Praticle 1: Data Preprocessing

```
In [ ]: import pandas as pd
         from sklearn.impute import SimpleImputer
         from sklearn.preprocessing import LabelEncoder, StandardScaler
In [17]: # Load the dataset
         data = pd.read csv('data house.csv')
In [19]: # Handling missing data
         imputer = SimpleImputer(missing values=np.nan, strategy='mean')
         imputer.fit(data[['bedrooms', 'bathrooms']])
         data[['bedrooms', 'bathrooms']] = imputer.transform(data[['bedrooms', 'ba
In [20]: # Encoding the independent variable
         floors_encoder = LabelEncoder()
         data['floors'] = floors encoder.fit transform(data['floors'])
         # Encoding the dependent variable
         price_encoder = LabelEncoder()
         data['price'] = price encoder.fit transform(data['price'])
In [22]: # Feature Scaling
         scaler = StandardScaler()
         num_cols = ['bedrooms', 'bathrooms', 'floors']
         data[num cols] = scaler.fit transform(data[num cols])
         data['price'] = scaler.fit transform(data[['price']])
In [23]: # Data Splitting
        X = data.drop('price', axis=1)
         y = data['price']
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
In [26]: print(X)
              bedrooms bathrooms floors
       0
             -0.398737 -1.447464 -0.915427
            -0.398737 0.175607 0.936506
       1
       2
            -1.473959 -1.447464 -0.915427
             0.676485 1.149449 -0.915427
       3
            -0.398737 -0.149007 -0.915427
                        . . .
       21609 0.676485 0.500221 0.936506
       21610 -1.473959 -1.772078 0.936506
       21611 -0.398737  0.500221  0.936506
       21612 -1.473959 -1.772078 0.936506
       [21613 rows x 3 columns]
In [27]: print(y)
```

```
1
              0.321547
      2
             -1.579385
      3
              0.602062
             0.198201
                . . .
      21608 -0.592611
      21609 -0.366807
      21610 -0.357854
      21611
            -0.366807
      21612 -0.805484
      Name: price, Length: 21613, dtype: float64
In [28]: print(X train)
            bedrooms bathrooms
                                floors
       6325 -0.398737 -0.473621 -0.915427
       13473 -1.473959 -1.447464 -0.915427
       17614 -0.398737 -1.447464 0.010539
       20868 -1.473959 0.500221 0.936506
                      . . .
       11964 -0.398737 -0.798235 -0.915427
       21575 -0.398737  0.500221  0.936506
       5390 -0.398737 0.500221 0.936506
      860 -2.549182 -1.772078 -0.915427
       15795 0.676485 0.500221 0.936506
       [17290 rows x 3 columns]
In [29]: print(X_test)
             bedrooms bathrooms floors
      735
            0.676485 0.175607 0.936506
            1.751707 1.149449 -0.915427
       2830
      4106 0.676485 0.500221 0.936506
       16218 -0.398737 1.798677 0.936506
       . . .
                         . . .
       . . .
                                  . . . .
      13674 -0.398737 -0.473621 -0.915427
       0.676485 0.500221 0.936506
      8805
       10168 0.676485 -0.473621 -0.915427
      2522 0.676485 0.500221 0.936506
       [4323 rows x 3 columns]
In [30]: print(y train)
            -0.805484
      6325
       13473 -1.228245
            -1.377455
      17614
       16970
            -1.035267
      20868
            0.054959
            -0.502091
      11964
      21575
            -0.370786
      5390
             0.484683
      860
             -1.295887
      15795 -0.858205
      Name: price, Length: 17290, dtype: float64
```

0

-1.411276

```
In [31]: print(y_test)
       735
               -0.565753
       2830
                1.402821
                1.686320
       4106
       16218
                1.970813
       19964
                0.973097
                  . . .
       13674 -0.720932
       20377 0.859697
       8805
               -1.058146
       10168
               0.605046
               -0.614495
       2522
       Name: price, Length: 4323, dtype: float64
```

Steps Followed:

- Handling missing data by imputing the missing values in the 'bedrooms' and 'bathrooms' columns with their respective means.
- Encoding the categorical variable 'floors' using LabelEncoder to assign unique numeric labels to each category.
- Encoding the dependent variable 'price' using LabelEncoder to convert it into numeric labels.
- Performing feature scaling on the numerical columns 'bedrooms', 'bathrooms',
 'floors', and the dependent variable 'price' using StandardScaler to standardize the
 values.
- Splitting the preprocessed data into training and test sets using train_test_split for further analysis or model training.

In []: