

# titanic-traindata

June 15, 2024

```
[1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
[133]: train_df=pd.read_csv("C:\\Users\\pandey\\Downloads\\titanic\\train.
↪csv")
```

```
[134]: train_df.head()
```

```
[134]: PassengerId  Survived  Pclass  \
0             1         0         3
1             2         1         1
2             3         1         3
3             4         1         1
4             5         0         3
```

```

                                Name      Sex  Age  SibSp  \
0                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      0   A/5 21171    7.2500   NaN        S
1      0   PC 17599   71.2833   C85        C
2      0  STON/O2. 3101282    7.9250   NaN        S
3      0    113803   53.1000  C123        S
4      0    373450    8.0500   NaN        S
```

```
[135]: train_df.drop(columns=("Cabin"), inplace=True)
```

```
[136]: train_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):
#   Column             Non-Null Count  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  Parch       891 non-null    int64
8  Ticket      891 non-null    object
9  Fare        891 non-null    float64
10 Embarked    889 non-null    object
dtypes: float64(2), int64(5), object(4)
memory usage: 76.7+ KB

```

```
[137]: train_df["Age"].fillna(train_df["Age"].mean(), inplace=True)
```

C:\Users\pandeyunny2315\AppData\Local\Temp\ipykernel\_7880\1036321305.py:1:  
FutureWarning: A value is trying to be set on a copy of a DataFrame or Series  
through chained assignment using an inplace method.  
The behavior will change in pandas 3.0. This inplace method will never work  
because the intermediate object on which we are setting values always behaves as  
a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using  
'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value)  
instead, to perform the operation inplace on the original object.

```
train_df["Age"].fillna(train_df["Age"].mean(), inplace=True)
```

```
[138]: train_df["Embarked"].fillna('S', inplace=True)
```

C:\Users\pandeyunny2315\AppData\Local\Temp\ipykernel\_7880\4256062730.py:1:  
FutureWarning: A value is trying to be set on a copy of a DataFrame or Series  
through chained assignment using an inplace method.  
The behavior will change in pandas 3.0. This inplace method will never work  
because the intermediate object on which we are setting values always behaves as  
a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using  
'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value)  
instead, to perform the operation inplace on the original object.

```
train_df["Embarked"].fillna('S', inplace=True)
```

```
[139]: train_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):
#   Column          Non-Null Count  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             891 non-null   float64
6   SibSp           891 non-null   int64
7   Parch           891 non-null   int64
8   Ticket          891 non-null   object
9   Fare            891 non-null   float64
10  Embarked        891 non-null   object
dtypes: float64(2), int64(5), object(4)
memory usage: 76.7+ KB
```

```
[140]: train_df['Survived']=train_df['Survived'].astype('category')
train_df['Age']=train_df['Age'].astype('int')
train_df['Embarked']=train_df['Embarked'].astype('category')
train_df['Sex']=train_df['Sex'].astype('category')
train_df['Pclass']=train_df['Pclass'].astype('category')
```

```
[141]: train_df.info()
```

