

# ADS Assignment 2

Name:Damini N

2. Load the dataset.

In [4]:

```
import pandas as pd

# Specify the path to your CSV file
csv_file_path = ('C:/Users/Damini N/OneDrive/Desktop/ADS/titanic.csv')

# Load the dataset from the CSV file
dataset = pd.read_csv(csv_file_path)
```

In [5]:

```
# Load the dataset without a header row
dataset = pd.read_csv(csv_file_path, header=None)
```

In [6]:

```
dataset
```

Out[6]:

	0	1	2	3	4	5	6	7	8	9	
0	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male
1	0	3	male	22.0	1	0	7.25	S	Third	man	True
2	1	1	female	38.0	1	0	71.2833	C	First	woman	False
3	1	3	female	26.0	0	0	7.925	S	Third	woman	False
4	1	1	female	35.0	1	0	53.1	S	First	woman	False
...	...	...	...	...	...	...	...	...	...	...	...
887	0	2	male	27.0	0	0	13.0	S	Second	man	True
888	1	1	female	19.0	0	0	30.0	S	First	woman	False
889	0	3	female	NaN	1	2	23.45	S	Third	woman	False
890	1	1	male	26.0	0	0	30.0	C	First	man	True
891	0	3	male	32.0	0	0	7.75	Q	Third	man	True

892 rows × 15 columns



In [7]:

```

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# Load the Titanic dataset
df = pd.read_csv('C:/Users/Damini N/OneDrive/Desktop/ADS/titanic.csv')

# Display the first few rows of the dataset
print(df.head())

```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class
0	0	3	male	22.0	1	0	7.2500	S	Third
1	1	1	female	38.0	1	0	71.2833	C	First
2	1	3	female	26.0	0	0	7.9250	S	Third
3	1	1	female	35.0	1	0	53.1000	S	First
4	0	3	male	35.0	0	0	8.0500	S	Third

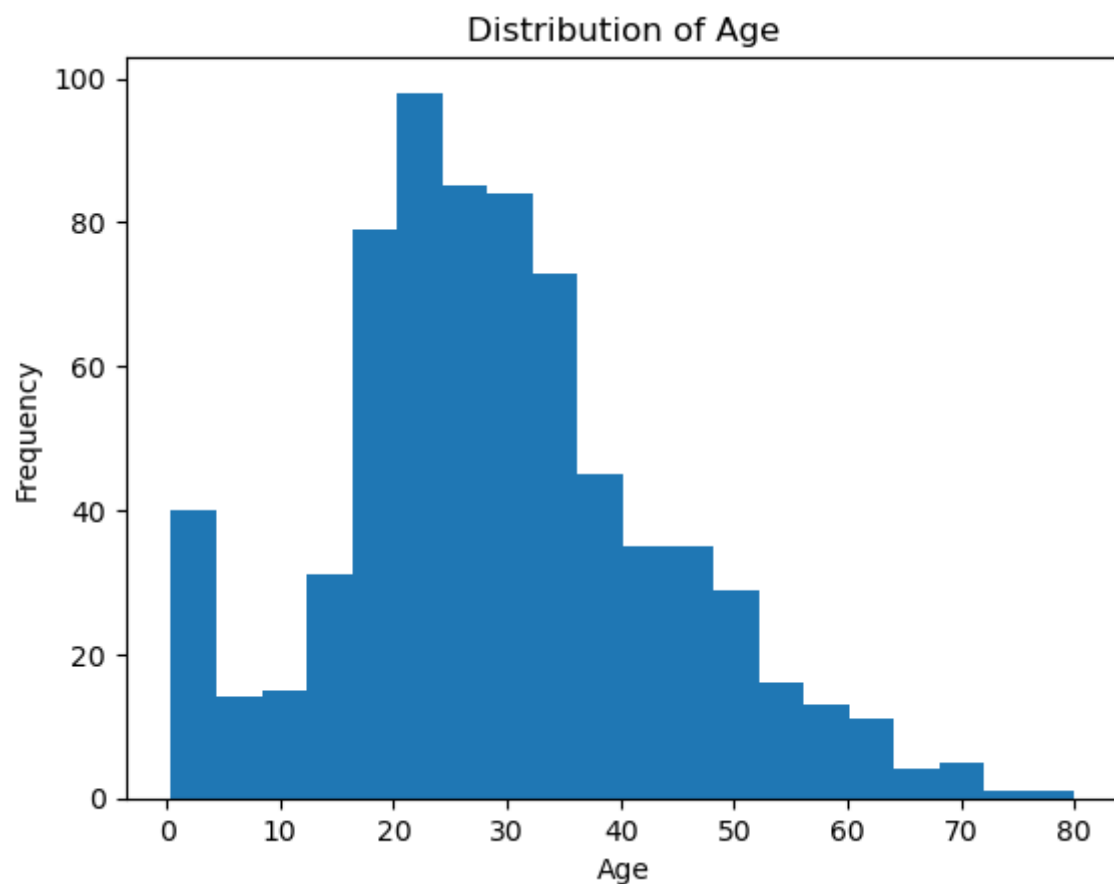
	who	adult_male	deck	embark_town	alive	alone
0	man	True	NaN	Southampton	no	False
1	woman	False	C	Cherbourg	yes	False
2	woman	False	NaN	Southampton	yes	True
3	woman	False	C	Southampton	yes	False
4	man	True	NaN	Southampton	no	True

