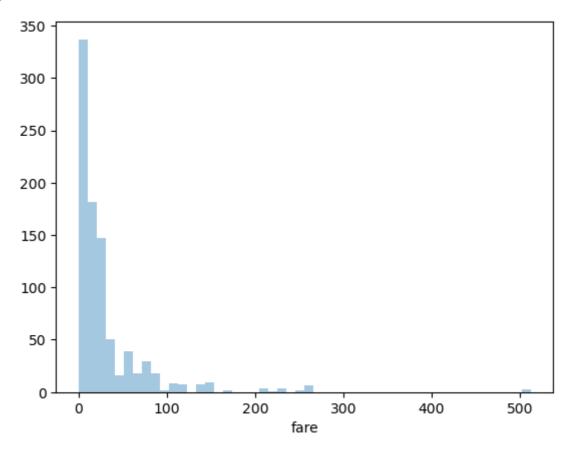
warnings.warn(msg, FutureWarning)

Out[4]: <AxesSubplot:xlabel='fare', ylabel='Density'>



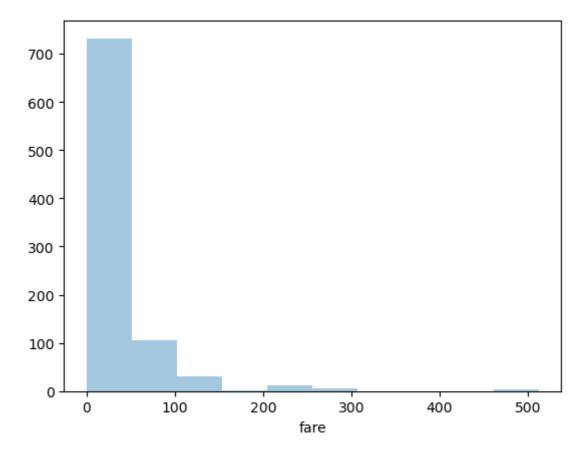
C:\Users\Lenovo\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureW arning: `distplot` is a deprecated function and will be removed in a future versio n. Please adapt your code to use either `displot` (a figure-level function with si milar flexibility) or `histplot` (an axes-level function for histograms). warnings.warn(msg, FutureWarning)

<AxesSubplot:xlabel='fare'> Out[5]:



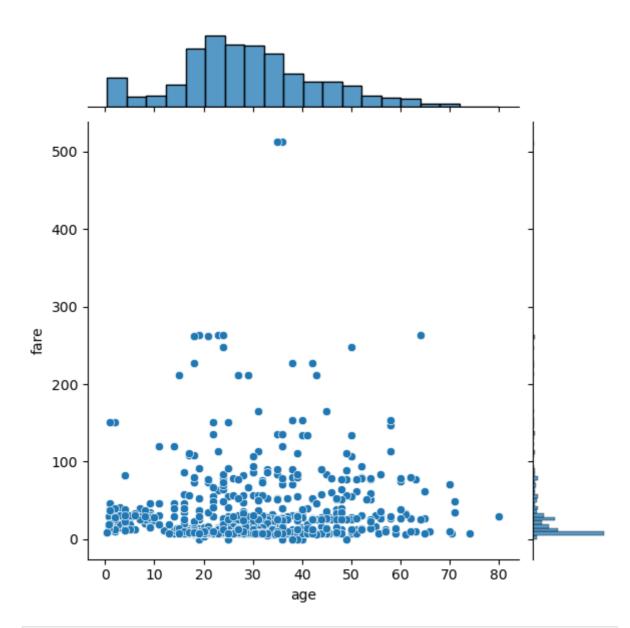
sns.distplot(dataset['fare'], kde=False, bins=10) In [6]:

<AxesSubplot:xlabel='fare'> Out[6]:



