

```
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	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.250
1		2	1	Cumings, Mrs. John Bradley (Florence Briggs Th... Heikkinen, Miss. Laina	female	38.0	1	0	PC 17599	71.283
2		3	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	26.0	0	0	STON/O2. 3101282	7.925
3		4	1	Allen, Mr. William Henry	male	35.0	1	0	113803	53.100
4		5	0			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	7.95
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)`