```
In [19]: import numpy as np
    import pandas as pd
    from sklearn.model_selection import train_test_split
    from sklearn.impute import SimpleImputer
    from sklearn.preprocessing import OneHotEncoder
    from sklearn.preprocessing import MinMaxScaler
    from sklearn.compose import ColumnTransformer

from sklearn.pipeline import Pipeline, make_pipeline
    from sklearn.feature_selection import SelectKBest, chi2
    from sklearn.tree import DecisionTreeClassifier
```

## Out[4]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
439	0	2	male	31.0	0	0	10.5000	S
687	0	3	male	19.0	0	0	10.1708	S
472	1	2	female	33.0	1	2	27.7500	S
67	0	3	male	19.0	0	0	8.1583	S

In [21]: data['Embarked'].nunique()

Out[21]: 3

In [5]: # doing train test and split of the data
X\_train,X\_test,y\_train,y\_test=train\_test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],test\_split(data.drop('Survived',axis=1),data['Survived'],data['Surviv

In [6]: X\_train.head()

## Out[6]:

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
331	1	male	45.5	0	0	28.5000	S
733	2	male	23.0	0	0	13.0000	S
382	3	male	32.0	0	0	7.9250	S
704	3	male	26.0	1	0	7.8542	S
813	3	female	6.0	4	2	31.2750	S

In [8]: y\_train.sample(5)

Out[8]: 598

195 1 428 0 376 1

403 0

0

Name: Survived, dtype: int64

```
In [13]: # filling the missing values through column transformer
          Transformer1=ColumnTransformer(transformers=[
              ('trans1_age',SimpleImputer(),[2]),
              ('trans2_embarked',SimpleImputer(strategy='most_frequent'), [6]) # idex or name le call gar
          ],remainder='passthrough')
In [14]: data.isna().sum()
Out[14]: Survived
          Pclass
                        0
                        0
          Sex
                      177
          Age
          SibSp
                        a
                        0
          Parch
          Fare
                        0
          Embarked
          dtype: int64
          One hot encoding
In [18]: Transformer2=ColumnTransformer(transformers=[
              ('onehotenconding', OneHotEncoder(sparse output=False,handle unknown='ignore'),[1,6])
          ], remainder='passthrough')
          Scaling
In [27]: Transformer3=ColumnTransformer(transformers=[('scale',MinMaxScaler(),slice(0,10))]) # 0 dekhi 8
          why did i write slice(0,10)? because out of 7 columns in X train, from column transformer 2 will be droped.
          remained 5 columns. Now from one hot encoding of sex and embarked 2 from sex and 3 from embarked will be
          formed adding total 5 columns to initial 5 columns which gives total of 10 columns so using 0, 10 in sciling
          Feature selection
In [28]: Transformer4=SelectKBest(score_func=chi2, k=8)
          ####Train the model
In [32]: Transformer5=DecisionTreeClassifier()
          Create a pipeline object
In [30]: pipe=Pipeline([('TRF1',Transformer1),
                         ('TRF2', Transformer2),
```

('TRF3', Transformer3), ('TRF4', Transformer4)])

## **Exploring pipe function**

```
In [ ]:
```