```
!pip install matplot
Requirement already satisfied: matplot in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (0.1.9)
Requirement already satisfied: pyloco>=0.0.134 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplot) (0.0.139)
Requirement already satisfied: matplotlib>=3.1.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplot) (3.7.3)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (1.1.0)
Requirement already satisfied: cycler>=0.10 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from
matplotlib >= 3.1.1 -> matplot) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (4.42.1)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\ilove\
appdata\local\programs\pvthon\pvthon311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (1.4.5)
Requirement already satisfied: numpy<2,>=1.20 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (1.25.2)
Requirement already satisfied: packaging>=20.0 in c:\users\ilove\
appdata\local\programs\pvthon\pvthon311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (23.1)
Requirement already satisfied: pillow>=6.2.0 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (10.0.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (3.1.1)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
matplotlib>=3.1.1->matplot) (2.8.2)
Requirement already satisfied: ushlex in c:\users\ilove\appdata\local\
programs\python\python311\lib\site-packages (from pyloco>=0.0.134-
>matplot) (0.99.1)
Requirement already satisfied: websocket-client in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
pyloco >= 0.0.134 -> matplot) (1.6.2)
Requirement already satisfied: twine in c:\users\ilove\appdata\local\
programs\python\python311\lib\site-packages (from pyloco>=0.0.134-
>matplot) (4.0.2)
Requirement already satisfied: typing in c:\users\ilove\appdata\local\
programs\python\python311\lib\site-packages (from pyloco>=0.0.134-
>matplot) (3.7.4.3)
Requirement already satisfied: SimpleWebSocketServer in c:\users\
```

```
ilove\appdata\local\programs\python\python311\lib\site-packages (from
pyloco >= 0.0.134 -> matplot) (0.1.2)
Requirement already satisfied: six>=1.5 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from python-
dateutil \ge 2.7 - matplotlib \ge 3.1.1 - matplot) (1.16.0)
Requirement already satisfied: pkginfo>=1.8.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from twine-
>pyloco>=0.0.134->matplot) (1.9.6)
Requirement already satisfied: readme-renderer>=35.0 in c:\users\
ilove\appdata\local\programs\python\python311\lib\site-packages (from
twine->pyloco>=0.0.134->matplot) (42.0)
Requirement already satisfied: requests>=2.20 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from twine-
>pyloco>=0.0.134->matplot) (2.31.0)
Requirement already satisfied: requests-toolbelt!=0.9.0,>=0.8.0 in c:\
users\ilove\appdata\local\programs\python\python311\lib\site-packages
(from twine->pyloco>=0.0.134->matplot) (1.0.0)
Requirement already satisfied: urllib3>=1.26.0 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from twine-
>pyloco>=0.0.134->matplot) (2.0.4)
Requirement already satisfied: importlib-metadata>=3.6 in c:\users\
ilove\appdata\local\programs\python\python311\lib\site-packages (from
twine->pyloco>=0.0.134->matplot) (6.8.0)
Requirement already satisfied: keyring>=15.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from twine-
>pyloco>=0.0.134->matplot) (24.2.0)
Requirement already satisfied: rfc3986>=1.4.0 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from twine-
>pyloco>=0.0.134->matplot) (2.0.0)
Requirement already satisfied: rich>=12.0.0 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from twine-
>pyloco>=0.0.134->matplot) (13.5.2)
Requirement already satisfied: zipp>=0.5 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from importlib-
metadata>=3.6->twine->pyloco>=0.0.134->matplot) (3.16.2)
Requirement already satisfied: jaraco.classes in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
keyring = 15.1 - twine - pyloco = 0.0.134 - matplot) (3.3.0)
Requirement already satisfied: pywin32-ctypes>=0.2.0 in c:\users\
ilove\appdata\local\programs\python\python311\lib\site-packages (from
keyring >= 15.1 - twine - pyloco >= 0.0.134 - matplot) (0.2.2)
Requirement already satisfied: nh3>=0.2.14 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from readme-
renderer>=35.0->twine->pyloco>=0.0.134->matplot) (0.2.14)
