```
1 import numpy as np
2 import pandas as pd
3 import seaborn as sns
4 import matplotlib.pyplot as plt
5
6 df = pd.read_csv('/content/Titanic-Dataset.csv')
7
```

exploring the dataset

1 df.head()

→		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	ıl.
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С	
	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 (Lilv Mav Peel)	female	35.0	1	0	113803	53.1000	C123	S	

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```
1 df.info()
2
```

<<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890

Data columns (total 12 columns): Non-Null Count Dtype PassengerId 891 non-null Survived 891 non-null int64 1 Pclass 891 non-null int64 3 891 non-null Name object 4 891 non-null Sex object 5 Age 714 non-null float64 6 SibSp 891 non-null int64 Parch 891 non-null int64 8 Ticket 891 non-null object 891 non-null float64 Fare 10 Cabin 204 non-null object 11 Embarked 889 non-null object dtypes: float64(2), int64(5), object(5) memory usage: 83.7+ KB

1 df.describe()

2 Passe

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare	
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000	11
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	

1 df.isnull().sum()

```
₹
                      0
      Passengerld
                      0
                      0
       Survived
         Pclass
         Name
                      0
          Sex
                      0
          Age
                    177
         SibSp
                      0
         Parch
         Ticket
                      0
         Fare
                      0
         Cabin
                    687
       Embarked
                      2
        na. intel
 1 df.nunique()
₹
                      a
      Passengerld 891
       Survived
                      2
        Pclass
                      3
         Name
                    891
          Sex
                      2
                     88
          Age
         SibSp
                      7
                      7
         Parch
         Ticket
                    681
         Fare
                    248
       Embarked
                      3
 1 df.duplicated().sum()
\rightarrow np.int64(0)
 1 df['Embarked'].value_counts()
\overline{\Rightarrow}
                 count
      Embarked
          s
                   644
         С
                   168
         Q
                    77
handling the missing data
```

```
1 #filling the age colum with the median value why median bacause the data may contain outlier values
2 df['Age'].fillna(df['Age'].median(),inplace=True)
3
4 #filling embarked with mode (mode can be used for categotical columns )
5 #(why[0]- mode return a frequent values to get the first frequent value we use this [0] )
6
7 df['Embarked'].fillna(df['Embarked'].mode()[0],inplace=True)
8
9 #drop cabin (too many null values)
```

Encoding the categorical variables

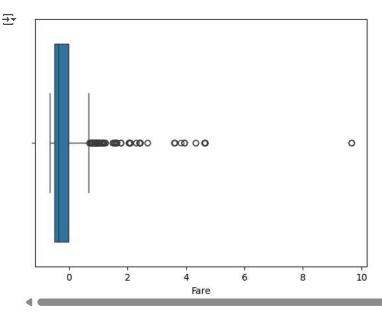
```
1 #label encoding sex column assigning o for male and 1 for female
2 df['Sex'] = df['Sex'].map({'male':0,'female':1})
3
4 #using one hot encoding because it is a location
5
6 df = pd.get_dummies(df,columns=['Embarked'],drop_first=True)
7
```

normalizing

```
1 #normalizing the age and fare column between 0 to 1 to improve the model performance
2 #using standardscaler we can use other preprocessors like minmaxscaler but it is more suitable in timeseries data,robustscaler when
3
4 from sklearn.preprocessing import StandardScaler
5 scaler = StandardScaler()
6 df[['Age', 'Fare']] = scaler.fit_transform(df[['Age', 'Fare']])
7
```

detect,remove outlier values

```
1 sns.boxplot(x=df['Fare'])
2 plt.show()
```



1 # there are several methods to remove outlier some of the methods like IQR z-score and isolation forest can be used for this data
2 #i am using the z-score method where - a value is more than 3 standard deviations from the mean, it's an outlier.
3
4 from scipy import stats
5 z = np.abs(stats.zscore(df['Fare']))

6 df = df[(z<3)]

1 df.head()

₹		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked_Q	Embarked_S	
	0	1	0	3	Braund, Mr. Owen Harris	NaN	-0.565736	1	0	A/5 21171	-0.502445	False	True	ıl.
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	NaN	0.663861	1	0	PC 17599	0.786845	False	False	
<					Hailkinan Miss					STON/O3				Þ
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covert the cleaned data to a csv file

1 #convert the data into csv usin the to_csv function in pandas $\,$

2

 ${\tt 3~df.to_csv("titanic_cleaned_data_by_sivashankar.csv",~index=False)}\\$