



Exploratory Data Analysis (EDA) on Titanic Dataset

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Reading csv file through file path.

```
[1]: import numpy as np
import pandas as pd
df=pd.read_csv("C:\\Users\\mayan\\Downloads\\titanic\\train.csv")
df.head()
```

```
[1]:
```

	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

Question 1. Use .describe(), .info(), .value_counts().

```
[4]: df.describe()
```

```
[4]:
```

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

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

```
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
```

[27]: df.value_counts()



[27]:

PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	
2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Thayer)	female	38.0	1	0	PC 17599	71.2833	C85	C	1
572	1	1	Appleton, Mrs. Edward Dale (Charlotte Lamson)	female	53.0	2	0	11769	51.4792	C101	S	1
578	1	1	Silvey, Mrs. William Baird (Alice Munger)	female	39.0	1	0	13507	55.9000	E44	S	1
582	1	1	Thayer, Mrs. John Borland (Marian Longstreth Morris)	female	39.0	1	1	17421	110.8833	C68	C	1
584	0	1	Ross, Mr. John Hugo	male	36.0	0	0	13049	40.1250	A10	C	1
...												
328	1	2	Ball, Mrs. (Ada E Hall)	female	36.0	0	0	28551	13.0000	D	S	1
330	1	1	Hippach, Miss. Jean Gertrude	female	16.0	0	1	111361	57.9792	B18	C	1
332	0	1	Partner, Mr. Austen	male	45.5	0	0	113043	28.5000	C124	S	1
333	0	1	Graham, Mr. George Edward	male	38.0	0	1	PC 17582	153.4625	C91	S	1
	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	C	1

Name: count, Length: 183, dtype: int64

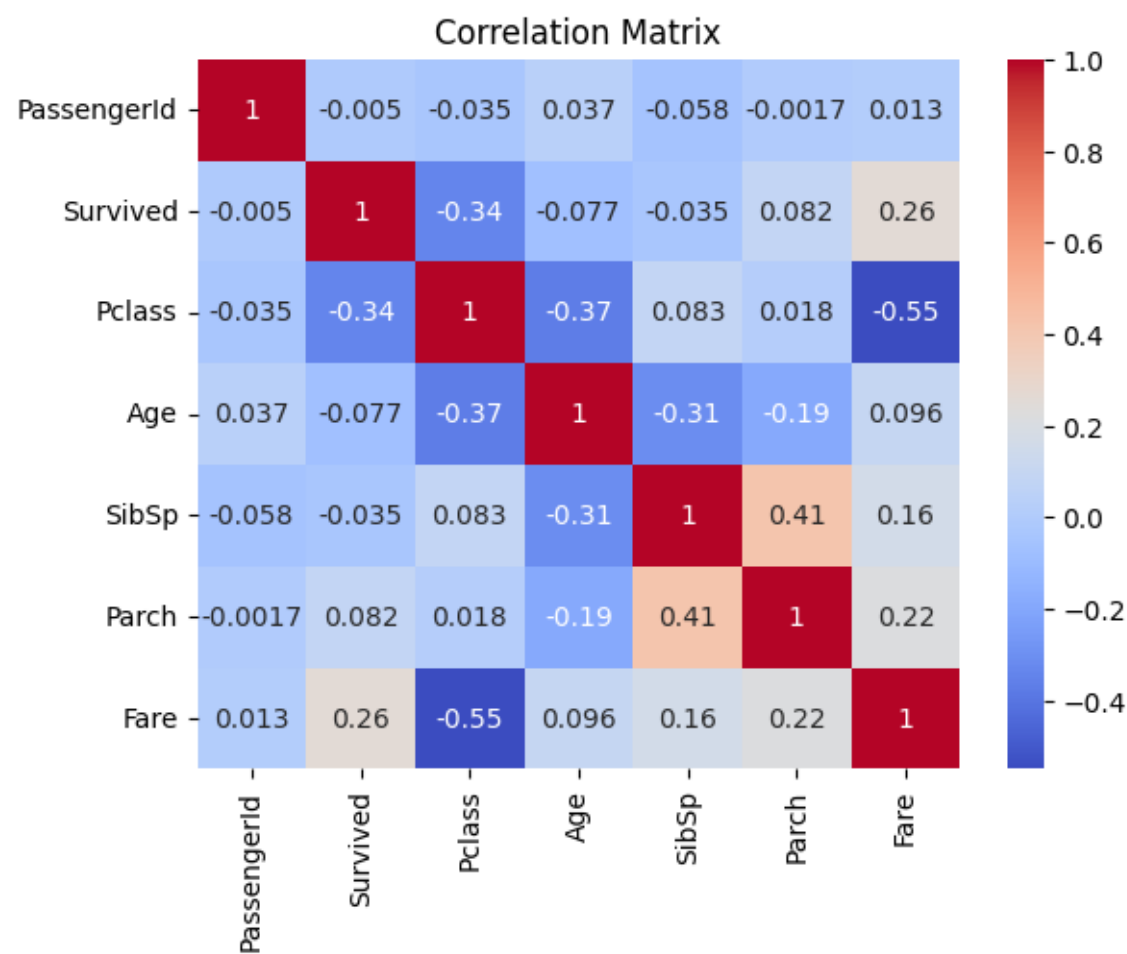
Expand Output

Question 2. Use sns.pairplot(), sns.heatmap() for visualization.

Pairplot :-

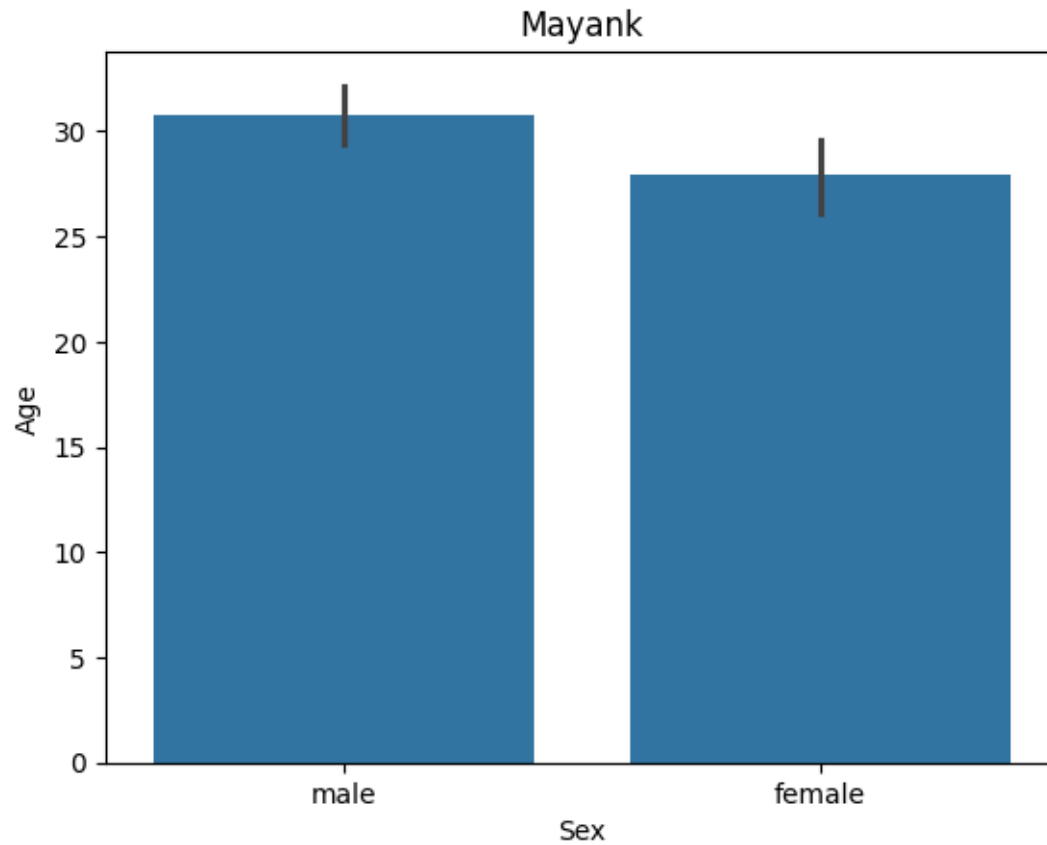


Heatmap :-



Question 3. Identify relationships and trends.

Barplot :-

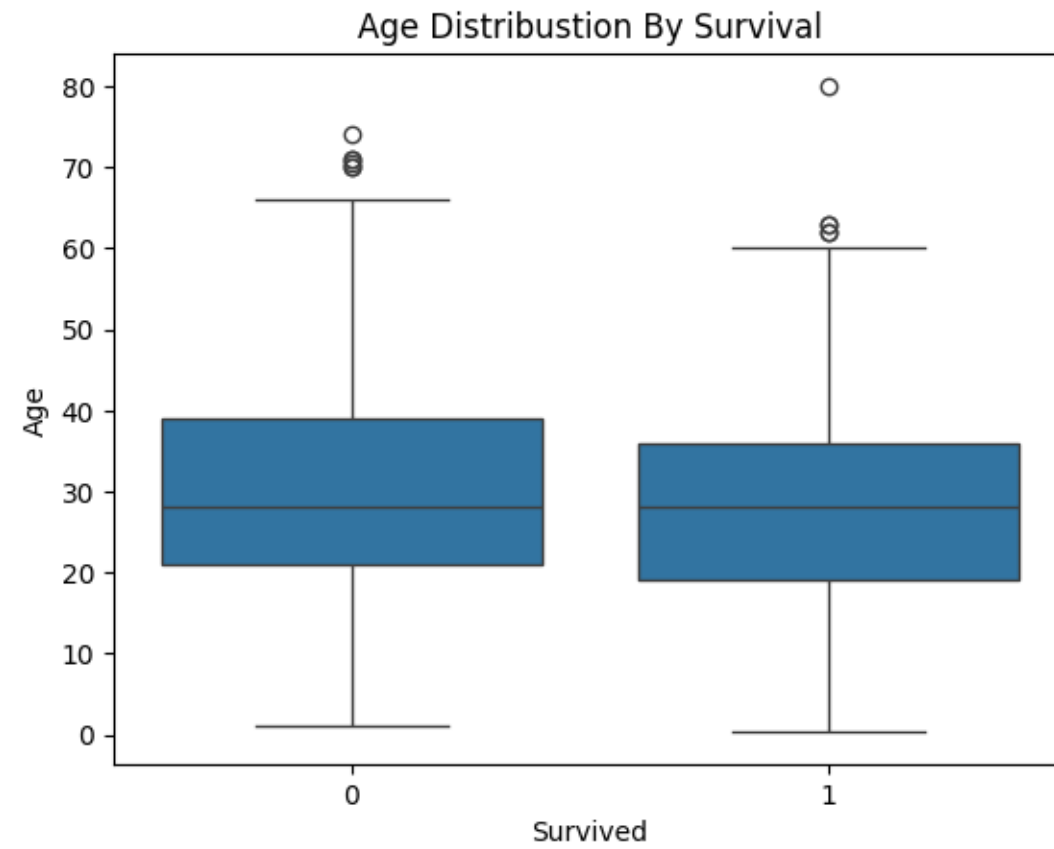


Question 4. Plot histograms, boxplots, scatterplots

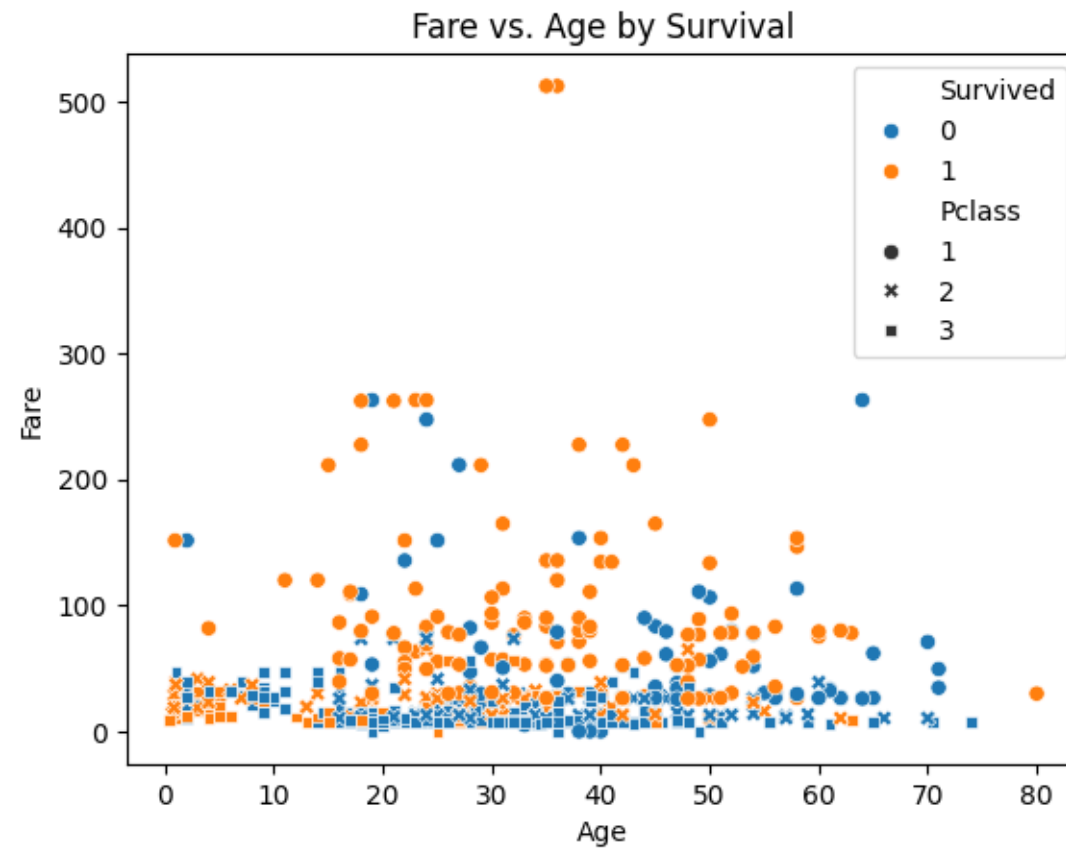
Histogram :-



Boxplot : -



Scatter Plot : -



Summary : -

Passengers in higher classes, who paid higher fares, were more likely to survive. Younger individuals also showed higher survival rates. Most passengers were concentrated in the 20-30 age range and paid lower fares, but survival was skewed towards those with higher fares and better class. Gender and age played roles in survival outcomes, with variability seen in the data.



Thank You