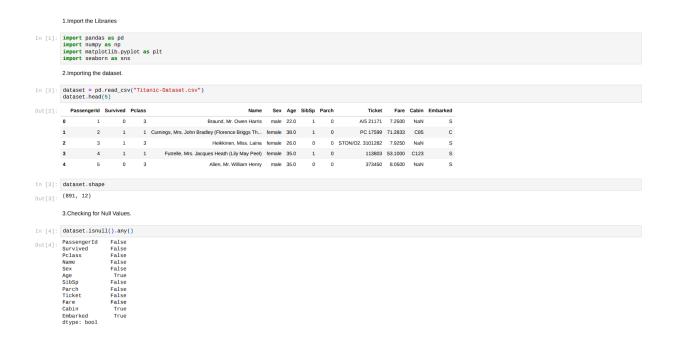
Reg. No.: 21BCE5770

Morning Batch(VIT Chennai)

Assignment 3: Data Preprocessing

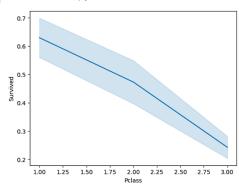


Reg. No.: 21BCE5770

Morning Batch(VIT Chennai)





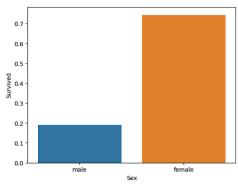


Reg. No.: 21BCE5770

Morning Batch(VIT Chennai)

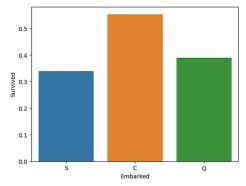
In [13]: #mostly female have survived.
sns.barplot(y=dataset['Survived'], x=dataset['Sex'], ci=0)

Out[13]: <Axes: xlabel='Sex', ylabel='Survived'>



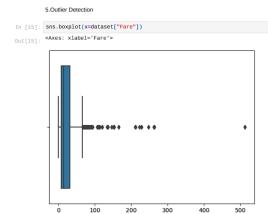
In [14]: #Mostly passengers with embarked C have mostly survived. sns.barplot(y=dataset['Survived'],x=dataset['Embarked'],ci=0)

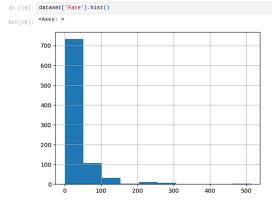
Out[14]: <Axes: xlabel='Embarked', ylabel='Survived'>



Reg. No.: 21BCE5770

Morning Batch(VIT Chennai)





```
In [17]: print('skewness value of Age: ',dataset['Age'].skew())
print('skewness value of Fare: ',dataset['Fare'].skew())
skewness value of Age: 0.5102446555756495
skewness value of Fare: 4.787316519674893
Inference: It shows that fare has outliers that are left skewed.

In [18]: #So we will use flooring and capping for removing outliers
Q1 = dataset['Fare'].quantile(0.25)
Q3 = dataset['Fare'].quantile(0.75)
IONET, whisker q1 = Q1 = (whisker_width*IOR)
upper_whisker = Q1 = (whisker_width*IOR)
upper_whisker = Q1 = (whisker_width*IOR)
upper_whisker = Q3 = (whisker_width*IOR)
dataset['Fare']=np.where(dataset['Fare'])

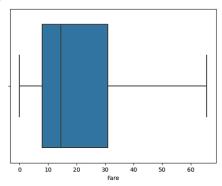
In [19]: sns.boxplot(x=dataset['Fare'])

Out[10]: <a href="Axes: xlabel='Fare'>"
Axes: xlabel='Fare'>"

Out[10]: <a href="Axes: xlabel='Fare'">Axes: xlabel='Fare'>"

Out[10]: <a href="Axes: xlabel='Fare'>"

Out[10]: *Axes: xlabel='Fare'
```



Inference: Hence We have successfully removed outliers.

Reg. No.: 21BCE5770

Morning Batch(VIT Chennai)

6.Splitting Dependent and Independent variables

```
In [20]: #dropping unecessary columns dataset.drop(['PassengerId','Name','Ticket','Cabin'],axis=1,inplace=True)
 In [21]: x=dataset.drop(columns=['Survived'])
y=dataset.iloc[:,0:1]
  In [22]: y.shape
 Out[22]: (891, 1)
               7.Perform Encoding
 In [23]: from sklearn.preprocessing import LabelEncoder le=LabelEncoder()
              x["Sex"]=le.fit_transform(x["Sex"])
sex_mapping = dict(zip(le.classes_, le.transform(le.classes_)))
print("Mapping for 'Sex' column:", sex_mapping)
               Mapping for 'Sex' column: {'female': 0, 'male': 1}
  In [24]: x["Embarked"]=le.fit_transform(x["Embarked"])
               embarked_mapping = dict(zip(le.classes_, le.transform(le.classes_)))
print("Mapping for 'Embarked' column:", embarked_mapping)
               Mapping for 'Embarked' column: {'C': 0, 'Q': 1, 'S': 2}
               8.Feature Scaling
  In [25]: from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
 In [26]: x=sc.fit_transform(x)
In [27]: x
[ 0.82737724, -1.35557354, -0.1846374, ..., 2.08089337, -0.02915533, 0.58595414], [ -1.5661693, 0.72769513, -0.25833799, ..., -0.47367361, 0.29082313, -1.4923022 ], [ 0.82737724, 0.727059513, 0.20276197, ..., -0.47367361, -0.79612661, -0.67817453]])
              9. Splitting Data into Train and Test
In [28]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=0)
In [29]: print(x_train.shape)
    print(y_train.shape)
    print(x_test.shape)
    print(x_test.shape)
              (623, 7)
(623, 1)
(268, 7)
(268, 7)
In [30]: x.shape
Out[30]: (891, 7)
```