PROJECT: Predicting House Prices Using Machine Learning

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Phase-03: Loading and Pre-Processing the Dataset

ABOUT THIS PHASE:

In this phase we need to do loading and pre-processing the datasets. Here I explain about what are the process to do this phase.

Step 1:

Import the dependencies

In this step we import the library files which are required to run this program code, like modules (numpy , pandas, sklearn , matpltlib, seaborn, XGBoost).

Step 2:

Impoting the dataset

In this step I import California_housing dataset form the sklearn module it is used to fetch the data and the data is used as the input of this project.

Step 3:

Loading the dataset into the pandas dataframe

In this step I load my data to the pandas data frame which gives the structure of our dataset

Step 4:

Checking the number of rows and column to the dataset

In this step I check the number of rows and column to the dataset and also check any missing values in my dataset

Step 5:

Statical measure of dataset

In this step to check some stats related to my dataset like (count, mean, minimum, maximum, standard deviation).

Step 6:

Generate heat map

In this step to generate heat map to know about my dataset clearly by using the module matplotlib.pyplot

Import the dependencies

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import sklearn.datasets
from sklearn.model_selection import train_test_split
from xgboost import XGBRegressor
from sklearn import metrics
```

Impoeting the california house prise dataset

```
from sklearn.datasets import fetch_california_housing
house_price_dataset = fetch_california_housing()
```

print(house_price_dataset)

```
→ {'data': array([[ 8.3252
                                   41.
                                                   6.98412698, ..., 2.55555556,
                       , -122.23
             37.88
                                    ],
          [ 8.3014
                                                           2.10984183,
                                          6.23813708, ...,
                       , -122.22
             37.86
                                     ],
                      , 52.
, -122.24
             7.2574
                                                           2.80225989,
                                          8.28813559, ...,
                                     ,
],
             37.85
             1.7
                       , 17.
                                          5.20554273, ...,
                                                           2.3256351 ,
                       , -121.22
             39.43
                       , 18.
             1.8672
                                          5.32951289, ...,
                                                             2.12320917,
                                     ,
],
             39.43
                       , -121.32
             2.3886
                          16.
                                          5.25471698, ...,
                                                             2.61698113,
             39.37
                       , -121.24
                                      ]]), 'target': array([4.526, 3.585, 3.521, ..., 0.923, 0.847, 0.894]), 'frame': None, 'target_na
   4
```

loading the dataset to the Pands DataFrame

 $house_price_dataframe = pd.DataFrame(house_price_dataset.data, columns = house_price_dataset.feature_names)$

print first 5 rows of our DataFrame house_price_dataframe.head()

| | MedInc | HouseAge | AveRooms | AveBedrms | Population | Ave0ccup | Latitude | Longitude |
|---|--------|----------|----------|-----------|------------|----------|----------|-----------|
| 0 | 8.3252 | 41.0 | 6.984127 | 1.023810 | 322.0 | 2.555556 | 37.88 | -122.23 |
| 1 | 8.3014 | 21.0 | 6.238137 | 0.971880 | 2401.0 | 2.109842 | 37.86 | -122.22 |
| 2 | 7.2574 | 52.0 | 8.288136 | 1.073446 | 496.0 | 2.802260 | 37.85 | -122.24 |
| 3 | 5.6431 | 52.0 | 5.817352 | 1.073059 | 558.0 | 2.547945 | 37.85 | -122.25 |
| 4 | 3.8462 | 52.0 | 6.281853 | 1.081081 | 565.0 | 2.181467 | 37.85 | -122.25 |

add the target column to the DataFrame

house_price_dataframe['price'] = house_price_dataset.target

house_price_dataframe.head()

| | MedInc | HouseAge | AveRooms | AveBedrms | Population | Ave0ccup | Latitude | Longitude | price |
|---|--------|----------|----------|-----------|------------|----------|----------|-----------|-------|
| 0 | 8.3252 | 41.0 | 6.984127 | 1.023810 | 322.0 | 2.555556 | 37.88 | -122.23 | 4.526 |
| 1 | 8.3014 | 21.0 | 6.238137 | 0.971880 | 2401.0 | 2.109842 | 37.86 | -122.22 | 3.585 |
| 2 | 7.2574 | 52.0 | 8.288136 | 1.073446 | 496.0 | 2.802260 | 37.85 | -122.24 | 3.521 |
| 3 | 5.6431 | 52.0 | 5.817352 | 1.073059 | 558.0 | 2.547945 | 37.85 | -122.25 | 3.413 |
| 4 | 3.8462 | 52.0 | 6.281853 | 1.081081 | 565.0 | 2.181467 | 37.85 | -122.25 | 3.422 |

checking the number of rows and columns in the data frame house_price_dataframe.shape

(20640, 9)

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#check for missing values
house_price_dataframe.isnull().sum()

MedInc 0 HouseAge 0 ${\sf AveRooms}$ 0 AveBedrms 0 Population 0 Ave0ccup Latitude 0 Longitude price dtype: int64

statical measure of the dataset
house_price_dataframe.describe()

| | MedInc | HouseAge | AveRooms | AveBedrms | Population | Ave0ccup | Latitude | Longitude | price |
|-------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|
| count | 20640.000000 | 20640.000000 | 20640.000000 | 20640.000000 | 20640.000000 | 20640.000000 | 20640.000000 | 20640.000000 | 20640.000000 |
| mean | 3.870671 | 28.639486 | 5.429000 | 1.096675 | 1425.476744 | 3.070655 | 35.631861 | -119.569704 | 2.068558 |
| std | 1.899822 | 12.585558 | 2.474173 | 0.473911 | 1132.462122 | 10.386050 | 2.135952 | 2.003532 | 1.153956 |
| min | 0.499900 | 1.000000 | 0.846154 | 0.333333 | 3.000000 | 0.692308 | 32.540000 | -124.350000 | 0.149990 |
| 25% | 2.563400 | 18.000000 | 4.440716 | 1.006079 | 787.000000 | 2.429741 | 33.930000 | -121.800000 | 1.196000 |
| 50% | 3.534800 | 29.000000 | 5.229129 | 1.048780 | 1166.000000 | 2.818116 | 34.260000 | -118.490000 | 1.797000 |
| 75% | 4.743250 | 37.000000 | 6.052381 | 1.099526 | 1725.000000 | 3.282261 | 37.710000 | -118.010000 | 2.647250 |
| max | 15.000100 | 52.000000 | 141.909091 | 34.066667 | 35682.000000 | 1243.333333 | 41.950000 | -114.310000 | 5.000010 |

underatanding various feature in the dataset

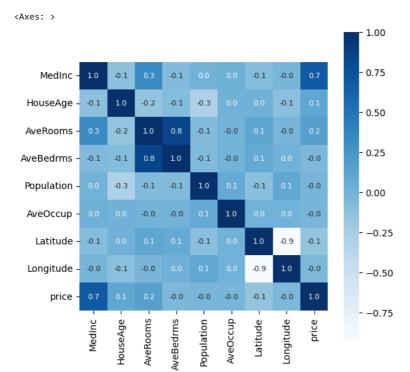
1.positive correlation 2.negative correlation

correlation = house_price_dataframe.corr()

constructing the heatmap

constructing the heatmap to understand the correlation plt.figure(figsize=(6,6))

 $sns.heatmap (correlation, cbar=True, square=True, fmt='.1f', annot=True, annot_kws=\{'size':8\}, cmap='Blues'\}$



splitting data and target

```
X = house_price_dataframe.drop(['price'], axis=1)
Y = house_price_dataframe['price']
print(X)
print(Y)
           MedInc HouseAge AveRooms AveBedrms Population AveOccup Latitude \
                                       1.023810
     0
           8.3252
                       41.0 6.984127
                                                      322.0 2.555556
           8.3014
                       21.0 6.238137
                                        0.971880
                                                      2401.0 2.109842
     1
                                      1.073446
1.073059
            7.2574
                       52.0 8.288136
                                                      496.0 2.802260
                                                                          37.85
     3
           5.6431
                       52.0 5.817352
                                                       558.0 2.547945
                                                                           37.85
                       52.0 6.281853 1.081081
     4
           3.8462
                                                      565.0 2.181467
                                                                          37.85
                       25.0 5.045455
                                                                          39.48
     20635 1.5603
                                       1.133333
                                                      845.0 2.560606
     20636 2.5568
                       18.0 6.114035
                                       1.315789
                                                      356.0 3.122807
                                                                          39.49
                                                      1007.0 2.325635
    20637 1.7000
                       17.0 5.205543
                                       1.120092
                                                                          39.43
     20638 1.8672
                       18.0 5.329513
                                       1.171920
                                                      741.0 2.123209
                                                                           39.43
    20639 2.3886
                       16.0 5.254717
                                       1.162264
                                                      1387.0 2.616981
                                                                          39.37
           Longitude
    0
              -122.23
     1
              -122.22
     2
              -122.24
     3
              -122.25
     4
              -122.25
              -121.09
     20635
     20636
              -121.21
     20637
              -121.22
     20638
              -121.32
     20639
              -121.24
     [20640 rows x 8 columns]
              4.526
     0
             3.585
     1
     2
              3.521
     3
              3.413
     4
              3.422
     20635
              0.781
     20636
              0.771
     20637
              0.923
     20638
              0.847
             0.894
     20639
    Name: price, Length: 20640, dtype: float64
```