

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

```
dt = sns.load_dataset("titanic")
```

```
dt.head()
```

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

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

```
dt.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   survived              891 non-null    int64
1   pclass                891 non-null    int64
2   sex                   891 non-null    object
3   age                   714 non-null    float64
4   sibsp                 891 non-null    int64
5   parch                 891 non-null    int64
6   fare                  891 non-null    float64
7   embarked              889 non-null    object
8   class                 891 non-null    category
9   who                   891 non-null    object
10  adult_male            891 non-null    bool
11  deck                  203 non-null    category
12  embark_town           889 non-null    object
13  alive                 891 non-null    object
```

```
14 alone      891 non-null    bool
dtypes: bool(2), category(2), float64(2), int64(4), object(5)
memory usage: 80.7+ KB
```

```
dt.describe()
```

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

```
dt.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
```

```
# Define highly contrasting colors
```

```
custom_palette = {0: 'red', 1: 'blue'} # 0: Not Survived (Red), 1: Survived (Blue)
```

```
# Create the box plot
```

```
plt.figure(figsize=(10, 6))
```

```

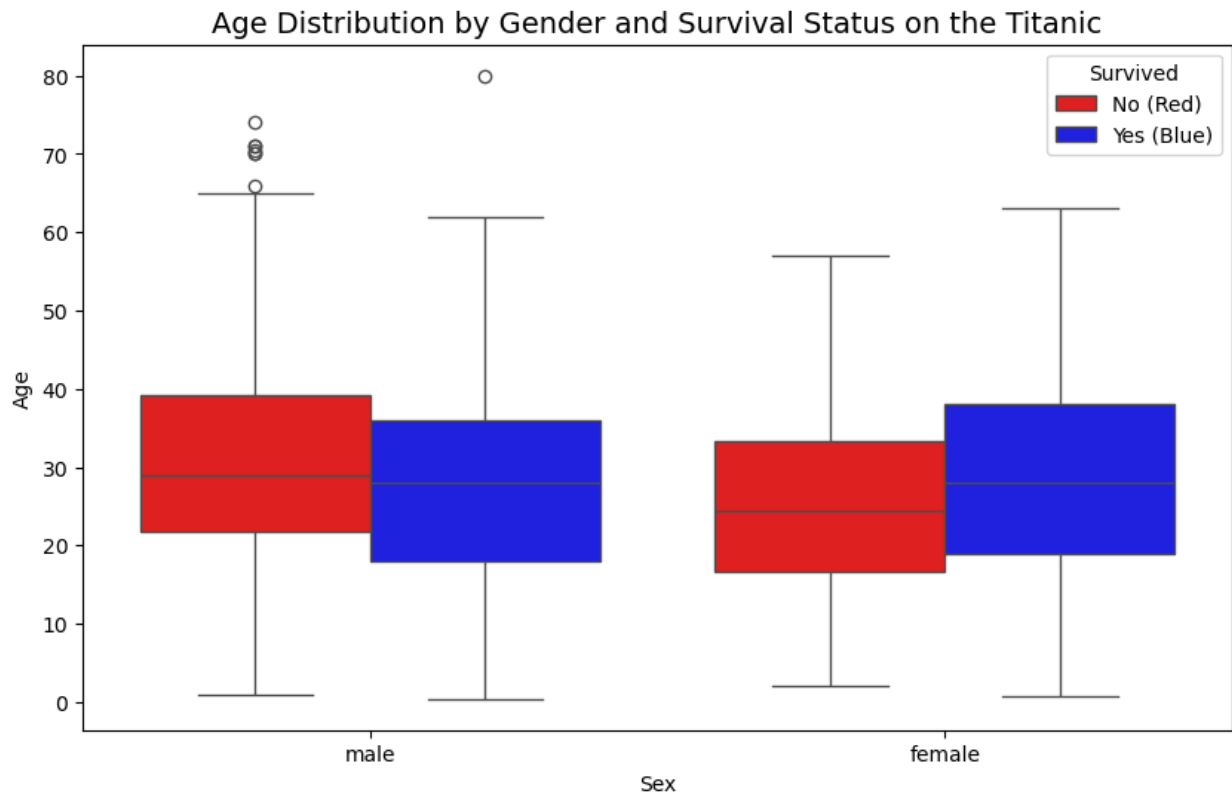
sns.boxplot(x='sex', y='age', hue='survived', data=dt,
palette=custom_palette)

# Fix legend labels properly
legend_labels = ['No (Red)', 'Yes (Blue)']
handles, labels = plt.gca().get_legend_handles_labels()
plt.legend(handles, legend_labels, title='Survived')

# Add title and labels
plt.title('Age Distribution by Gender and Survival Status on the
Titanic', fontsize=14)
plt.xlabel('Sex')
plt.ylabel('Age')

# Show plot
plt.show()

```



OBSERVATION:- The boxplot reveals that survival on the Titanic was influenced by both gender and age. Female passengers had a higher survival rate compared to males, as indicated by the larger presence of the blue (survived) boxes among females. In both genders, survivors tended to be younger, with the median age of survivors slightly lower than that of non-survivors. Notably, many children, especially young girls, survived, supporting the “women and children first” policy. In contrast, a significant number of males did not survive, particularly adults. The presence of outliers also shows that a few elderly individuals were among both the survivors and non-survivors.

```

# Define distinct colors for survival status
custom_palette = {0: 'red', 1: 'blue'} # 0: Did not survive, 1: Survived

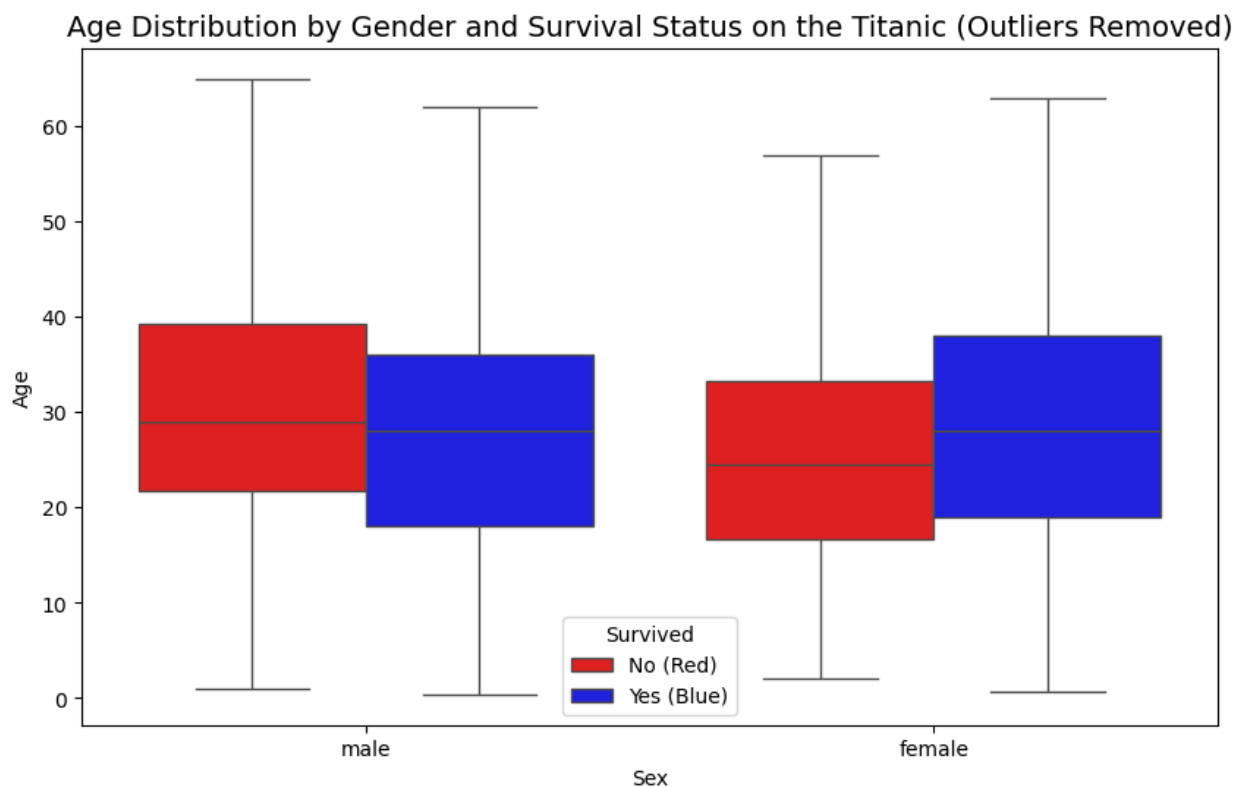
# Create the box plot without outliers
plt.figure(figsize=(10, 6))
sns.boxplot(x='sex', y='age', hue='survived', data=dt,
            palette=custom_palette, showfliers=False)

# Fix legend labels properly
legend_labels = ['No (Red)', 'Yes (Blue)']
handles, labels = plt.gca().get_legend_handles_labels()
plt.legend(handles, legend_labels, title='Survived')

# Add title and labels
plt.title('Age Distribution by Gender and Survival Status on the Titanic (Outliers Removed)', fontsize=14)
plt.xlabel('Sex')
plt.ylabel('Age')

# Show plot
plt.show()

```



OBSERVATION:- This boxplot shows the age of men and women on the Titanic and whether they survived, but without showing the extreme age values (outliers). From the graph, we can see that more women survived than men. Also, people who survived were usually younger than

those who didn't. Removing the outliers makes it easier to see the main age group for each category and better understand the pattern of survival.

```
# Count of survival status grouped by gender
survival_counts = dt.groupby(['sex', 'survived']).size().unstack()

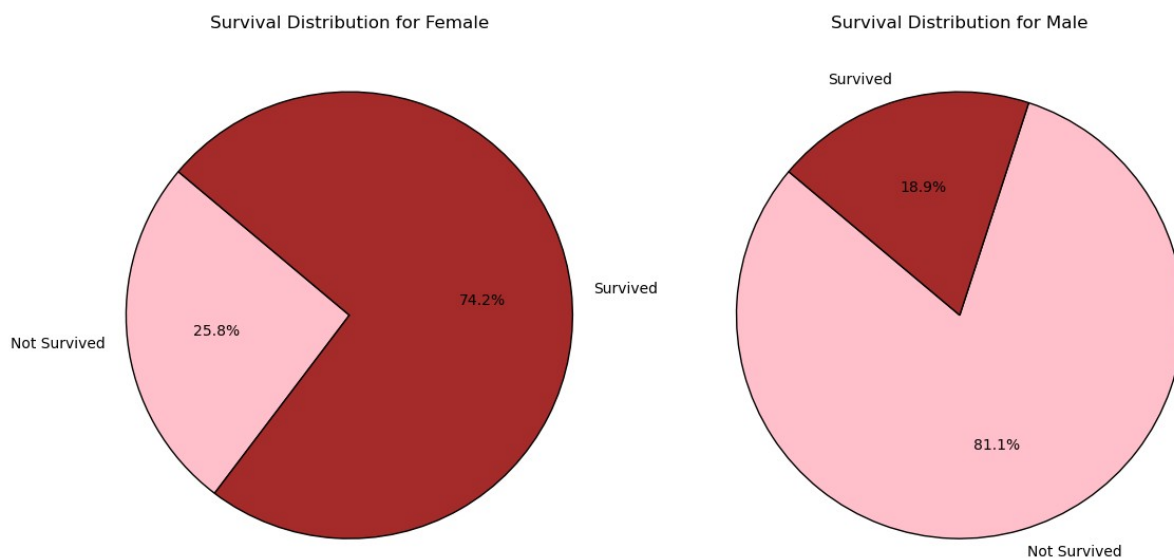
# Define colors for better visualization
colors = ['pink', 'brown'] # pink: Not survived, brown: Survived

# Create the pie charts
fig, axes = plt.subplots(1, 2, figsize=(12, 6))

# Plot each pie chart by gender
for i, gender in enumerate(survival_counts.index):
    axes[i].pie(survival_counts.loc[gender],
                labels=['Not Survived', 'Survived'],
                autopct='%1.1f%%',
                colors=colors,
                startangle=140,
                wedgeprops={'edgecolor': 'black'})

    axes[i].set_title(f'Survival Distribution for {gender.capitalize()}')

plt.tight_layout()
plt.show()
```



OBSERVATION:- The pie charts show that more women survived the Titanic than men. In the chart for women, the bigger part is for "Survived," while in the men's chart, the bigger part is for "Not Survived." This means women had a better chance of survival. It clearly shows that during the rescue, women were given more priority than men, which follows the rule of "women and children first."