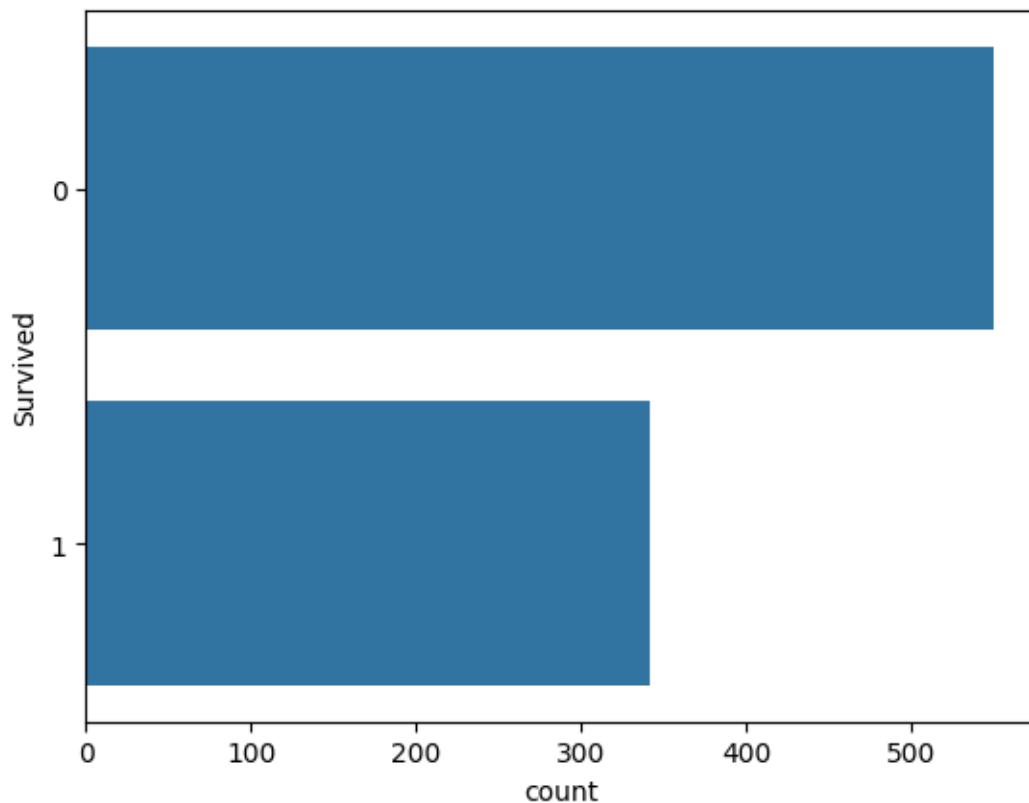
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):
#   Column          Non-Null Count  Dtype
---  -
0   PassengerId     891 non-null   int64
1   Survived        891 non-null   category
2   Pclass          891 non-null   category
3   Name            891 non-null   object
4   Sex             891 non-null   category
5   Age             891 non-null   int32
6   SibSp           891 non-null   int64
7   Parch           891 non-null   int64
8   Ticket          891 non-null   object
9   Fare            891 non-null   float64
10  Embarked        891 non-null   category
dtypes: category(4), float64(1), int32(1), int64(3), object(2)
memory usage: 49.4+ KB
```

```
[142]: death_percent=round((train_df['Survived'].value_counts().values[0]/891)*100)
death_percent
```

```
[142]: 62
```

```
[143]: sns.countplot(train_df['Survived'])
death_percent=round((train_df['Survived'].value_counts().values[0]/891)*100)
print('Out of 891 {} people deid in the accident'.format(death_percent))
```

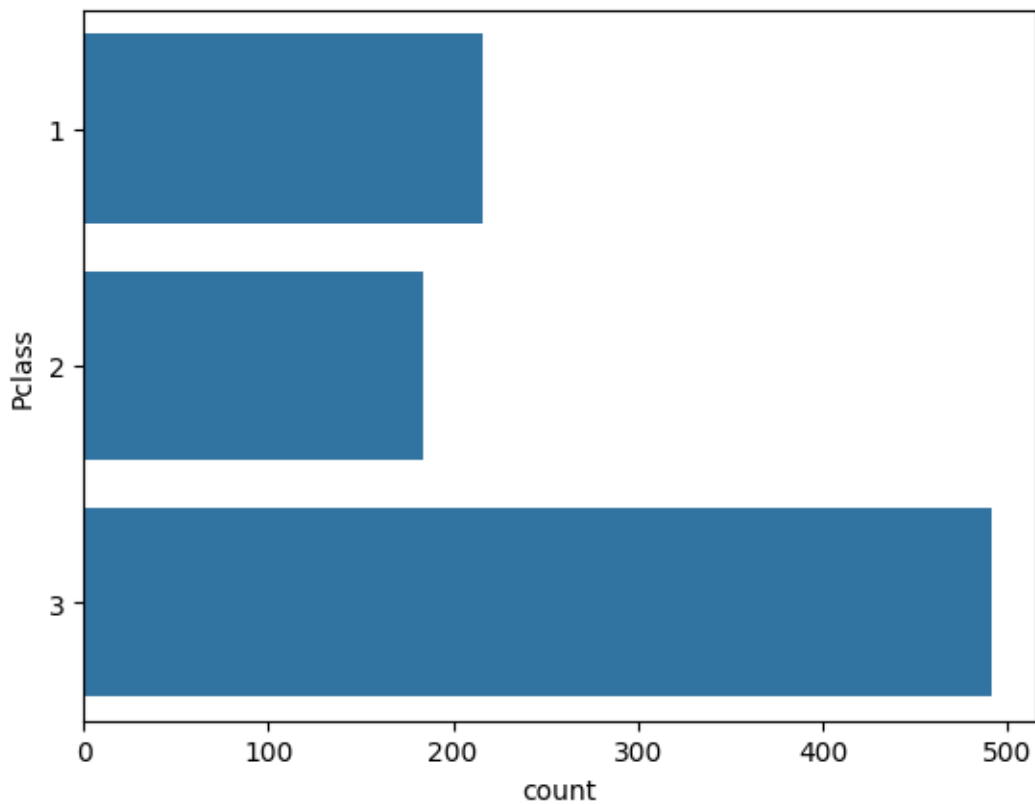
Out of 891 62 people deid in the accident



