```
import pandas as pd
from sklearn.model selection import train test split
from sklearn.preprocessing import MinMaxScaler
# a) Read the data with pandas and describe the data
data = pd.read csv("/content/housing.csv")
data = data.drop('ocean_proximity',axis=1)
print("Description of the data:")
print(data.describe())
# b) Find data type and shape of each column
print("\nData types of each column:")
print(data.dtypes)
print("\nShape of the data:")
print(data.shape)
# c) Find the null values and fill them with mean of that column
null values = data.isnull().sum()
print("\nNull values in the data:")
print(null values)
# Fill null values with mean
data.fillna(data.mean(), inplace=True)
# d) Find features and target variables
features = data.drop(columns=['median house value'])
target = data['median house value']
print(features)
print(target)
# e) Split the data into train and test
X train, X test, y train, y test = train test split(features, target,
test size=0.2, random state=42)
# f) Normalize the data with min-max scaling
scaler = MinMaxScaler()
X train scaled = scaler.fit transform(X train)
X test scaled = scaler.transform(X test)
print("\nThe Normalized Data:\n")
print("Train Scaled\n")
print(X train scaled)
print("\nTest Scaled\n")
print(X test scaled)
Description of the data:
          longitude
                         latitude housing median age
                                                        total rooms \
count 20640.000000 20640.000000
                                         20640.000000 20640.000000
mean
        -119.569704
                        35.631861
                                            28.639486
                                                        2635.763081
                                            12.585558
std
           2.003532
                         2.135952
                                                        2181.615252
min
        -124.350000
                        32.540000
                                             1.000000
                                                           2.000000
```

25% -121.800000 33.930000 18.000000 1447.750000 50% -118.490000 34.260000 29.000000 2127.000000 77.000000 37.000000 3148.000000 3148.000000 3148.000000 3148.000000 3148.000000 3148.000000 000000									
count 20433.000000 20640.000000 20640.000000 20640.000000 3.870671 std 421.385670 1132.462122 382.329753 1.899822 min 1.000000 3.000000 1.000000 0.499900 25% 296.000000 787.000000 280.000000 2.563400 50% 435.000000 1725.000000 409.000000 3.534800 75% 647.000000 35682.000000 605.000000 4.743250 6445.000000 35682.000000 6082.000000 15.000100 median_house_value custofilias custofilias custofilias 6082.000000 15.000100 median_house_value custofilias custofilias	50% 75%	-118.490000 -118.010000	34.260000 37.710000	29.000 37.000	000 2127.000000 000 3148.000000				
count	mean std min 25% 50% 75%	$204\overline{3}3.000000$ 537.870553 421.385070 1.000000 296.000000 435.000000 647.000000	20640.000000 1425.476744 1132.462122 3.000000 787.000000 1166.000000 1725.000000	20640.000000 499.539680 382.329753 1.000000 280.000000 409.000000 605.000000	20640.000000 3.870671 1.899822 0.499900 2.563400 3.534800 4.743250				
longitude float64 latitude float64 housing_median_age float64 total_rooms float64 total_bedrooms float64 population float64 median_income float64 median_house_value float64 dtype: object Shape of the data: (20640, 9) Null values in the data: longitude 0 latitude 0 housing_median_age 0 total_rooms 0 total_bedrooms 207 population 0 households 0 median_income 0	count 20640.000000 mean 206855.816909 std 115395.615874 min 14999.000000 25% 119600.000000 50% 179700.000000 75% 264725.000000								
Null values in the data: longitude 0 latitude 0 housing_median_age 0 total_rooms 0 total_bedrooms 207 population 0 households 0 median_income 0	longitude float64 latitude float64 housing_median_age float64 total_rooms float64 total_bedrooms float64 population float64 households float64 median_income float64 median_house_value float64								
<pre>longitude</pre>									
	longit latitu housin total_ total_ popula househ median	ude de g_median_age rooms bedrooms tion olds _income	0 0 0 0 207 0 0						

dtype:	int64 longitude	latitudo h	ousing median age	total rooms
total	bedrooms \	tatitude II	ousing_median_age	totat_rooms
0 _	-122.23	37.88	41.0	880.0
129.0 1 1106.0	-122.22	37.86	21.0	7099.0
2 190.0	-122.24	37.85	52.0	1467.0
3	-122.25	37.85	52.0	1274.0
235.0	-122.25	37.85	52.0	1627.0
280.0				
20635	-121.09	39.48	25.0	1665.0
374.0 20636	-121.21	39.49	18.0	697.0
150.0 20637	-121.22	39.43	17.0	2254.0
485.0 20638	-121.32	39.43	18.0	1860.0
409.0 20639 616.0	-121.24	39.37	16.0	2785.0
010.0				
0 1 2 3 4	population 322.0 2401.0 496.0 558.0 565.0	households 126.0 1138.0 177.0 219.0 259.0	-8.3252 8.3014 7.2574 5.6431	
20635 20636 20637 20638 20639	845.0 356.0 1007.0 741.0 1387.0	330.0 114.0 433.0 349.0 530.0	2.5568 1.7000 1.8672	
[20640 0 1 2 3	rows x 8 co 452600.0 358500.0 352100.0 341300.0 342200.0		2.5000	
20635 20636 20637	78100.0 77100.0 92300.0			

```
20638
          84700.0
20639
          89400.0
Name: median house value, Length: 20640, dtype: float64
The Normalized Data:
Train Scaled
[[0.72908367 0.01702128 0.62745098 ... 0.06437961 0.10228581
0.190321511
 [0.61653386 \ 0.12978723 \ 0.94117647 \ \dots \ 0.0367443 \ 0.12415721
0.228452021
 [0.38545817 0.22446809 0.05882353 ... 0.02556125 0.05508962
0.25216204]
 [0.59462151 0.15744681 0.68627451 ... 0.04913254 0.08649893
0.167894241
 [0.23804781 \ 0.53510638 \ 0.2745098 \ \dots \ 0.04972112 \ 0.09176122
0.359946761
 [0.19223108 0.55531915 1. ... 0.07332044 0.20407828
0.1431428511
Test Scaled
[[0.53187251 0.37340426 0.47058824 ... 0.03893046 0.0588719
0.081467841
 [0.48705179 \ 0.27553191 \ 0.56862745 \ \dots \ 0.04377925 \ 0.09587239
0.140094621
 [0.19023904 0.55851064 1. ... 0.03663219 0.15819766 0.2055282
 [0.22908367 0.50638298 0.47058824 ... 0.04433981 0.09324124
0.60205376]
 [0.45717131 0.44893617 0.68627451 ... 0.03430589 0.07778326
0.157590931
 [0.59561753 \ 0.17765957 \ 0.31372549 \ \dots \ 0.04675019 \ 0.07350765
0.21049365]]
```