3.Perform Below Visualizations. • Univariate Analysis • Bi - Variate Analysis • Multi - Variate Analysis

In [12]:

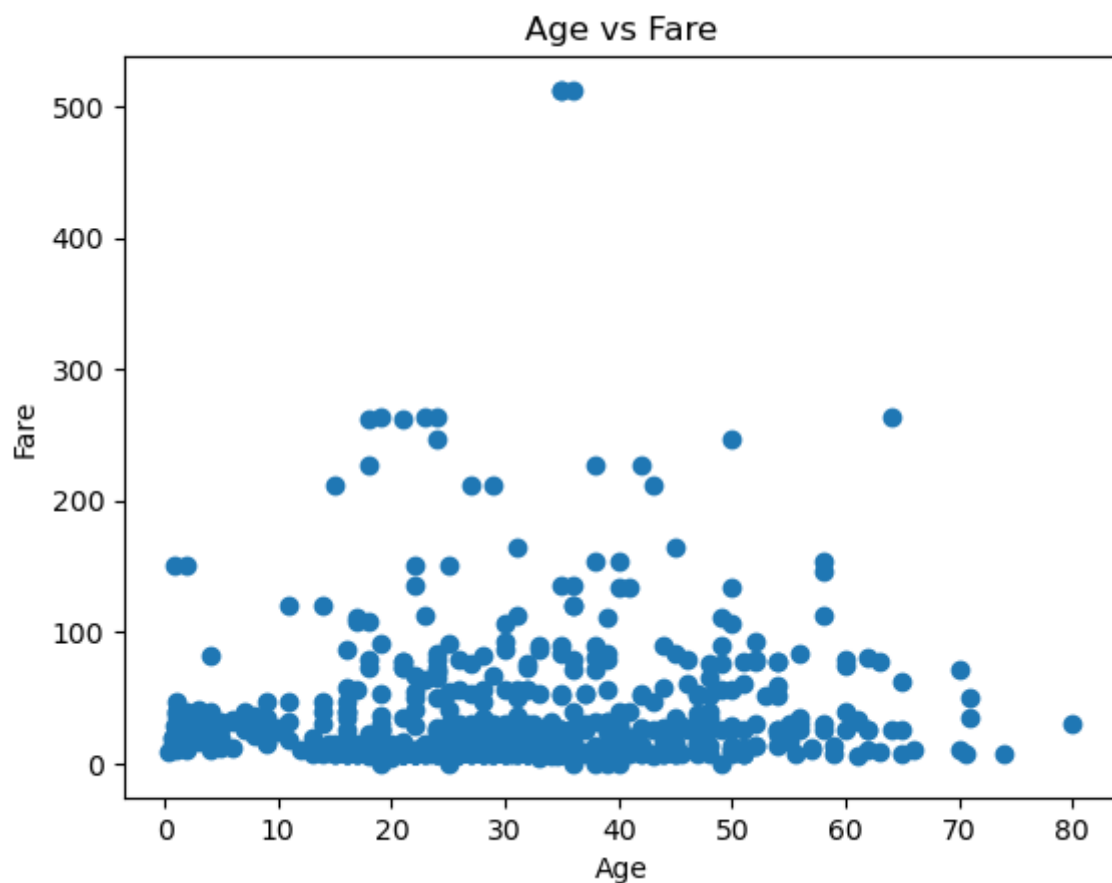
```
import matplotlib.pyplot as plt

# Plotting histogram for age
plt.hist(df['age'].dropna(), bins=20)
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.title('Distribution of Age')
plt.show()
```



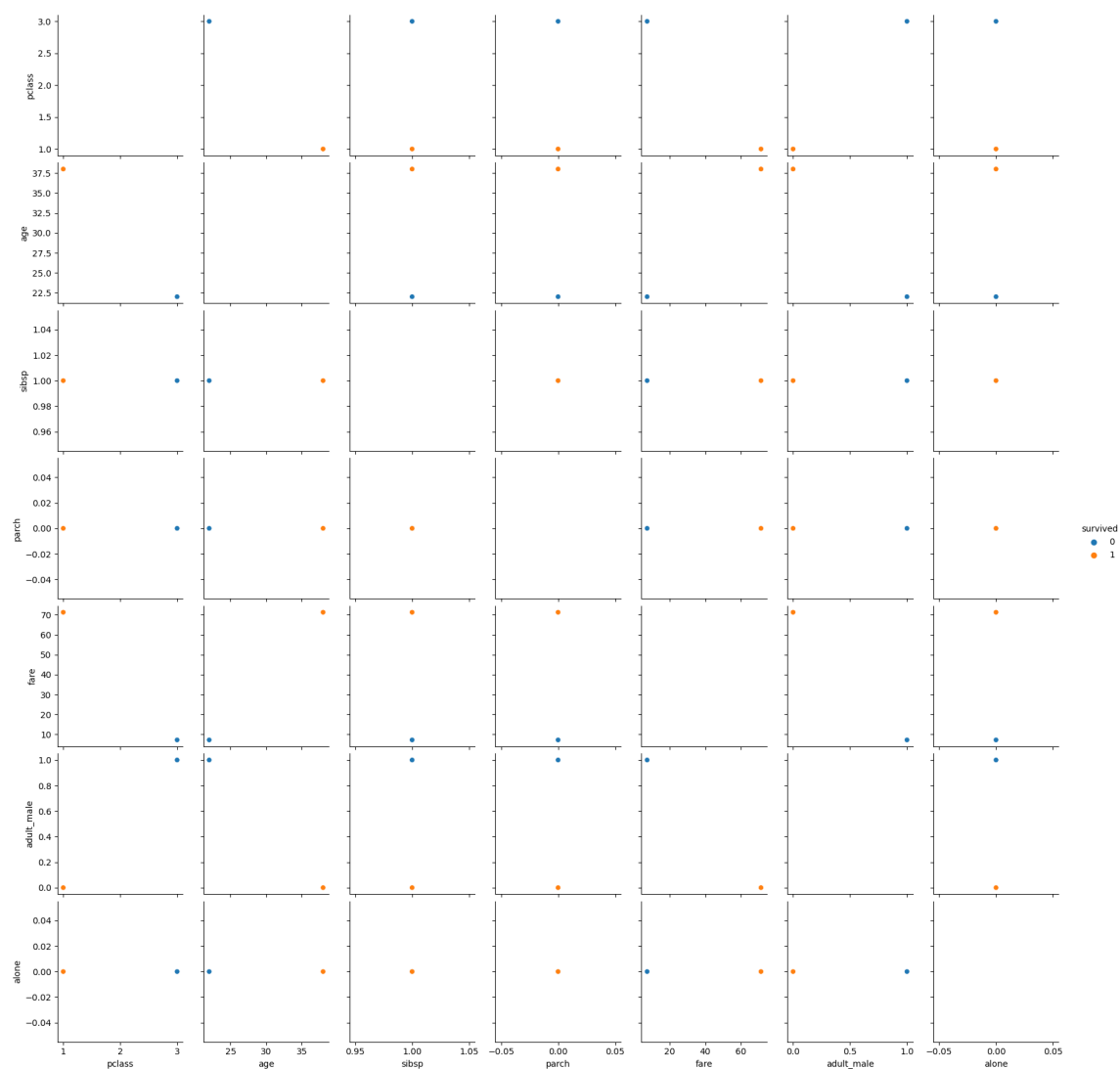
In [13]:

```
# Plotting scatter plot for age and fare  
plt.scatter(df['age'], df['fare'])  
plt.xlabel('Age')  
plt.ylabel('Fare')  
plt.title('Age vs Fare')  
plt.show()
```



In [15]:

```
# Multivariate analysis using seaborn  
sns.pairplot(df, hue='survived')  
plt.show()
```



In [16]:

```

import pandas as pd

# Create a DataFrame from the given dataset
data = {
    'survived': [0, 1],
    'pclass': [3, 1],
    'sex': ['male', 'female'],
    'age': [22, 38],
    'sibsp': [1, 1],
    'parch': [0, 0],
    'fare': [7.25, 71.2833],
    'embarked': ['S', 'C'],
    'class': ['Third', 'First'],
    'who': ['man', 'woman'],
    'adult_male': [True, False],
    'deck': ['', 'C'],
    'embark_town': ['Southampton', 'Cherbourg'],
    'alive': ['no', 'yes'],
    'alone': [False, False]
}

df = pd.DataFrame(data)

# Compute descriptive statistics
statistics = df.describe()

# Display the statistics
print(statistics)