```
[144]: sns.countplot(train_df['Pclass'])
print((train_df['Pclass'].value_counts()))
print(((train_df['Pclass'].value_counts()/891)*100))
```

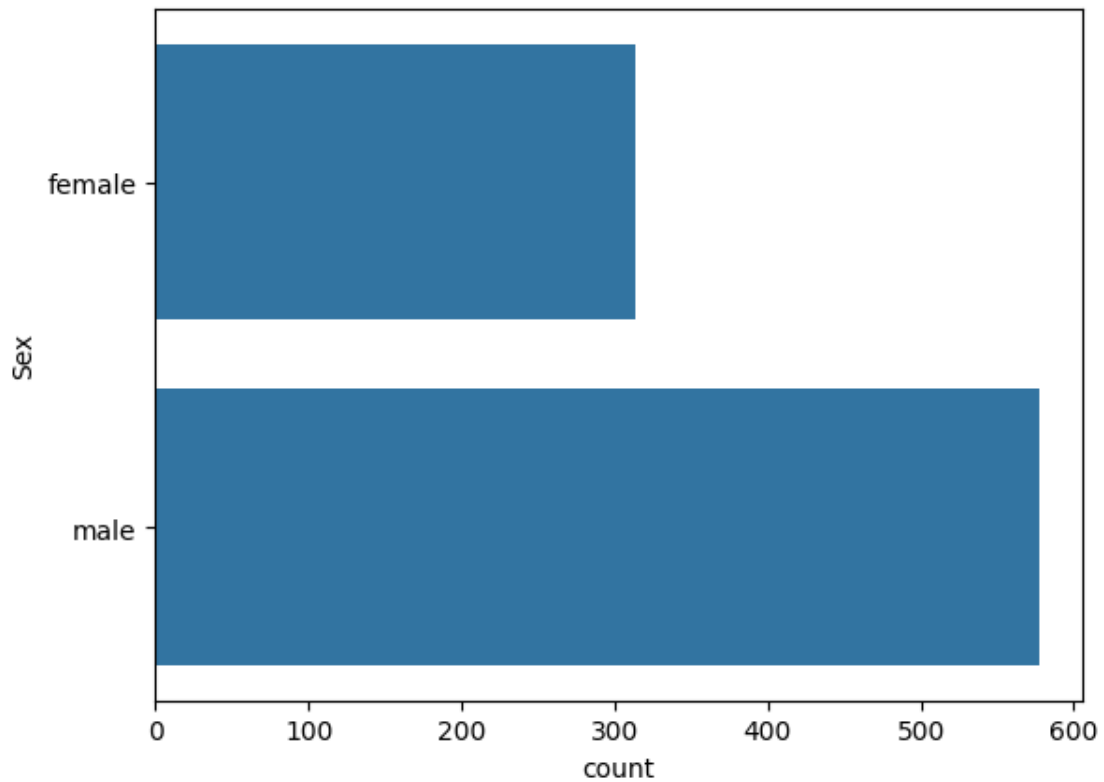
```
Pclass
3    491
1    216
2    184
Name: count, dtype: int64
Pclass
3    55.106622
```

