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PRESENTED BY:

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## Titanic Data Analysis

#### Introduction

The Titanic dataset contains data about the passengers who were aboard the RMS Titanic when it sank on April 15, 1912. The dataset includes various features such as passenger class, age, sex, fare, and survival status. The goal of this analysis is to explore and visualize the dataset to understand factors that influenced survival rates.

## **Import Libraries**

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

## Load Dataset

```
df = pd.read_csv('dataset/train.csv')
```

## **Data Cleaning**

### **Initial Inspection**

```
df.head(2)
   PassengerId Survived Pclass \
0
1
                                                Name
                                                         Sex
                                                                Age
SibSp \
                                                        male 22.0
                             Braund, Mr. Owen Harris
1
  Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
1
1
   Parch
             Ticket
                        Fare Cabin Embarked
0
                      7.2500
       0
         A/5 21171
                               NaN
                                          S
         PC 17599 71.2833
                               C85
# Check rows and columns
df.shape
```

```
(891, 12)
# Check datatypes of data
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
 0
                                   int64
1
     Survived
                  891 non-null
                                   int64
 2
     Pclass
                  891 non-null
                                   int64
 3
     Name
                  891 non-null
                                   object
 4
     Sex
                  891 non-null
                                   object
 5
     Age
                  714 non-null
                                   float64
 6
     SibSp
                  891 non-null
                                   int64
 7
     Parch
                  891 non-null
                                   int64
 8
     Ticket
                  891 non-null
                                   object
 9
     Fare
                  891 non-null
                                   float64
 10
                  204 non-null
    Cabin
                                   object
11
     Embarked
                  889 non-null
                                   object
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

**Note:** The dataset contains **891** entries and **12** columns, including 'PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp', 'Parch', 'Ticket', 'Fare', 'Cabin', and 'Embarked'.

# Handling Missing Values

```
df.isna().sum()
PassengerId
                  0
Survived
                  0
Pclass
Name
Sex
                  0
                177
Age
SibSp
                  0
Parch
                  0
Ticket
                  0
Fare
Cabin
                687
Embarked
dtype: int64
```

**Note:** The **Age** and **Embarked** columns have missing values. The **Cabin** column has a high number of missing values of around 687.

#### Handling Cabin Column

```
# Drop the 'Cabin' column due to a high number of missing values
df = df.drop('Cabin', axis=1)
```

#### Handling Age Column

**Note:** As we cannot fill Age through mean or mode b/c age of persons may very on it, so remove it again from dataset

```
# Drop 'Age' column due to its large number of missing values of
around more than 170+
df = df.drop('Age',axis=1)
df.isna().sum()
PassengerId
Survived
Pclass
               0
Name
Sex
               0
SibSp
               0
Parch
               0
Ticket
               0
Fare
Embarked
dtype: int64
```

#### Handling Embarked Column

```
df['Embarked'].head()

0    S
1    C
2    S
3    S
4    S
Name: Embarked, dtype: object
```

Note: Now drop null values, as we only have two null values so they don't effect our dataset

```
# Drop rows with missing 'Embarked' values
df.dropna(subset=['Embarked'], inplace=True)
```

```
df.head(2)
  PassengerId Survived Pclass \
0
1
                                                       Sex SibSp
                                              Name
Parch \
                            Braund, Mr. Owen Harris
                                                      male
  Cumings, Mrs. John Bradley (Florence Briggs Th... female
1
0
     Ticket
                Fare Embarked
  A/5 21171
              7.2500
 PC 17599 71.2833
```

# Removing Unnecessary Columns

**Note:** As we know that PassengerId, Name and Ticket doesnot tell us whether a person survive or not, so drop them from dataset

```
def remove_columns(df, columns_to_remove):
    Remove specified columns from the DataFrame if they exist.

Parameters:
    df (pd.DataFrame): The DataFrame from which columns will be removed.
    columns_to_remove (list of str): List of column names to remove.

Returns:
    pd.DataFrame: The DataFrame with the specified columns removed.
    """
# List of columns to remove
    columns_to_remove = [col for col in columns_to_remove if col in df.columns]

# Drop the columns if they exist df = df.drop(columns=columns_to_remove, axis=1)
    return df

df = remove_columns(df, ['Name', 'Fare', 'PassengerId', 'Fare', 'Ticket'])
```

```
df.columns
Index(['Survived', 'Pclass', 'Sex', 'SibSp', 'Parch', 'Embarked'],
dtype='object')
df.shape
(889, 6)
df.head(3)
   Survived
             Pclass
                         Sex
                               SibSp
                                      Parch Embarked
0
                        male
          0
                                           0
                                   1
1
           1
                   1
                      female
                                   1
                                           0
                                                     C
2
           1
                   3
                      female
                                   0
                                           0
```

## Data Exploration

#### **Descriptive Statistics**

```
# Get descriptive statistics of the dataset
df.describe()
         Survived
                        Pclass
                                                   Parch
                                      SibSp
       889.000000
                    889.000000
                                 889,000000
                                             889.000000
count
                                               0.382452
mean
         0.382452
                      2.311586
                                   0.524184
std
         0.486260
                      0.834700
                                   1.103705
                                               0.806761
         0.000000
                      1.000000
                                   0.000000
                                               0.000000
min
                                               0.000000
25%
         0.000000
                      2.000000
                                   0.000000
50%
         0.000000
                      3.000000
                                   0.000000
                                               0.00000
75%
         1.000000
                      3.000000
                                   1.000000
                                               0.000000
         1.000000
                      3.000000
                                   8.000000
                                               6.000000
max
```

**Note:** This provides a summary of numerical features in the dataset, including counts, means, standard deviations, and ranges.

## Data Visualization

```
Font Dicts For Different Purpose

title_fontdict = {
   'fontsize': 20,
   'fontweight': 'bold',
```

```
'family': 'serif',
    'color': 'darkblue',
    'verticalalignment': 'top'
}
xlabel fontdict = {
    'fontsize': 14,
    'fontweight': 'medium',
    'family': 'sans-serif',
    'color': 'darkgreen',
    'horizontalalignment': 'right'
}
ylabel_fontdict = {
    'fontsize': 14,
    'fontweight': 'medium',
    'family': 'sans-serif',
    'color': 'darkred',
    'horizontalalignment': 'left'
}
ticklabel_fontdict = {
    'fontsize': 12,
    'fontweight': 'light',
    'family': 'monospace',
    'color': 'black'
}
legend_fontdict = {
    'fontsize': 12,
    'fontweight': 'normal',
    'family': 'serif',
    'color': 'purple'
}
grid_fontdict = {
    'fontsize': 10,
    'fontweight': 'light',
    'family': 'sans-serif',
    'color': 'gray'
}
```

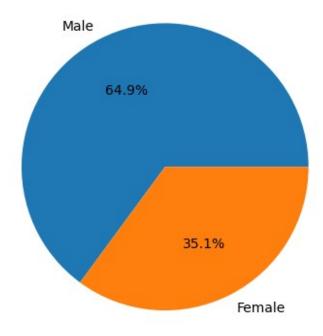
#### Distribution of Gender

```
g_counts = df['Sex'].value_counts()

# Pie chart showing the distribution of male and female passengers
plt.pie(g counts,labels=g counts.index.str.title(), autopct='%1.1f%%')
```

```
plt.title("Distribution Of Gender", fontdict=title_fontdict)
plt.show()
```

## **Distribution Of Gender**



**Note:** This chart illustrates the proportion of male and female passengers aboard the Titanic.

## Survivors by Gender

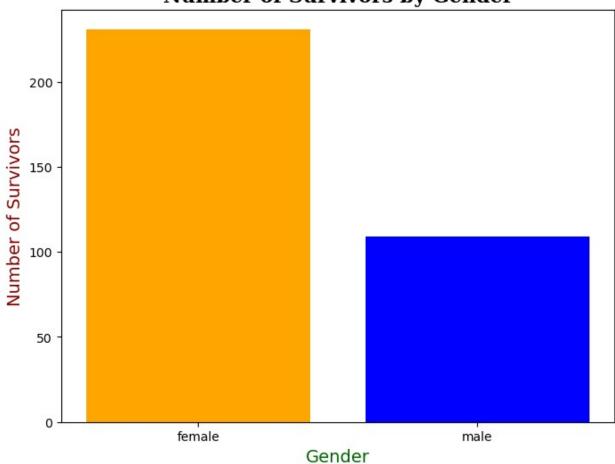
```
# Group by 'Sex' and sum 'Survived' to get the number of survivors by
gender
survival_by_gender = df.groupby('Sex')['Survived'].sum()

# Plotting
plt.figure(figsize=(8, 6))
plt.bar(survival_by_gender.index, survival_by_gender.values,
color=['orange', 'blue'])

# Adding title and labels
plt.title('Number of Survivors by Gender', fontsize=16,
fontweight='bold', family='serif')
plt.xlabel('Gender', fontdict=xlabel_fontdict)
```

```
plt.ylabel('Number of Survivors', fontdict=ylabel_fontdict)
# Show the plot
plt.show()
```





Note: This chart displays the number of survivors categorized by gender.

```
Sex
female 231
male 109
Name: Survived, dtype: int64
survival_by_gender.index
Index(['female', 'male'], dtype='object', name='Sex')
```

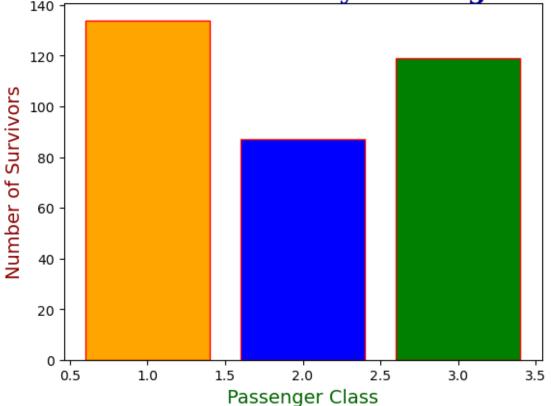
```
survival_by_gender.values
array([231, 109], dtype=int64)
```

# Survivors by Passenger Class

```
# Group by 'Pclass' and sum 'Survived' to get the number of survivors
by class
pclass_survived = df.groupby('Pclass')['Survived'].sum()

# Bar chart showing the number of survivors by passenger class
plt.bar(pclass_survived.index, pclass_survived.values,
color=('orange','blue','green'), edgecolor='red')
plt.title("Number of Survivors by Passenger Class",
fontdict=title_fontdict)
plt.xlabel("Passenger Class", fontdict=xlabel_fontdict)
plt.ylabel("Number of Survivors", fontdict=ylabel_fontdict)
plt.show()
```





**Note:** This chart shows the distribution of survivors across different passenger classes.

## Passenger Class Distribution by Gender

```
# Group by 'Sex' and 'Pclass', then count occurrences
pclass_gender_counts = df.groupby(['Sex', 'Pclass']).size().unstack()

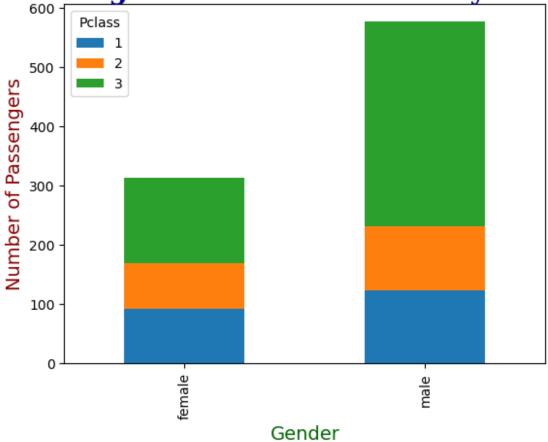
# Plotting
plt.figure(figsize=(10, 7))
pclass_gender_counts.plot(kind='bar', stacked=True)

# Adding title and labels
plt.title('Passenger Class Distribution by Gender',
fontdict=title_fontdict)
plt.xlabel('Gender', fontdict=xlabel_fontdict)
plt.ylabel('Number of Passengers', fontdict=ylabel_fontdict)

# Show the plot
plt.show()
```

<Figure size 1000x700 with 0 Axes>

# Passenger Class Distribution by Gender



**Note:** This stacked bar chart illustrates the **distribution of passengers** across different classes, **separated by gender**.

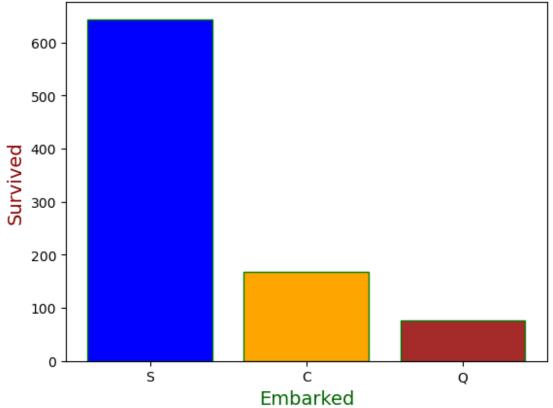
pclass_gender_counts			
Pclass Sex	1	2	3
female male	92 122	76 108	144 347

## Survival Rate by Embarked Location

```
# Count the number of passengers by embarkation point
e_count = df['Embarked'].value_counts()

# Bar chart showing the number of passengers embarking at different
locations
plt.bar(e_count.index,
e_count.values,edgecolor='green',color=('blue','orange','brown'))
plt.title("Survival On The Basis Of Embarked",
fontdict=title_fontdict)
plt.xlabel("Embarked", fontdict=xlabel_fontdict)
plt.ylabel("Survived", fontdict=ylabel_fontdict)
plt.show()
```

## Survival On The Basis Of Embarked



**Note:** This chart shows the **number of passengers** boarding from each **embarkation** point.

## Distribution of Siblings Aboard

```
# Plotting histogram
plt.figure(figsize=(8, 6))
plt.hist(df['SibSp'], bins=range(df['SibSp'].min(), df['SibSp'].max()
+ 2), color='orange', edgecolor='black')
plt.title('Distribution of Siblings Aboard', fontdict=title_fontdict)
plt.xlabel('Number of Siblings', fontdict=xlabel_fontdict)
plt.ylabel('Number of Passengers', fontdict=ylabel_fontdict)
plt.grid(True)
plt.show()
```



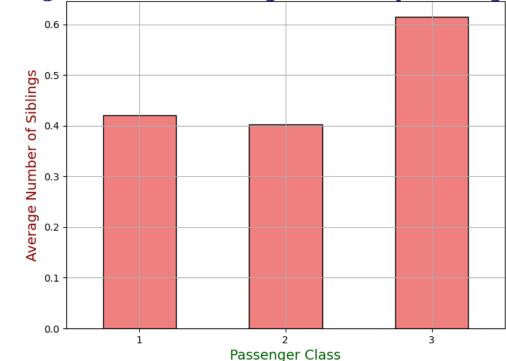
**Note:** This histogram displays how many siblings each passenger had aboard.

## Average Number of Siblings by Passenger Class

```
# Group by 'Pclass' and calculate average 'SibSp'
average_sibsp_by_pclass = df.groupby('Pclass')['SibSp'].mean()

# Plotting bar chart
plt.figure(figsize=(8, 6))
average_sibsp_by_pclass.plot(kind='bar', color='lightcoral',
edgecolor='black')
plt.title('Average Number of Siblings Aboard by Passenger Class',
fontdict=title_fontdict)
plt.xlabel('Passenger Class', fontdict=xlabel_fontdict)
plt.ylabel('Average Number of Siblings', fontdict=ylabel_fontdict)
plt.ylabel('Average Number of Siblings', fontdict=ylabel_fontdict)
plt.grid(True)
plt.grid(True)
plt.show()
```

#### **Average Number of Siblings Aboard by Passenger Class**



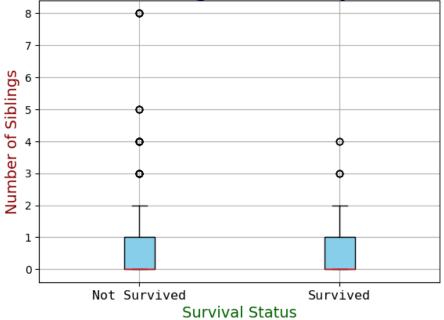
**Note:** This chart shows the average number of siblings aboard for each passenger class.

## Siblings Aboard by Survival Status

```
# Plotting box plot
plt.figure(figsize=(8, 6))
```

```
df.boxplot(column='SibSp', by='Survived', grid=False,
color='darkblue', patch artist=True,
           boxprops=dict(facecolor='skyblue', color='black'),
           whiskerprops=dict(color='black'),
           capprops=dict(color='black'),
           medianprops=dict(color='red'))
plt.title('Distribution of Siblings Aboard by Survival Status',
fontdict=title fontdict)
plt.suptitle('') # Remove default title
plt.xlabel('Survival Status', fontdict=xlabel fontdict)
plt.ylabel('Number of Siblings', fontdict=ylabel_fontdict)
plt.xticks([1, 2], ['Not Survived', 'Survived'],
fontdict=ticklabel fontdict)
plt.grid(True)
plt.show()
<Figure size 800x600 with 0 Axes>
```

#### Distribution of Siblings Aboard by Survival Status



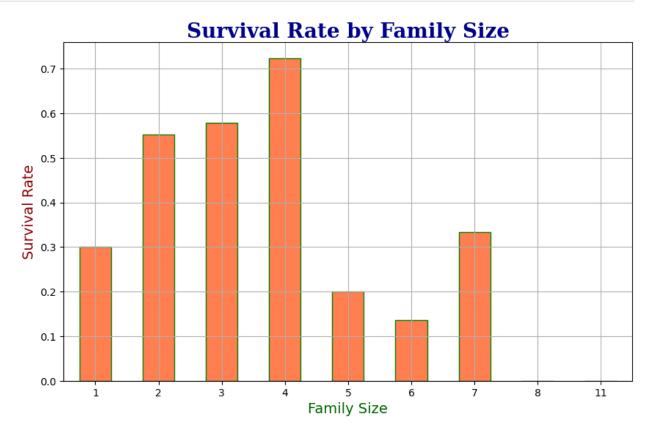
**Note:** This box plot visualizes the **distribution** of **siblings** aboard, comparing survivors to non-survivors.

## Fare Distribution

```
# Create a new feature for family size
df['FamilySize'] = df['SibSp'] + df['Parch'] + 1

# Group by family size and calculate survival rate
family_size_survival = df.groupby('FamilySize')['Survived'].mean()

# Bar chart showing survival rates by family size
plt.figure(figsize=(10, 6))
family_size_survival.plot(kind='bar', color='coral',
edgecolor='green')
plt.title('Survival Rate by Family Size', fontdict=title_fontdict)
plt.xlabel('Family Size', fontdict=xlabel_fontdict)
plt.ylabel('Survival Rate', fontdict=ylabel_fontdict)
plt.xticks(rotation=0)
plt.grid(True)
plt.show()
```



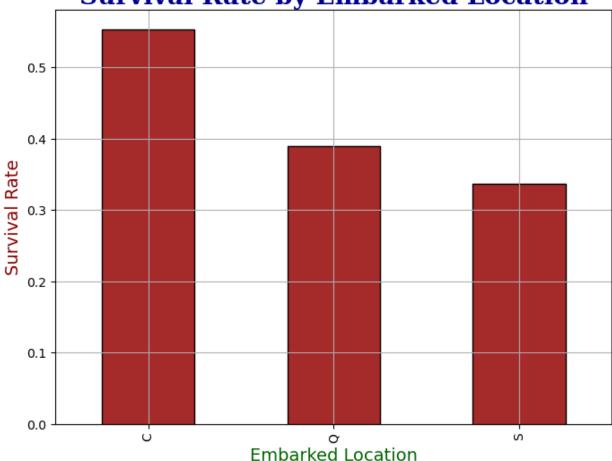
**Note:** Examining survival rates based on family size can provide insights into the impact of having family aboard.

# Detailed Analysis of Embarked Locations

```
# Survival rate by embarkation point
embarked_survival = df.groupby('Embarked')['Survived'].mean()

# Bar chart showing survival rates by embarkation point
plt.figure(figsize=(8, 6))
embarked_survival.plot(kind='bar', color='brown', edgecolor='black')
plt.title('Survival Rate by Embarked Location',
fontdict=title_fontdict)
plt.xlabel('Embarked Location', fontdict=xlabel_fontdict)
plt.ylabel('Survival Rate', fontdict=ylabel_fontdict)
plt.grid(True)
plt.show()
```



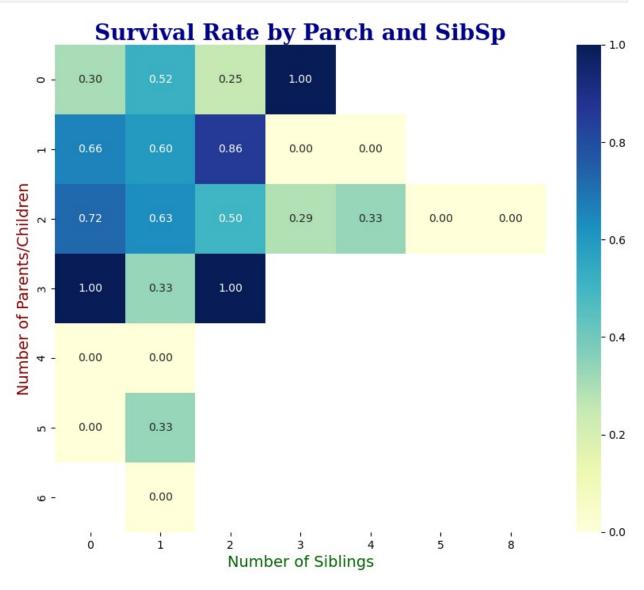


**Note:** This analysis provides survival rates for each embarkation point, giving additional context to previous charts.

## Survival by Parch and SibSp

```
# Group by 'Parch' and 'SibSp' and calculate survival rates
parch_sibsp_survival = df.groupby(['Parch', 'SibSp'])
['Survived'].mean().unstack()

# Heatmap showing survival rates by number of parents/children and
siblings/spouses
plt.figure(figsize=(10, 8))
sns.heatmap(parch_sibsp_survival, annot=True, cmap='YlGnBu',
fmt='.2f')
plt.title('Survival Rate by Parch and SibSp', fontdict=title_fontdict)
plt.xlabel('Number of Siblings', fontdict=xlabel_fontdict)
plt.ylabel('Number of Parents/Children', fontdict=ylabel_fontdict)
plt.show()
```



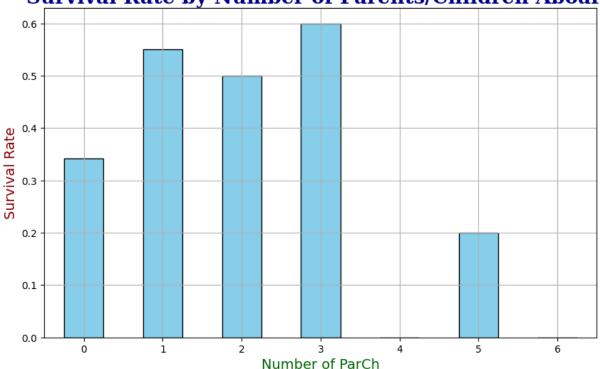
**Note:** This heatmap provides a detailed view of how survival rates vary with the number of parents/children and siblings/spouses.

# Survival Rate by Parch

```
# Group by 'Parch' and calculate survival rate
parch_survival = df.groupby('Parch')['Survived'].mean()

# Plotting bar chart for survival rates by 'Parch'
plt.figure(figsize=(10, 6))
parch_survival.plot(kind='bar', color='skyblue', edgecolor='black')
plt.title('Survival Rate by Number of Parents/Children Aboard',
fontdict=title_fontdict)
plt.xlabel('Number of ParCh', fontdict=xlabel_fontdict)
plt.ylabel('Survival Rate', fontdict=ylabel_fontdict)
plt.xticks(rotation=0)
plt.grid(True)
plt.show()
```





**Note:** This bar chart shows how the survival rate changes with the number of parents or children aboard.

# Summary and Conclusion

In this analysis, we explored the Titanic dataset to understand various factors affecting survival rates. Key findings include:

- The majority of passengers were male.
- Female passengers had a higher survival rate compared to male passengers.
- Passengers in higher classes had better survival rates.
- Passengers embarking from locations C,Q and S had distinct patterns in terms of survival.