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ADS Assignment 2

Titanic Ship Case Study

Problem Description: On April 15, 1912, during her maiden voyage, the Titanic sank after colliding with an iceberg, killing 1502 out of 2224 passengers and crew. Translated 32% survival rate.

- One of the reasons that the shipwreck led to such loss of life was that there were not enough lifeboats for the passengers and crew.
- Although there was some element of luck involved in surviving the sinking, some groups of people were more likely to survive than others, such as women, children, and the upperclass.

The problem associated with the Titanic dataset is to predict whether a passenger survived the disaster or not. The dataset contains various features such as passenger class, age, gender, cabin, fare, and whether the passenger had any siblings or spouses on board. These features can be used to build a predictive model to determine the likelihood of a passenger surviving the disaster. The dataset offers opportunities for feature engineering, data visualization, and model selection, making it a valuable resource for developing and testing data analysis and machine learning skills.

Perform Below Tasks to complete the assignment:-

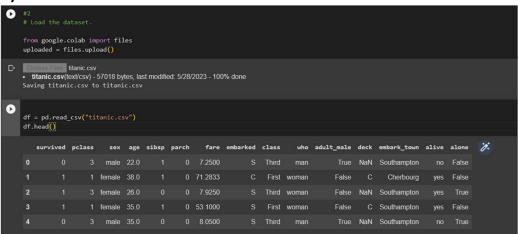
- 1. Download the dataset: Dataset
- 2. Load the dataset.
- 3. Perform Below Visualizations.
 - Univariate Analysis
 - Bi Variate Analysis
 - Multi Variate Analysis
- 4. Perform descriptive statistics on the dataset.
- 5. Handle the Missing values.
- 6. Find the outliers and replace the outliers
- 7. Check for Categorical columns and perform encoding.
- 8. Split the data into dependent and independent variables.
- 9. Scale the independent variables
- 10. Split the data into training and testing

1)

```
#1
# Download the dataset: Dataset

import numpy as np
import pandas as pd
import seaborn as sns
```

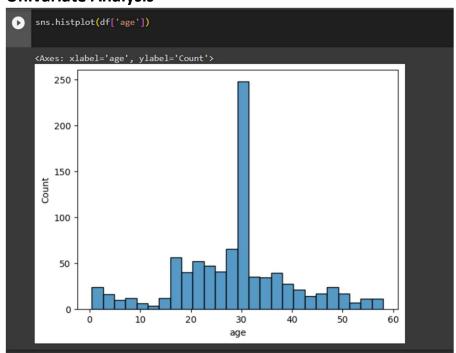
2)

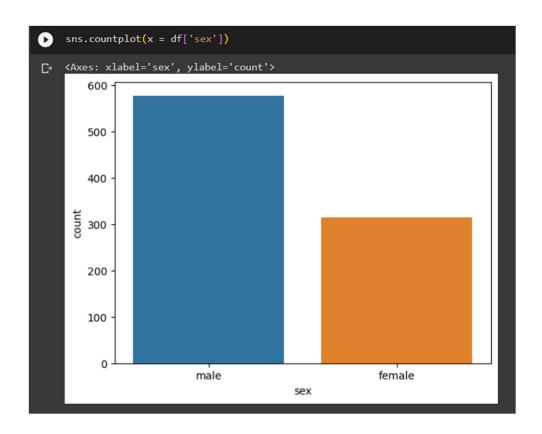


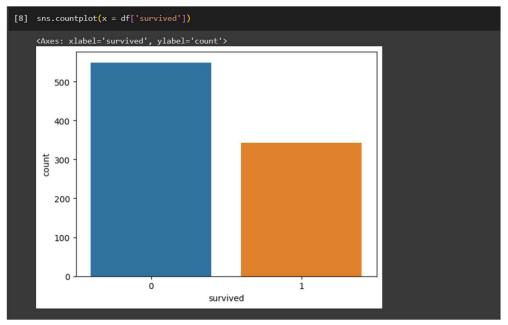
```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):
                Non-Null Count Dtype
# Column
0 survived 891 non-null
                                int64
    pclass
                 891 non-null
                                 int64
                 891 non-null
                                object
    sex
                 714 non-null
                                float64
    age
                 891 non-null
    parch
                 891 non-null
                                int64
  fare
                 891 non-null
                                float64
    embarked
                 889 non-null
                                object
   class
                 891 non-null
                                object
 9 who
                 891 non-null
                                object
 10 adult_male 891 non-null
                                bool
11 deck 203 non-null
12 embark_town 889 non-null
                                object
                                object
 13 alive
                 891 non-null
                                object
14 alone
                891 non-null
                                bool
dtypes: bool(2), float64(2), int64(4), object(7)
memory usage: 92.4+ KB
```

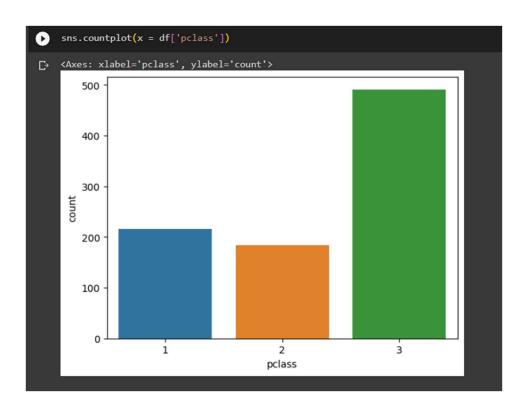
3)

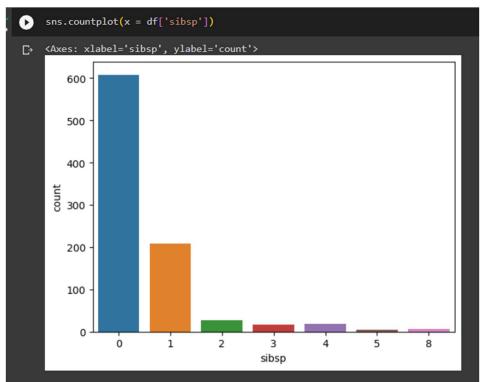
• Univariate Analysis

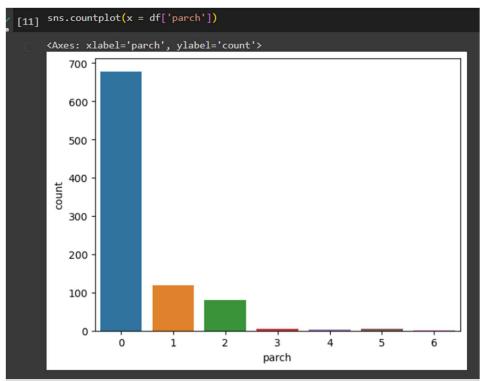


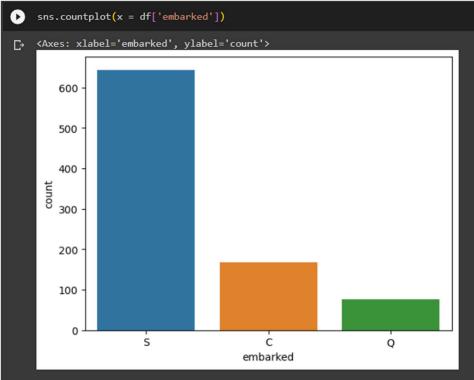






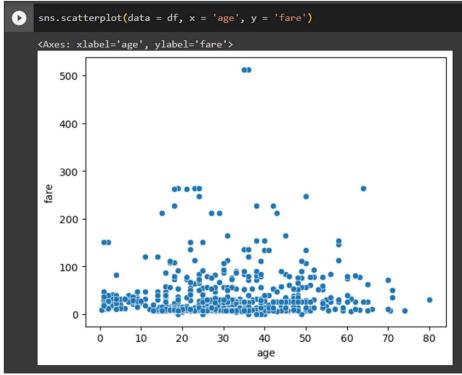


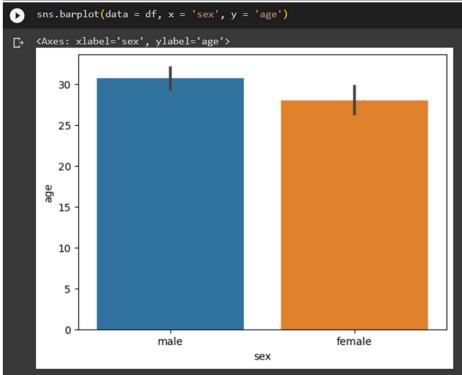


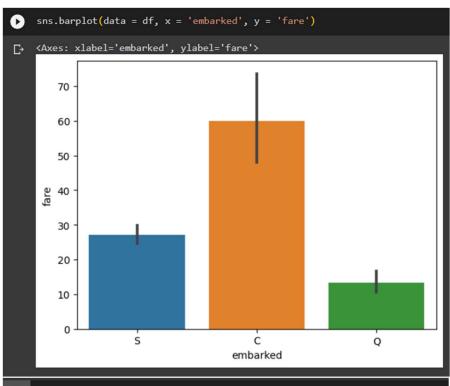


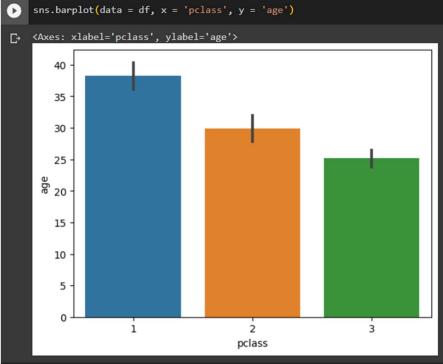
bivariate analysis

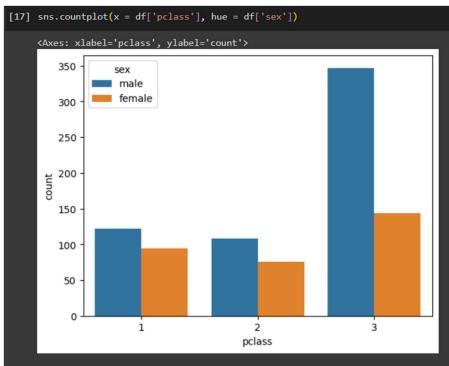
•

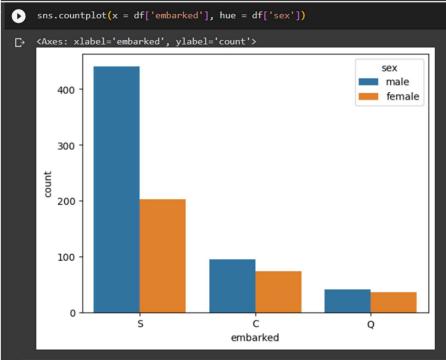


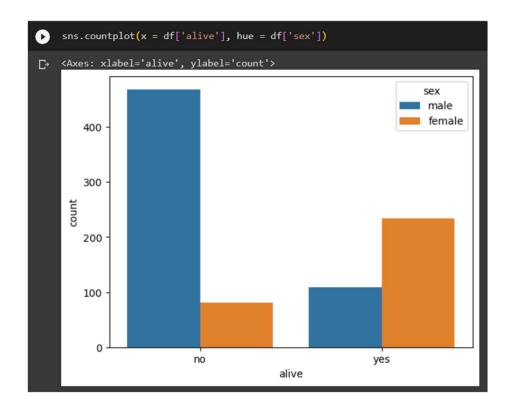




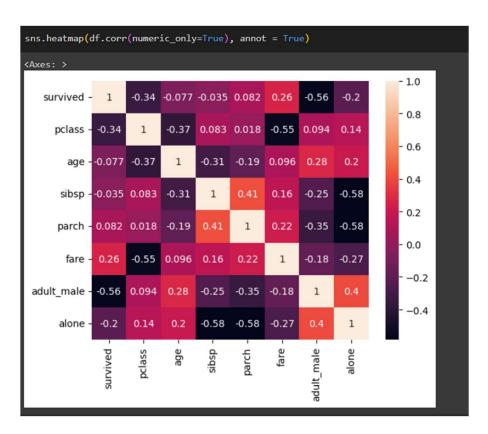








multivariate analysis



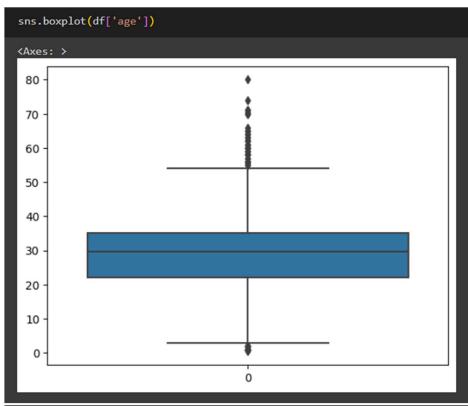
4)

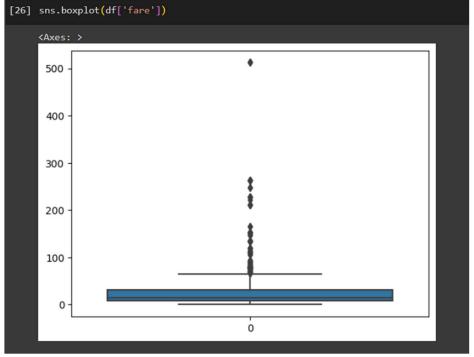
df.describe()							
	survived	pclass	age	sibsp	parch	fare	%
count	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000	
mean	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208	
std	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429	
min	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000	
25%	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400	
50%	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200	
75%	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000	
max	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200	

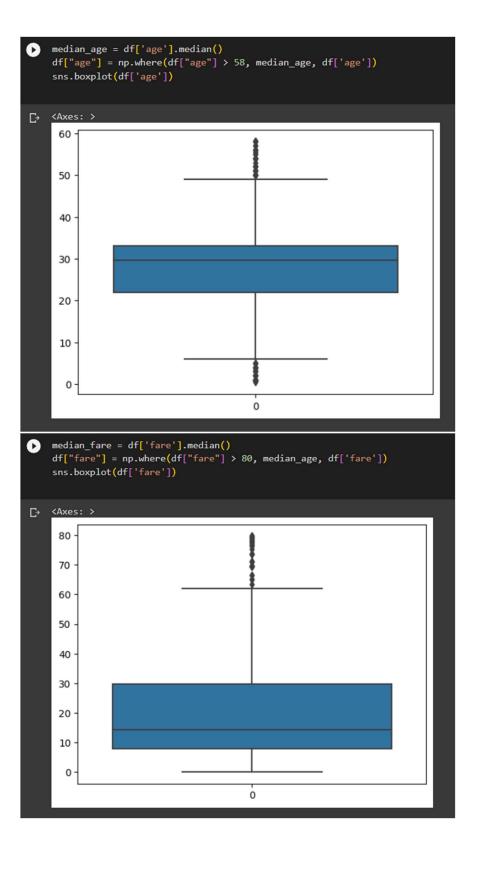
5)

```
df.isnull().sum()
                 0
survived
pclass
                 0
sex
               177
age
sibsp
                 0
parch
                 0
fare
embarked
class
who
                 0
adult_male
                 0
deck
               688
embark_town
alive
alone
                 0
dtype: int64
```

```
df.dropna(subset=['embark_town'], how='all', inplace = True)
df['age'] = df['age'].fillna(df['age'].mean())
    df.drop(['deck'], axis = 1,inplace = True)
    df.isnull().sum()
pclass
    age
    sibsp
    parch
    fare
    embarked
    class
    who
    adult_male
    embark_town
    alive
    alone
    dtype: int64
```







2

3

4

yes

yes

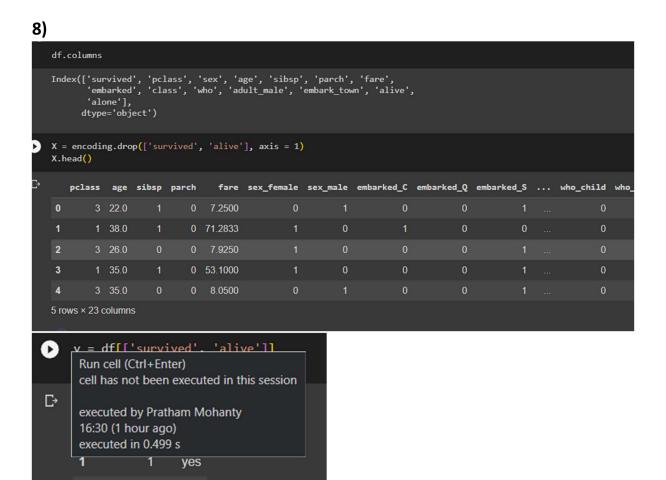
no

0

```
#7
# Check for Categorical columns and perform encoding.

from sklearn.preprocessing import OneHotEncoder encoding = pd.get_dummies(df, columns = ['sex', 'embarked', 'class', 'who', 'adult_male', 'embark_town', 'alone'])

survived pclass age sibsp parch fare alive sex_female sex_male embarked_C ... who_child who_man who_woman accompany and the process of the part of the part of the process of the part of the pa
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10)