```
1    24.242424
2    20.650954
Name: count, dtype: float64
```



```
[58]: sns.countplot(train_df['Sex'])
      print((train_df['Sex'].value_counts()))
      print((train_df['Sex'].value_counts()/891)*100)
```

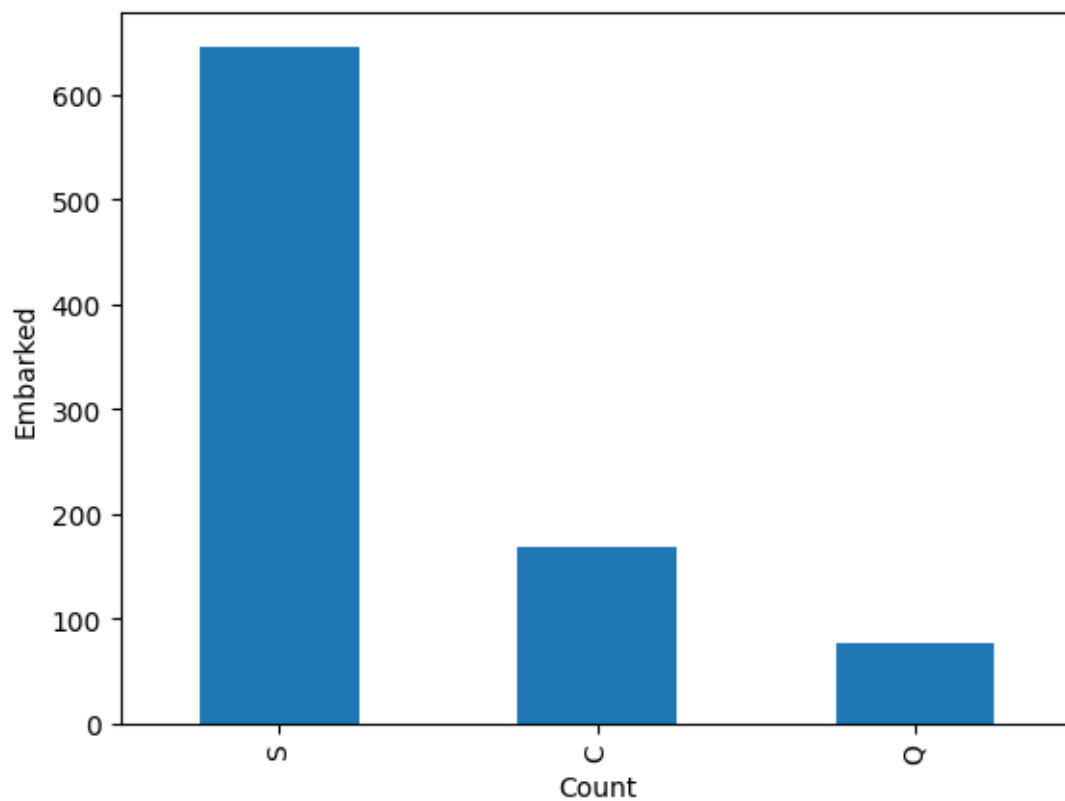
```
Sex
male    577
female  314
Name: count, dtype: int64
Sex
male    64.758698
female  35.241302
Name: count, dtype: float64
```