Requirement already satisfied: docutils>=0.13.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
readme-renderer>=35.0->twine->pyloco>=0.0.134->matplot) (0.20.1)
Requirement already satisfied: Pygments>=2.5.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
```

```
readme-renderer>=35.0->twine->pyloco>=0.0.134->matplot) (2.16.1)
Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\
ilove\appdata\local\programs\python\python311\lib\site-packages (from
reguests>=2.20->twine->pyloco>=0.0.134->matplot) (3.2.0)
Requirement already satisfied: idna<4,>=2.5 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from
reguests>=2.20->twine->pyloco>=0.0.134->matplot) (3.4)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
requests>=2.20->twine->pyloco>=0.0.134->matplot) (2023.7.22)
Requirement already satisfied: markdown-it-py>=2.2.0 in c:\users\
ilove\appdata\local\programs\python\python311\lib\site-packages (from
rich>=12.0.0->twine->pyloco>=0.0.134->matplot) (3.0.0)
Requirement already satisfied: mdurl~=0.1 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from markdown-it-
py = 2.2.0 - rich = 12.0.0 - twine - pyloco = 0.0.134 - matplot) (0.1.2)
Requirement already satisfied: more-itertools in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
jaraco.classes->keyring>=15.1->twine->pyloco>=0.0.134->matplot)
(10.1.0)
!pip install scikit-learn
Collecting scikit-learn
 Obtaining dependency information for scikit-learn from
https://files.pythonhosted.org/packages/77/85/bff3a1e818ec6aa3dd466ff4
f4b0a727db9fdb41f2e849747ad902ddbe95/scikit learn-1.3.0-cp311-cp311-
win amd64.whl.metadata
 Downloading scikit learn-1.3.0-cp311-cp311-win amd64.whl.metadata
(11 \text{ kB})
Requirement already satisfied: numpy>=1.17.3 in c:\users\ilove\
appdata\local\programs\pvthon\pvthon311\lib\site-packages (from
scikit-learn) (1.25.2)
Requirement already satisfied: scipy>=1.5.0 in c:\users\ilove\appdata\
local\programs\python\python311\lib\site-packages (from scikit-learn)
(1.11.2)
Requirement already satisfied: joblib>=1.1.1 in c:\users\ilove\
appdata\local\programs\python\python311\lib\site-packages (from
scikit-learn) (1.1.1)
Collecting threadpoolctl>=2.0.0 (from scikit-learn)
 Obtaining dependency information for threadpoolctl>=2.0.0 from
https://files.pythonhosted.org/packages/81/12/fd4dea011af9d69e1cad05c7
5f3f7202cdcbeac9b712eea58ca779a72865/threadpoolctl-3.2.0-py3-none-
any.whl.metadata
 Downloading threadpoolctl-3.2.0-py3-none-any.whl.metadata (10.0 kB)
Downloading scikit learn-1.3.0-cp311-cp311-win amd64.whl (9.2 MB)
   ----- 0.0/9.2 MB ? eta -:--:--
    ----- 0.2/9.2 MB 3.5 MB/s eta
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  -- ----- 0.5/9.2 MB 5.7 MB/s eta
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0:00:02
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 ----- 2.0/9.2 MB 8.3 MB/s eta
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 ----- 2.5/9.2 MB 8.9 MB/s eta
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 ----- 3.3/9.2 MB 8.9 MB/s eta
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 ----- 4.0/9.2 MB 9.0 MB/s eta
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 ----- 4.4/9.2 MB 9.1 MB/s eta
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 ----- 4.8/9.2 MB 9.3 MB/s eta
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 ----- 5.2/9.2 MB 8.9 MB/s eta
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 ----- 5.6/9.2 MB 9.1 MB/s eta
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 ----- 6.0/9.2 MB 8.9 MB/s eta
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 ----- 6.5/9.2 MB 9.2 MB/s eta
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 ----- 6.9/9.2 MB 9.2 MB/s eta
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 ----- 7.5/9.2 MB 9.2 MB/s eta
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 ----- 7.8/9.2 MB 9.1 MB/s eta
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 ----- 8.3/9.2 MB 9.1 MB/s eta
0:00:01
 ----- 8.8/9.2 MB 9.3 MB/s eta
0:00:01
 ----- 9.0/9.2 MB 9.2 MB/s eta
0:00:01
 ----- 9.2/9.2 MB 9.0 MB/s eta
0:00:01
 ----- 9.2/9.2 MB 8.4 MB/s eta
0:00:00
Downloading threadpoolctl-3.2.0-py3-none-any.whl (15 kB)
Installing collected packages: threadpoolctl, scikit-learn
Successfully installed scikit-learn-1.3.0 threadpoolctl-3.2.0
import pandas as pd
import numpy as np
import seaborn as sns
```

```
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
%matplotlib inline
```