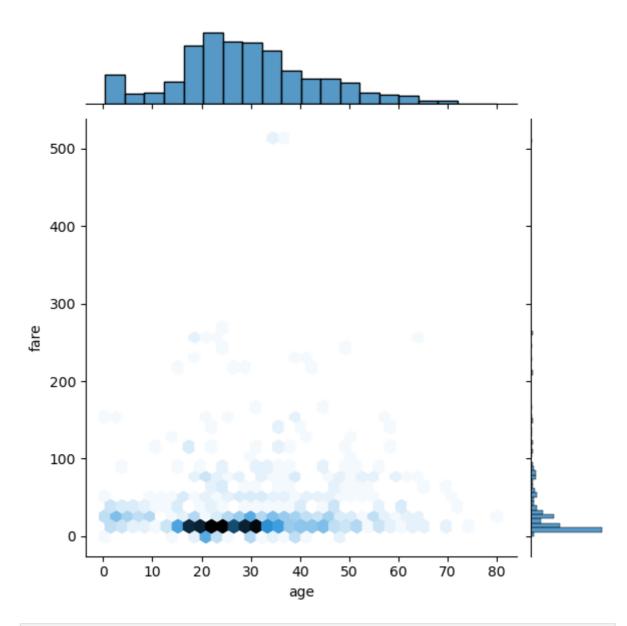
In [7]: sns.jointplot(x='age', y='fare', data=dataset)

Out[7]: <seaborn.axisgrid.JointGrid at 0x27a9bc08ac0>



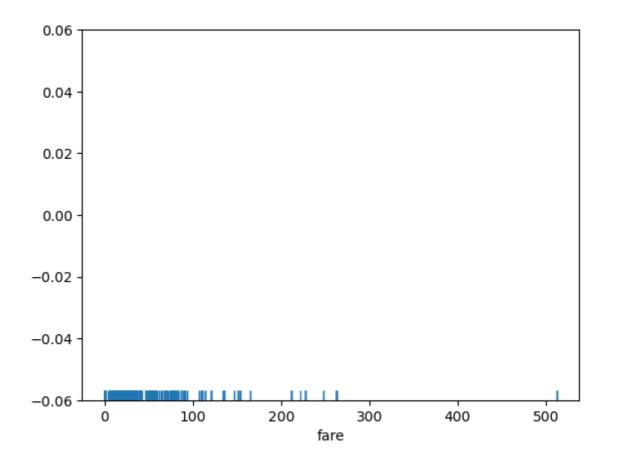
In [8]: sns.jointplot(x='age', y='fare', data=dataset, kind='hex')

Out[8]: <seaborn.axisgrid.JointGrid at 0x27a9c0adeb0>



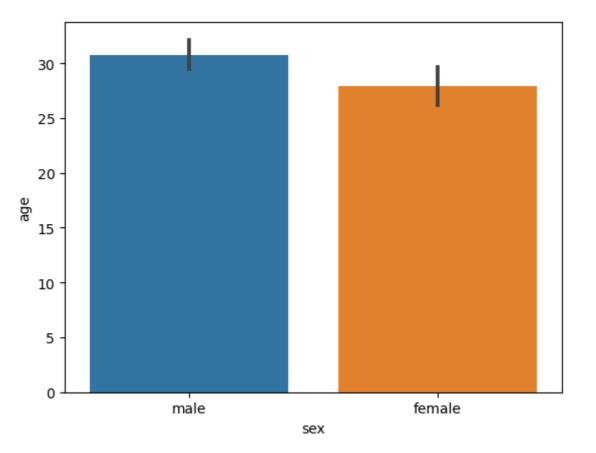
In [9]: sns.rugplot(dataset['fare'])

Out[9]: <AxesSubplot:xlabel='fare'>



In [10]: sns.barplot(x='sex', y='age', data=dataset)

Out[10]: <AxesSubplot:xlabel='sex', ylabel='age'>



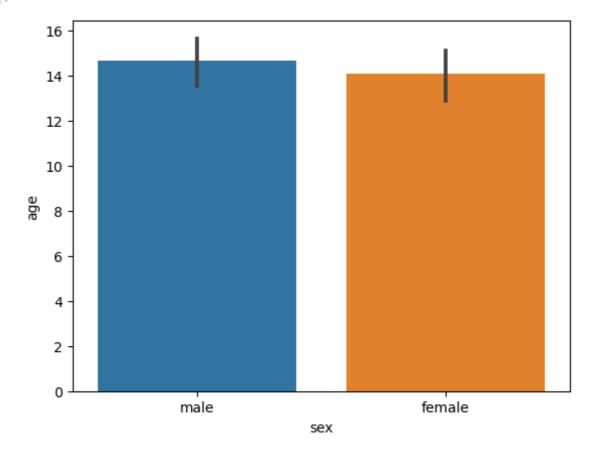
In [11]: import numpy as np

import mathletlib nyplot as nlt

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js

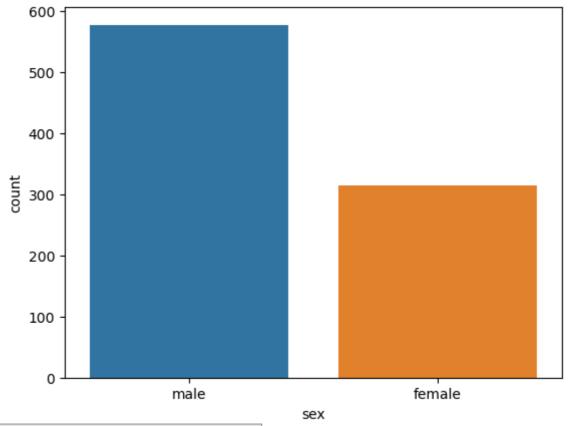
```
sns.barplot(x='sex', y='age', data=dataset, estimator=np.std)
```

Out[11]: <AxesSubplot:xlabel='sex', ylabel='age'>



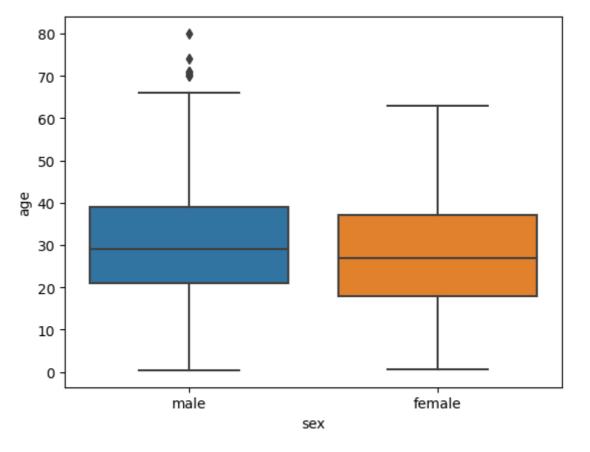
In [12]: sns.countplot(x='sex', data=dataset)

Out[12]: <AxesSubplot:xlabel='sex', ylabel='count'>



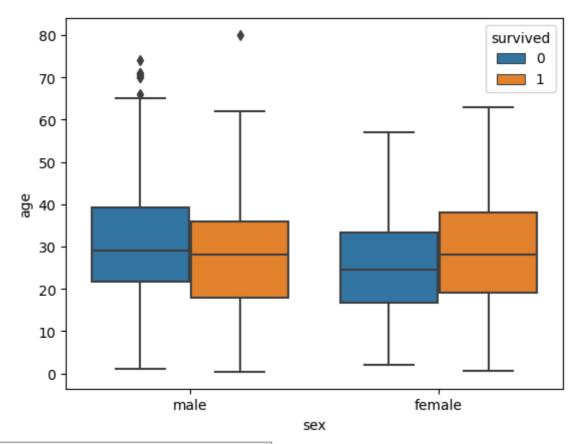
```
In [13]: sns.boxplot(x='sex', y='age', data=dataset)
```

Out[13]: <AxesSubplot:xlabel='sex', ylabel='age'>



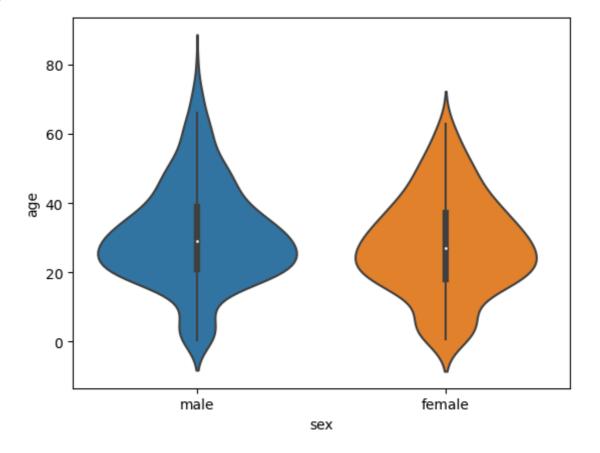
In [14]: sns.boxplot(x='sex', y='age', data=dataset, hue="survived")