```
[79]: print((train_df['Embarked'].value_counts()))
print((train_df['Embarked'].value_counts()/891)*100)
Embarked_counts=train_df['Embarked'].value_counts()
Embarked_counts.plot(kind='bar')
plt.xlabel('Count')
plt.ylabel('Embarked')
plt.show()
```

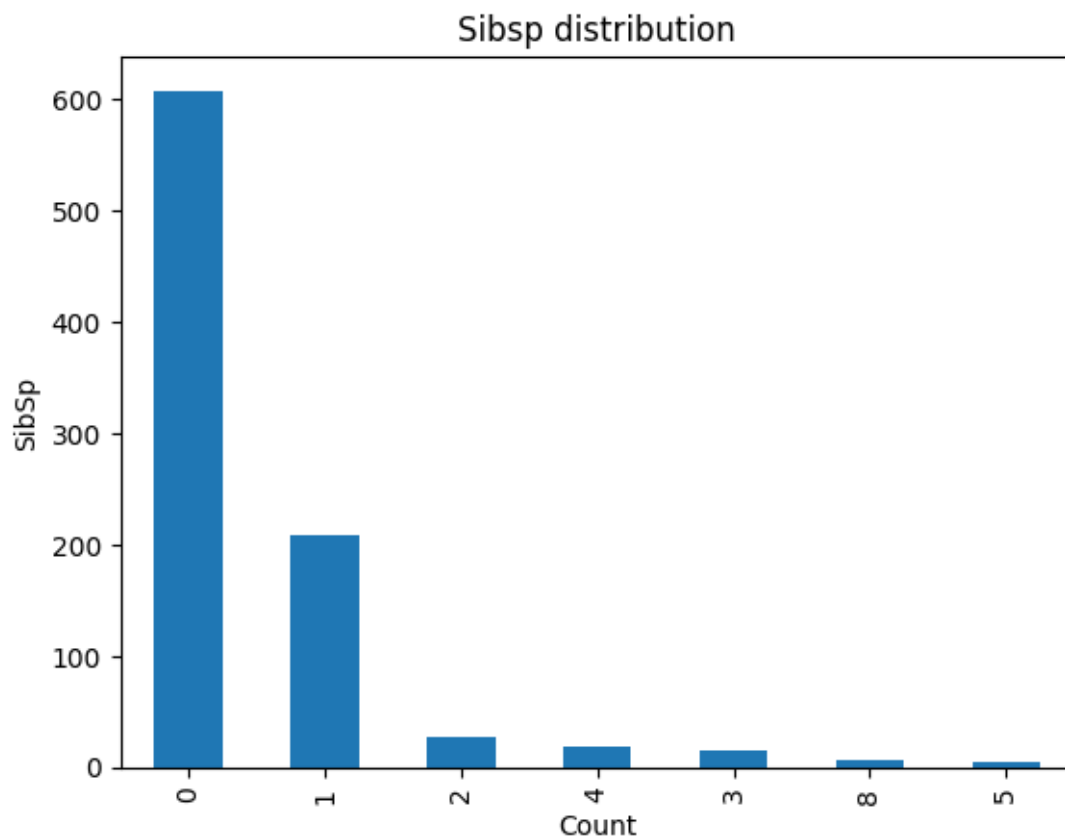
```
Embarked
S    646
C    168
Q     77
Name: count, dtype: int64

Embarked
S    72.502806
C    18.855219
Q     8.641975
Name: count, dtype: float64
```



```
[93]: print((train_df['SibSp'].value_counts()/891)*100)
SibSp_counts=train_df['SibSp'].value_counts() #.plot(kind='barh')
SibSp_counts.plot(kind='bar')
plt.xlabel('Count')
plt.ylabel('SibSp')
plt.title('Sibsp distribution')
plt.show()
```

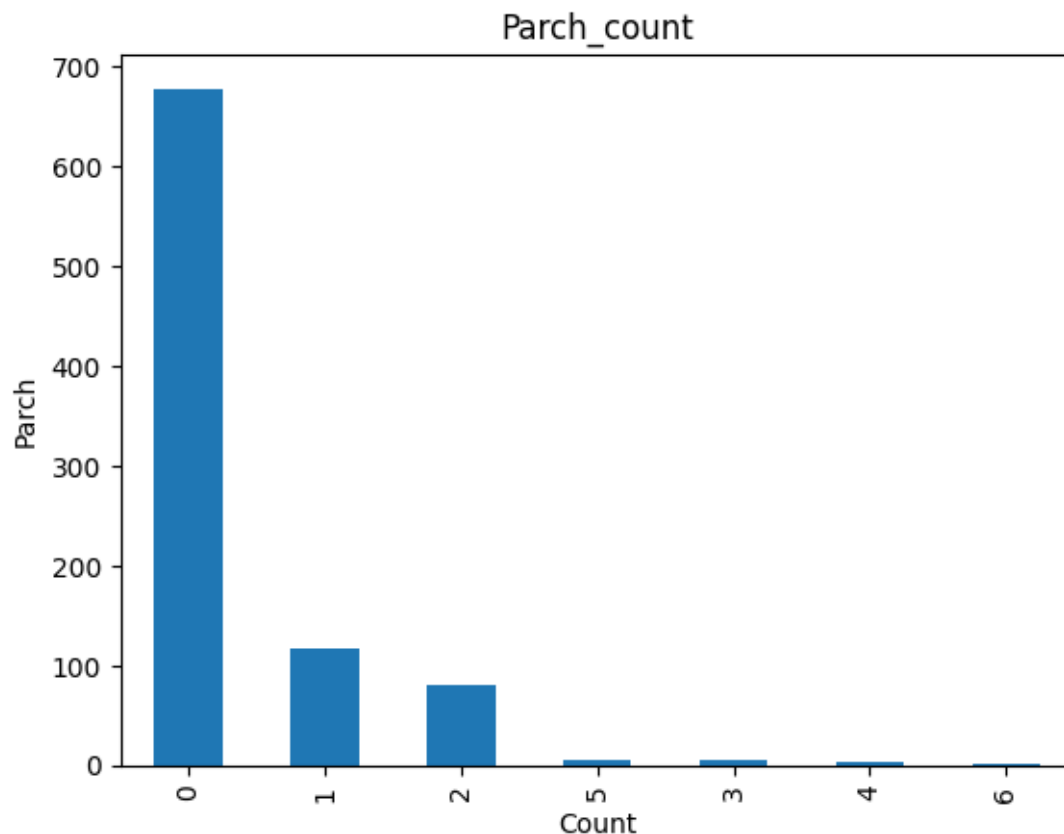
```
SibSp
0    68.237935
1    23.456790
2     3.142536
4     2.020202
3     1.795735
8     0.785634
5     0.561167
Name: count, dtype: float64
```



```
[94]: print((train_df['Parch'].value_counts()/891)*100)
Parch_count=(train_df['Parch'].value_counts())
Parch_count.plot(kind='bar')
plt.xlabel('Count')
plt.ylabel('Parch')
plt.title('Parch_count')
plt.show()
```

```
Parch
0    76.094276
1    13.243547
2     8.978676
5     0.561167
3     0.561167
4     0.448934
6     0.112233
Name: count, dtype: float64
```





```
[96]: sns.distplot(train_df['Age'])
print(train_df['Age'].skew())
print(train_df['Age'].kurt())
```

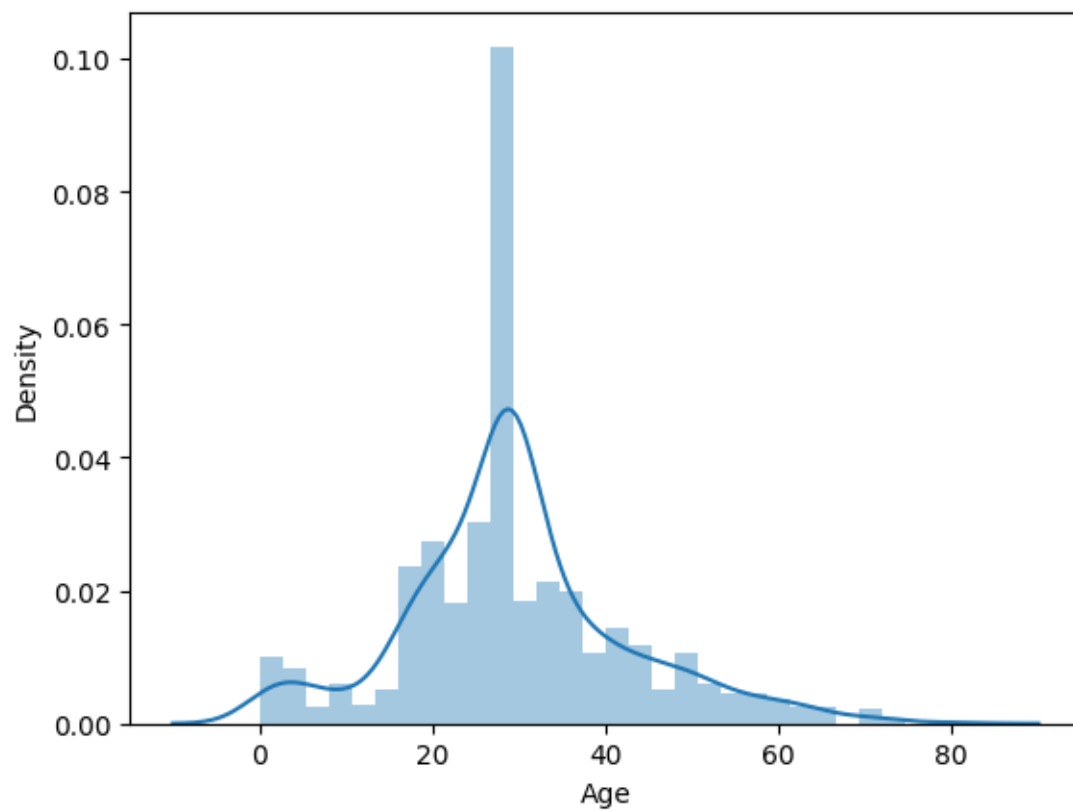
C:\Users\pandeyunny2315\AppData\Local\Temp\ipykernel\_7880\101640707.py:1:  
UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

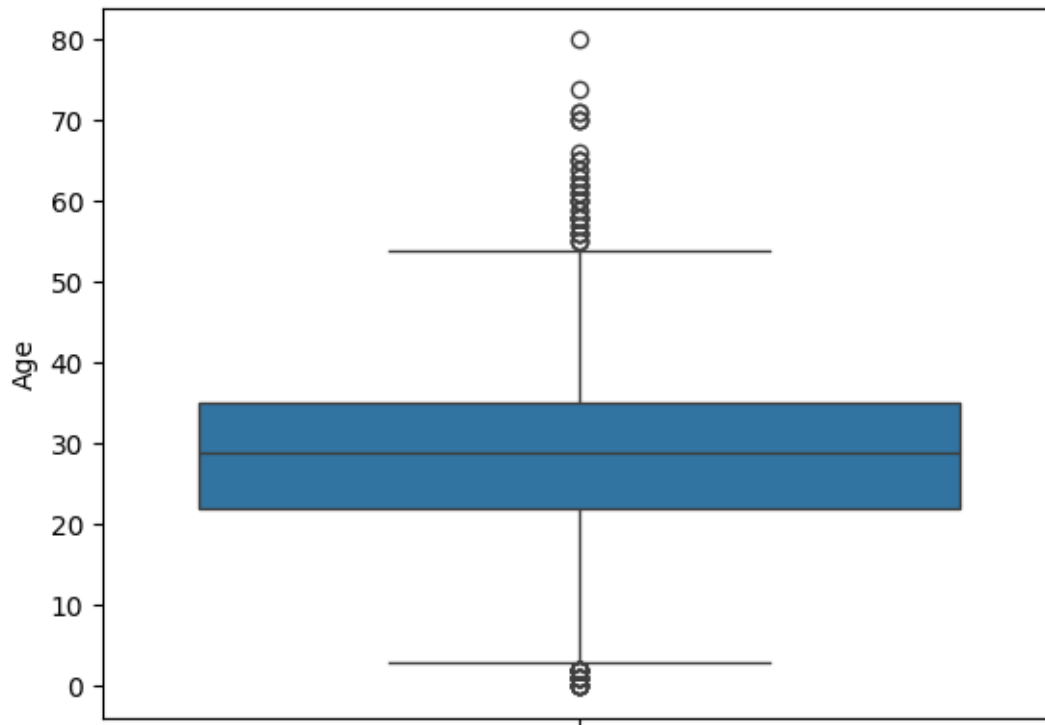
For a guide to updating your code to use the new functions, please see  
<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(train_df['Age'])
0.45956263424701577
0.9865867453652877
```



```
[97]: sns.boxplot(train_df['Age'])
```

```
[97]: <Axes: ylabel='Age'>
```



```
[174]: def Family_type(number):
        if number==0:
            return 'Alone'
        elif number > 0 and number <= 4:
            return 'Medium'
        else:
            return 'Large'
```

```
[182]: train_df['Family_size'].sample(10)
```

```
[182]: 11      0
        96      0
        98      1
       634      5
        94      0
       260      0
       736      4
       221      0
       768      1
       675      0
        Name: Family_size, dtype: int64
```

```
[178]: #train_df['Family_size']= train_df['Parch'] + train_df['SibSp']  
(train_df['Family_size'] > 4).sum()
```

```
[178]: 47
```

```
[186]: train_df['Family_type']=train_df['Family_size'].apply(Family_type)
```

```
[187]: train_df['Family_size'].value_counts()
```

```
[187]: Family_size  
0      537  
1      161  
2      102  
3       29  
5       22  
4       15  
6       12  
10       7  
7        6  
Name: count, dtype: int64
```

```
[188]: train_df['Family_type'].value_counts()
```

```
[188]: Family_type  
Alone      537  
Medium    307  
Large      47  
Name: count, dtype: int64
```

```
[199]: train_df.drop(columns={'SibSp','Parch','Family_size'}, inplace=True)
```

```
[200]: train_df.sample(5)
```

```
[200]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	\
450	451	0	2	West, Mr. Edwy Arthur	male	36	
656	657	0	3	Radeff, Mr. Alexander	male	29	
56	57	1	2	Rugg, Miss. Emily	female	21	
443	444	1	2	Reynaldo, Ms. Encarnacion	female	28	
810	811	0	3	Alexander, Mr. William	male	26	

	Ticket	Fare	Embarked	Family_type
450	C.A. 34651	27.7500	S	Medium
656	349223	7.8958	S	Alone
56	C.A. 31026	10.5000	S	Alone
443	230434	13.0000	S	Alone
810	3474	7.8875	S	Alone

