

Titanic Dataset: Exploratory Data Analysis

Generated on 2025-08-19 14:41:58

Using uploaded titanic.csv (Kaggle Titanic train.csv)

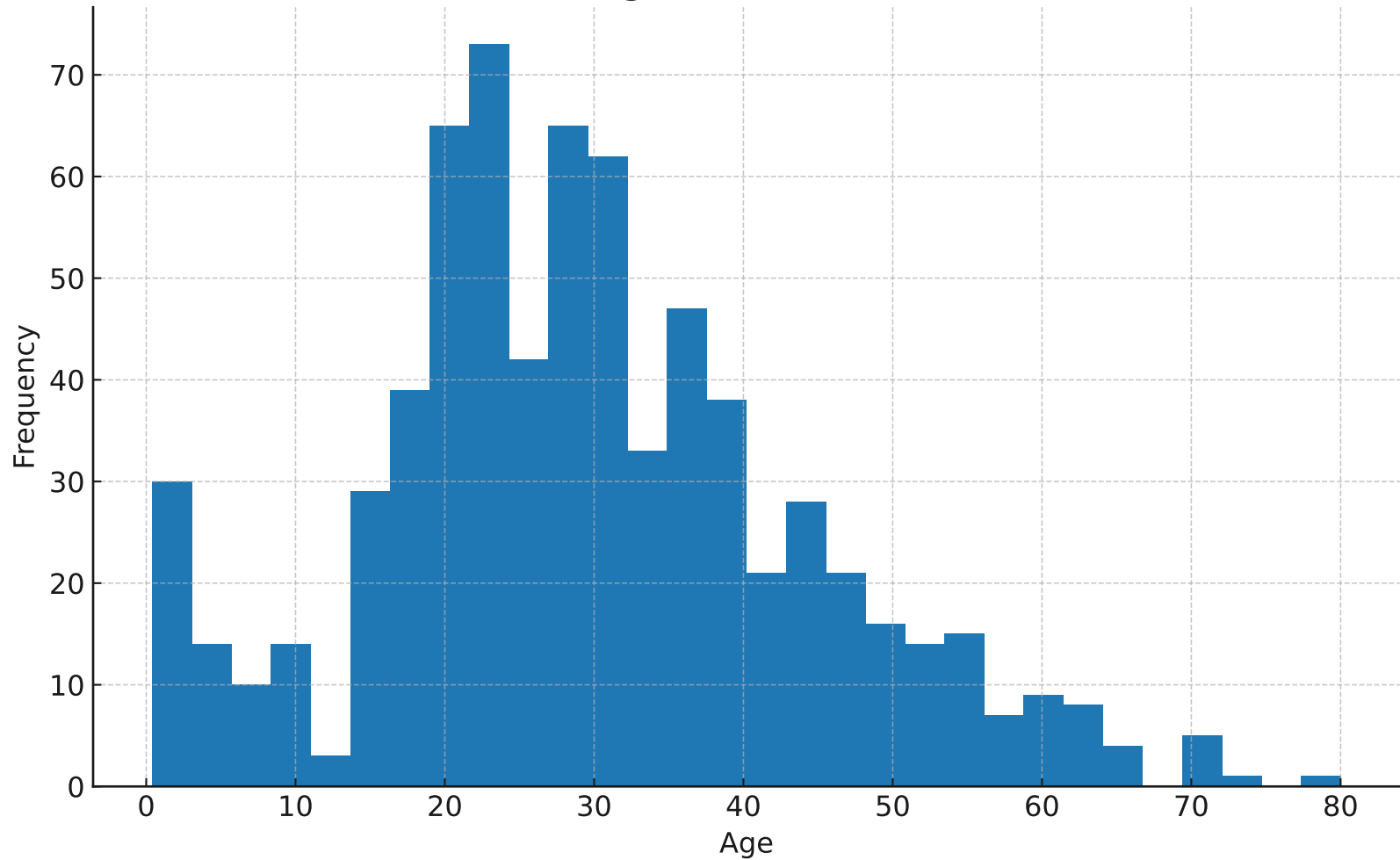
Rows: 891, Columns: 12

```
=== .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
```

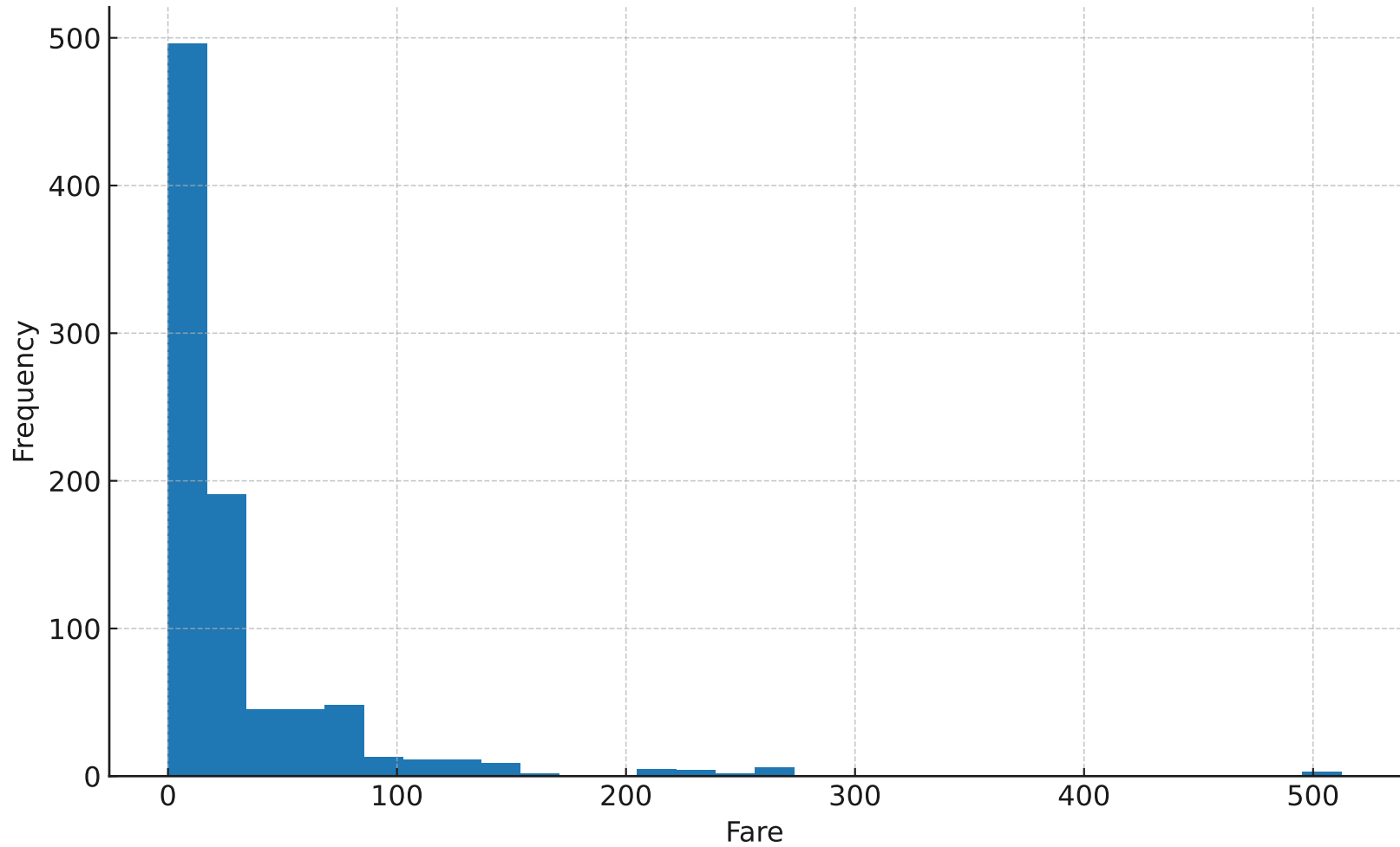
```
=== Missing Values ===
Cabin      687
Age        177
Embarked    2
PassengerId 0
Survived    0
Pclass      0
Name        0
Sex         0
SibSp       0
Parch       0
Ticket      0
Fare        0
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
count	891.000000	891.000000	891.000000	891	891	714.000000	891.000000	891.000000	891	891.000000	204	88
unique	NaN	NaN	NaN	891	2	NaN	NaN	NaN	681	NaN	147	
top	NaN	NaN	NaN	Braund, Mr. Owen Harris	male	NaN	NaN	NaN	347082	NaN	B96 B98	
freq	NaN	NaN	NaN	1	577	NaN	NaN	NaN	7	NaN	4	64
mean	446.000000	0.383838	2.308642	NaN	NaN	29.699118	0.523008	0.381594	NaN	32.204208	NaN	NaN
std	257.353842	0.486592	0.836071	NaN	NaN	14.526497	1.102743	0.806057	NaN	49.693429	NaN	NaN
min	1.000000	0.000000	1.000000	NaN	NaN	0.420000	0.000000	0.000000	NaN	0.000000	NaN	NaN
25%	223.500000	0.000000	2.000000	NaN	NaN	20.125000	0.000000	0.000000	NaN	7.910400	NaN	NaN
50%	446.000000	0.000000	3.000000	NaN	NaN	28.000000	0.000000	0.000000	NaN	14.454200	NaN	NaN
75%	668.500000	1.000000	3.000000	NaN	NaN	38.000000	1.000000	0.000000	NaN	31.000000	NaN	NaN
max	891.000000	1.000000	3.000000	NaN	NaN	80.000000	8.000000	6.000000	NaN	512.329200	NaN	NaN

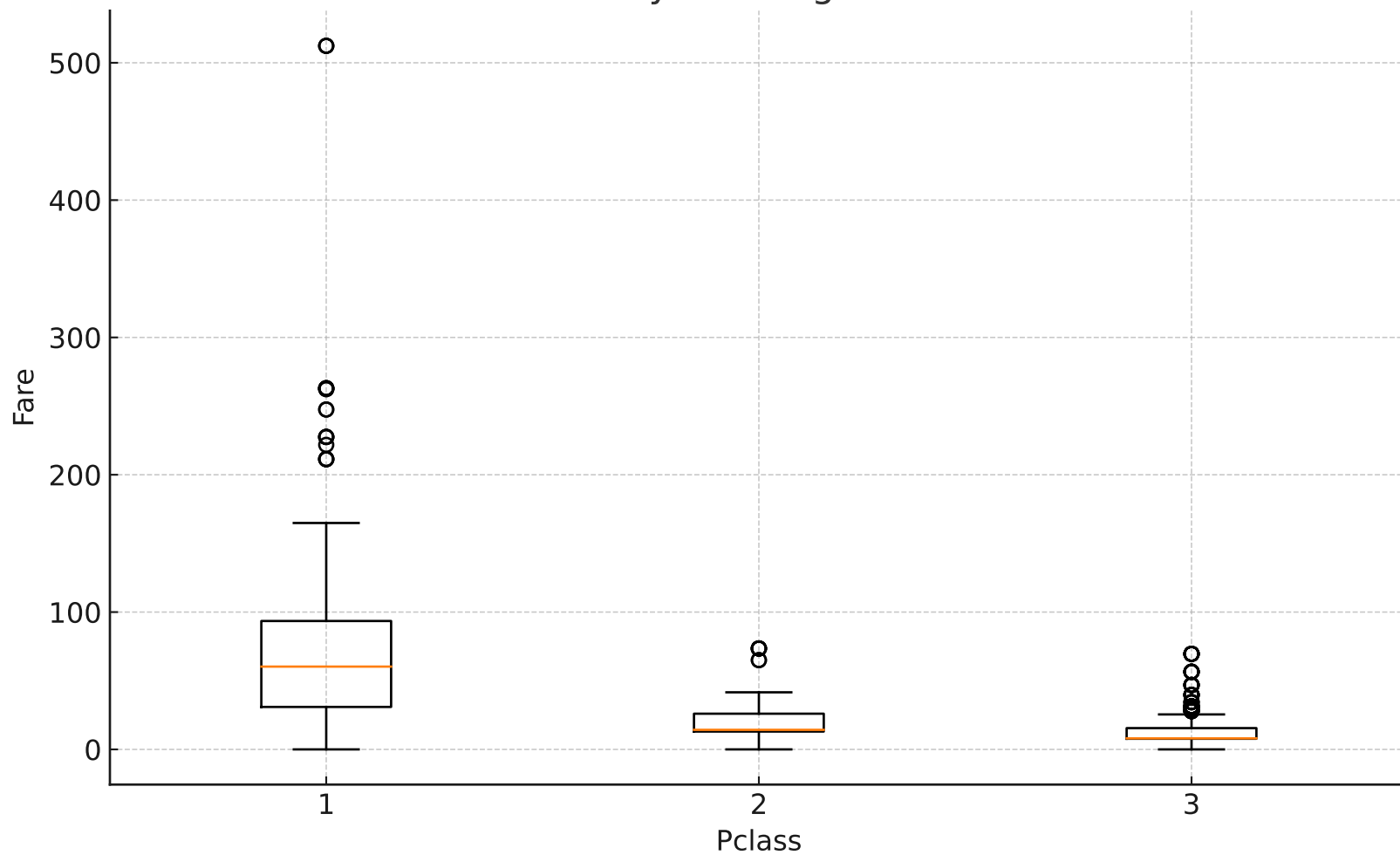
Age Distribution



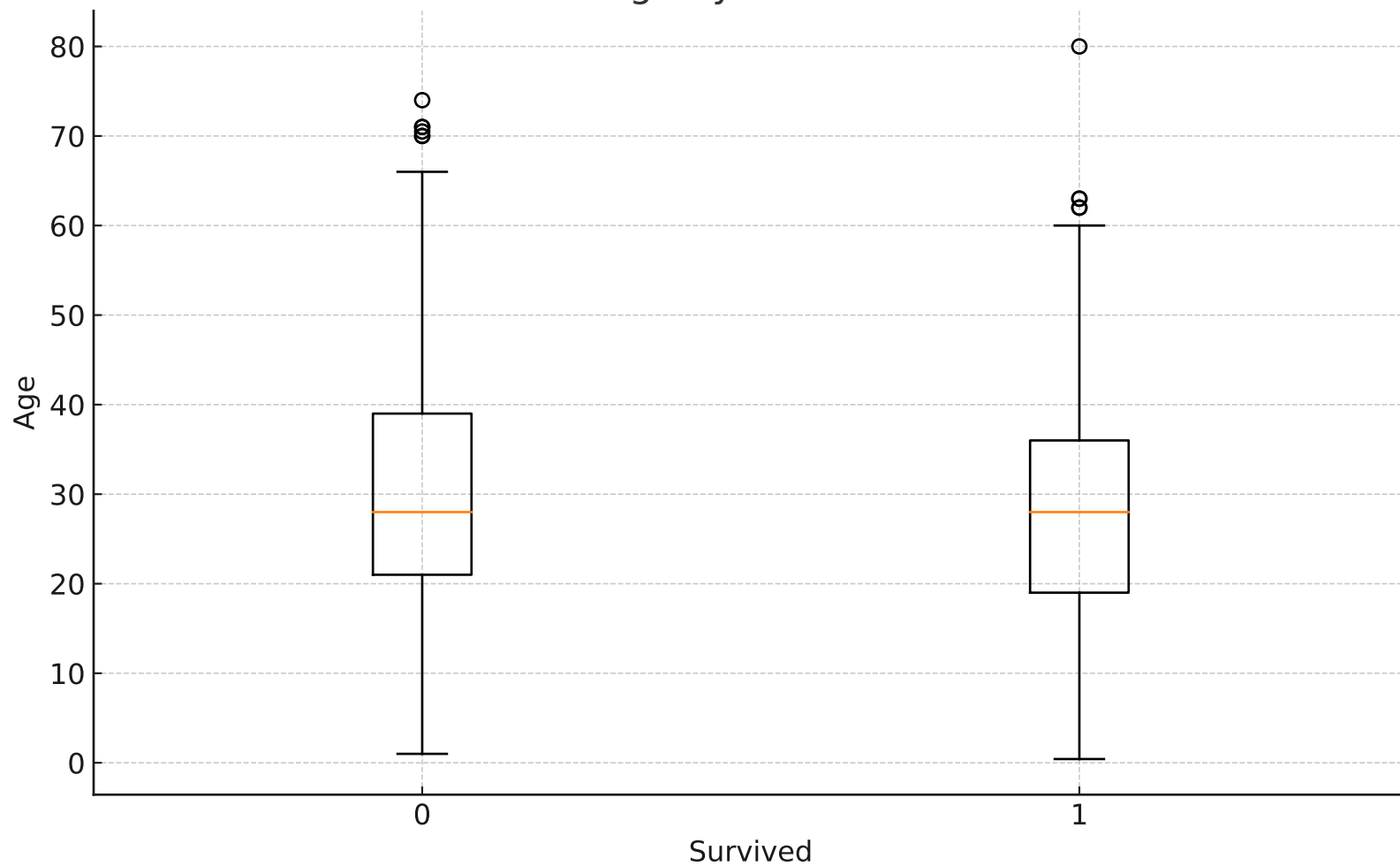
Fare Distribution



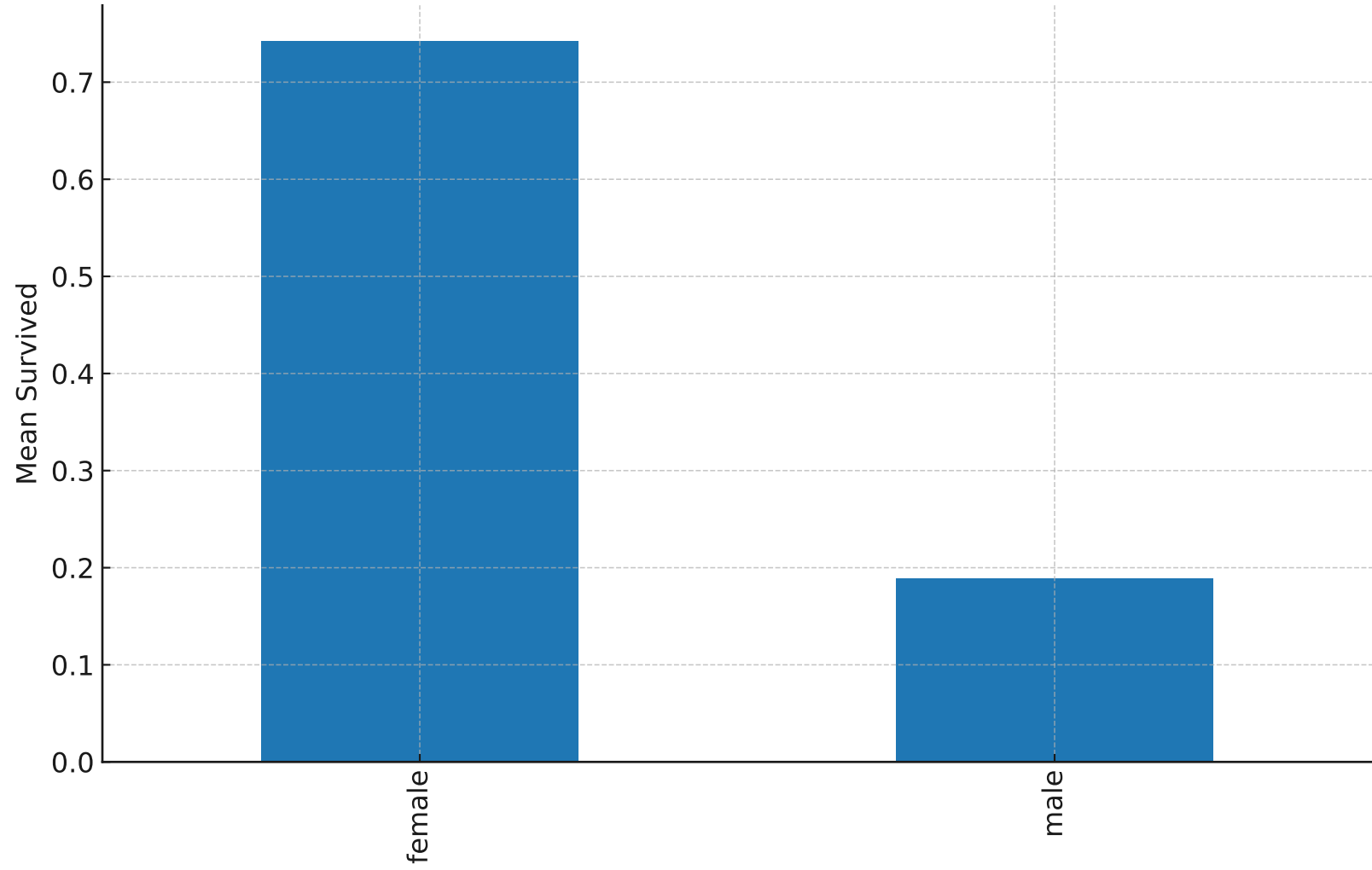
Fare by Passenger Class



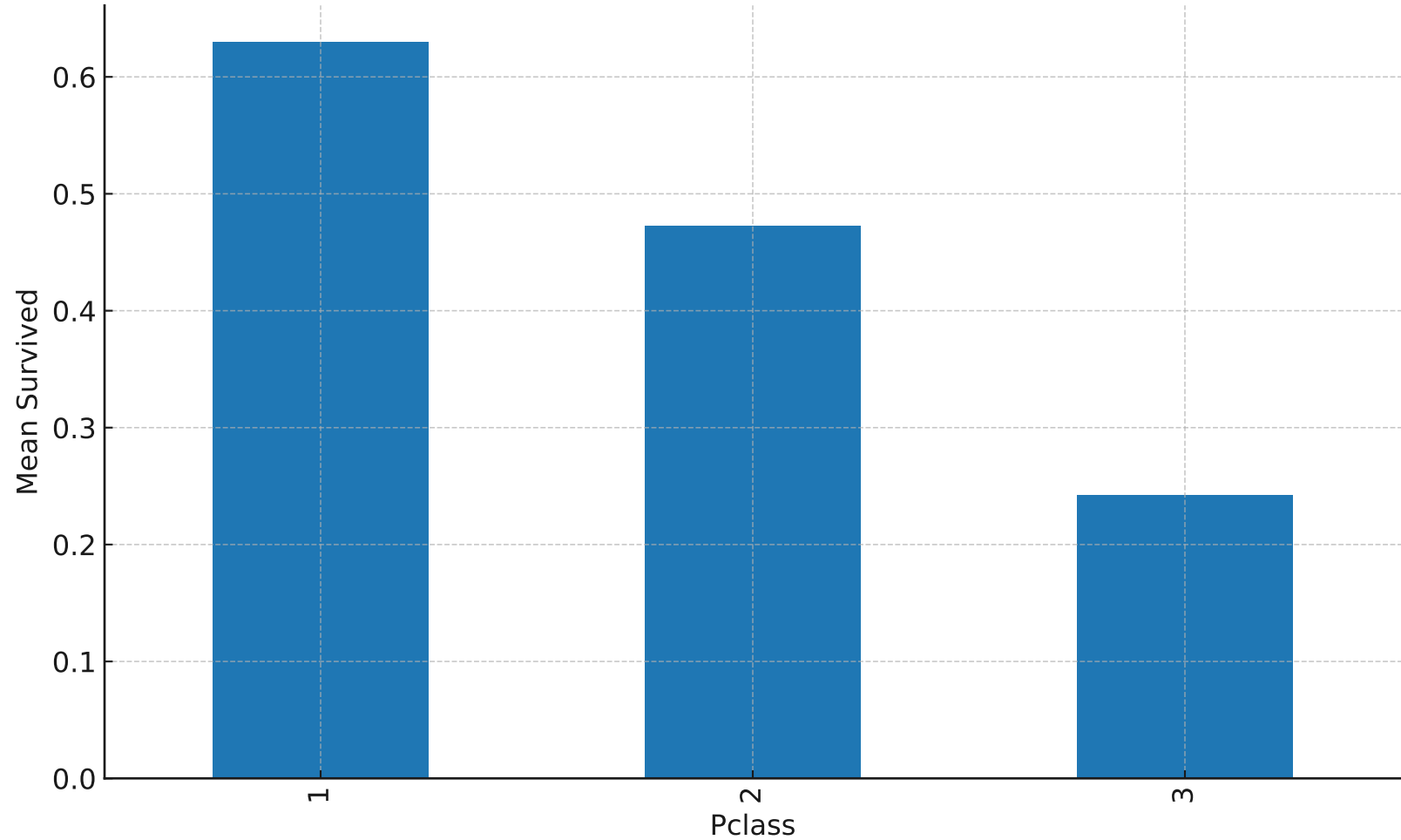
Age by Survival



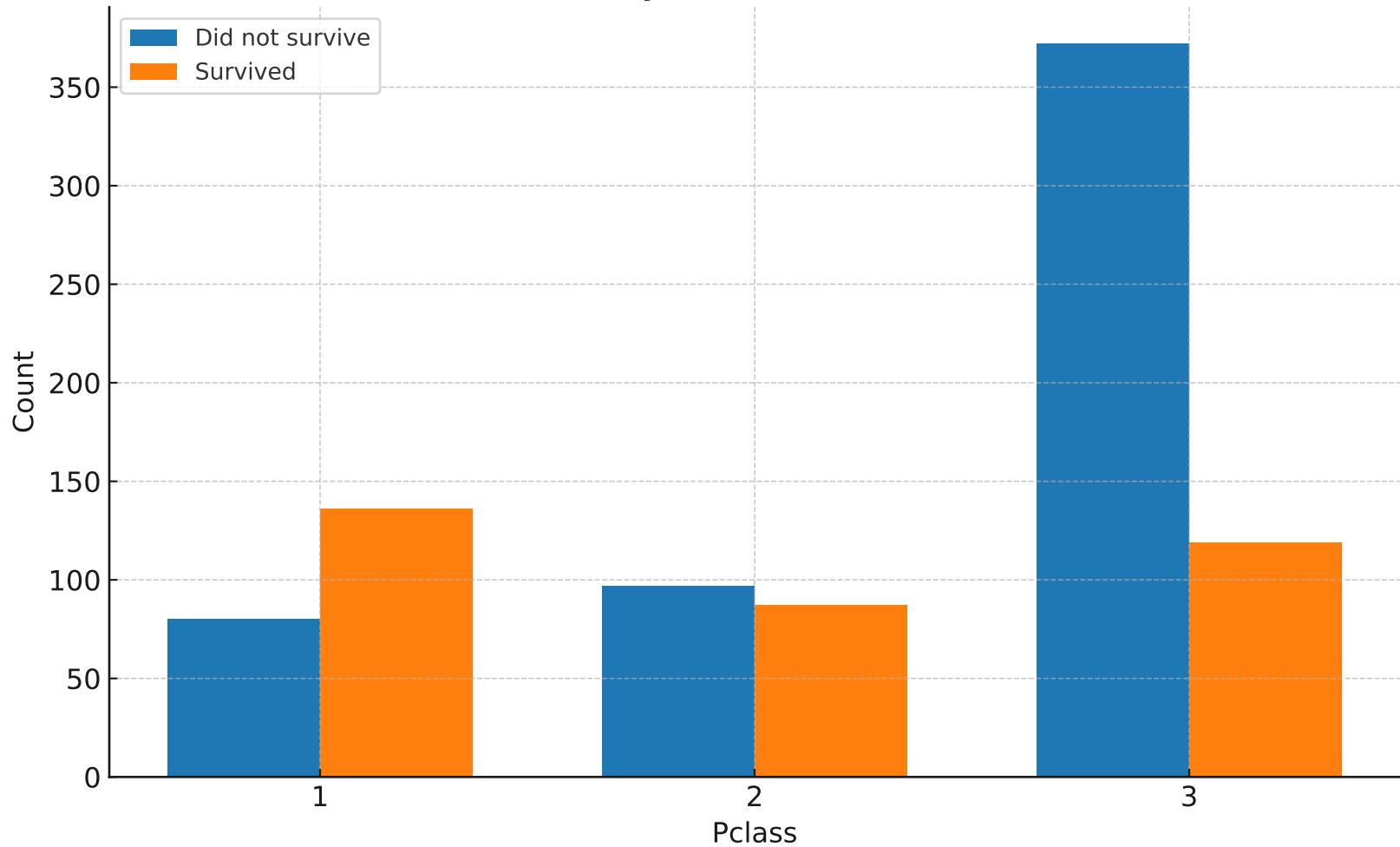
Survival Rate by Sex



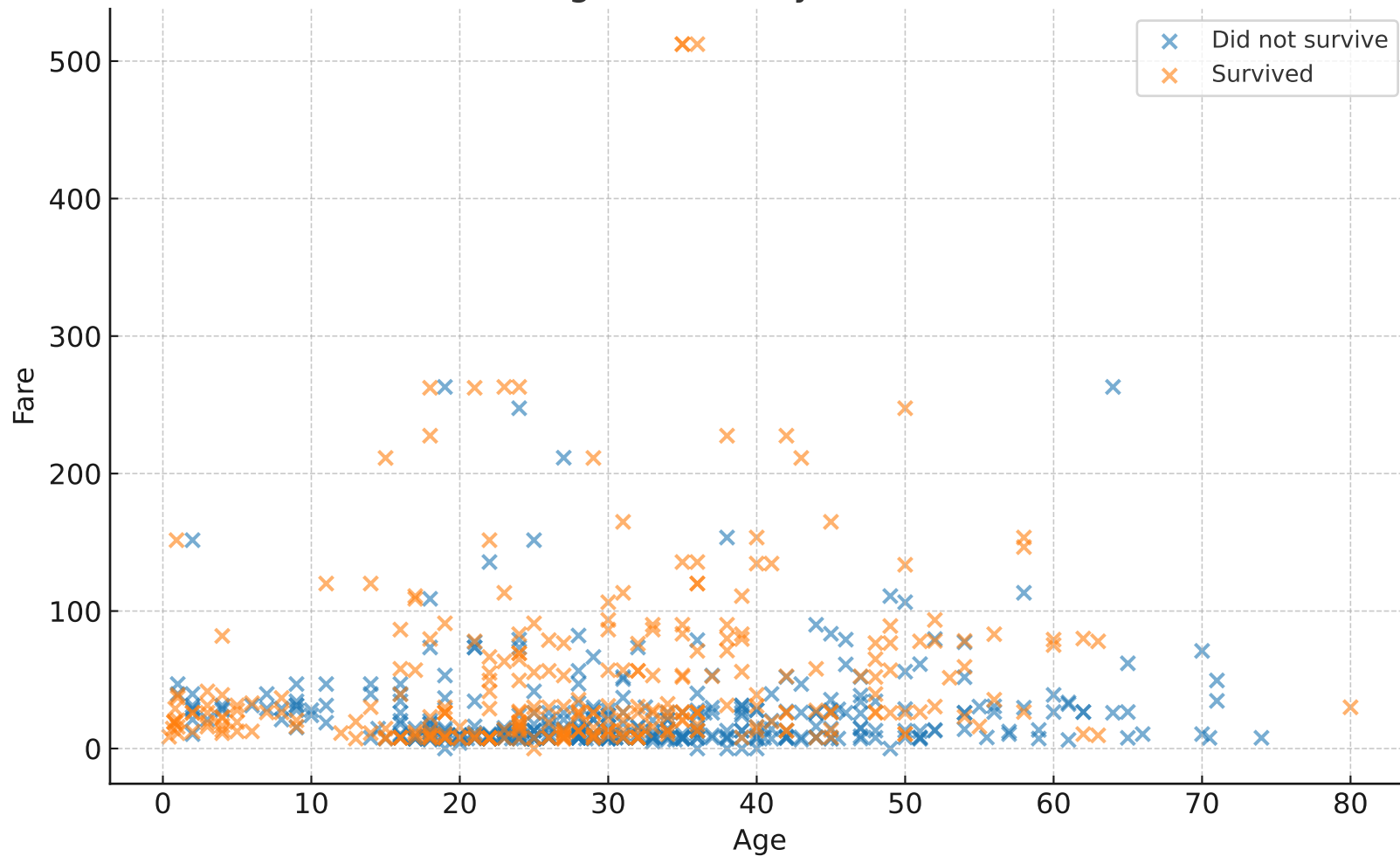
Survival Rate by Pclass



Counts by Pclass and Survival



Age vs Fare by Survival



Correlation Matrix (Numeric)



- Age: skewed toward young adults; missing values exist.
- Fare: right-skewed; wide spread especially in 1st class.
- Fare vs Pclass: strong separation, higher class => higher fare.
- Age vs Survival: children had higher survival than some adults.
- Survival Rate by Sex: females survived at higher rates.
- Survival Rate by Pclass: higher classes show higher survival.
- Counts: 3rd class had many non-survivors.
- Age vs Fare scatter: survivors more common at higher fares.
- Correlations: Survived positively linked to Fare, negatively to Pclass.

Summary of Findings:

- Survival higher among females and higher passenger classes.
- Fare (wealth) is associated with better survival.
- Children had better chances of survival.
- Missing Age/Embarked values need imputation for modeling.
- Next steps: feature engineering (family size, titles, cabin decks), preprocessing for ML models.