Out[14]: <AxesSubplot:xlabel='sex', ylabel='age'>

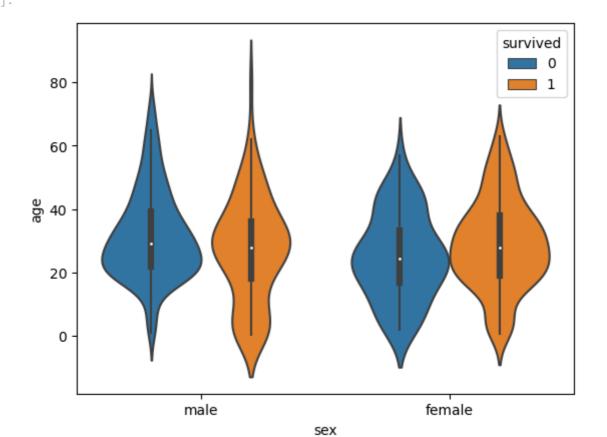


```
In [15]: sns.violinplot(x='sex', y='age', data=dataset)
```

Out[15]: <AxesSubplot:xlabel='sex', ylabel='age'>

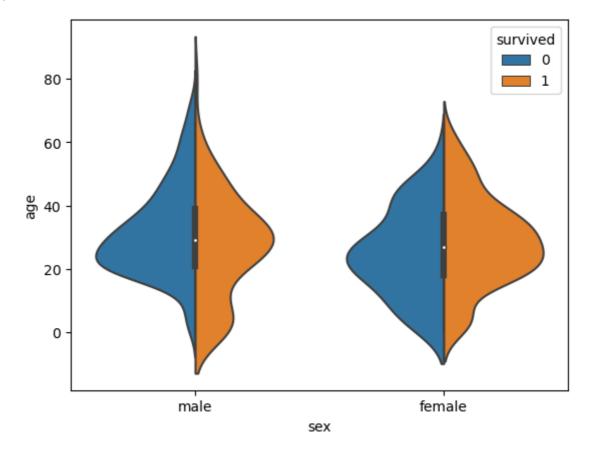


In [16]: sns.violinplot(x='sex', y='age', data=dataset, hue='survived')
Out[16]: <AxesSubplot:xlabel='sex', ylabel='age'>



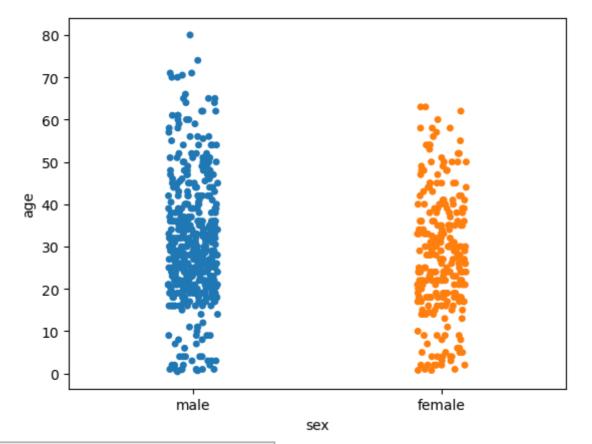
```
In [17]: sns.violinplot(x='sex', y='age', data=dataset, hue='survived', split=True)
```

Out[17]: <AxesSubplot:xlabel='sex', ylabel='age'>



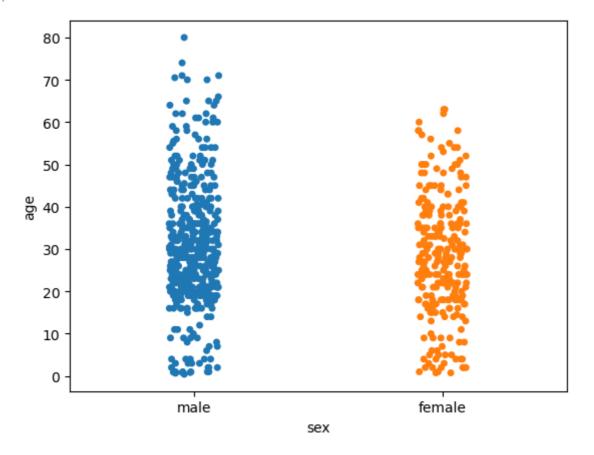
In [18]: sns.stripplot(x='sex', y='age', data=dataset)

Out[18]: <AxesSubplot:xlabel='sex', ylabel='age'>

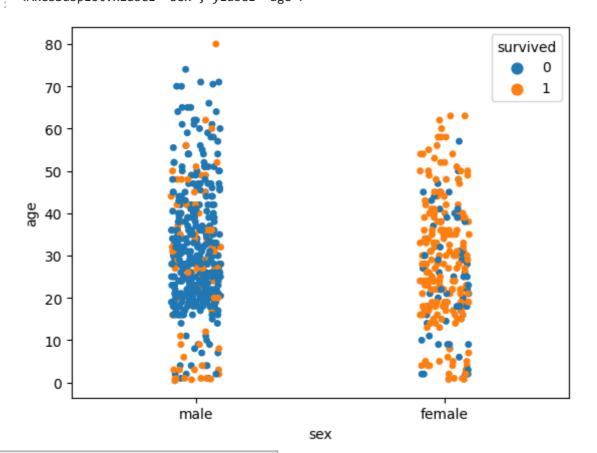


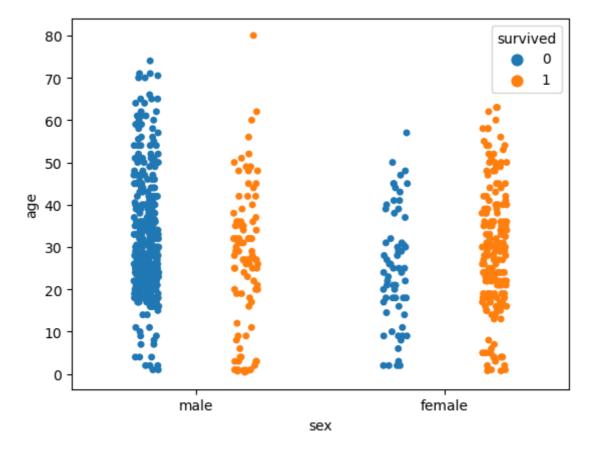
```
In [19]: sns.stripplot(x='sex', y='age', data=dataset, jitter=True)
```

Out[19]: <AxesSubplot:xlabel='sex', ylabel='age'>



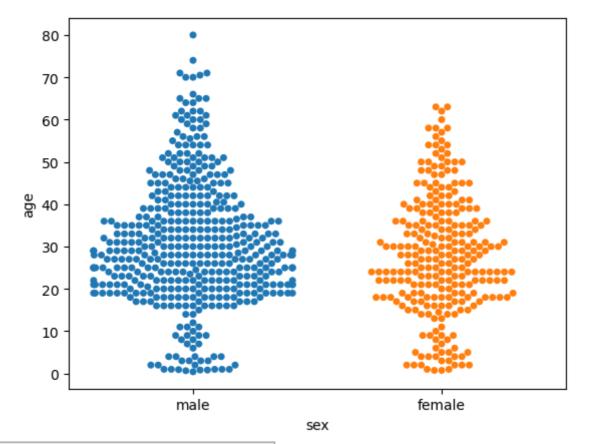
In [20]: sns.stripplot(x='sex', y='age', data=dataset, jitter=True, hue='survived')
Out[20]: <AxesSubplot:xlabel='sex', ylabel='age'>





