SMART BRIDGE INTERNSHIP APPLIED DATA SCIENCE ASSIGNMENT-2

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```
In [1]: import pandas as pd
data=pd.read_csv('titanic.csv')
```

```
In [2]: print(data)
```

```
survived pclass
                                                              age sibsp parch
                                                                                                     fare embarked

        male
        22.0
        1
        0
        7.2500
        S
        Third

        emale
        38.0
        1
        0
        71.2833
        C
        First

        emale
        26.0
        0
        0
        7.9250
        S
        Third

        emale
        35.0
        1
        0
        53.1000
        S
        First

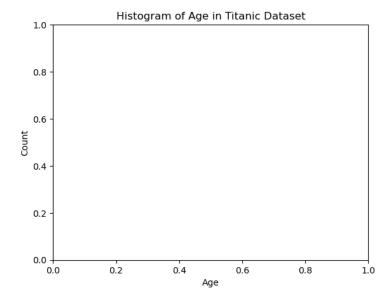
                         0 3
                                      1 female 38.0
        1
                            1
                                         3 female 26.0
                                       1 female 35.0
                who adult_male deck embark_town alive
                                                                                                                alone
                                         True NaN Southampton no False
0
                man
           woman
                                       False C
                                                                        Cherbourg yes False
```

```
In [3]: import matplotlib.pyplot as plt
In [4]: age_column = data['age']
In [5]: plt.hist(age_column, bins=20, edgecolor='black')
```

```
In [5]: plt.hist(age_column, bins=20, edgecolor='black')
<BarContainer object of 20 artists>)
      100
      80
      60
       40
      20
              10
                   20
                       30
                                50
                                              80
                            40
                                     60
                                         70
```

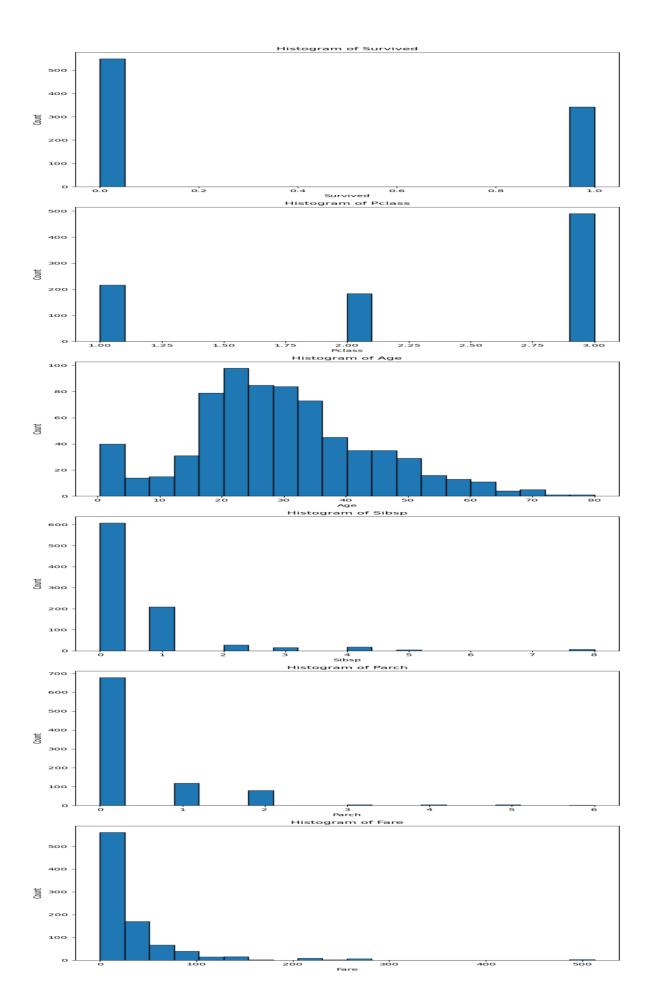
```
In [6]: # Set the labels and title
plt.xlabel('Age')
plt.ylabel('Count')
plt.title('Histogram of Age in Titanic Dataset')
```

Out[6]: Text(0.5, 1.0, 'Histogram of Age in Titanic Dataset')



In [7]: plt.show()

```
In [8]: import matplotlib.pyplot as plt
         # Assuming 'data' is your DataFrame
         # List of columns to create histograms for
columns = ['survived', 'pclass', 'age', 'sibsp', 'parch', 'fare']
         # Set up the figure and subplots
         fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))
         # Create histograms for each column
         for i, column in enumerate(columns):
            # Select the column
             data_column = data[column]
             # Create the histogram
axes[i].hist(data_column, bins=20, edgecolor='black')
             # Set the labels and title for each subplot
             axes[i].set_xlabel(column.capitalize())
             axes[i].set_ylabel('Count')
axes[i].set_title(f'Histogram of {column.capitalize()}')
         # Adjust the spacing between subplots
         plt.tight_layout()
         # Display the histograms
plt.show()
```



```
In [9]: import pandas as pd
import matplotlib.pyplot as plt

# Assuming 'data' is your DataFrame

# List of columns to create bar charts for
columns = ['survived', 'pclass', 'sex', 'age', 'sibsp', 'parch', 'fare', 'embarked', 'class', 'who', 'adult_male', 'deck', 'embarked'
# Set up the figure and subplots
fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))

# Create bar charts for each column
for i, column in enumerate(columns):
# Select the column
column_data = data[column]

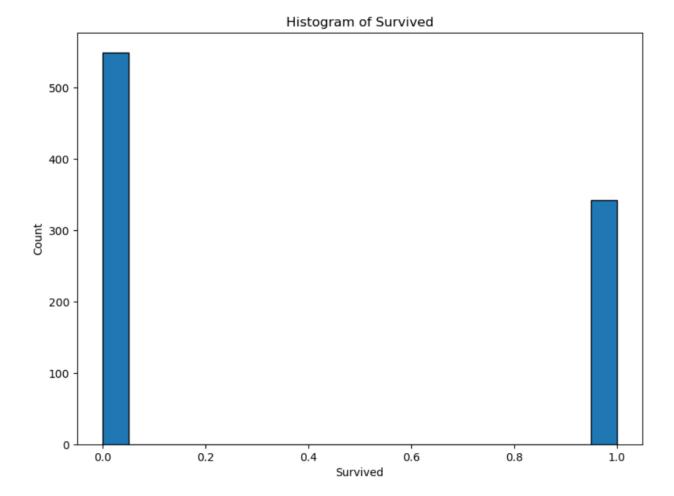
# Calculate the frequencies or counts
counts = column_data.value_counts()

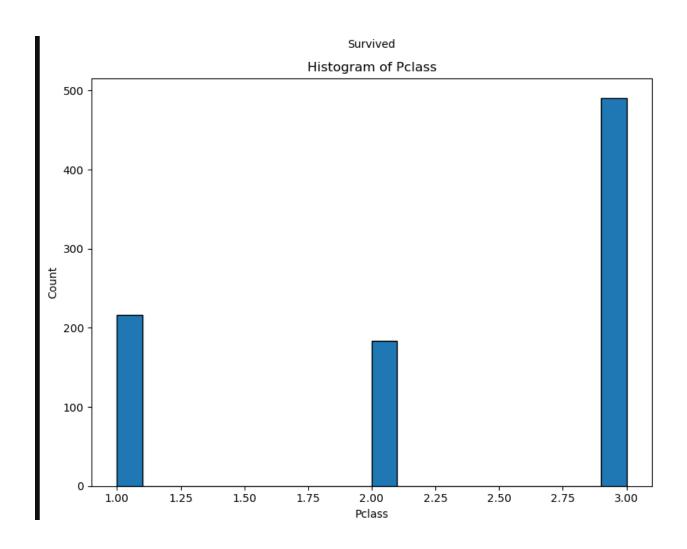
# Create the bar chart
axes[i].bar(counts.index, counts.values)