```
[192]: pd.crosstab(train_df['Family_type'],train_df['Survived']).apply(lambda r:
↳round((r/r.sum()*100, 1), axis=1)
```

```
[192]: Survived      0      1
Family_type
Alone           69.6   30.4
Large           85.1   14.9
Medium          44.0   56.0
```

```
[197]: pd.crosstab(train_df['Sex'],train_df['Survived']).apply(lambda r: round((r/r.
↳sum()*100, 1), axis=1)
```

```
[197]: Survived      0      1
Sex
female          25.8   74.2
male             81.1   18.9
```

```
[198]: pd.crosstab(train_df['Pclass'], train_df['Survived']).apply(lambda r: round((r/
↳r.sum()*100, 1), axis=1)
```

```
[198]: Survived      0      1
Pclass
1           37.0   63.0
2           52.7   47.3
3           75.8   24.2
```

```
[203]: pd.crosstab(train_df['Embarked'],train_df['Survived']).apply(lambda r: round((r/
↳r.sum()*100,1), axis=1)
```

```
[203]: Survived      0      1
Embarked
C           44.6   55.4
Q           61.0   39.0
S           66.1   33.9
```

```
[204]: train_df['Embarked'].value_counts()
```

```
[204]: Embarked
S      646
C      168
Q       77
Name: count, dtype: int64
```

```
[209]: # Chances of female survived is higher than male survived as you can see 74.2%
↳are females and only 18.9% of mens were survived.
# People travelling in pclass 1 are having more likely to survive than pclass 3
↳and pclass 2.
```

```
# somehow people travelling "Medium group" have more chances of surviving than  
↪ people travelling "Alone" are with "Large group".  
# people going to C are more likely to survived.  
# people in the range of 20-40 had a higher chance of not surviving.
```

[ ]: