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

df = pd.read\_csv("/content/drive/MyDrive/DATASETS/Titanic-Dataset.csv")

from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

df.head()

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Ca
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	1
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	(
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	1
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	С
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	1

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):

Ducu	COTAMILIS (COC	ar re coramiis).							
#	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	Cabin	204 non-null	object						
11	Embarked	889 non-null	object						
dtyp	<pre>dtypes: float64(2), int64(5), object(5)</pre>								
memory usage: 83.7+ KB									

df.describe()

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare	$\blacksquare$
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000	ıl.
mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208	
std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429	
min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000	
25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400	
50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200	
75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000	
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200	

df.corr()

<ipython-input-10-2f6f6606aa2c>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future ver
 df.corr()

```
        PassengerId
        Survived
        Pclass
        Age
        SibSp
        Parch
        Fare

        PassengerId
        1.000000
        -0.005007
        -0.035144
        0.036847
        -0.057527
        -0.001652
        0.012658
        1

        Survived
        -0.005007
        1.000000
        -0.338481
        -0.077221
        -0.035322
        0.081629
        0.257307
```

df.corr().Survived.sort\_values(ascending = False)

Fare 0.257307
Parch 0.081629
PassengerId -0.005007
SibSp -0.035322
Age -0.077221
Pclass -0.338481
Name: Survived, dtype: float64

## df.isnull().any()

False PassengerId Survived False Pclass False False Name Sex False Age True SibSp False Parch False Ticket False False Fare Cabin True Embarked True dtype: bool

sum(df.Cabin.isnull())

687

sum(df.Age.isnull())

177

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

sum(df.Embarked.isnull())

2

df.describe()

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare	
count	891.000000	891.000000	891.000000	891.000000	891.000000	891.000000	891.000000	11.
mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208	
std	257.353842	0.486592	0.836071	13.002015	1.102743	0.806057	49.693429	
min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000	
25%	223.500000	0.000000	2.000000	22.000000	0.000000	0.000000	7.910400	
50%	446.000000	0.000000	3.000000	29.699118	0.000000	0.000000	14.454200	
75%	668.500000	1.000000	3.000000	35.000000	1.000000	0.000000	31.000000	
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200	

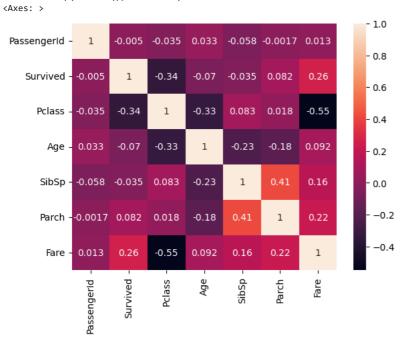
plt.scatter(df["Fare"],df["Survived"])

<matplotlib.collections.PathCollection at 0x7c90b8ff11b0>

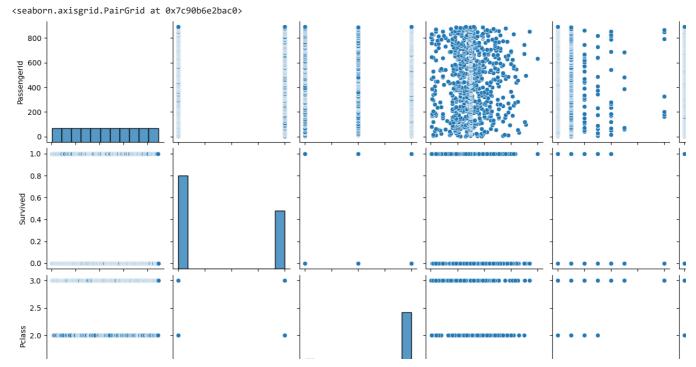


sns.heatmap(df.corr(),annot=True)

<ipython-input-21-8df7bcac526d>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future ver sns.heatmap(df.corr(),annot=True)



sns.pairplot(df)

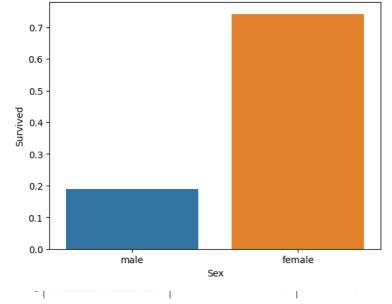


sns.barplot(x=df["Sex"],y=df["Survived"],ci=0)

<ipython-input-22-8ae461271d98>:1: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

sns.barplot(x=df["Sex"],y=df["Survived"],ci=0)
<Axes: xlabel='Sex', ylabel='Survived'>



sns.barplot(x=df["Embarked"],y=df["Survived"],ci=0)

<ipython-input-23-d5b0276940a6>:1: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

sns.barplot(x=df["Embarked"],y=df["Survived"],ci=0)
<Axes: xlabel='Embarked', ylabel='Survived'>

0.5 -

sns.barplot(x=df["Parch"],y=df["Survived"],ci=0)

<ipython-input-26-a1496fefeaf8>:1: FutureWarning:

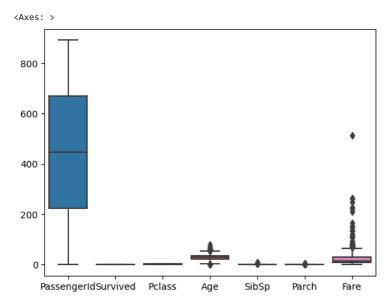
The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

sns.barplot(x=df["Parch"],y=df["Survived"],ci=0)
<Axes: xlabel='Parch', ylabel='Survived'>

0.6 - 0.5 - 0.4 - PANA 0.3 - 0.2 - 0.1 - 0.0 - 0 1 2 3 4 5 6

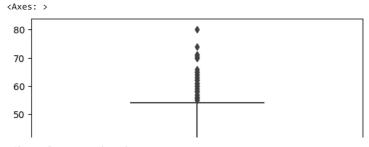
Parch

sns.boxplot(df)



sns.boxplot(df.Age)

1



Q1 = df['Age'].quantile(0.25) Q3 = df['Age'].quantile(0.75)

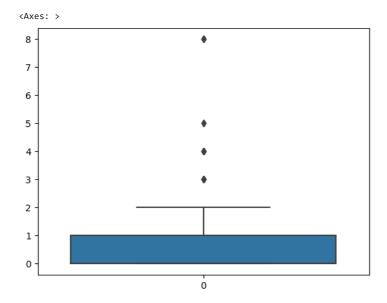
IQR = Q3 - Q1

threshold = 1.5 \* IQR

sns.boxplot(df.Age)

<Axes: >
50 40 30 20 10 -

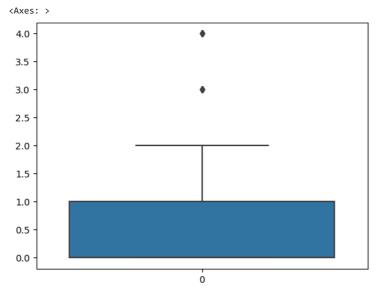
sns.boxplot(df.SibSp)



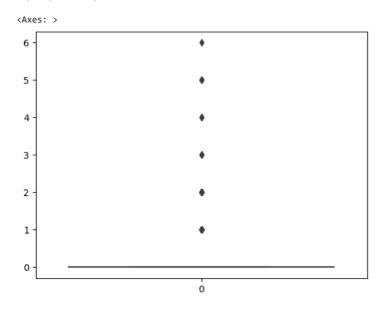
p99 = df.SibSp.quantile(0.99)

df = df[df.SibSp < p99]

sns.boxplot(df.SibSp)



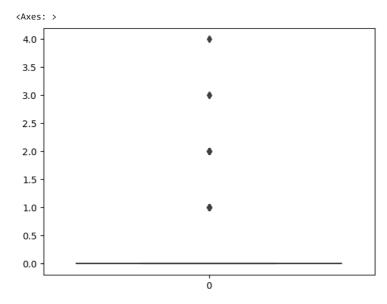
sns.boxplot(df.Parch)



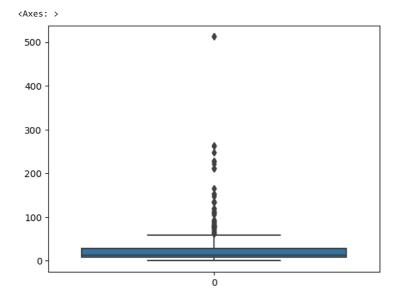
p99 = df.Parch.quantile(0.99)

df = df[df.Parch < p99]

sns.boxplot(df["Parch"])



sns.boxplot(df["Fare"])

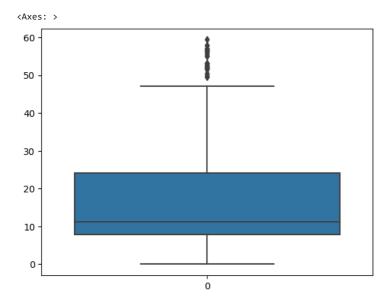


```
Q1 = df['Fare'].quantile(0.25)
Q3 = df['Fare'].quantile(0.75)

IQR = Q3 - Q1
threshold = 1.5 * IQR

df = df[(df['Fare'] >= Q1 - threshold) & (df['Fare'] <= Q3 + threshold)]</pre>
```

sns.boxplot(df.Fare)



x = df.drop(columns=["Survived","PassengerId","Name","Ticket","Cabin"],axis=1) # Independent variables should be in df or 2d array x.head()

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	
0	3	male	22.000000	1	0	7.2500	S	ılı
2	3	female	26.000000	0	0	7.9250	S	
3	1	female	35.000000	1	0	53.1000	S	
4	3	male	35.000000	0	0	8.0500	S	
5	3	male	29.699118	0	0	8.4583	Q	

```
y = pd.Series(df["Survived"])
```

y.head()

x\_Scaled.head()

```
Name: Survived, dtype: int64
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
x["Sex"] = le.fit_transform(x["Sex"])
x.head()
                         Age SibSp Parch
                                                              \blacksquare
        Pclass Sex
                                            Fare Embarked
     0
                1 22.000000
                                        0 7.2500
                                                              th
     2
                 0 26.000000
             3
                                  0
                                        0 7.9250
                                                          S
                 0 35.000000
                                        0 53.1000
                                                          S
                1 35.000000
                                 0
                                        0 8.0500
                                                          S
     4
             3
             3
                1 29.699118
                                 0 0 8.4583
                                                         Q
print(le.classes_)
    ['female' 'male']
mapping=dict(zip(le.classes_,range(len(le.classes_))))
mapping
    {'female': 0, 'male': 1}
le1 = LabelEncoder()
x["Embarked"] = le1.fit_transform(x["Embarked"])
x.head()
                        Age SibSp Parch
                                                              \blacksquare
        Pclass Sex
                                           Fare Embarked
     0
                1 22.000000
                                        0 7.2500
                                                          2
                                                              th
                                                          2
     2
                 0 26.000000
             3
                                  0
                                        0 7.9250
                                                          2
             1
                 0 35.000000
                                 1
                                        0 53.1000
     4
             3
                1 35.000000
                                  0
                                        0 8.0500
                                                          2
            3
                1 29.699118
                                 0
                                        0 8.4583
                                                          1
print(le1.classes_)
    ['C' 'Q' 'S']
mapping1=dict(zip(le1.classes_,range(len(le1.classes_))))
mapping1
    {'C': 0, 'Q': 1, 'S': 2}
from \ sklearn.preprocessing \ import \ MinMaxScaler
ms = MinMaxScaler()
x_Scaled = pd.DataFrame(ms.fit_transform(x),columns = x.columns)
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	$\blacksquare$
0	1.0	1.0	0.372549	0.25	0.0	0.122054	1.0	ıl.
1	1.0	0.0	0.450980	0.00	0.0	0.133418	1.0	
2	0.0	0.0	0.627451	0.25	0.0	0.893939	1.0	
3	1.0	1.0	0.627451	0.00	0.0	0.135522	1.0	

from sklearn.model\_selection import train\_test\_split

 $x\_train, x\_test, y\_train, y\_test = train\_test\_split(x\_Scaled, y, test\_size = 0.2, random\_state = 0)$ 

print(x\_train.shape,x\_test.shape,y\_train.shape,y\_test.shape)

(564, 7) (142, 7) (564,) (142,)

