→ BHARAT INTERN TASK-2

TITANIC CLASSIFICATION

import pandas as pd import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

titanic = pd.read_csv(r"Titanic-Dataset.csv")

titanic

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S	ılı
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С	7
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S	
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	S	
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	S	
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	S	

titanic.isnull().sum()

PassengerId Survived 0 Pclass 0 0 Name 0 Sex 177 Age SibSp 0 Parch 0 Ticket 0 Fare Cabin 687 Embarked 2 dtype: int64

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

titanic["Embarked"].fillna(titanic["Embarked"].mode()[0],inplace=True)

titanic.isnull().sum()

PassengerId Survived Pclass Name 0 Sex 0 Age SibSp 0 Parch 0 Ticket 0 Fare 0 Cabin 687 Embarked dtype: int64

titanic

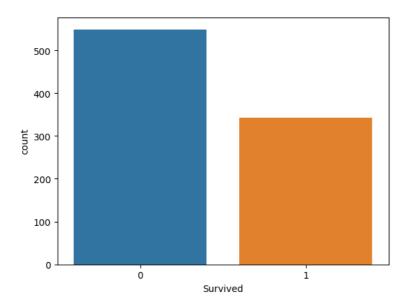
	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	
0	1	0	3	Braund, Mr. Owen Harris	male	22.000000	1	0	A/5 21171	7.2500	NaN	S	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.000000	1	0	PC 17599	71.2833	C85	С	
2	3	1	3	Heikkinen, Miss. Laina	female	26.000000	0	0	STON/O2. 3101282	7.9250	NaN	S	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.000000	1	0	113803	53.1000	C123	S	
4	5	0	3	Allen, Mr. William Henry	male	35.000000	0	0	373450	8.0500	NaN	S	

titanic.drop(["Name"],axis=1,inplace=True)

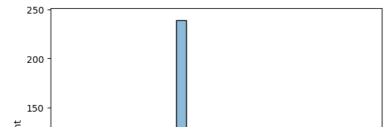
titanic.describe()

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
count	891.000000	891.000000	891.000000	891.000000	891.000000	891.000000	891.000000
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

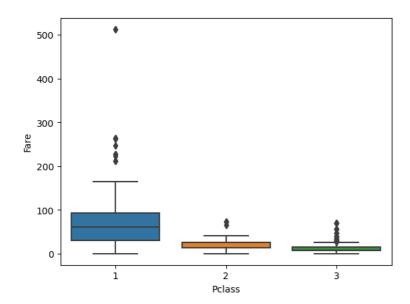
sns.countplot(x="Survived",data = titanic)
plt.show()



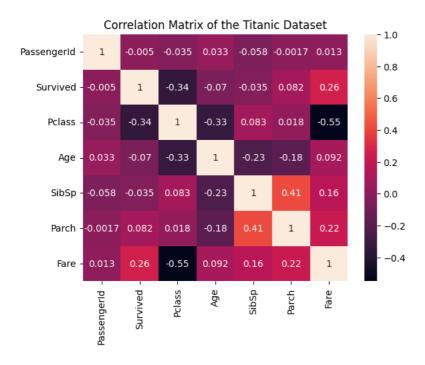
sns.histplot(titanic["Age"],bins=30,kde=True)
plt.show()



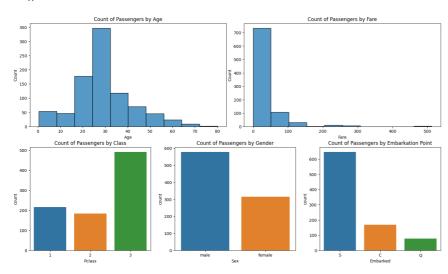
sns.boxplot(x="Pclass",y="Fare",data=titanic)
plt.show()



correlation_matrix = titanic.select_dtypes(include=['number']).corr()
sns.heatmap(correlation_matrix,annot=True)
plt.title("Correlation Matrix of the Titanic Dataset")
plt.show()



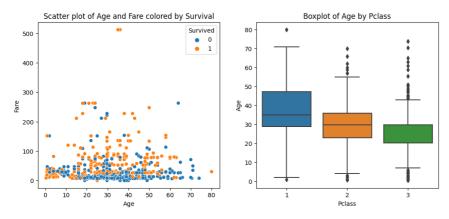
```
plt.figure(figsize=(18,10))
plt.subplot(2,4,(1,2))
sns.histplot(x="Age",bins=10,data=titanic)
plt.title("Count of Passengers by Age")
plt.subplot(2,4,(3,4))
sns.histplot(x="Fare",bins=10,data=titanic)
plt.title("Count of Passengers by Fare")
plt.subplot(2,3,4)
sns.countplot(x='Pclass', data=titanic)
plt.title('Count of Passengers by Class')
plt.subplot(2,3,5)
sns.countplot(x='Sex', data=titanic)
plt.title('Count of Passengers by Gender')
plt.subplot(2,3,6)
sns.countplot(x='Embarked', data=titanic)
plt.title('Count of Passengers by Embarkation Point')
plt.show()
```



```
plt.figure(figsize=(12,5))
plt.subplot(1,2,1)
sns.scatterplot(x="Age",y="Fare",data = titanic,hue = "Survived")
plt.title('Scatter plot of Age and Fare colored by Survival')

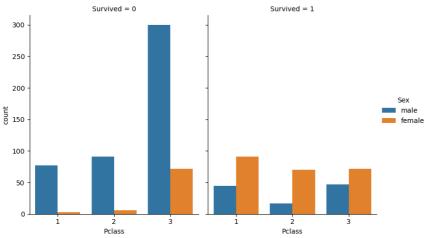
plt.subplot(1,2,2)
sns.boxplot(x="Pclass",y="Age",data=titanic)
plt.title('Boxplot of Age by Pclass')

plt.show()
```



sns.catplot(x="Pclass",hue="Sex",col="Survived",data=titanic,kind ='count',height =5, aspect=0.8)
plt.suptitle("Survival Count by Passenger Class and Gender", y=1.05, fontsize=16)
plt.show()





sns.pairplot(titanic[['Age', 'Fare', 'SibSp', 'Parch', 'Survived']],hue = "Survived",height =2)
plt.suptitle("Multivariate Analysis of Survivors", y=1.02, fontsize=16)
plt.show()

```
Multivariate Analysis of Survivors
         80
         60
       96 40
         20
          n
        500
        400
        300
       Fare
        200
        100
          8
        Sp
import pandas as pd
```

import numpy as np

from sklearn.model_selection import train_test_split

from sklearn.ensemble import RandomForestClassifier

 $from \ sklearn.metrics \ import \ accuracy_score, classification_report$

from sklearn.preprocessing import LabelEncoder

일 | •• •) ٨ • • • label_encoder = LabelEncoder()

titanic['Embarked'] = label_encoder.fit_transform(titanic['Embarked'])

titanic['Sex']=label_encoder.fit_transform(titanic['Sex'])

titanic.drop("Ticket",axis=1,inplace=True)

titanic.drop("PassengerId",axis=1,inplace=True)

titanic

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Cabin	Embarked	
0	0	3	1	22.000000	1	0	7.2500	NaN	2	th
1	1	1	0	38.000000	1	0	71.2833	C85	0	+/
2	1	3	0	26.000000	0	0	7.9250	NaN	2	
3	1	1	0	35.000000	1	0	53.1000	C123	2	
4	0	3	1	35.000000	0	0	8.0500	NaN	2	
886	0	2	1	27.000000	0	0	13.0000	NaN	2	
887	1	1	0	19.000000	0	0	30.0000	B42	2	
888	0	3	0	29.699118	1	2	23.4500	NaN	2	
889	1	1	1	26.000000	0	0	30.0000	C148	0	
890	0	3	1	32.000000	0	0	7.7500	NaN	1	

891 rows × 9 columns

from sklearn.ensemble import RandomForestClassifier from sklearn.model_selection import train_test_split

```
# Assuming 'titanic' is your DataFrame
```

x = titanic.drop("Survived", axis=1)

x = x.select_dtypes(include=['int64', 'float64']) # Keep only numeric columns

y = titanic['Survived']

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)

model = RandomForestClassifier(n_estimators=100, random_state=42) model.fit(x_train, y_train)

RandomForestClassifier RandomForestClassifier(random_state=42)

 $y_pred=model.predict(x_test)$

print("Accuracy score :",accuracy_score(y_test,y_pred)*100,"%")

print("Report :\n",classification_report(y_test,y_pred))

Accuracy score : 81.00558659217877~%

Report :

кероге .	precision	recall	f1-score	support
0	0.82	0.87	0.84	105
1	0.79	0.73	0.76	74
accuracy macro avg	0.81	0.80	0.81 0.80	179 179
weighted avg	0.81	0.81	0.81	179