

```
In [2]: import pandas as pd
import numpy as np

# Load Titanic dataset
df = pd.read_csv("train.csv")

# Display the first few rows and info
display(df.head())
df.info()
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.250
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.283
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.925
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.100
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.050

```
<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 [10]: # 1. Handling missing values

# Convert to numeric first (results in float64 due to NaNs and decimals)
df['Age'] = pd.to_numeric(df['Age'], errors='coerce')

# To convert to 'Int64', we must first handle the decimals.
# We'll round them to the nearest whole number, then cast.
df['Age'] = df['Age'].round(0).astype('Int64')

# Fill missing 'Embarked' with the most frequent value
```

```
df['Embarked'] = df['Embarked'].fillna(df['Embarked'].mode()[0])
```

```
# Fill missing 'Fare' with the median value  
df['Fare'] = df['Fare'].fillna(df['Fare'].median())
```

```
In [12]: # 2. Convert non-numeric columns to category types to save memory  
df['Pclass'] = df['Pclass'].astype('category')  
df['Survived'] = df['Survived'].astype('category')  
df['Sex'] = df['Sex'].astype('category')  
df['Embarked'] = df['Embarked'].astype('category')
```

```
In [28]: # 3. Check the data types after the conversion and filling missing values  
print("Data types after conversion:")  
df.dtypes
```

Data types after conversion:

```
Out[28]: PassengerId      int64  
Survived      category  
Pclass        category  
Name          object  
Sex           category  
Age           Int64  
SibSp         int64  
Parch         int64  
Ticket        object  
Fare          float64  
Cabin         object  
Embarked      category  
Norm_Fare     float64  
Standardized_Age Float64  
dtype: object
```

```
In [30]: # 4. Apply Aggregation: Example for 'Age' and 'Fare'  
age_stats = df['Age'].agg(['mean', 'std', 'min', 'max'])  
fare_stats = df['Fare'].agg(['mean', 'std', 'min', 'max'])  
  
# Display the summary statistics  
print("\nAge Stats:")  
age_stats  
print("\nFare Stats:")  
fare_stats
```

Age Stats:

Fare Stats:

```
Out[30]: mean      32.204208  
std       49.693429  
min        0.000000  
max      512.329200  
Name: Fare, dtype: float64
```

```
In [32]: # 5. Apply Normalization (Scaling)  
# Normalize 'Fare' column: scale the 'Fare' to [0, 1] using Min-Max normalization  
min_fare = df['Fare'].min()  
max_fare = df['Fare'].max()  
df['Norm_Fare'] = (df['Fare'] - min_fare) / (max_fare - min_fare)
```

```
In [34]: # 6. Apply Standardization (Z-score scaling)  
# Standardize the 'Age' column: mean = 0, std = 1  
age_mean = df['Age'].mean()  
age_std = df['Age'].std()  
df['Standardized_Age'] = (df['Age'] - age_mean) / age_std
```

```
In [36]: # 7. Display first few rows of the transformed dataframe
```

```
print("\nFirst few rows of the transformed dataframe:")
df.head()
```

First few rows of the transformed dataframe:

```
Out[36]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22	1	0	A/5 21171	7.25
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...)	female	38	1	0	PC 17599	71.28
2	3	1	3	Heikkinen, Miss. Laina	female	26	0	0	STON/O2. 3101282	7.92
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35	1	0	113803	53.10
4	5	0	3	Allen, Mr. William Henry	male	35	0	0	373450	8.05

```
In [39]: # 8. Optional: Save the transformed dataframe to a new CSV file
df.to_csv('titanic_transformed.csv', index=False)
```