EDA ON Titanic Survival Predictions

Load the Titanic dataset

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
from flask import Flask, render_template, request
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report

titanic_df = pd.read_csv('/content/Titanic-Dataset.csv')
```

Display the first few rows of the dataset

```
print(titanic_df.head())
       PassengerId Survived Pclass \
               1
                                1
                                                         Sex Age SibSp \
    0
                               Braund, Mr. Owen Harris
                                                        male 22.0
       Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
    2
                                Heikkinen, Miss. Laina female 26.0
    3
            Futrelle, Mrs. Jacques Heath (Lily May Peel) female 35.0
                              Allen, Mr. William Henry
                                                       male 35.0
       Parch
                      Ticket
                                 Fare Cabin Embarked
                    A/5 21171 7.2500
                     PC 17599 71.2833
                                       C85
                                                 C
          0 STON/02. 3101282 7.9250 NaN
                                                 S
                      113803 53.1000 C123
                      373450 8.0500 NaN
```

info about the dataset

```
print(titanic df.info())
<<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 891 entries, 0 to 890
     Data columns (total 12 columns):
         Column
                      Non-Null Count Dtype
         PassengerId 891 non-null
                                      int64
     1
         Survived
                      891 non-null
                                      int64
     2
         Pclass
                      891 non-null
                                      int64
     3
         Name
                      891 non-null
                                      object
     4
                      891 non-null
         Sex
                                      object
     5
                      714 non-null
                                      float64
          Age
         SibSp
                      891 non-null
                                      int64
     7
                      891 non-null
         Parch
                                      int64
     8
         Ticket
                      891 non-null
                                      object
     9
         Fare
                       891 non-null
                                      float64
     10
         Cabin
                       204 non-null
                                      object
     11 Embarked
                      889 non-null
                                      object
     dtypes: float64(2), int64(5), object(5)
     memory usage: 83.7+ KB
     None
```

Describe the numerical features

```
print(titanic_df.describe())
₹
            PassengerId
                           Survived
                                         Pclass
                                                                  SibSp
            891.000000 891.000000
                                    891.000000
                                                714.000000 891.000000
     count
                                                  29.699118
     mean
            446.000000
                          0.383838
                                       2.308642
                                                               0.523008
             257.353842
                          0.486592
                                       0.836071
                                                  14.526497
                                                               1.102743
     std
     min
              1.000000
                          0.000000
                                       1.000000
                                                   0.420000
                                                               0.000000
     25%
             223.500000
                          0.000000
                                       2.000000
                                                  20.125000
                                                               0.000000
     50%
             446.000000
                          0.000000
                                       3.000000
                                                  28.000000
                                                               0.000000
     75%
             668.500000
                          1.000000
                                       3.000000
                                                  38.000000
                                                               1.000000
             891.000000
                          1.000000
                                       3.000000
                                                  80.000000
                                                               8.000000
     max
                 Parch
                              Fare
     count
           891.000000
                        891.000000
              0.381594
                        32.204208
     mean
              0.806057
                         49.693429
     std
     min
              0.000000
                         0.000000
     25%
              0.000000
                         7.910400
              0.000000
                        14.454200
     75%
                        31.000000
              0.000000
              6.000000 512.329200
     max
```

Check the unique values for categorical features

```
'Heikkinen, Miss. Laina' 'Futrelle, Mrs. Jacques Heath (Lily May Peel)'
'Allen, Mr. William Henry' 'Moran, Mr. James' 'McCarthy, Mr. Timothy J'
'Palsson, Master. Gosta Leonard'
'Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)'
'Nasser, Mrs. Nicholas (Adele Achem)' 'Sandstrom, Miss. Marguerite Rut'
'Bonnell, Miss. Elizabeth' 'Saundercock, Mr. William Henry'
'Andersson, Mr. Anders Johan' 'Vestrom, Miss. Hulda Amanda Adolfina'
'Hewlett, Mrs. (Mary D Kingcome) ' 'Rice, Master. Eugene'
'Williams, Mr. Charles Eugene'
'Vander Planke, Mrs. Julius (Emelia Maria Vandemoortele)'
'Masselmani, Mrs. Fatima' 'Fynney, Mr. Joseph J' 'Beesley, Mr. Lawrence'
'McGowan, Miss. Anna "Annie" 'Sloper, Mr. William Thompson'
'Palsson, Miss. Torborg Danira'
'Asplund, Mrs. Carl Oscar (Selma Augusta Emilia Johansson)'
'Emir, Mr. Farred Chehab' 'Fortune, Mr. Charles Alexander'
'O\'Dwyer, Miss. Ellen "Nellie"' 'Todoroff, Mr. Lalio'
'Uruchurtu, Don. Manuel E'
'Spencer, Mrs. William Augustus (Marie Eugenie)'
'Glynn, Miss. Mary Agatha' 'Wheadon, Mr. Edward H'
'Meyer, Mr. Edgar Joseph' 'Holverson, Mr. Alexander Oskar'
'Mamee, Mr. Hanna' 'Cann, Mr. Ernest Charles'
'Vander Planke, Miss. Augusta Maria' 'Nicola-Yarred, Miss. Jamila'
'Ahlin, Mrs. Johan (Johanna Persdotter Larsson)'
'Turpin, Mrs. William John Robert (Dorothy Ann Wonnacott)'
'Kraeff, Mr. Theodor' 'Laroche, Miss. Simonne Marie Anne Andree'
'Devaney, Miss. Margaret Delia' 'Rogers, Mr. William John'
'Lennon, Mr. Denis' "O'Driscoll, Miss. Bridget" 'Samaan, Mr. Youssef'
'Arnold-Franchi, Mrs. Josef (Josefine Franchi)'
'Panula, Master. Juha Niilo' 'Nosworthy, Mr. Richard Cater'
'Harper, Mrs. Henry Sleeper (Myna Haxtun)'
'Faunthorpe, Mrs. Lizzie (Elizabeth Anne Wilkinson)'
'Ostby, Mr. Engelhart Cornelius' 'Woolner, Mr. Hugh' 'Rugg, Miss. Emily'
'Novel, Mr. Mansouer' 'West, Miss. Constance Mirium'
'Goodwin, Master. William Frederick' 'Sirayanian, Mr. Orsen'
'Icard, Miss. Amelie' 'Harris, Mr. Henry Birkhardt'
'Skoog, Master, Harald' 'Stewart, Mr. Albert A'
'Moubarek, Master. Gerios' 'Nye, Mrs. (Elizabeth Ramell)'
'Crease, Mr. Ernest James' 'Andersson, Miss. Erna Alexandra'
'Kink, Mr. Vincenz' 'Jenkin, Mr. Stephen Curnow'
'Goodwin, Miss. Lillian Amy' 'Hood, Mr. Ambrose Jr'
'Chronopoulos, Mr. Apostolos' 'Bing, Mr. Lee' 'Moen, Mr. Sigurd Hansen'
'Staneff, Mr. Ivan' 'Moutal, Mr. Rahamin Haim'
'Caldwell, Master. Alden Gates' 'Dowdell, Miss. Elizabeth'
'Waelens, Mr. Achille' 'Sheerlinck, Mr. Jan Baptist'
'McDermott, Miss. Brigdet Delia' 'Carrau, Mr. Francisco M'
'Ilett, Miss. Bertha'
'Backstrom, Mrs. Karl Alfred (Maria Mathilda Gustafsson)'
'Ford, Mr. William Neal' 'Slocovski, Mr. Selman Francis'
'Fortune, Miss. Mabel Helen' 'Celotti, Mr. Francesco'
'Christmann, Mr. Emil' 'Andreasson, Mr. Paul Edvin'
'Chaffee, Mr. Herbert Fuller' 'Dean, Mr. Bertram Frank'
'Coxon, Mr. Daniel' 'Shorney, Mr. Charles Joseph'
'Goldschmidt, Mr. George B' 'Greenfield, Mr. William Bertram'
'Doling, Mrs. John T (Ada Julia Bone)' 'Kantor, Mr. Sinai'
'Petranec, Miss. Matilda' 'Petroff, Mr. Pastcho ("Pentcho")'
```

Understand the target variable (Survived) distribution

```
print(titanic_df['Survived'].value_counts())
```

```
Survived
0 549
1 342
Name: count, dtype: int64
```

Handle missing values

Feature Engineering

```
titanic df['FamilySize'] = titanic df['SibSp'] + titanic df['Parch'] + 1
```

Encode categorical variables

```
Heikkinen, Miss. Laina
                                                     0 26.0
3
       Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                     0 35.0
                                                                 1
                                                                        0
4
                          Allen, Mr. William Henry
                                                     1 35.0
            Ticket
                      Fare Embarked FamilySize
         A/5 21171 7.2500
1
          PC 17599 71.2833
                                              2
  STON/02. 3101282 7.9250
            113803 53.1000
                                              2
3
            373450 8.0500
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):
    Column
                 Non-Null Count Dtype
                 -----
    PassengerId 891 non-null
    Survived
                891 non-null
    Pclass
                 891 non-null
                                int64
 3
                 891 non-null
                                object
                 891 non-null
 5
                 891 non-null
                                float64
    Age
    SibSp
                 891 non-null
                                int64
 7
    Parch
                 891 non-null
                                int64
 8
    Ticket
                 891 non-null
                                object
 9
                 891 non-null
    Fare
                                float64
 10 Embarked
                 891 non-null
                                int64
 11 FamilySize 891 non-null
dtypes: float64(2), int64(8), object(2)
memory usage: 83.7+ KB
None
```

Correlation analysis

```
numerical features = titanic df.select dtypes(include=[np.number])
correlation_matrix = numerical_features.corr()
print(correlation_matrix['Survived'].sort_values(ascending=False))
→ Survived
                   1.000000
     Fare
                    0.257307
     Parch
                    0.081629
     PassengerId
                   -0.005007
     SibSp
                   -0.035322
     Age
                   -0.077221
                   -0.338481
     Pclass
     Name: Survived, dtype: float64
```

Analyze survival rate by different features

Survival rate by Pclass

```
print(titanic_df.groupby('Pclass')['Survived'].mean())
```

```
Pclass
    1     0.629630
    2     0.472826
    3     0.242363
    Name: Survived, dtype: float64
```

Survival rate by Sex

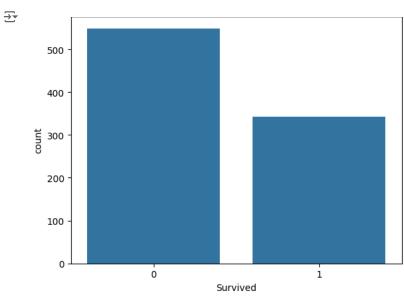
```
print(titanic_df.groupby('Sex')['Survived'].mean())

Sex
    female    0.742038
    male    0.188908
    Name: Survived, dtype: float64
```

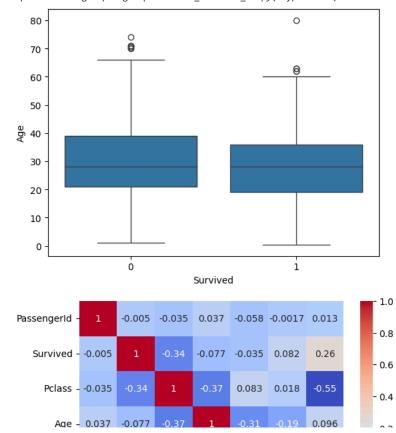
Survival rate by Embarked

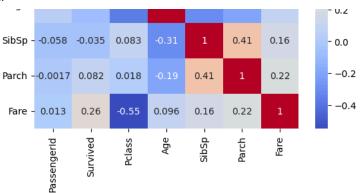
Visualizations

```
sns.countplot(x='Survived', data=titanic_df)
plt.show()
sns.boxplot(x='Survived', y='Age', data=titanic_df)
plt.show()
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
plt.show()
```



/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)





Analyze survival rate based on title extracted from the Name column

```
titanic_df['Title'] = titanic_df['Name'].str.extract(' ([A-Za-z]+)\.', expand=False)
print(titanic_df.groupby('Title')['Survived'].mean())
Title
     Capt
                 0.000000
     Col
                 0.500000
     Countess
                 1.000000
                 0.000000
                 0.428571
     Jonkheer
                 0.000000
     Lady
                 1.000000
    Major
                 0.500000
    Master
                 0.575000
    Miss
                 0.697802
    Mlle
                 1.000000
    Mme
                 1.000000
    Mr
                 0.156673
    Mrs
                 0.792000
    Ms
                 1.000000
                 0.000000
     Rev
     Sir
                 1.000000
```

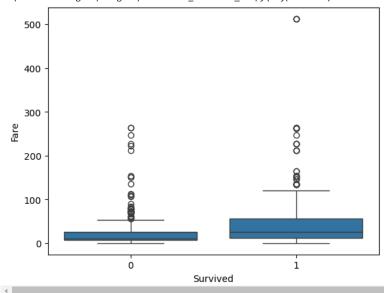
Investigate the relationship between Fare and survival rate

```
print(titanic_df.groupby('Survived')['Fare'].mean())
sns.boxplot(x='Survived', y='Fare', data=titanic_df)
plt.show()
```

Name: Survived, dtype: float64

```
→ Survived
        22.117887
        48.395408
   Name: Fare, dtype: float64
```

/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)



Create new features based on existing ones

```
titanic_df['AgeGroup'] = pd.cut(titanic_df['Age'], bins=[0, 18, 30, 50, float('inf')], labels=['Child', 'Young Adult', 'Adult', 'Senior'])
titanic_df['FareRange'] = pd.qcut(titanic_df['Fare'], q=4, labels=['Low', 'Medium', 'High', 'Very High'])
```

Interaction term between Pclass and Fare

```
titanic_df['PclassFare'] = titanic_df['Pclass'] * titanic_df['Fare']
```

Explore different machine learning models and compare their performance

```
X = titanic df[['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', 'Embarked', 'FamilySize', 'Title', 'AgeGroup', 'FareRange', 'PclassFare']]
y = titanic_df['Survived']
X = pd.get dummies(X, columns=['Title', 'AgeGroup', 'FareRange'], dummy na=True)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Define features (X) and target (y)

```
X = titanic_df.drop('Survived', axis=1)
y = titanic_df['Survived']
```

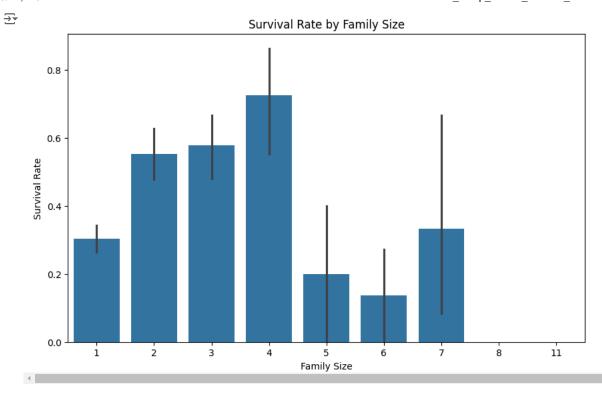
Split data into training and testing sets

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Survival Rate by Age Group

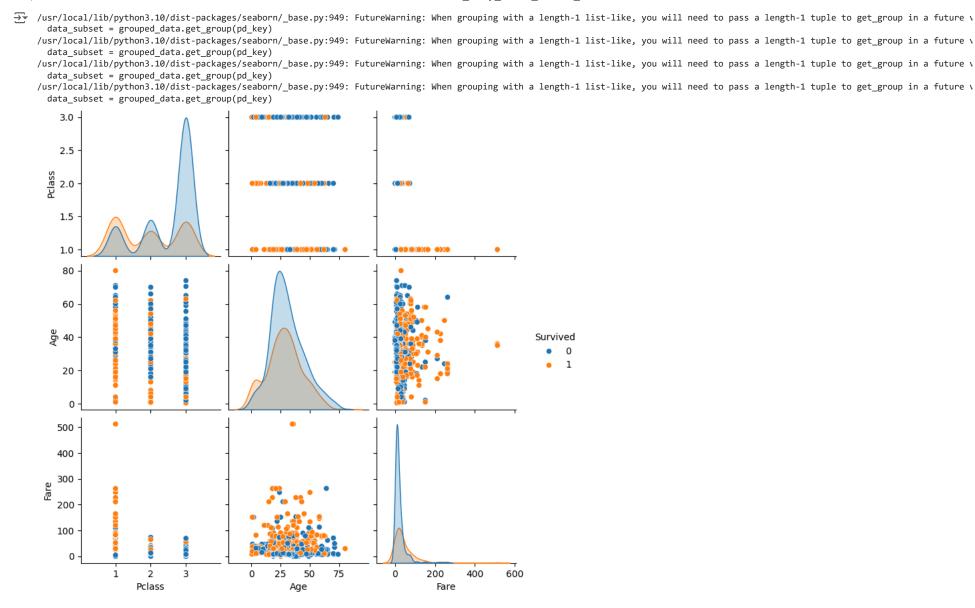
Survival Rate by Family Size

```
titanic_df['FamilySize'] = titanic_df['SibSp'] + titanic_df['Parch'] + 1
plt.figure(figsize=(10, 6))
sns.barplot(x='FamilySize', y='Survived', data=titanic_df)
plt.title('Survival Rate by Family Size')
plt.xlabel('Family Size')
plt.ylabel('Survival Rate')
plt.show()
```



Pairplot for selected features

sns.pairplot(titanic_df[['Survived', 'Pclass', 'Sex', 'Age', 'Fare']], hue='Survived')
plt.show()

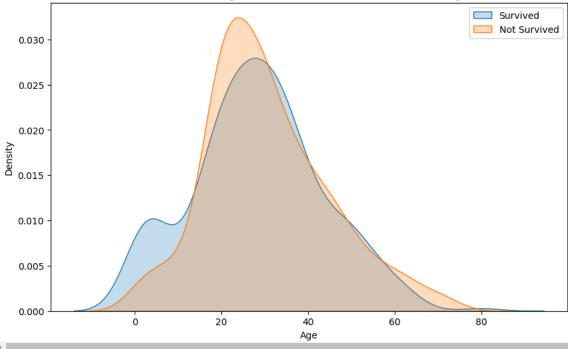


Distribution of Age for Survived and Not Survived passengers

```
plt.figure(figsize=(10, 6))
sns.kdeplot(titanic_df['titanic_df['Survived'] == 1]['Age'], label='Survived', shade=True)
sns.kdeplot(titanic_df[titanic_df['Survived'] == 0]['Age'], label='Not Survived', shade=True)
plt.title('Distribution of Age for Survived and Not Survived Passengers')
```

```
plt.xlabel('Age')
plt.ylabel('Density')
plt.legend()
plt.show()
```

Distribution of Age for Survived and Not Survived Passengers

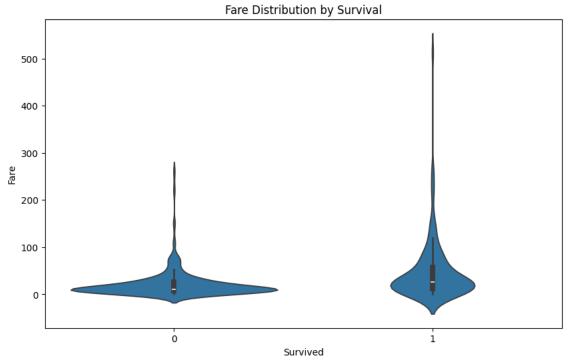


Violin plot of Fare by Survived

```
plt.figure(figsize=(10, 6))
sns.violinplot(x='Survived', y='Fare', data=titanic_df
plt.title('Fare Distribution by Survival')
plt.xlabel('Survived')
plt.ylabel('Fare')
plt.show()
```

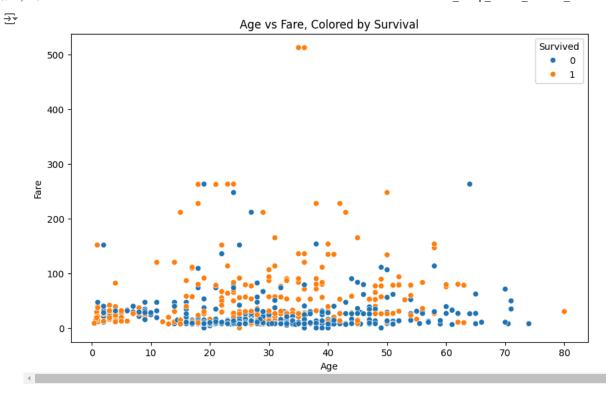
/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key)

/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key)



Scatter plot of Age vs Fare, colored by Survived

```
plt.figure(figsize=(10, 6))
sns.scatterplot(x='Age', y='Fare', hue='Survived', data=titanic_df)
plt.title('Age vs Fare, Colored by Survival')
plt.xlabel('Age')
plt.ylabel('Fare')
plt.show()
```



Analyze Survival Rate by Ticket Class and Sex Combined

Name: Survived, dtype: float64

0.500000

0.135447

3_female

3_male

Investigate Survival Rate by Age and Fare Interaction

```
titanic_df['AgeRange'] = pd.cut(titanic_df['Age'], bins=[0, 10, 20, 30, 40, 50, float('inf')])
titanic_df['FareRange'] = pd.qcut(titanic_df['Fare'], q=4, labels=['Low', 'Medium', 'High', 'Very High'])
print(titanic_df.groupby(['AgeRange', 'FareRange'])['Survived'].mean())
```

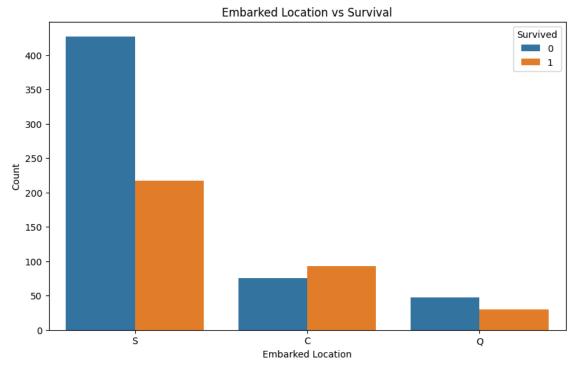
```
→ AgeRange
                  FareRange
    (0.0, 10.0]
                                    NaN
                               0.875000
                  Medium
                  High
                               0.617647
                               0.454545
                  Very High
    (10.0, 20.0]
                  Low
                               0.250000
                               0.365854
                               0.470588
                  High
                  Very High
                               0.520000
    (20.0, 30.0]
                  Low
                               0.238806
                               0.250000
                  Medium
                               0.488372
                  High
                  Very High
                               0.636364
    (30.0, 40.0]
                               0.038462
                  Low
                               0.325581
                  Medium
                  High
                               0.463415
                  Very High
                               0.777778
    (40.0, 50.0]
                               0.000000
                               0.368421
                  Medium
                  High
                               0.400000
                  Very High
                               0.516129
    (50.0, inf]
                  Low
                               0.000000
                               0.153846
                  Medium
                               0.352941
                  High
                  Very High
                               0.518519
    Name: Survived, dtype: float64
```

<ipython-input-3-dbd9974b9a4a>:3: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain of print(titanic_df.groupby(['AgeRange', 'FareRange'])['Survived'].mean())

Explore the distribution of Embarked locations for survived and non-survived passengers

```
plt.figure(figsize=(10, 6))
sns.countplot(x='Embarked', hue='Survived', data=titanic_df)
plt.title('Embarked Location vs Survival')
plt.xlabel('Embarked Location')
plt.ylabel('Count')
plt.show()
```

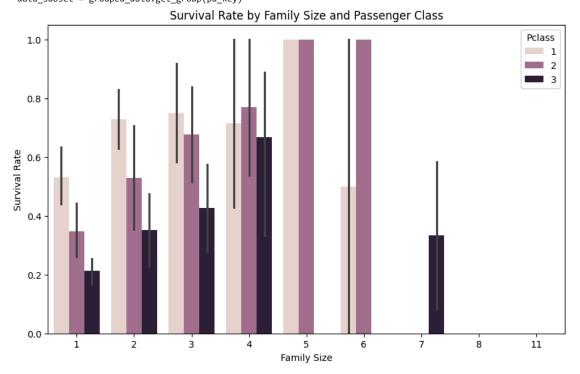
🚁 /usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key) /usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future \text{\chi} data_subset = grouped_data.get_group(pd_key)



Analyze the relationship between Family Size and Survival Rate for different passenger classes

```
titanic_df['FamilySize'] = titanic_df['SibSp'] + titanic_df['Parch'] + 1
plt.figure(figsize=(10, 6))
sns.barplot(x='FamilySize', y='Survived', hue='Pclass', data=titanic_df)
plt.title('Survival Rate by Family Size and Passenger Class')
plt.xlabel('Family Size')
plt.ylabel('Survival Rate')
plt.show()
```

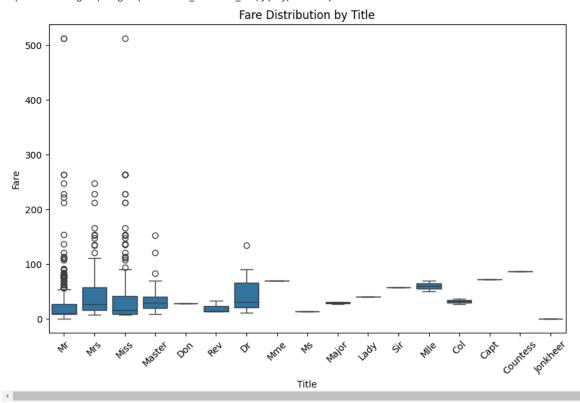
🚁 /usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key) /usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future \text{\chi} data_subset = grouped_data.get_group(pd_key) /usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future \text{\chi} data_subset = grouped_data.get_group(pd_key)



Explore the relationship between Title and Fare

```
titanic_df['Title'] = titanic_df['Name'].str.extract(' ([A-Za-z]+)\.', expand=False)
plt.figure(figsize=(10, 6))
sns.boxplot(x='Title', y='Fare', data=titanic_df)
plt.title('Fare Distribution by Title')
plt.xlabel('Title')
plt.ylabel('Fare')
plt.xticks(rotation=45)
plt.show()
```

/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)



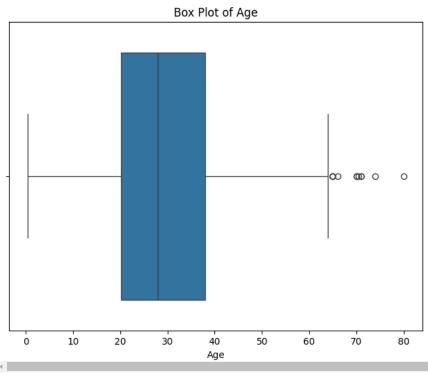
Create a correlation matrix including the new features created in the EDA process

```
numerical_features = titanic_df.select_dtypes(include=['number'])
correlation_matrix = numerical_features.corr()
print(correlation_matrix['Survived'].sort_values(ascending=False))
    Survived
                    1.000000
     Fare
                    0.257307
     Parch
                    0.081629
                    0.016639
     FamilySize
     PassengerId
                   -0.005007
     SibSp
                   -0.035322
                   -0.077221
     Pclass
                   -0.338481
     Name: Survived, dtype: float64
```

Check for outliers in Age using a box plot

```
plt.figure(figsize=(8, 6))
sns.boxplot(x='Age', data=titanic_df)
plt.title('Box Plot of Age')
plt.show()
```

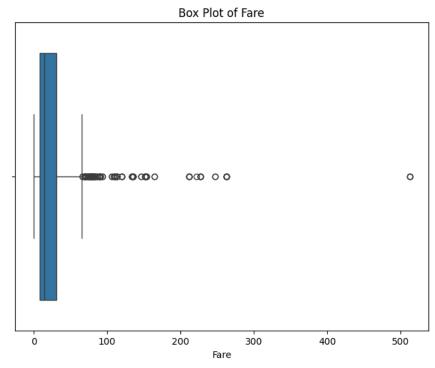
/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)



Check for outliers in Fare using a box plot

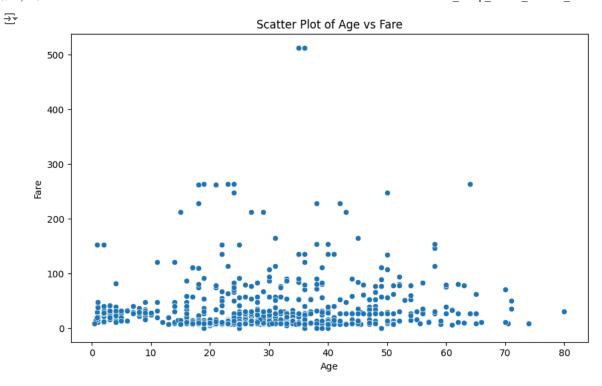
```
plt.figure(figsize=(8, 6))
sns.boxplot(x='Fare', data=titanic_df)
plt.title('Box Plot of Fare')
plt.show()
```

/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)



scatter plot to visualize outliers in Age and Fare together

```
plt.figure(figsize=(10, 6))
sns.scatterplot(x='Age', y='Fare', data=titanic_df)
plt.title('Scatter Plot of Age vs Fare')
plt.show()
```



IQR (Interquartile Range) to identify outliers more precisely

```
Q1_age = titanic_df['Age'].quantile(0.25)
Q3_age = titanic_df['Age'].quantile(0.75)
IQR_age = Q3_age - Q1_age
lower_bound_age = Q1_age - 1.5 * IQR_age
upper_bound_age = Q3_age + 1.5 * IQR_age

Q1_fare = titanic_df['Fare'].quantile(0.25)
Q3_fare = titanic_df['Fare'].quantile(0.75)
IQR_fare = Q3_fare - Q1_fare
lower_bound_fare = Q1_fare - 1.5 * IQR_fare
upper_bound_fare = Q3_fare + 1.5 * IQR_fare
```

Identify outliers for Age and Fare

```
outliers_age = titanic_df[(titanic_df['Age'] < lower_bound_age) | (titanic_df['Age'] > upper_bound_age)]
outliers_fare = titanic_df[(titanic_df['Fare'] < lower_bound_fare) | (titanic_df['Fare'] > upper_bound_fare)]
print("Outliers in Age:")
print(outliers_age)
print("\nOutliers in Fare:")
```

```
print(outliers_fare)

print(titanic_df.groupby(['Embarked', 'Pclass'])['Survived'].mean())
print(titanic_df.groupby(['Embarked', 'Fare'])['Survived'].mean())
print(titanic_df.groupby(['Embarked', 'Pclass', 'Fare'])['Survived'].mean())
```



```
7.2292 0.266667
7.8958 0.200000
....
S 153.4625 0.666667
```

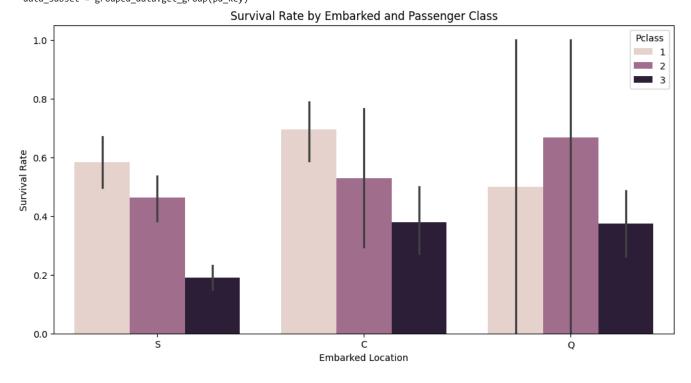
Visualize the survival rate by Embarked, Pclass, and Fare

```
plt.figure(figsize=(12, 6))
sns.barplot(x='Embarked', y='Survived', hue='Pclass', data=titanic_df)
plt.title('Survival Rate by Embarked and Passenger Class')
plt.xlabel('Embarked Location')
plt.ylabel('Survival Rate')
plt.show()
```

/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key)

/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key)

/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future v data_subset = grouped_data.get_group(pd_key)



```
plt.figure(figsize=(12, 6))
sns.boxplot(x='Embarked', y='Fare', hue='Survived', data=titanic_df)
plt.title('Fare Distribution by Embarked Location and Survival')
plt.xlabel('Embarked Location')
plt.ylabel('Fare')
```

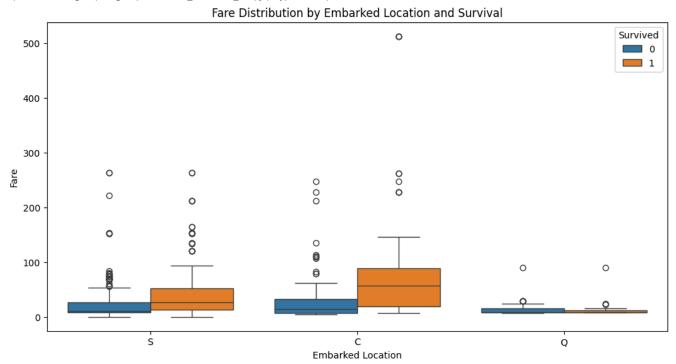
htr.zuom()

/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future \(\cdot\) data_subset = grouped_data.get_group(pd_key)

/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)

/usr/local/lib/python3.10/dist-packages/seaborn/_base.py:949: FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future \(\text{data}_subset = \text{grouped_data.get_group}(\text{pd_key}) \)

/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:640: FutureWarning: SeriesGroupBy.grouper is deprecated and will be removed in a future version of pandas. positions = grouped.grouper.result_index.to_numpy(dtype=float)



Define features (X) and target (y)

X = titanic_df.drop('Survived', axis=1)
y = titanic df['Survived']

Convert categorical features to numerical using one-hot encoding

X = pd.get_dummies(X, columns=['Sex', 'Embarked', 'Title', 'AgeGroup', 'FareRange'], dummy_na=True)