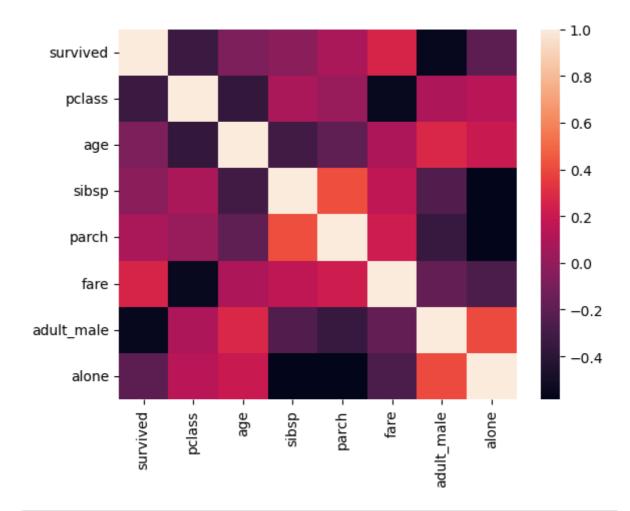


Out[25]: <AxesSubplot:xlabel='sex', ylabel='age'>



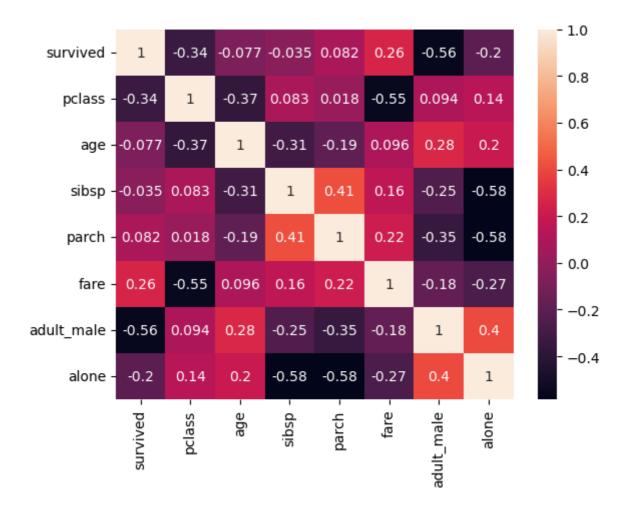
```
sns.matrixplot(x='sex', y='age', data=dataset)
In [26]:
           AttributeError
                                                          Traceback (most recent call last)
           ~\AppData\Local\Temp\ipykernel_14272\2259532084.py in <module>
           ----> 1 sns.matrixplot(x='sex', y='age', data=dataset)
          AttributeError: module 'seaborn' has no attribute 'matrixplot'
In [27]:
          dataset.corr()
Out[27]:
                       survived
                                    pclass
                                                         sibsp
                                                                   parch
                                                                               fare adult_male
                                                                                                    alone
                                                age
                      1.000000
             survived
                                -0.338481
                                           -0.077221
                                                     -0.035322
                                                                0.081629
                                                                           0.257307
                                                                                      -0.557080 -0.203367
               pclass
                      -0.338481
                                 1.000000
                                           -0.369226
                                                      0.083081
                                                                0.018443
                                                                          -0.549500
                                                                                      0.094035
                                                                                                 0.135207
                 age
                      -0.077221
                                 -0.369226
                                            1.000000
                                                     -0.308247
                                                                -0.189119
                                                                           0.096067
                                                                                      0.280328
                                                                                                 0.198270
                      -0.035322
                                 0.083081
                                           -0.308247
                                                      1.000000
                                                                0.414838
                                                                           0.159651
                                                                                      -0.253586
                                                                                               -0.584471
                sibsp
                       0.081629
                                                                           0.216225
                                                                                                -0.583398
               parch
                                 0.018443
                                           -0.189119
                                                      0.414838
                                                                1.000000
                                                                                      -0.349943
                 fare
                       0.257307
                                -0.549500
                                            0.096067
                                                      0.159651
                                                                0.216225
                                                                           1.000000
                                                                                      -0.182024 -0.271832
           adult_male -0.557080
                                 0.094035
                                            0.280328
                                                     -0.253586
                                                               -0.349943
                                                                         -0.182024
                                                                                      1.000000
                                                                                                 0.404744
                alone
                     -0.203367
                                 0.135207
                                            0.198270
                                                     -0.584471
                                                               -0.583398
                                                                         -0.271832
                                                                                      0.404744
                                                                                                 1.000000
           corr=dataset.corr()
In [28]:
In [31]:
           sns.heatmap(corr)
           <AxesSubplot:>
```

Out[31]:



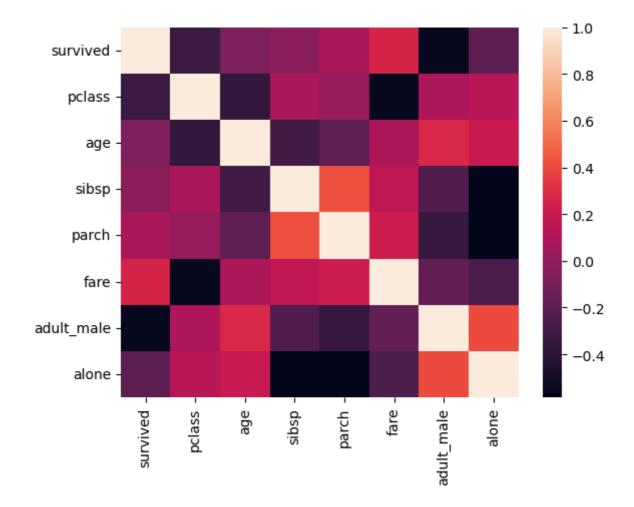
In [33]: sns.heatmap(corr, annot=True)

Out[33]: <AxesSubplot:>



In [34]: sns.heatmap(corr)

Out[34]: <AxesSubplot:>



In [ ]: