

NIKHIL MALVI B 50


EXPERIMENT 1

▼ Titanic Dataset Preprocessing


This code performs preprocessing steps on the Titanic dataset, including:

1. **Data Loading:** Loads the Titanic dataset from a remote URL using pandas.
2. **Exploratory Data Analysis (EDA):**
 - Checks for missing values in each column.
 - Calculates descriptive statistics.
 - Examines data types and the dataset's dimensions.
3. **Data Type Conversion:**
 - Converts relevant columns ('Survived', 'Pclass', 'Sex', 'Embarked') to categorical data types for better analysis.
4. **Data Cleaning and Normalization:**
 - Fills missing values in the 'Age' column with the median age.
 - Applies Z-score normalization to the 'Fare' column, scaling it to have zero mean and unit variance.
5. **Output:**
 - Prints the results of the missing data check, descriptive statistics, data types, and dataset dimensions.
 - Prints the final data types after conversion and cleaning.

```
1 import pandas as pd
2
3 url = 'https://raw.githubusercontent.com/datasciencedojo/datasets/master/titanic.csv'
4 titanic_df = pd.read_csv(url) # Corrected line: removed extra 'lad this' and closed parenthesis
5 titanic_df.head(5)
```



	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
				Futrelle, Mrs. Jacques Heath (Lilv								




Next steps:

[Generate code with titanic_df](#)

[View recommended plots](#)

[New interactive sheet](#)

```
1 #Check for missing values
2 missing_data = titanic_df.isnull().sum()
3 # Get initial statistics for the dataset
4 data_description = titanic_df.describe()
5 # Check the data types and dimensions
6 data_types = titanic_df.dtypes
7 dimensions = titanic_df.shape
8 # Output
9 print("Missing Data:\n", missing_data)
10 print("\nData Description:\n", data_description)
11 print("\nData Types:\n", data_types)
12 print("\nDataset Dimensions:", dimensions)
```



```
Missing Data:
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
```

Data Description:

	PassengerId	Survived	Pclass	Age	SibSp \
count	891.000000	891.000000	891.000000	714.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	0.523008
std	257.353842	0.486592	0.836071	14.526497	1.102743
min	1.000000	0.000000	1.000000	0.420000	0.000000
25%	223.500000	0.000000	2.000000	20.125000	0.000000
50%	446.000000	0.000000	3.000000	28.000000	0.000000
75%	668.500000	1.000000	3.000000	38.000000	1.000000
max	891.000000	1.000000	3.000000	80.000000	8.000000

	Parch	Fare
count	891.000000	891.000000
mean	0.381594	32.204208
std	0.806057	49.693429
min	0.000000	0.000000
25%	0.000000	7.910400
50%	0.000000	14.454200
75%	0.000000	31.000000
max	6.000000	512.329200

Data Types:

```
PassengerId    int64
Survived        int64
Pclass          int64
Name            object
Sex             object
Age            float64
SibSp           int64
Parch           int64
Ticket          object
Fare            float64
Cabin           object
Embarked        object
dtype: object
```

Dataset Dimensions: (891, 12)

```
1 # Check for categorical variables and convert them if needed
2 titanic_df['Survived'] = titanic_df['Survived'].astype('category') # Corrected column name to 'Survived'
3 titanic_df['Pclass'] = titanic_df['Pclass'].astype('category')
4 titanic_df['Sex'] = titanic_df['Sex'].astype('category')
5 titanic_df['Embarked'] = titanic_df['Embarked'].astype('category')
6 # Normalize continuous numerical columns (Age, Fare, etc.)
7 titanic_df['Age'] = titanic_df['Age'].fillna(titanic_df['Age'].median()) # Replace missing values with the median
8 titanic_df['Fare'] = (titanic_df['Fare'] - titanic_df['Fare'].mean()) / titanic_df['Fare'].std() # Z-score normalization
9 # Output the final data types after conversion
10 print("\nFinal Data Types after Conversion:\n", titanic_df.dtypes)
```



Final Data Types after Conversion:

```
PassengerId    int64
Survived        category
Pclass          category
Name            object
Sex             category
Age            float64
SibSp           int64
Parch           int64
Ticket          object
Fare            float64
Cabin           object
Embarked        category
dtype: object
```

```
1 titanic_df_encoded = pd.get_dummies(titanic_df, columns=['Sex',
2 'Embarked'], drop_first=True)
3 # Display the first few rows
4 print(titanic_df_encoded) # Removed the extra indent before this line.
```



```
PassengerId  Survived  Pclass  \
0            1         0        3
1            2         1        1
2            3         1        3
3            4         1        1
4            5         0        3
..          ...      ...      ...
```

```

886      887      0      2
887      888      1      1
888      889      0      3
889      890      1      1
890      891      0      3

```

```

      Name      Age  SibSp  Parch  \
0      Braund, Mr. Owen Harris  22.0      1      0
1  Cumings, Mrs. John Bradley (Florence Briggs Th...  38.0      1      0
2      Heikkinen, Miss. Laina  26.0      0      0
3  Futrelle, Mrs. Jacques Heath (Lily May Peel)  35.0      1      0
4      Allen, Mr. William Henry  35.0      0      0
..      ...      ...      ...      ...
886      Montvila, Rev. Juozas  27.0      0      0
887      Graham, Miss. Margaret Edith  19.0      0      0
888  Johnston, Miss. Catherine Helen "Carrie"  28.0      1      2
889      Behr, Mr. Karl Howell  26.0      0      0
890      Dooley, Mr. Patrick  32.0      0      0

```

```

      Ticket      Fare  Cabin  Sex_male  Embarked_Q  Embarked_S
0      A/5 21171 -0.502163   NaN      True      False      True
1      PC 17599  0.786404   C85      False      False      False
2  STON/O2. 3101282 -0.488580   NaN      False      False      True
3      113803  0.420494   C123      False      False      True
4      373450 -0.486064   NaN      True      False      True
..      ...      ...      ...      ...      ...
886      211536 -0.386454   NaN      True      False      True
887      112053 -0.044356   B42      False      False      True
888      W./C. 6607 -0.176164   NaN      False      False      True
889      111369 -0.044356   C148      True      False      False
890      370376 -0.492101   NaN      True      True      False

```

[891 rows x 13 columns]