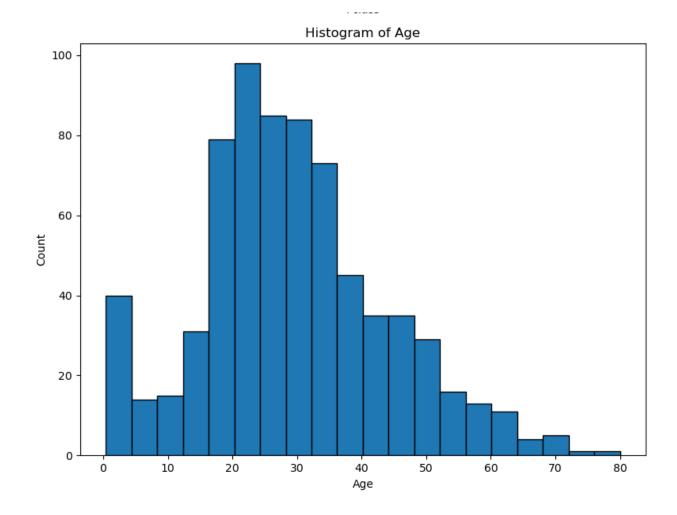
# Set the labels and title for each subplot|
axes[i].set_xlabel(column.capitalize())
axes[i].set_xlabel(column.capitalize())
axes[i].set_xlabel(column.capitalize())
# Adjust the spacing between subplots
plt.tight_layout()

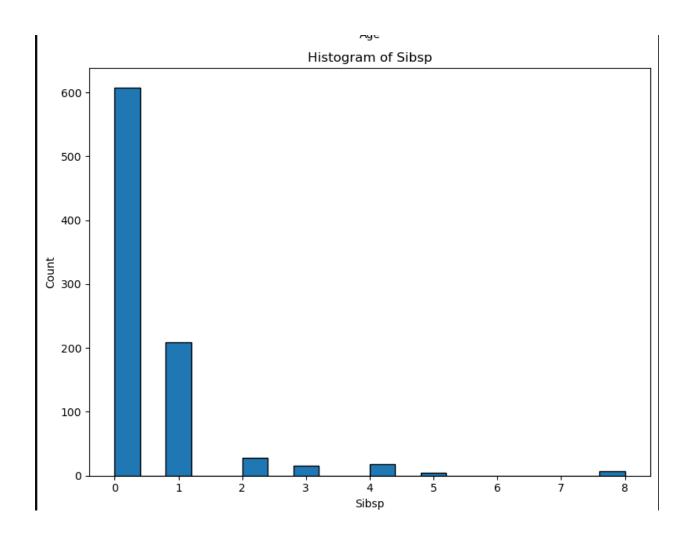
# Display the bar charts
plt.show()
```

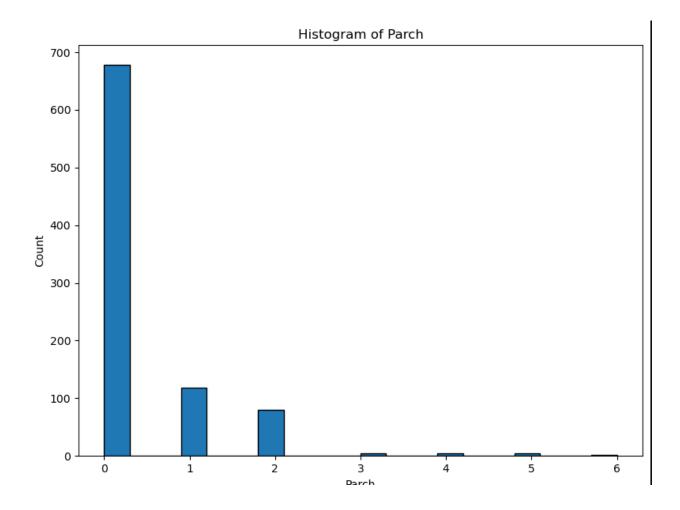


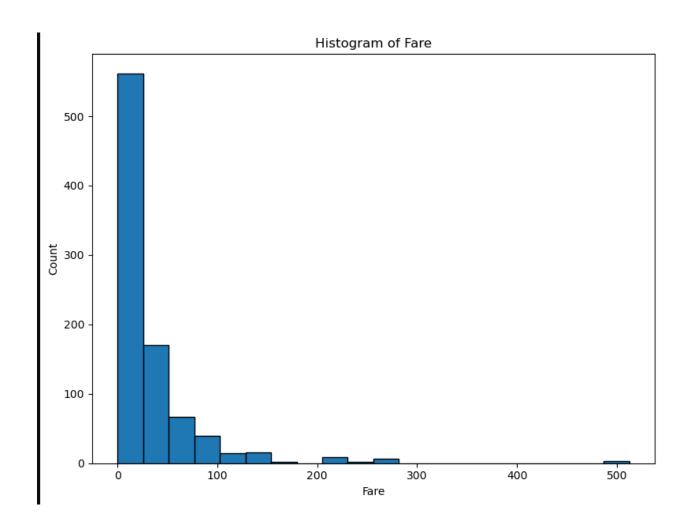




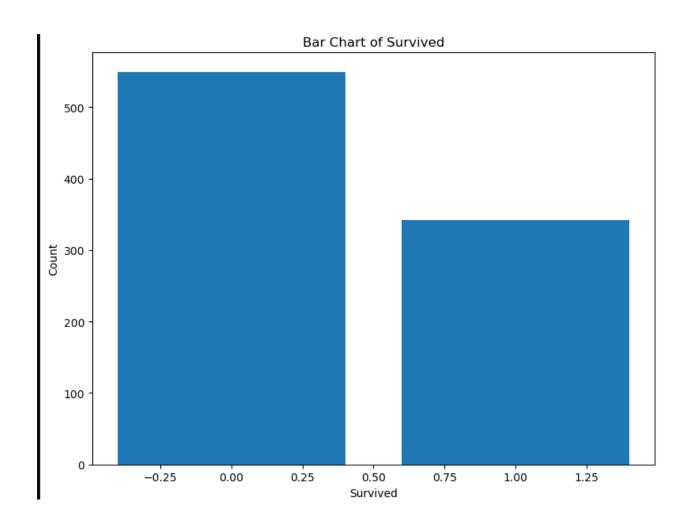


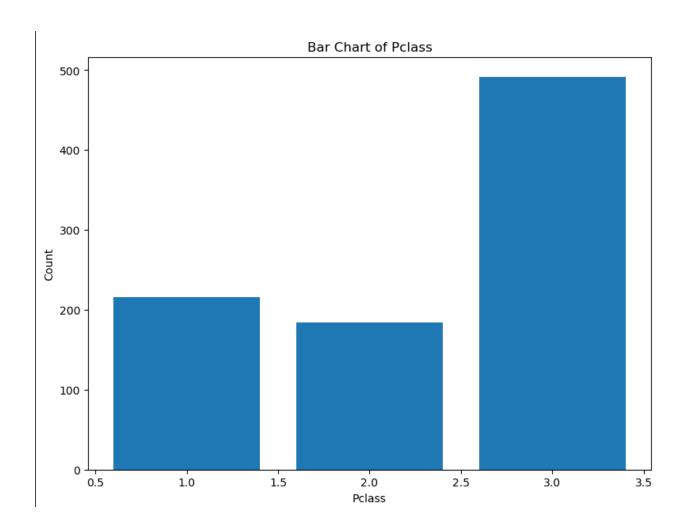


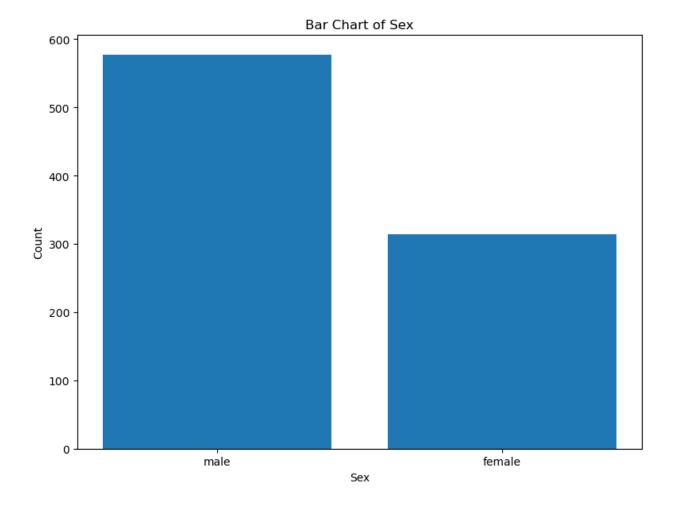


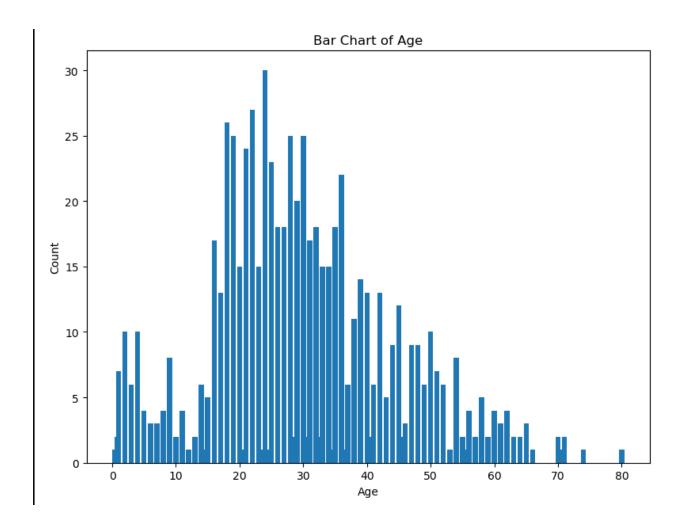


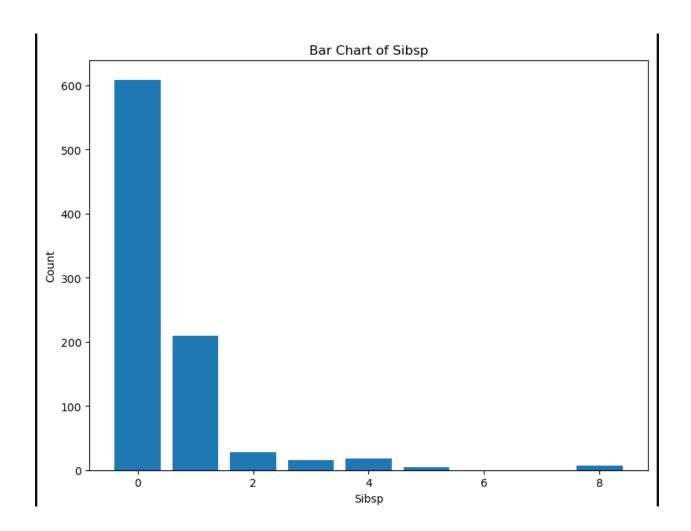
```
In [9]: import pandas as pd
        import matplotlib.pyplot as plt
        # Assuming 'data' is your DataFrame
        # List of columns to create bar charts for
        columns = ['survived', 'pclass', 'sex', 'age', 'sibsp', 'parch', 'fare', 'embarked', 'class', 'who', 'a
        # Set up the figure and subplots
        fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))
        # Create bar charts for each column
        for i, column in enumerate(columns):
            # Select the column
            column_data = data[column]
            # Calculate the frequencies or counts
            counts = column_data.value_counts()
            # Create the bar chart
            axes[i].bar(counts.index, counts.values)
            # Set the labels and title for each subplot
            axes[i].set_xlabel(column.capitalize())
            axes[i].set_ylabel('Count')
            axes[i].set_title(f'Bar Chart of {column.capitalize()}')
        # Adjust the spacing between subplots
plt.tight_layout()
        # Display the bar charts
        plt.show()
```

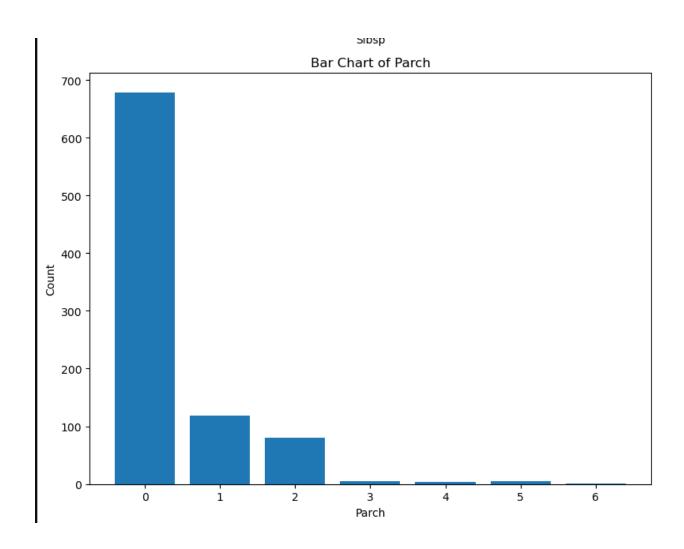


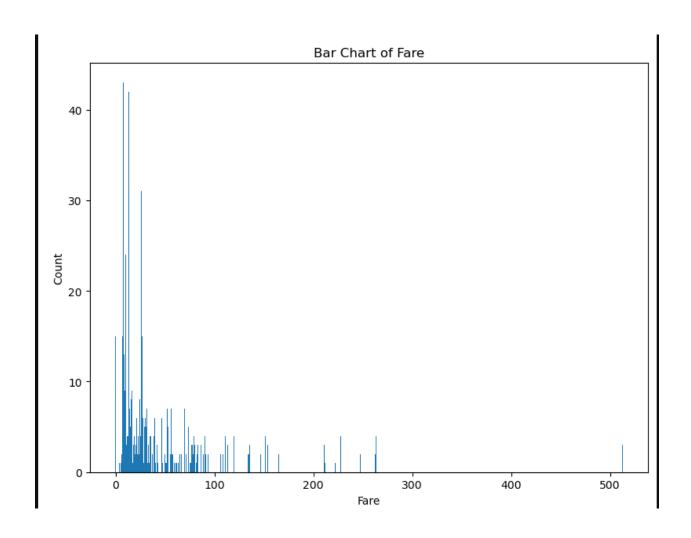


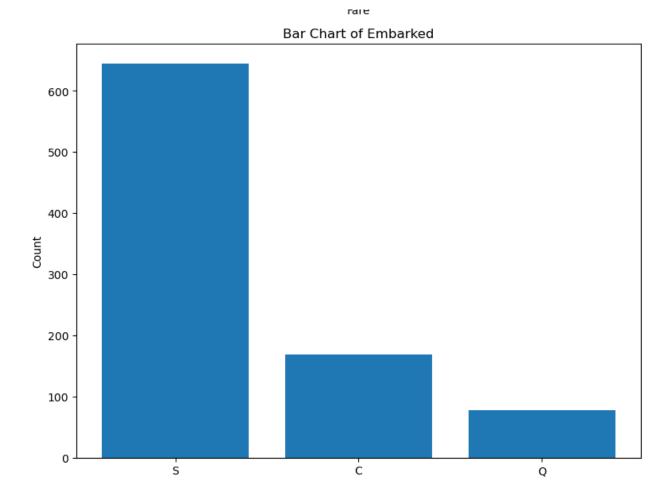




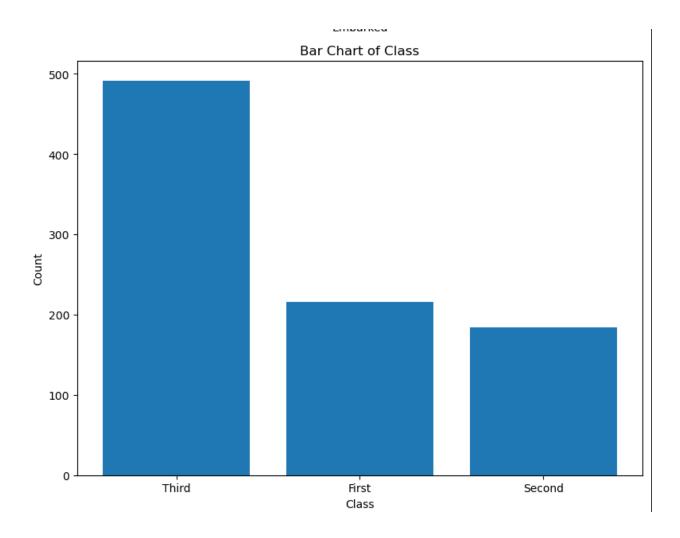


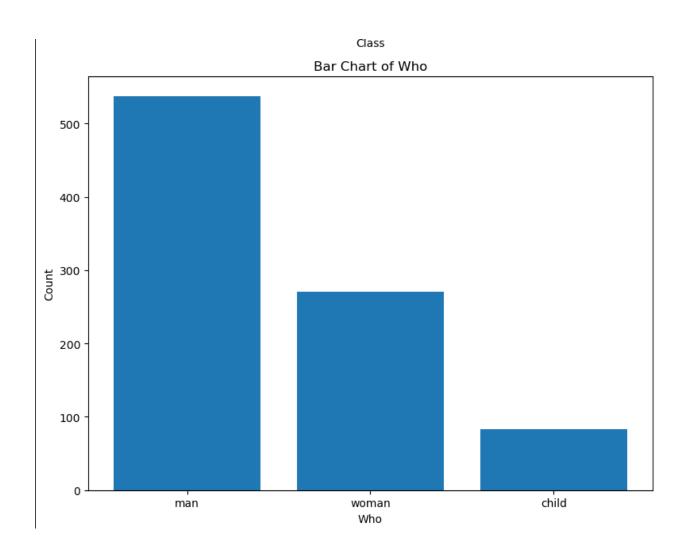


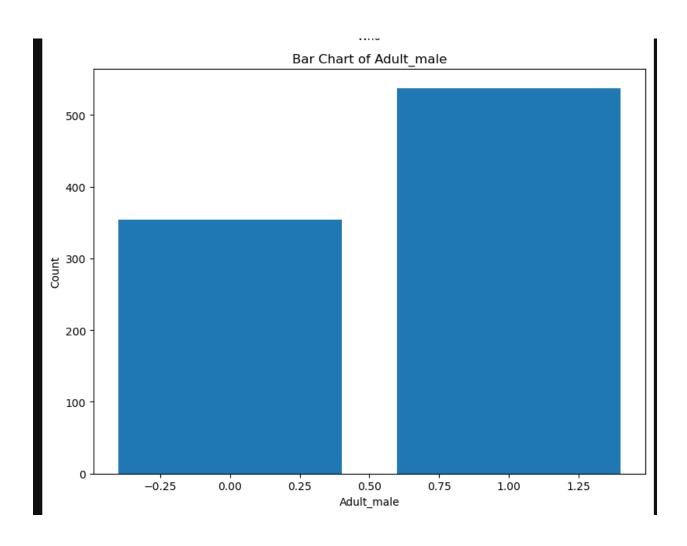


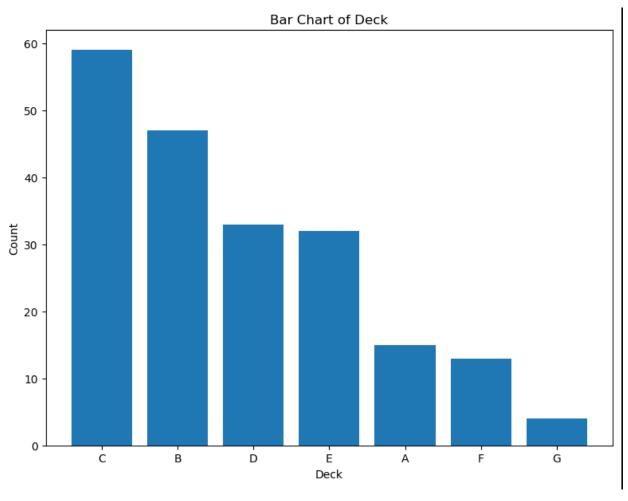


Embarked

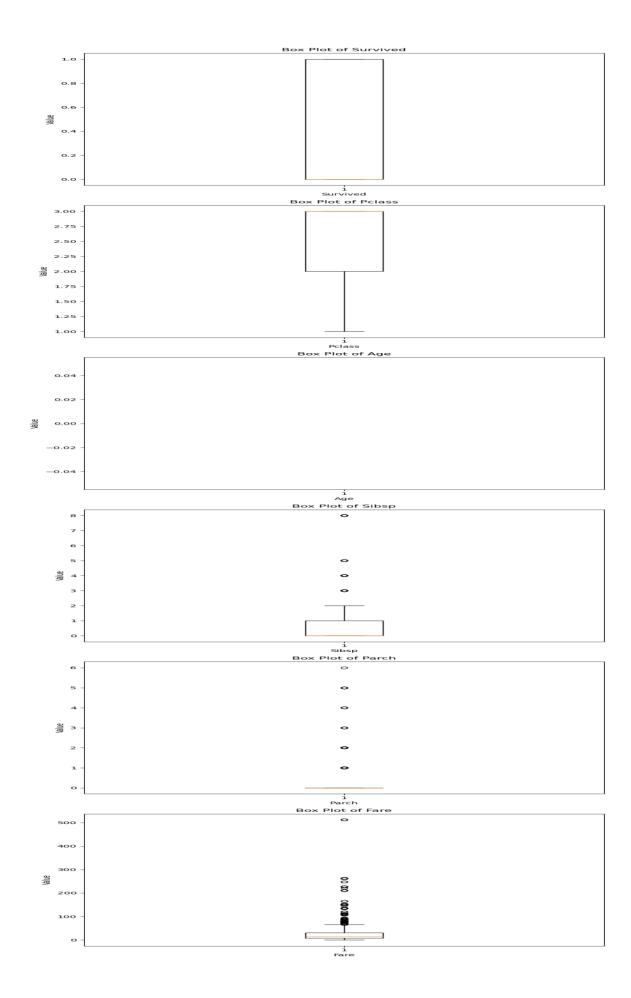




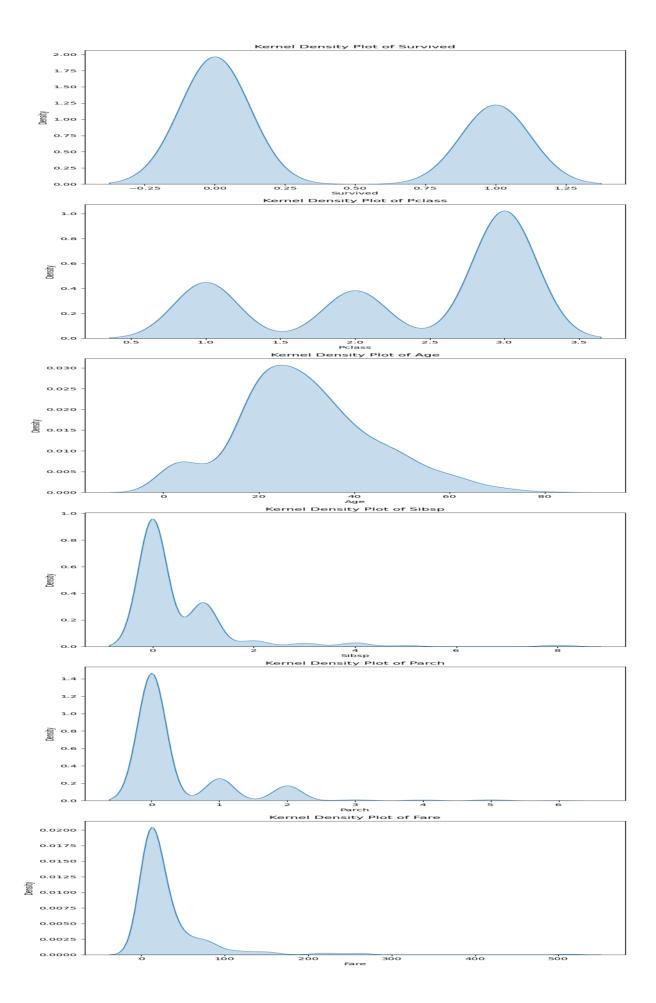




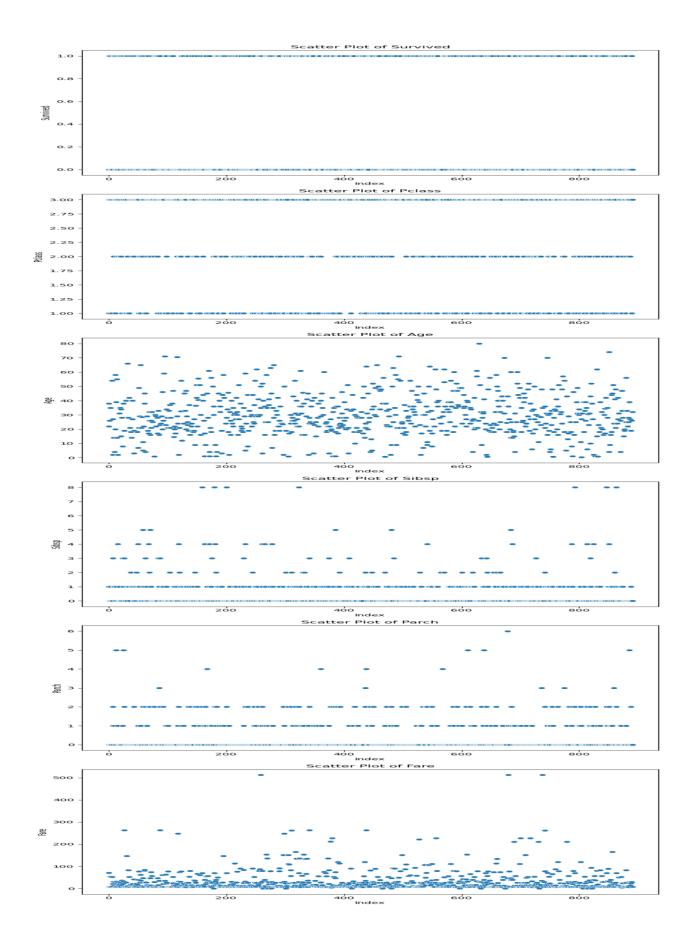
```
In [11]: import pandas as pd
          import matplotlib.pyplot as plt
          # Assuming 'data' is your DataFrame
          # List of columns to create box plots for
columns = ['survived', 'pclass', 'age', 'sibsp', 'parch', 'fare']
          # Set up the figure and subplots
fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))
          # Create box plots for each column
          for i, column in enumerate(columns):
              # Select the column
               column_data = data[column]
               # Create the box plot
               axes[i].boxplot(column_data)
               # Set the labels and title for each subplot
               axes[i].set_xlabel(column.capitalize())
               axes[i].set_ylabel('Value')
axes[i].set_title(f'Box Plot of {column.capitalize()}')
           # Adjust the spacing between subplots
          plt.tight_layout()
          # Display the box plots
          plt.show()
```



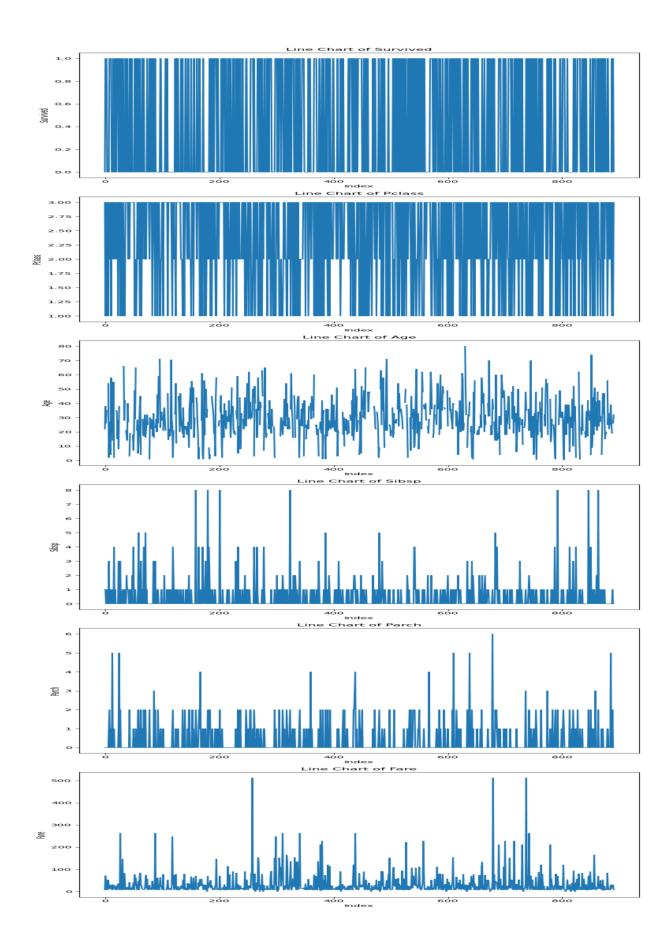
```
In [14]: import pandas as pd
          import seaborn as sns
          import matplotlib.pyplot as plt
         # Assuming 'data' is your DataFrame
         # List of columns to create KDE plots for
columns = ['survived', 'pclass', 'age', 'sibsp', 'parch', 'fare']
         # Set up the figure and subplots
         fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))
          # Create KDE plots for each column
         for i, column in enumerate(columns):
              # Select the column
             column_data = data[column]
             # Create the KDE plot
             sns.kdeplot(column_data, ax=axes[i], fill=True)
             # Set the labels and title for each subplot
             axes[i].set_xlabel(column.capitalize())
             axes[i].set_ylabel('Density')
              axes[i].set_title(f'Kernel Density Plot of {column.capitalize()}')
          # Adjust the spacing between subplots
         plt.tight_layout()
         # Display the KDE plots
         plt.show()
```



```
In [15]: import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         # Assuming 'data' is your DataFrame
         # List of columns to create scatter plots for
         columns = ['survived', 'pclass', 'age', 'sibsp', 'parch', 'fare']
         # Set up the figure and subplots
         fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))
         # Create scatter plots for each column
         for i, column in enumerate(columns):
            # Select the column
            column_data = data[column]
            # Generate x-coordinates for scatter plot
            x = range(len(column_data))
            # Create the scatter plot
            sns.scatterplot(x=x, y=column_data, ax=axes[i])
            # Set the labels and title for each subplot
            axes[i].set_xlabel('Index')
             axes[i].set_ylabel(column.capitalize())
             axes[i].set_title(f'Scatter Plot of {column.capitalize()}')
         # Adjust the spacing between subplots
         plt.tight_layout()
         # Display the scatter plots
         plt.show()
```

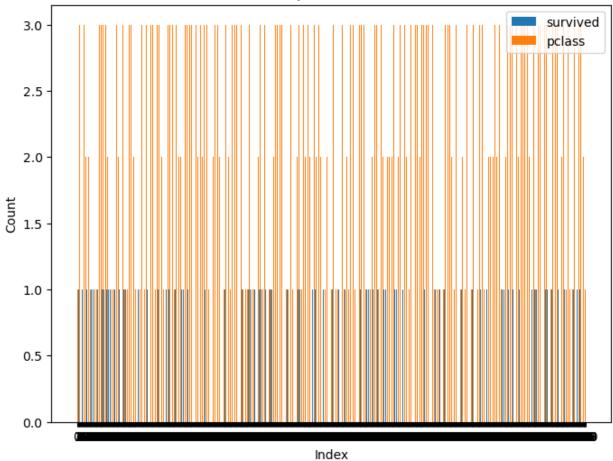


```
In [16]: import pandas as pd
         import matplotlib.pyplot as plt
         # Assuming 'data' is your DataFrame
         # List of columns to create line charts for
         columns = ['survived', 'pclass', 'age', 'sibsp', 'parch', 'fare']
         # Set up the figure and subplots
         fig, axes = plt.subplots(nrows=len(columns), ncols=1, figsize=(8, 6 * len(columns)))
         # Create line charts for each column
         for i, column in enumerate(columns):
             # Select the column
             column_data = data[column]
             # Generate x-coordinates for line chart
            x = range(len(column_data))
            # Create the line chart
            axes[i].plot(x, column_data)
             # Set the labels and title for each subplot
             axes[i].set_xlabel('Index')
             axes[i].set_ylabel(column.capitalize())
             axes[i].set_title(f'Line Chart of {column.capitalize()}')
         # Adjust the spacing between subplots
         plt.tight_layout()
         # Display the line charts
         plt.show()
```

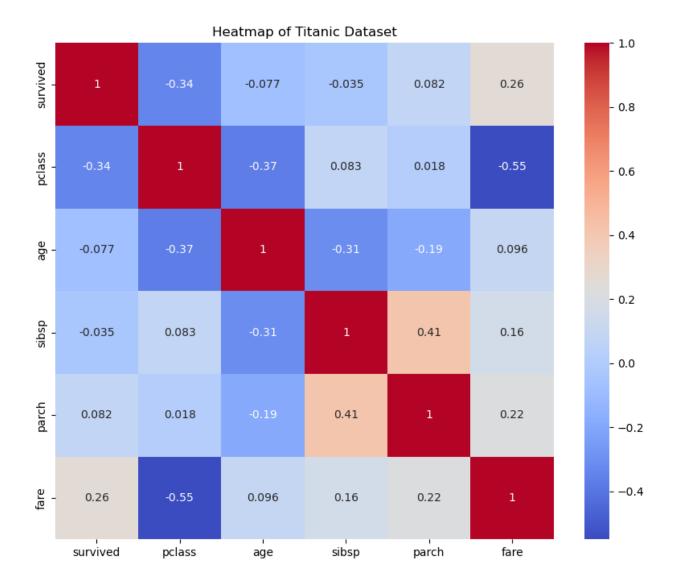


```
In [17]: import pandas as pd
         import matplotlib.pyplot as plt
         # Assuming 'data' is your DataFrame
         # List of columns for the bar chart
         columns = ['survived', 'pclass']
         # Set up the figure and subplots
         fig, ax = plt.subplots(figsize=(8, 6))
         # Set the positions and width for the bars
         positions = range(len(data))
         width = 0.35
         # Create the bar chart
         for i, column in enumerate(columns):
             # Select the column
             column_data = data[column]
            # Generate the x-coordinates for the bars
            x = [pos + width * i for pos in positions]
             # Create the bars
             ax.bar(x, column_data, width, label=column)
         # Set the labels and title
         ax.set_xlabel('Index')
         ax.set_ylabel('Count')
         ax.set_title('Grouped Bar Chart')
         # Set the x-axis ticks and labels
         ax.set_xticks([pos + width for pos in positions])
         ax.set_xticklabels(data.index)
         # Add a Legend
         ax.legend()
         # Display the bar chart
         plt.show()
```





```
In [19]: import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         # Assuming 'data' is your DataFrame
         # Select numeric columns for correlation calculation
         numeric_columns = data.select_dtypes(include='number')
         # Compute the correlation matrix
         correlation_matrix = numeric_columns.corr()
         # Set up the figure and axes
         fig, ax = plt.subplots(figsize=(10, 8))
         # Create the heatmap
         sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', ax=ax)
         # Set the title
         ax.set_title('Heatmap of Titanic Dataset')
         # Display the heatmap
         plt.show()
```



```
In [20]: import pandas as pd
import matplotlib.pyplot as plt

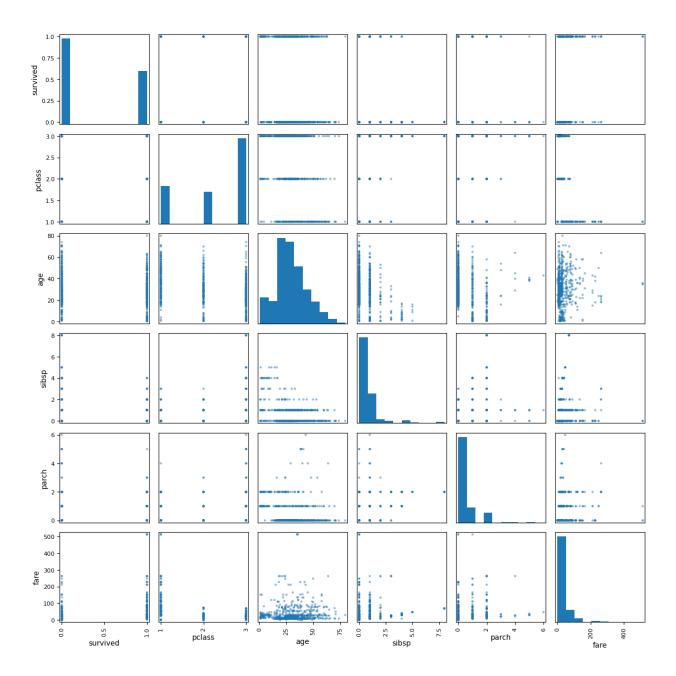
# Assuming 'data' is your DataFrame

# Select the columns for the scatter plot matrix
columns = ['survived', 'pclass', 'age', 'sibsp', 'parch', 'fare']

# Create the scatter plot matrix
scatter_matrix = pd.plotting.scatter_matrix(data[columns], figsize=(12, 12))

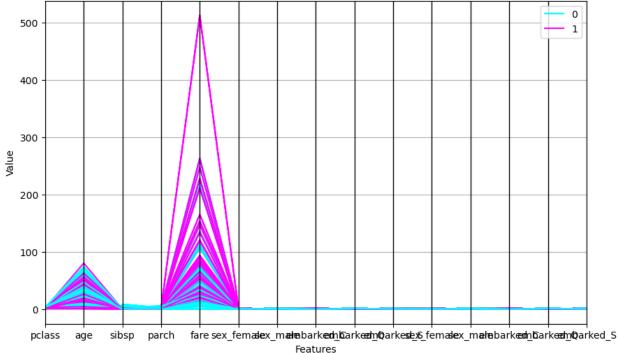
# Adjust the spacing between subplots
plt.tight_layout()

# Display the scatter plot matrix
plt.show()
```



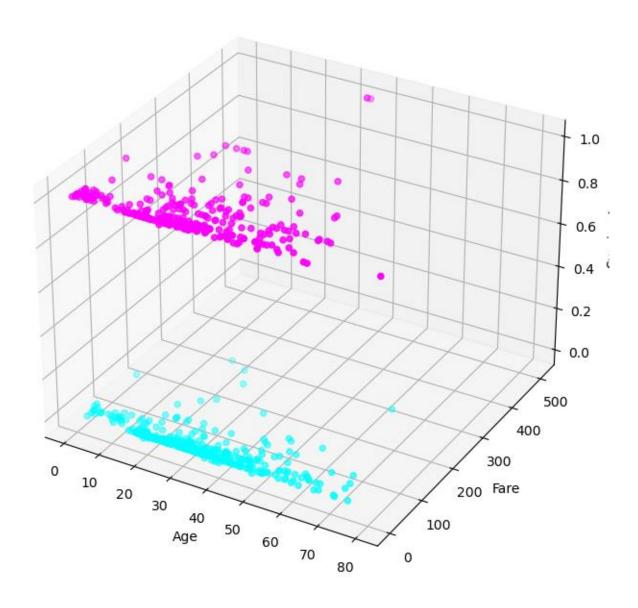
```
In [26]: import pandas as pd
          import matplotlib.pyplot as plt
         from sklearn.preprocessing import LabelEncoder
         # Assuming 'data' is your DataFrame
         # Select the columns for the Parallel Coordinates Plot
         columns = ['pclass', 'sex', 'age', 'sibsp', 'parch', 'fare', 'embarked']
         # Encode the 'survived' column
         label_encoder = LabelEncoder()
         data['survived_encoded'] = label_encoder.fit_transform(data['survived'])
         # Encode categorical columns using one-hot encoding
         categorical_columns = ['sex', 'embarked']
data_encoded = pd.get_dummies(data[columns + categorical_columns])
         # Merge the encoded columns with the target column
         data_final = pd.concat([data_encoded, data['survived_encoded']], axis=1)
         # Create the Parallel Coordinates Plot using pandas.plotting
         plt.figure(figsize=(10, 6))
         pd.plotting.parallel_coordinates(data_final, 'survived_encoded', colormap='cool')
         plt.title('Parallel Coordinates Plot of Titanic Dataset')
         plt.xlabel('Features')
plt.ylabel('Value')
         plt.legend()
         # Display the Parallel Coordinates Plot
         plt.show()
```





```
In [28]: import pandas as pd
         import matplotlib.pyplot as plt
         from mpl_toolkits.mplot3d import Axes3D
         # Assuming 'data' is your DataFrame
         # Select the columns for the 3D scatter plot
         columns = ['age', 'fare', 'survived']
         # Create a subset of the data with the selected columns
         subset = data[columns]
         # Remove rows with missing values
         subset = subset.dropna()
         # Create a 3D scatter plot
         fig = plt.figure(figsize=(10, 8))
         ax = fig.add_subplot(111, projection='3d')
         ax.scatter(subset['age'], subset['fare'], subset['survived'], c=subset['survived'], cmap='cool')
         # Set labels for each axis
         ax.set_xlabel('Age')A
         ax.set_ylabel('Fare')
ax.set_zlabel('Survived')
         # Set the title of the plot
         plt.title('3D Scatter Plot of Titanic Dataset')
         # Show the plot
         plt.show()
```

3D Scatter Plot of Titanic Dataset



```
In [29]: import pandas as pd
import plotly.express as px

# Assuming 'data' is your DataFrame

# Select the columns for the treemap
columns = ['survived', 'pclass', 'sex', 'age', 'sibsp', 'parch', 'fare', 'embark'

# Create a subset of the data with the selected columns
subset = data[columns]

# Remove rows with missing values
subset = subset.dropna()

# Create the treemap
fig = px.treemap(subset, path=columns)

# Set the title of the treemap
fig.update_layout(title='Treemap of Titanic Dataset')

# Show the treemap
fig.show()
```

4. Perform descriptive statistics on the dataset

```
In [30]: import pandas as pd
        # Assuming 'data' is your DataFrame
        # Perform descriptive statistics on the dataset
        statistics = data.describe(include='all')
        # Print the descriptive statistics
        print(statistics)
                                                                     parch \
                survived
                           pclass sex
                                              age
                                                          sibsp
        count 891.000000 891.000000
                                      891 714.000000 891.000000 891.000000
                     NaN
        unique
                           NaN
                                      2
                                                           NaN
                                             NaN
                                                                       NaN
        top
                      NaN
                                NaN male
                                                 NaN
                                                            NaN
                                                                       NaN
        freq
                     NaN
                                NaN 577
                                                 NaN
                                                            NaN
                                                                       NaN
                                      NaN 29.699118
NaN 14.526497
                 0.383838
                            2.308642
                                                       0.523008
                                                                  0.381594
        mean
        std
                 0.486592
                            0.836071
                                                      1.102743
                                                                  0.806057
                 0.000000
                            1.000000
                                           0.420000 0.000000
                                                                 0.000000
        min
                                      NaN
                 0.000000
                            2.000000
                                           20.125000
                                                       0.000000
                                                                  0.000000
        25%
                                      NaN
        50%
                 0.000000
                            3.000000
                                      NaN
                                           28.000000
                                                       0.000000
                                                                  0.000000
        75%
                 1.000000
                            3.000000
                                      NaN 38.000000
                                                      1.000000
                                                                  0.000000
        max
                 1.000000
                           3.000000 NaN 80.000000
                                                       8.000000
                                                                 6.000000
                     fare embarked class who adult_male deck embark_town alive \
        count
               891.000000
                              889
                                    891 891
                                                 891 203
                                                                   889 891
        unique
                     NaN
                              3
                                     3
                                         3
                                                    2
                                                        7
                                                                    3
                                                                          2
        top
                      NaN
                                S
                                  Third man
                                                  True
                                                        C
                                                            Southampton
                                                                         no
                      NaN
                              644
                                   491 537
                                                  537
                                                       59
                                                                   644
                                                                        549
        freq
                32.204208
                                                                        NaN
        mean
                             NaN
                                    NaN NaN
                                                   NaN NaN
                                                                   NaN
        std
                49.693429
                              NaN
                                    NaN NaN
                                                   NaN NaN
                                                                   NaN
                                                                        NaN
        min
                0.000000
                              NaN
                                    NaN NaN
                                                   NaN NaN
                                                                   NaN
                                                                        NaN
                 7.910400
                                    NaN NaN
                                                   NaN NaN
        25%
                              NaN
                                                                   NaN
                                                                        NaN
        50%
                14.454200
                              NaN
                                    NaN NaN
                                                   NaN NaN
                                                                   NaN
                                                                        NaN
        75%
                31.000000
                                                  NaN NaN
                                                                   NaN
                                                                        NaN
                              NaN
                                   NaN NaN
               512.329200
                                   NaN NaN
                                                  NaN NaN
                                                                   NaN
                                                                        NaN
        max
                            NaN
              alone sex encoded survived encoded
        count
                891 891.000000
                                      891.000000
                            NaN
        unique
```

5. Handle the Missing values

```
In [31]: import pandas as pd

# Load the Titanic dataset
data = pd.read_csv('titanic.csv')

# Check for missing values
print(data.isnull().sum())

# Drop rows with missing values
data = data.dropna()

# Fill missing values with a specific value
data['age'] = data['age'].fillna(data['age'].mean())
data['embarked'] = data['embarked'].fillna(data['embarked'].mode()[0])

# Perform linear interpolation to fill missing fare values
data['fare'] = data['fare'].interpolate(method='linear')

# Drop columns with a high percentage of missing values
data = data.drop('deck', axis=1)

# Check for missing values again to confirm
print(data.isnull().sum())
```

survived	0
pclass	0
sex	0
age	177
sibsp	0
parch	0
fare	0
embarked	2
class	0
who	0
adult_male	0
deck	688
embark_town	2
alive	0
alone	0
dtype: int64	
survived	0
pclass	0
sex	0
age	0
sibsp	0
parch	0
fare	0
embarked	0
class	0
who	0
adult_male	0
embark_town	0
alive _	0
alone	0
dtype: int64	

6. Find the outliers and replace the outliers

```
In [33]: import pandas as pd
         import numpy as np
         from scipy import stats
         # Load the Titanic dataset
         data = pd.read_csv('titanic.csv')
         # Identify outliers using z-score
         z_scores = np.abs(stats.zscore(data['fare']))
threshold = 3
         outliers = np.where(z_scores > threshold)
         # Replace outliers with the median value
         median_fare = data['fare'].median()
         data.loc[outliers[0], 'fare'] = median_fare
         # Check for outliers again to confirm
         z_scores_after = np.abs(stats.zscore(data['fare']))
         new_outliers = np.where(z_scores_after > threshold)
         print("Number of outliers after replacement:", len(new_outliers[0]))
         Number of outliers after replacement: 22
```

```
In [34]: import pandas as pd
import numpy as np
from scipy import stats

# Load the Titanic dataset
data = pd.read_csv('titanic.csv')

# Calculate z-scores for the 'fare' column
z_scores = np.abs(stats.zscore(data['fare']))

# Set the threshold for identifying outliers
threshold = 3

# Find the outliers based on the z-scores
outliers = data[z_scores > threshold]

# Print the outliers
print("Outliers in the 'fare' column:")
print(outliers)
```

```
Outliers in the 'fare' column:
                             age sibsp parch
    survived pclass
                     sex
                                                   fare embarked class \
                                           2 263.0000
          0
                       male 19.0
                                                             S First
27
                 1
                                    3
                                      3
                                             2 263.0000
                                                                S First
88
           1
                   1 female 23.0
118
           0
                   1
                     male 24.0
                                      0
                                            1 247.5208
                                                               C First
          1
                   1 female 35.0
258
                                     0
                                            0 512.3292
                                                               C First
                 1 female 35.0 0 1 247.5292
1 female 50.0 0 1 247.5208
1 female 18.0 2 2 262.3750
1 female 24.0 3 2 263.0000
1 male 27.0 0 2 211.5000
1 female 42.0 0 0 227.5250
1 male 64.0 1 4 263.0000
1 male NaN 0 0 221.7792
299
          1
                                                               C First
                                                               C First
311
           1
                                                               S First
           1
341
377
           0
                                                               C First
           1
380
                                                               C First
438
          0
                                                               S First
527
          0
                                                               S First
557
          0
                   1 male NaN
                                     0
                                            0 227.5250
                                                               C First
                       male 36.0
                                     0
                                            1 512.3292
                                                               C First
679
          1
                   1
                 1 male 36.0 0 1 512.3292
1 female 15.0 0 1 211.3375
1 female 18.0 1 0 227.5250
1 female 38.0 0 0 227.5250
1 female 29.0 0 0 211.3375
1 male 35.0 0 0 512.3292
1 female 21.0 2 2 262.3750
                                                               S First
689
           1
700
           1
                                                                C First
716
          1
                                                               C First
                                                               S First
730
          1
737
           1
                                                               C First
742
           1
                                                               C First
779
           1
                   1 female 43.0
                                      0
                                            1 211.3375
                                                               S First
                            I TOMBLE 45.0
                                                                            5 (14)
                                                        1 211,0070
               who adult_male deck embark_town alive alone
        27
               man
                         True C Southampton
                                                  no False
        88
                         False
                                  C Southampton
                                                  yes False
             woman
                                                   no False
        118
               man
                          True
                                В
                                     Cherbourg
        258
             woman
                         False NaN
                                       Cherbourg
                                                  yes
                                                        True
                                                  yes False
        299
                         False B
                                       Cherbourg
             woman
        311 woman
                        False B
                                       Cherbourg yes False
        341 woman
                        False C Southampton
                                                  yes False
        377
                                  C
                                       Cherbourg
              man
                          True
                                                    no False
        380 woman
                         False NaN
                                       Cherbourg yes
                                                        True
        438
                                C Southampton
             man
                         True
                                                  no False
        527
             man
                         True
                                C Southampton
                                                  no
                                                        True
        557
                         True NaN
                                       Cherbourg
                                                  no
                                                        True
               man
                                                  yes False
        679
                          True
                                В
               man
                                        Cherbourg
        689 child
                         False B Southampton
                                                  yes False
        700 woman
                        False C
                                     Cherbourg
                                                  yes False
                                       Cherbourg
                         False C
        716 woman
                                                  yes
                                                        True
        730 woman
                         False B Southampton
                                                   yes
                                                        True
        737
                          True B
                                       Cherbourg
                                                        True
               man
                                                  yes
        742 woman
                         False B
                                        Cherbourg yes False
        779
             woman
                         False
                                  B Southampton yes False
```

^{7.} Check for Categorical columns and perform encoding

```
In [35]: import pandas as pd

# Load the Titanic dataset
data = pd.read_csv('titanic.csv')

# Check for categorical columns
categorical_columns = data.select_dtypes(include=['object']).columns

# Perform encoding for categorical columns
data_encoded = pd.get_dummies(data, columns=categorical_columns)

# Print the encoded dataset
print("Encoded dataset:")
print(data_encoded.head())
```

```
Encoded dataset:
  survived pclass age sibsp parch fare
0 3 22.0 1 0 7.2500
1 1 38.0 1 0 71.2833
                                       fare adult_male alone \
                                                   True False
                                                   False False
1
               3 26.0 0 0 7.9250
1 35.0 1 0 53.1000
                                                 False True
3
        1
                                                 False False
4
         0
               3 35.0 0
                                 0 8.0500
                                                   True True
   sex_female sex_male ... deck_C deck_D deck_E deck_F deck_G \
                 1 ...
                            0 0
0
                                               0
                                                       0
                                                              0
        0
                    0 ...
                                        0
1
           1
                                1
                                      0
2
           1
                    0 ...
                                0
                                              0
                                                       0
                                                              0
3
                    0 ...
                                1
                                       0
                                               0
                                                       0
                                                              0
           1
4
                    1 ...
                                        0
                                               0
                                                       0
   {\tt embark\_town\_Cherbourg} \quad {\tt embark\_town\_Queenstown} \quad {\tt embark\_town\_Southampton}
                     0
1
                     1
                                            0
                                                                   0
                                            0
2
                                                                   1
                     Ø
3
                     0
                                            0
                                                                   1
4
                     0
                                            0
                                                                   1
   alive_no alive_yes
0
       1
1
         0
                   1
2
         0
                   1
3
         0
                   1
4
         1
                   a
[5 rows x 31 columns]
```

8. Split the data into dependent and independent variables.

```
In [36]: import pandas as pd

# Load the Titanic dataset
data = pd.read_csv('titanic.csv')

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

# Print the shapes of the variables
print("Independent variables shape:", X.shape)
print("Dependent variable shape:", y.shape)
Independent variables shape: (891, 14)
Dependent variable shape: (891,)
```

```
In [37]: print(X)
                        ss sex age sibsp parch fare embarked class who
3 male 22.0 1 0 7.2500 S Third man
1 female 38.0 1 0 71.2833 C First woman
                   pclass
                                                                                                     who \
            1
                       3 female 26.0 0 0 7.9250 S Third woman
1 female 35.0 1 0 53.1000 S First woman
3 male 35.0 0 0 8.0500 S Third man
            2
            3
                      2 male 27.0 0 0 13.0000 S Second man
1 female 19.0 0 0 30.0000 S First woman
3 female NaN 1 2 23.4500 S Third woman
1 male 26.0 0 0 30.0000 C First man
3 male 32.0 0 0 7.7500 Q Third man
            886
            888
            889
            890
                  adult_male deck embark_town alive alone
            0
                        True NaN Southampton no False
                        False C Cherbourg yes False
False NaN Southampton yes True
            1
            2
            3
                        False C Southampton yes False
                          True NaN Southampton no
            4
                         True NaN Southampton no True
            886
                         False B Southampton yes
False NaN Southampton no
            887
                                                            no False
            888
            229
                          True C
                                          Cherbourg yes True
            890
                          True NaN Queenstown no True
            [891 rows x 14 columns]
```

```
In [39]: print(y)
         1
                1
         2
                1
         3
         4
                0
         886
               0
         887
                1
         888
         889
                1
         890
         Name: survived, Length: 891, dtype: int64
```

9. Scale the independent variables

```
In [43]: from sklearn.preprocessing import StandardScaler, OneHotEncoder
         from sklearn.compose import ColumnTransformer
         # Load the Titanic dataset
         data = pd.read_csv('titanic.csv')
         # Split into dependent and independent variables
         X = data.drop('survived', axis=1) # Independent variables (features)
         y = data['survived'] # Dependent variable (target)
         # Identify the categorical columns
         categorical_cols = X.select_dtypes(include=['object']).columns
         # Perform one-hot encoding on categorical columns
         encoder = OneHotEncoder(drop='first')
         X_encoded = encoder.fit_transform(X[categorical_cols]).toarray()
         encoded_cols = encoder.get_feature_names_out(categorical_cols)
         X_encoded = pd.DataFrame(X_encoded, columns=encoded_cols)
         # Concatenate encoded columns with remaining columns
         X_encoded = pd.concat([X_encoded, X.drop(categorical_cols, axis=1)], axis=1)
         # Scale the independent variables
         scaler = StandardScaler()
         X_scaled = scaler.fit_transform(X_encoded)
         # Print the scaled independent variables
         print(X_scaled)
         [[ 0.73769513 -0.30756234  0.61930636  ... -0.50244517  0.81192233
           -1.2316449 ]
          [-1.35557354 -0.30756234 -1.61470971 ... 0.78684529 -1.2316449
           -1.2316449 ]
          [-1.35557354 -0.30756234  0.61930636  ... -0.48885426 -1.2316449
           0.81192233]
          [-1.35557354 -0.30756234 0.61930636 ... -0.17626324 -1.2316449
           -1.2316449
          [ 0.73769513 -0.30756234 -1.61470971 ... -0.04438104  0.81192233
            0.81192233]
          [ 0.73769513 3.25137334 -1.61470971 ... -0.49237783 0.81192233
            0.81192233]]
```

10. Split the data into training and testing

```
In [44]: from sklearn.model_selection import train_test_split

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

# Print the shapes of the training and testing sets
print("Training set shape:", X_train.shape, y_train.shape)
print("Testing set shape:", X_test.shape, y_test.shape)
Training set shape: (712, 26) (712,)
Testing set shape: (179, 26) (179,)
```

```
In [45]: print("Training set:")
       print(X_train)
       print(y_train)
       print("Testing set:")
       print(X_test)
       print(y_test)
       Training set:
       0.81192233]
        [ 0.73769513 -0.30756234  0.61930636  ... -0.38667072  0.81192233
         0.81192233]
        [ 0.73769513 -0.30756234  0.61930636  ... -0.48885426  0.81192233
         0.81192233]
        [ 0.73769513 -0.30756234  0.61930636  ... -0.36435545  0.81192233
        -1.2316449 ]
       [ 0.73769513 -0.30756234  0.61930636  ...  0.90773798  0.81192233
        -1.2316449 ]]
       331
            0
       733
            0
       382
            0
       704
            0
       813
            0
       106
           1
       270
            0
       860
            0
       435
            1
            0
       102
       Name: survived, Length: 712, dtype: int64
       Testing set:
       [[ 0.73769513 -0.30756234 -1.61470971 ... -0.34145224  0.81192233
        -1.2316449
       [ 0.73769513 -0.30756234  0.61930636  ... -0.43700744  0.81192233
         0.81192233]
```