```

	survived	pclass	age	sibsp	parch	fare
count	2.000000	2.000000	2.000000	2.0	2.0	2.000000
mean	0.500000	2.000000	30.000000	1.0	0.0	39.266650
std	0.707107	1.414214	11.313708	0.0	0.0	45.278381
min	0.000000	1.000000	22.000000	1.0	0.0	7.250000
25%	0.250000	1.500000	26.000000	1.0	0.0	23.258325
50%	0.500000	2.000000	30.000000	1.0	0.0	39.266650
75%	0.750000	2.500000	34.000000	1.0	0.0	55.274975
max	1.000000	3.000000	38.000000	1.0	0.0	71.283300

5. Handle the Missing values.

In [17]:

```
df.dropna(inplace=True)
```

6. Find the outliers and replace the outliers

In [22]:

```

import pandas as pd
import numpy as np

# Create a DataFrame from the given dataset
data = {
    'survived': [0, 1],
    'pclass': [3, 1],
    'sex': ['male', 'female'],
    'age': [22, 38],
    'sibsp': [1, 1],
    'parch': [0, 0],
    'fare': [7.25, 71.2833],
    'embarked': ['S', 'C'],
    'class': ['Third', 'First'],
    'who': ['man', 'woman'],
    'adult_male': [True, False],
    'deck': ['', 'C'],
    'embark_town': ['Southampton', 'Cherbourg'],
    'alive': ['no', 'yes'],
    'alone': [False, False]
}

df = pd.DataFrame(data)

# Identify outliers using the IQR method
Q1 = df['fare'].quantile(0.25)
Q3 = df['fare'].quantile(0.75)
IQR = Q3 - Q1

lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

outliers = df[(df['fare'] < lower_bound) | (df['fare'] > upper_bound)]
print("Outliers:")
print(outliers)

# Replace outliers with a specified value
replacement_value = df['fare'].median()
df.loc[(df['fare'] < lower_bound) | (df['fare'] > upper_bound), 'fare'] = replacement_value

print("\nAfter replacing outliers:")
print(df)

```

Outliers:

Empty DataFrame

Columns: [survived, pclass, sex, age, sibsp, parch, fare, embarked, class, who, adult\_male, deck, embark\_town, alive, alone]

Index: []

After replacing outliers:

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	\
0	0	3	male	22	1	0	7.2500	S	Third	
1	1	1	female	38	1	0	71.2833	C	First	

	who	adult_male	deck	embark_town	alive	alone
0	man	True		Southampton	no	False
1	woman	False	C	Cherbourg	yes	False

In [23]:

```
import pandas as pd
from sklearn.preprocessing import OneHotEncoder

# Create a DataFrame from the given dataset
data = {
    'survived': [0, 1],
    'pclass': [3, 1],
    'sex': ['male', 'female'],
    'age': [22, 38],
    'sibsp': [1, 1],
    'parch': [0, 0],
    'fare': [7.25, 71.2833],
    'embarked': ['S', 'C'],
    'class': ['Third', 'First'],
    'who': ['man', 'woman'],
    'adult_male': [True, False],
    'deck': ['', 'C'],
    'embark_town': ['Southampton', 'Cherbourg'],
    'alive': ['no', 'yes'],
    'alone': [False, False]
}

df = pd.DataFrame(data)

# Identify categorical columns
categorical_cols = df.select_dtypes(include=['object']).columns.tolist()
print("Categorical columns:", categorical_cols)

# Perform one-hot encoding
encoder = OneHotEncoder(sparse=False, drop='first')
encoded_cols = pd.DataFrame(encoder.fit_transform(df[categorical_cols]))
encoded_cols.columns = encoder.get_feature_names(categorical_cols)

# Replace categorical columns with encoded columns
df.drop(columns=categorical_cols, inplace=True)
df = pd.concat([df, encoded_cols], axis=1)

# Display the encoded dataset
print("\nEncoded dataset:")
print(df)
```



Categorical columns: ['sex', 'embarked', 'class', 'who', 'deck', 'embark\_town', 'alive']

Encoded dataset:

	survived	pclass	age	sibsp	parch	fare	adult_male	alone	sex_male
0	0	3	22	1	0	7.2500	True	False	
1	1	1	38	1	0	71.2833	False	False	

	embarked_S	class_Third	who_woman	deck_C	embark_town_Southampton
0	1.0	1.0	0.0	0.0	1.0
1	0.0	0.0	1.0	1.0	0.0

	alive_yes
0	0.0
1	1.0

C:\Users\Damini N\anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureWarning: Function get\_feature\_names is deprecated; get\_feature\_names is deprecated in 1.0 and will be removed in 1.2. Please use get\_feature\_names\_out instead.

warnings.warn(msg, category=FutureWarning)

In [24]:

```

import pandas as pd

# Create a DataFrame from the given dataset
data = {
    'survived': [0, 1],
    'pclass': [3, 1],
    'sex': ['male', 'female'],
    'age': [22, 38],
    'sibsp': [1, 1],
    'parch': [0, 0],
    'fare': [7.25, 71.2833],
    'embarked': ['S', 'C'],
    'class': ['Third', 'First'],
    'who': ['man', 'woman'],
    'adult_male': [True, False],
    'deck': ['', 'C'],
    'embark_town': ['Southampton', 'Cherbourg'],
    'alive': ['no', 'yes'],
    'alone': [False, False]
}

df = pd.DataFrame(data)

# Split into features (independent variables) and target (dependent variable)
X = df.drop('survived', axis=1)
y = df['survived']

# Display the independent variables
print("Independent Variables (X):")
print(X)

# Display the dependent variable
print("\nDependent Variable (y):")
print(y)

```

Independent Variables (X):

	pclass	sex	age	sibsp	parch	fare	embarked	class	who	\
0	3	male	22	1	0	7.2500	S	Third	man	
1	1	female	38	1	0	71.2833	C	First	woman	

	adult_male	deck	embark_town	alive	alone
0	True		Southampton	no	False
1	False	C	Cherbourg	yes	False

Dependent Variable (y):

```

0    0
1    1
Name: survived, dtype: int64

```

In [25]:

```

import pandas as pd
from sklearn.preprocessing import MinMaxScaler

# Create a DataFrame from the given dataset
data = {
    'survived': [0, 1],
    'pclass': [3, 1],
    'sex': ['male', 'female'],
    'age': [22, 38],
    'sibsp': [1, 1],
    'parch': [0, 0],
    'fare': [7.25, 71.2833],
    'embarked': ['S', 'C'],
    'class': ['Third', 'First'],
    'who': ['man', 'woman'],
    'adult_male': [True, False],
    'deck': ['', 'C'],
    'embark_town': ['Southampton', 'Cherbourg'],
    'alive': ['no', 'yes'],
    'alone': [False, False]
}

df = pd.DataFrame(data)

# Select only the numeric columns to scale
numeric_cols = df.select_dtypes(include='number').columns.tolist()

# Scale the numeric columns using Min-Max scaling
scaler = MinMaxScaler()
df[numeric_cols] = scaler.fit_transform(df[numeric_cols])

# Display the scaled dataset
print(df)

```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	wh
0	0.0	1.0	male	0.0	0.0	0.0	0.0	S	Third	ma
1	1.0	0.0	female	1.0	0.0	0.0	1.0	C	First	woma

	adult_male	deck	embark_town	alive	alone
0	True		Southampton	no	False
1	False	C	Cherbourg	yes	False

10. Split the data into training and testing

In [26]:

```
import pandas as pd
from sklearn.model_selection import train_test_split

# Create a DataFrame from the given dataset
data = {
    'survived': [0, 1],
    'pclass': [3, 1],
    'sex': ['male', 'female'],
    'age': [22, 38],
    'sibsp': [1, 1],
    'parch': [0, 0],
    'fare': [7.25, 71.2833],
    'embarked': ['S', 'C'],
    'class': ['Third', 'First'],
    'who': ['man', 'woman'],
    'adult_male': [True, False],
    'deck': ['', 'C'],
    'embark_town': ['Southampton', 'Cherbourg'],
    'alive': ['no', 'yes'],
    'alone': [False, False]
}

df = pd.DataFrame(data)

# Split the dataset into features (X) and target (y)
X = df.drop('survived', axis=1)
y = df['survived']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Display the shapes of the resulting sets
print("Training set - X shape:", X_train.shape)
print("Training set - y shape:", y_train.shape)
print("Testing set - X shape:", X_test.shape)
print("Testing set - y shape:", y_test.shape)
```

```
Training set - X shape: (1, 14)
Training set - y shape: (1,)
Testing set - X shape: (1, 14)
Testing set - y shape: (1,)
```

In [ ]: