

1. Use the inbuilt dataset 'titanic' as used in the above problem. Plot a box plot for distribution of age with respect to each gender along with the information about whether they survived or not. (Column names : 'sex' and 'age')
2. Write observations on the inference from the above statistics.

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import warnings
warnings.filterwarnings("ignore")
%matplotlib inline
```

Load data and basic stats

```
In [2]: df = pd.read_csv("train.csv")
```

```
In [3]: df.shape
```

```
Out[3]: (891, 12)
```

```
In [4]: df
```

```
Out[4]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S
...
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	S
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	S
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	S
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	C
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN	Q

891 rows × 12 columns

```
In [5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  -
0   PassengerId     891 non-null    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  Cabin           204 non-null    object
11  Embarked        889 non-null    object
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

```
In [6]: df.describe()
```

Out[6]:	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
	count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000
	mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594
	std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057
	min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000
	25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000
	50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000
	75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000
	max	891.000000	1.000000	3.000000	80.000000	8.000000	512.329200

```
In [7]: df.isna().sum()
```

```
Out[7]: PassengerId      0
Survived      0
Pclass        0
Name          0
Sex           0
Age          177
SibSp         0
Parch         0
Ticket        0
Fare          0
Cabin        687
Embarked      2
dtype: int64
```

```
In [8]: df["Age"] = df["Age"].fillna(df["Age"].mean())
```

```
In [9]: df.isna().sum()
```

```
Out[9]: PassengerId      0
Survived      0
Pclass        0
Name          0
Sex           0
Age           0
SibSp         0
Parch         0
Ticket        0
Fare          0
Cabin        687
Embarked      2
dtype: int64
```

Visualization

```
In [10]: def fun1(value):
         if (value == "male"):
             return 1
         else:
             return 0
```

```
In [11]: def fun2(value):
         if (value == 'S'):
             return 0
         elif (value == 'C'):
             return 1
         elif (value == 'Q'):
             return 2
         else:
             return 0
```

```
In [12]: df["Sex"] = df["Sex"].apply(fun1)
```

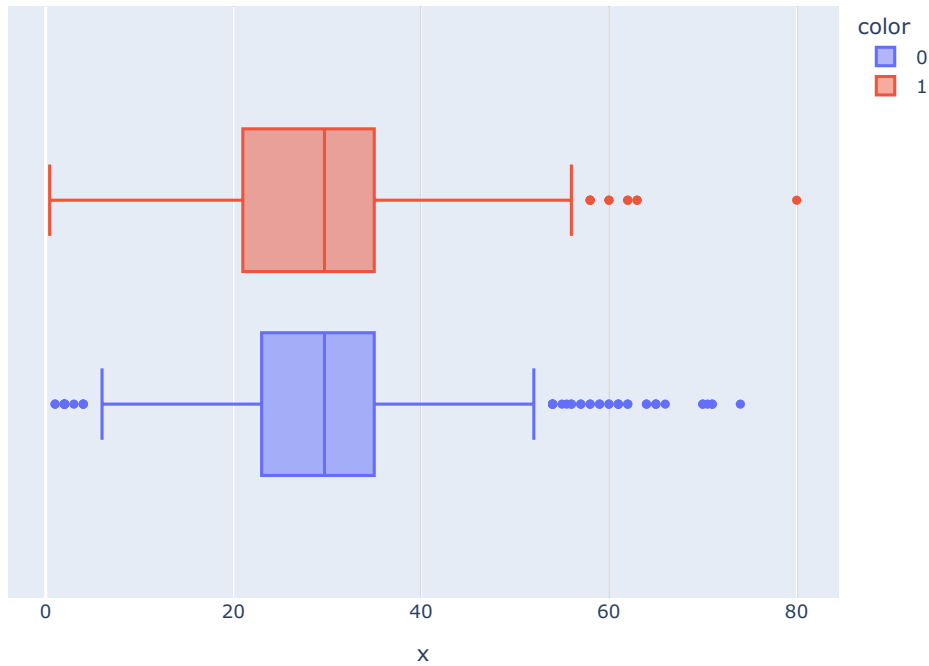
```
In [13]: df["Embarked"] = df["Embarked"].apply(fun2)
```

```
In [14]: df = df.drop("Cabin", axis=1)
```

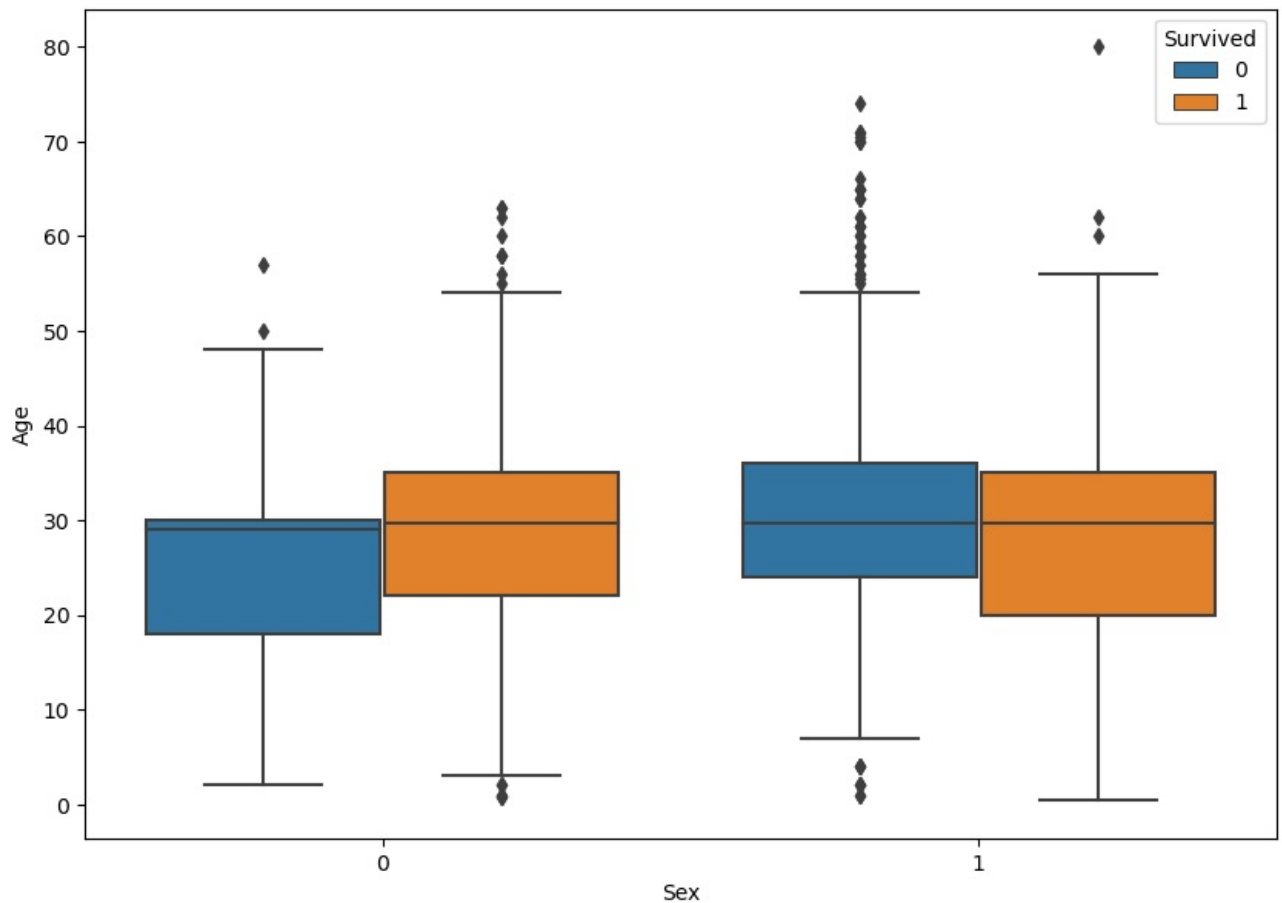
```
In [15]: df.shape
```

```
Out[15]: (891, 11)
```

```
In [16]: px.box(df["Sex"], df["Age"], color=df["Survived"])
```



```
In [17]: plt.figure(figsize=(10,7))
box = sns.boxplot(df,x="Sex", y="Age", hue="Survived")
plt.show()
```



This code will display a box plot showing the distribution of ages with respect to gender and survival status. You can observe trends like whether there are age differences between survivors and non-survivors, or if gender has a distinct influence on survival outcomes.