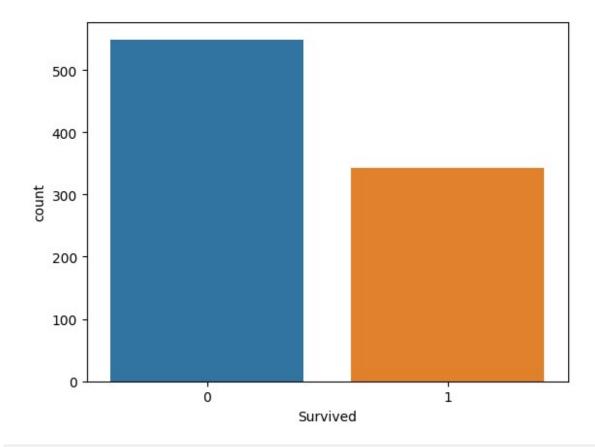
loading the dataset

```
train = pd.read csv('train.csv')
test = pd.read csv('test.csv')
train.head()
                Survived
                           Pclass
   PassengerId
             1
                                3
             2
                                1
1
                        1
2
             3
                        1
                                3
3
             4
                        1
                                1
4
                                                  Name
                                                            Sex
                                                                  Age
SibSp \
                              Braund, Mr. Owen Harris
                                                           male 22.0
1
1
   Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
1
2
                               Heikkinen, Miss. Laina female 26.0
0
3
        Futrelle, Mrs. Jacques Heath (Lily May Peel) female 35.0
1
4
                             Allen, Mr. William Henry
                                                           male 35.0
0
   Parch
                     Ticket
                                Fare Cabin Embarked
0
                 A/5 21171
                                        NaN
                                                   S
       0
                              7.2500
                   PC 17599
                             71.2833
                                                   C
1
                                        C85
2
       0
                                                   S
          STON/02. 3101282
                              7.9250
                                        NaN
                                                   S
3
       0
                     113803
                             53.1000
                                       C123
       0
                     373450
                              8.0500
                                        NaN
# statistical info
train.describe()
       PassengerId
                                      Pclass
                       Survived
                                                      Age
                                                                SibSp \
                                              714.000000
count
        891.000000
                     891.000000
                                 891.000000
                                                           891.000000
        446.000000
                       0.383838
                                    2.308642
                                               29.699118
                                                             0.523008
mean
std
        257.353842
                       0.486592
                                    0.836071
                                               14.526497
                                                             1.102743
min
          1.000000
                       0.000000
                                    1.000000
                                                0.420000
                                                             0.000000
        223.500000
                       0.000000
                                    2.000000
                                               20.125000
                                                             0.000000
25%
```

```
50%
        446.000000
                       0.000000
                                    3.000000
                                               28.000000
                                                             0.000000
75%
                                               38.000000
        668.500000
                       1.000000
                                    3.000000
                                                             1.000000
        891.000000
                       1.000000
                                    3,000000
                                               80,000000
                                                             8.000000
max
            Parch
                          Fare
       891.000000
                    891.000000
count
         0.381594
                     32.204208
mean
std
         0.806057
                     49.693429
         0.000000
                      0.000000
min
                      7.910400
25%
         0.000000
50%
         0.000000
                     14.454200
                     31.000000
75%
         0.000000
max
         6.000000
                    512.329200
# datatype info
train.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):
                   Non-Null Count
     Column
                                    Dtype
0
     PassengerId
                   891 non-null
                                    int64
 1
     Survived
                   891 non-null
                                    int64
 2
     Pclass
                   891 non-null
                                    int64
 3
     Name
                   891 non-null
                                    object
 4
     Sex
                   891 non-null
                                    object
5
     Age
                   714 non-null
                                    float64
 6
     SibSp
                   891 non-null
                                    int64
 7
                   891 non-null
                                    int64
     Parch
 8
     Ticket
                   891 non-null
                                    object
 9
                   891 non-null
                                    float64
     Fare
 10
     Cabin
                   204 non-null
                                    object
 11
    Embarked
                   889 non-null
                                    object
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

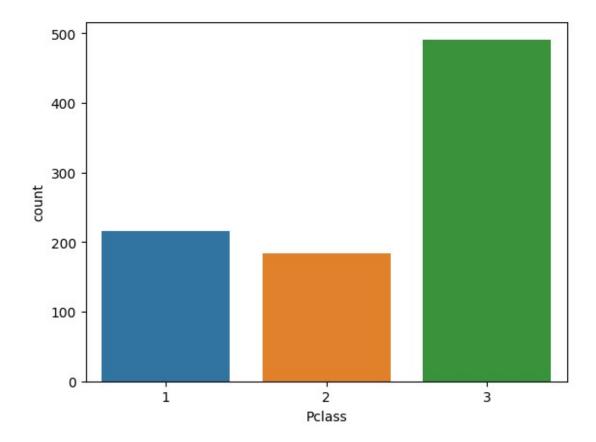
exploratory data analysis

```
sns.countplot(x = 'Survived', data = train)
<Axes: xlabel='Survived', ylabel='count'>
```



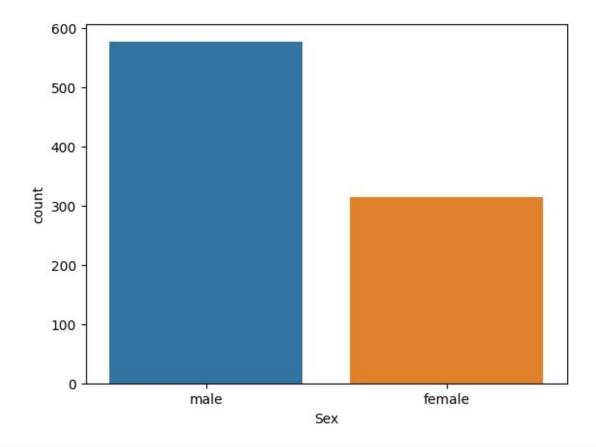
sns.countplot(x = 'Pclass', data = train)

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



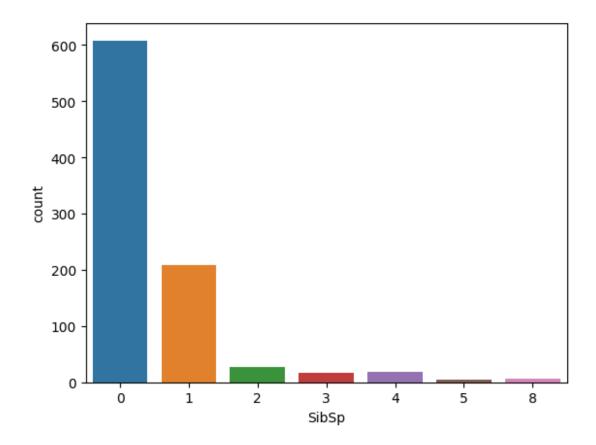
sns.countplot(x = 'Sex', data = train)

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



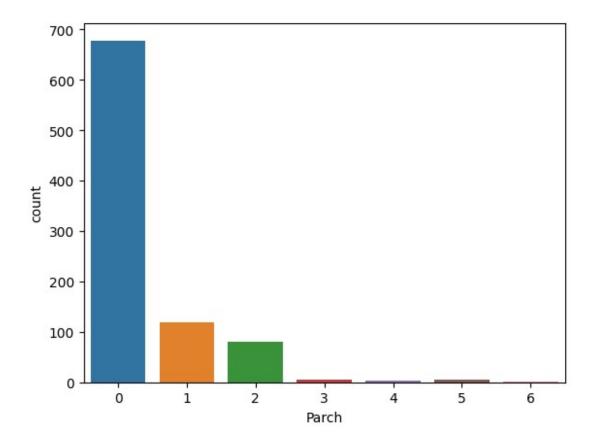
sns.countplot(x = 'SibSp', data = train)

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



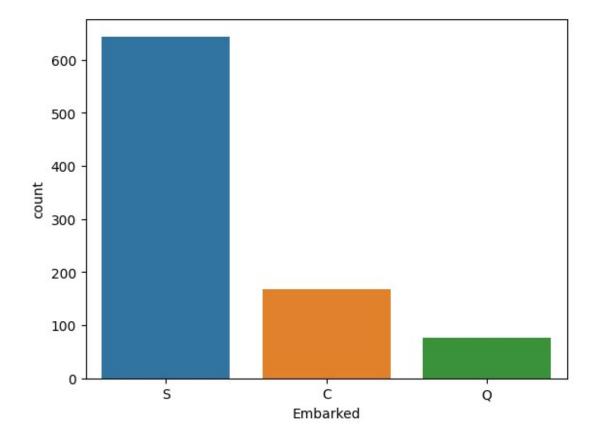
sns.countplot(x = 'Parch', data = train)

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



sns.countplot(x = 'Embarked', data = train)

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



numerical attributes

. . .

In summary, the main difference between countplot and distplot is the type of data they are designed to visualize:

countplot is for categorical data and shows the count of unique categories.

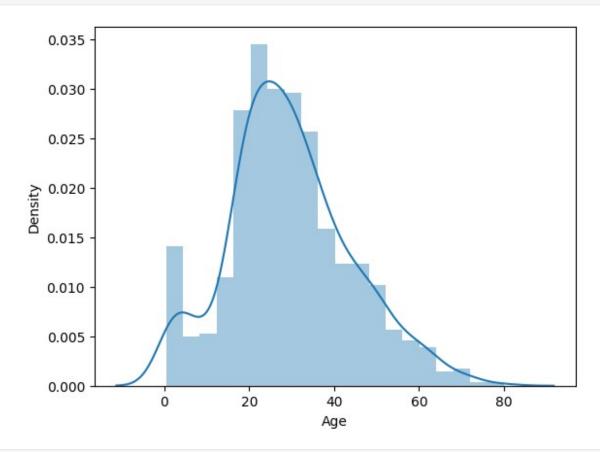
distplot is for continuous numerical data and shows the distribution of values along with an estimation of the probability density function.

You should choose the appropriate plot based on the nature of your data and the specific insights you want to gain from your visualization.

'\nIn summary, the main difference between countplot and distplot is the type of data they \nare designed to visualize:\n\ncountplot is for categorical data and shows the count of unique categories.\n\ndistplot is for continuous numerical data and shows the distribution of values along with \nan estimation of the probability density function.\n\nYou should choose the appropriate plot based on the nature of your data and the specific \ninsights you want to gain from your visualization.\ n'

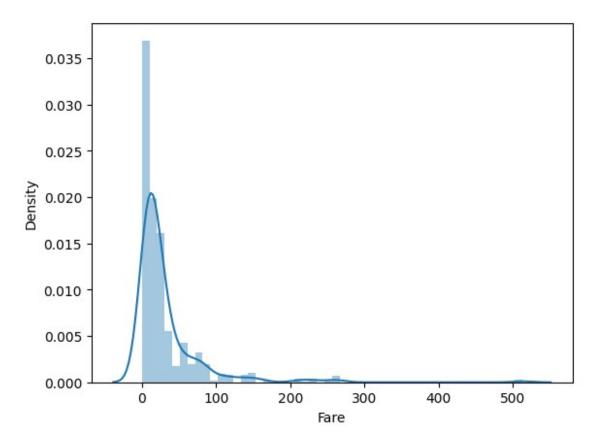
sns.distplot(train['Age'])

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

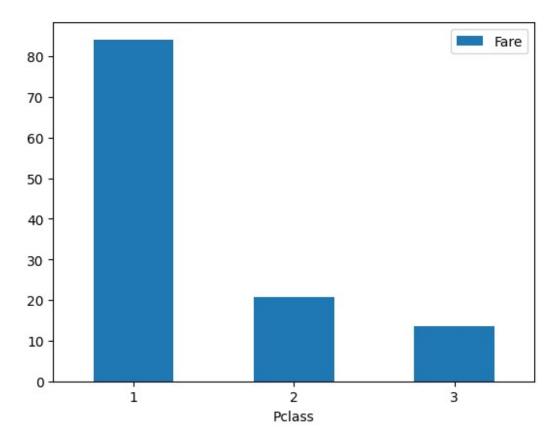


sns.distplot(train['Fare'])

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

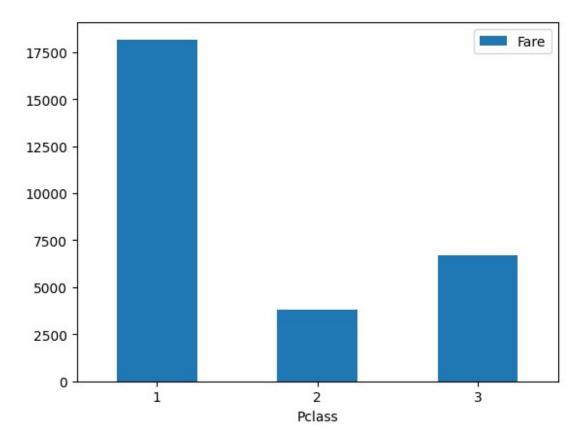


```
class_fare = train.pivot_table(index = 'Pclass', values = 'Fare')
class_fare.plot(kind = 'bar')
plt.xlabel('Pclass')
plt.xticks(rotation = 0)
plt.show()
```

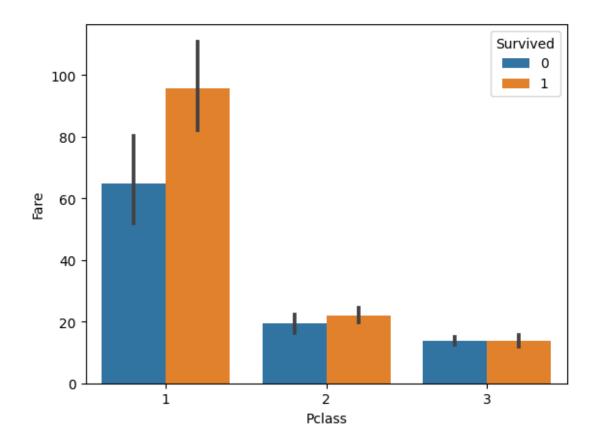


```
# this is for the total fare

class_fare = train.pivot_table(index = 'Pclass', values = 'Fare',
    aggfunc = np.sum)
    class_fare.plot(kind = 'bar')
    plt.xlabel('Pclass')
    plt.xticks(rotation = 0)
    plt.show()
```



```
sns.barplot(data = train, x = 'Pclass',y = 'Fare', hue = 'Survived')
# this is for making the barplot
# hue here is used to color the barplot
<Axes: xlabel='Pclass', ylabel='Fare'>
```

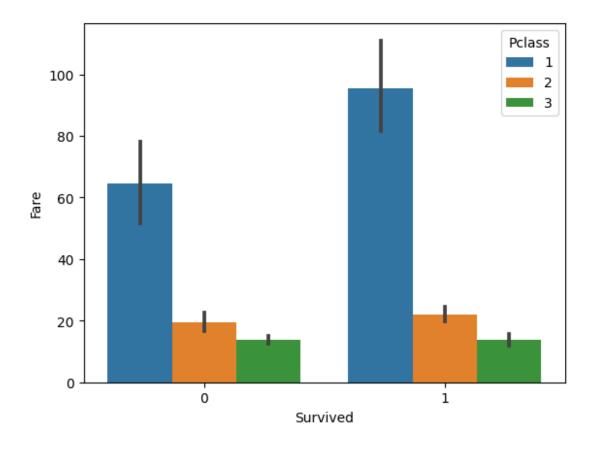


sns.barplot(data = train, x = 'Survived', y = 'Fare', hue = 'Pclass')

what this barplot tells us is that the number of people who survived and whom did not

survive based on their class and fare

<Axes: xlabel='Survived', ylabel='Fare'>



data pre processing

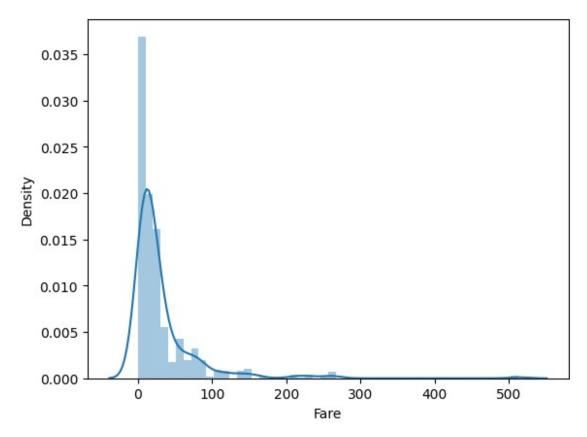
```
train_len = len(train)
# combine two data frames
df = pd.concat([train, test], axis = 0)
df = df.reset index(drop = True )
df.head()
   PassengerId
                Survived
                          Pclass \
0
                      0.0
             1
                                3
1
             2
                      1.0
                                1
2
             3
                      1.0
                                3
3
             4
                                1
                      1.0
                                3
                      0.0
                                                  Name
                                                           Sex
                                                                 Age
SibSp \
                              Braund, Mr. Owen Harris
                                                          male 22.0
1
   Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
1
2
                               Heikkinen, Miss. Laina female 26.0
0
```

```
3
        Futrelle, Mrs. Jacques Heath (Lily May Peel) female 35.0
1
4
                             Allen, Mr. William Henry
                                                          male 35.0
0
   Parch
                     Ticket
                                Fare Cabin Embarked
0
                 A/5 21171
                              7.2500
                                       NaN
                                                   S
       0
                                                   C
                  PC 17599
                                        C85
1
       0
                             71,2833
2
          STON/02. 3101282
                              7.9250
                                                   S
       0
                                       NaN
                                                   S
3
       0
                     113803
                             53.1000
                                      C123
                                                   S
4
       0
                     373450
                              8.0500
                                       NaN
df.tail()
      PassengerId
                    Survived
                              Pclass
                                                                Name
Sex \
1304
             1305
                         NaN
                                   3
                                                 Spector, Mr. Woolf
male
1305
             1306
                         NaN
                                   1
                                      Oliva y Ocana, Dona. Fermina
female
1306
             1307
                         NaN
                                      Saether, Mr. Simon Sivertsen
male
1307
                                                Ware, Mr. Frederick
             1308
                         NaN
male
1308
             1309
                         NaN
                                           Peter, Master. Michael J
male
                    Parch
                                       Ticket
                                                    Fare Cabin Embarked
       Age
            SibSp
                                                                       S
1304
                                    A.5. 3236
                                                  8.0500
       NaN
                                                           NaN
1305 39.0
                                     PC 17758
                                                108.9000
                                                          C105
                                                                       C
1306 38.5
                0
                        0
                           SOTON/0.Q. 3101262
                                                  7.2500
                                                           NaN
                                                                       S
                                                                       S
1307
                                       359309
                                                  8.0500
       NaN
                        0
                                                           NaN
1308
       NaN
                                          2668
                                                 22.3583
                                                           NaN
                                                                       C
# find the null values
train len
891
df.isnull().sum()
PassengerId
                   0
Survived
                418
Pclass
                   0
                   0
Name
```

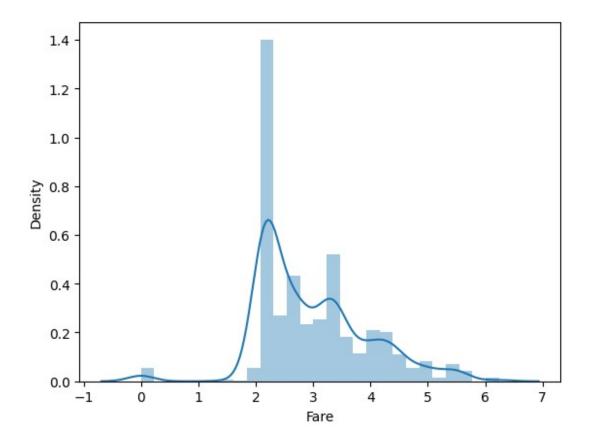
```
Sex
                263
Age
SibSp
                  0
Parch
                  0
Ticket
Fare
                  1
Cabin
               1014
Embarked
dtype: int64
# dropping the coloumn cabin
df = df.drop(columns = ['Cabin'], axis = 1)
df['Age'].mean()
29.881137667304014
df['Embarked'].mode()[0]
# filling the missing values using mean of that column
df['Age'] = df['Age'].fillna(df['Age'].mean())
df['Fare'] = df['Fare'].fillna(df['Fare'].mean())
df['Embarked'] = df['Embarked'].fillna(df['Embarked'].mode()[0])
```

log transformation for uniform data distribution

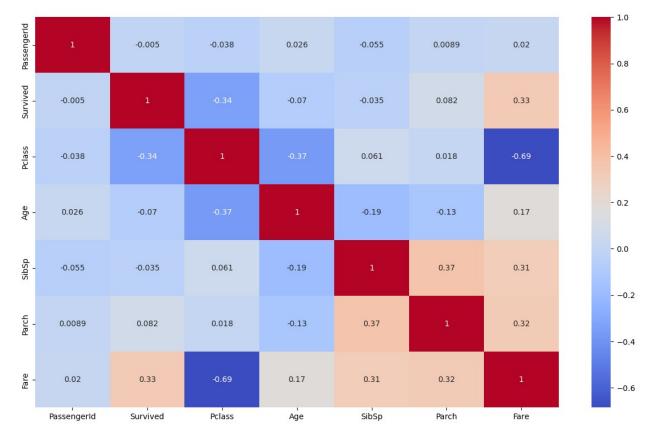
```
sns.distplot(train['Fare'])
<Axes: xlabel='Fare', ylabel='Density'>
```

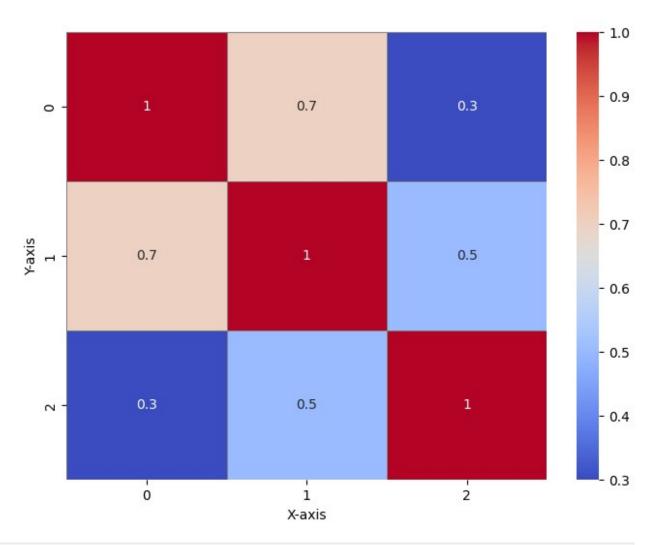


```
df['Fare'] = np.log(df['Fare'] + 1)
# what we are doing here is that we are replacing the null values with
mean
sns.distplot(df['Fare'])
<Axes: xlabel='Fare', ylabel='Density'>
```



co relation matrix





df	.head()								
0 1 2 3 4	Passenge	rId 1 2 3 4 5	Survived 0.0 1.0 1.0 1.0	Pclass 3 1 3 1 3	\				
Si	bSp \						Name	Sex	Age
0	.БЭР (Braun	d, N	Mr. Owen	Harris	male	22.0
1	Cumings,	Mrs.	John Bra	dley (Flo	rend	ce Briggs	Th	female	38.0
2				Heik	kine	en, Miss.	Laina	female	26.0
3	Fut	relle	, Mrs. Ja	cques Hea	th	(Lily May	Peel)	female	35.0

```
4
                           Allen, Mr. William Henry
                                                      male 35.0
0
   Parch
                   Ticket
                               Fare Embarked
0
                A/5 21171 2.110213
      0
                                           C
1
      0
                 PC 17599 4.280593
2
                                           S
      0
         STON/02. 3101282 2.188856
3
                                           S
      0
                   113803 3.990834
                   373450 2.202765
      0
df = df.drop(columns = ['Name', 'Ticket'], axis = 1)
df.head()
   PassengerId Survived Pclass Sex Age SibSp Parch
/
0
                    0.0
                              3
                                   male 22.0
                                                         0 2.110213
                                                   1
                    1.0
                                female 38.0
                                                            4.280593
1
                              1
                    1.0
                              3 female 26.0
                                                          0 2.188856
                    1.0
                                 female 35.0
                                                            3.990834
3
                    0.0
                                   male 35.0
                              3
                                                         0 2.202765
  Embarked
0
        S
        C
1
2
        S
3
        S
4
```

label encoding

```
from sklearn.preprocessing import LabelEncoder

cols = ['Sex', 'Embarked']
le = LabelEncoder()
# what we are doing here is that we are creating an instance of the label encoder class

for col in cols:
    df[col] = le.fit_transform(df[col])
df.head()

PassengerId Survived Pclass Sex Age SibSp Parch Fare Embarked
```

0	1	0.0	3	1	22.0	1	0	2.110213
2	2	1 0	1	0	20.0	1	0	4 200502
0	Z	1.0	1	0	38.0	1	0	4.280593
2	3	1.0	3	0	26.0	0	0	2.188856
2	4	1 0	1	^	25 0	-	0	2 000024
3 2	4	1.0	T	0	35.0	T	0	3.990834
4	5	0.0	3	1	35.0	0	0	2.202765
2								
<pre># now here what we can see is that the values of sex and embarked which were strings # before have been converted to numerical values which can now be used to make and visuali # properly</pre>								

train test split

```
train = df.iloc[:train len, :]
test = df.iloc[train_len:, :]
train.head()
   PassengerId Survived Pclass
                                   Sex
                                          Age SibSp Parch
                                                                  Fare
Embarked
0
                      0.0
                                3
                                         22.0
                                                          0
                                                              2.110213
             1
2
1
             2
                      1.0
                                1
                                     0
                                        38.0
                                                              4.280593
0
2
             3
                      1.0
                                3
                                        26.0
                                                              2.188856
2
3
                      1.0
                                        35.0
                                                              3.990834
2
             5
4
                      0.0
                                3
                                     1 35.0
                                                              2.202765
2
test.head()
     PassengerId
                  Survived Pclass Sex Age SibSp Parch
                                                                    Fare
891
             892
                                           34.5
                                                     0
                                                                2.178064
                        NaN
892
             893
                        NaN
                                           47.0
                                                     1
                                                                2.079442
893
             894
                        NaN
                                           62.0
                                                                2.369075
                                        1
             895
894
                        NaN
                                        1 27.0
                                                                2.268252
```

```
895
            896
                      NaN
                                3
                                     0 22.0
                                                  1
                                                         1 2.586824
     Embarked
891
892
           2
893
           1
           2
894
895
           2
X = train.drop(columns = ['PassengerId', 'Survived'], axis = 1)
y = train['Survived']
X.head()
   Pclass Sex Age SibSp Parch
                                             Embarked
                                       Fare
        3
            1
               22.0
                         1
                                0 2.110213
                                                    2
               38.0
                                0 4.280593
                                                    0
1
        1
                         1
2
        3
                                                    2
            0 26.0
                         0
                                0 2.188856
3
        1
               35.0
                         1
                                0 3.990834
                                                    2
            0
4
                                                    2
            1
               35.0
                         0
                                0 2.202765
```

model training

```
from sklearn.model selection import train test split, cross val score
# classify coloumn
def classify(model):
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size = 0.25, random_state = 42)
    model.fit(X train, y train)
    print('Accuracy: ', model.score(X test, y test))
    score = cross_val_score(model, X, y, cv = 5)
    print('CV Score: ', np.mean(score))
from sklearn.linear model import LogisticRegression
model = LogisticRegression()
classify(model)
Accuracy: 0.8071748878923767
CV Score: 0.7833971502102819
from sklearn.tree import DecisionTreeClassifier
model = DecisionTreeClassifier()
classify(model)
           0.726457399103139
Accuracy:
CV Score: 0.7677044755508129
```

```
from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
classify(model)
        0.8026905829596412
Accuracy:
CV Score: 0.815956311593748
!pip install xgboost
Collecting xgboost
 Obtaining dependency information for xgboost from
https://files.pythonhosted.org/packages/32/10/4689bda37403f7dd029d550c
4446e0097c2f33b8ae877b235e76d5c49bc2/xgboost-2.0.0-py3-none-
win amd64.whl.metadata
 Downloading xgboost-2.0.0-py3-none-win amd64.whl.metadata (2.0 kB)
Requirement already satisfied: numpy in c:\users\ilove\appdata\local\
programs\python\python311\lib\site-packages (from xgboost) (1.25.2)
Requirement already satisfied: scipy in c:\users\ilove\appdata\local\
programs\python\python311\lib\site-packages (from xgboost) (1.11.2)
Downloading xgboost-2.0.0-py3-none-win_amd64.whl (99.7 MB)
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Installing collected packages: xgboost
Successfully installed xgboost-2.0.0
from sklearn.svm import SVC
model = SVC()
classify(model)
Accuracy:
        0.6188340807174888
CV Score: 0.6835289686774214
from sklearn.neighbors import KNeighborsClassifier
model = KNeighborsClassifier()
classify(model)
Accuracy:
        0.7802690582959642
CV Score: 0.7744397715146569
import xgboost as xgb
from xgboost import XGBClassifier
model = XGBClassifier()
classify(model)
        0.7847533632286996
Accuracy:
CV Score: 0.8148327160881301
!pip install lightgbm catboost
Collecting lightabm
 Obtaining dependency information for lightgbm from
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https://files.pythonhosted.org/packages/b3/f8/ee33e36194eb03a76eccf3ad
ac3fba51f0e56fbd20609bb531659d48d3cb/lightgbm-4.1.0-py3-none-
win amd64.whl.metadata
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d2a63a3f1d2066c2f058b2d974ce5479e32e/catboost-1.2.1.1-cp311-cp311-
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Requirement already satisfied: scipy in c:\users\ilove\appdata\local\
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Requirement already satisfied: six in c:\users\ilove\appdata\local\
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Requirement already satisfied: contourpy>=1.0.1 in c:\users\ilove\
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3210ffa136a367751e454214013b441c4575/tenacity-8.2.3-py3-none-
anv.whl.metadata
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Downloading lightgbm-4.1.0-py3-none-win amd64.whl (1.3 MB)
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Downloading tenacity-8.2.3-py3-none-any.whl (24 kB)
Installing collected packages: tenacity, graphviz, plotly, lightgbm,
catboost
Successfully installed catboost-1.2.1.1 graphviz-0.20.1 lightgbm-4.1.0
plotly-5.17.0 tenacity-8.2.3
from lightgbm import LGBMClassifier
from catboost import CatBoostClassifier
from lightgbm import LGBMClassifier
model = LGBMClassifier(verbose = -1)
classify(model)
Accuracy: 0.8116591928251121
CV Score: 0.8238277572029377
from catboost import CatBoostClassifier
model = CatBoostClassifier(verbose = 0)
classify(model)
Accuracy: 0.8295964125560538
CV Score: 0.8226790534178645
```