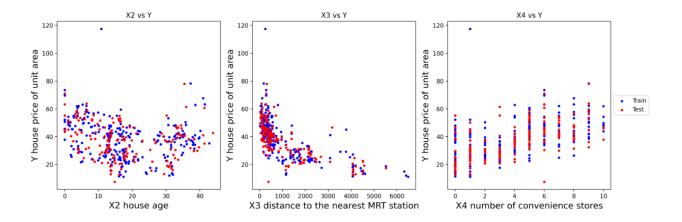
1.

```
# 1. Dataset Loading & splitting:
# randomly pick 60% training data、40% test data
train_set = df.sample(frac=0.6, random_state = 50) # random_state用來固定同一組隨機數據
test_set = df.drop(train_set.index)
```

2.



3.

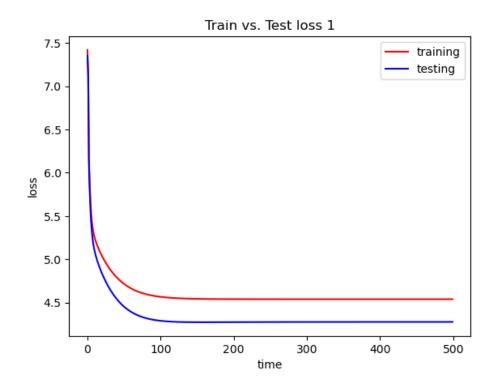
```
# 3. Define loss function (Mean Square Error)
# Using basic mathematical operations in NumPy

def mse_loss(error):
    MSE = np.mean(np.square(error))
    return MSE

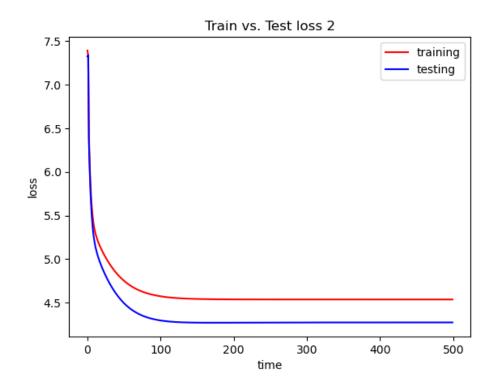
def error_func(predict, real):
    error = predict - real
    return error

def y_pred(X, beta):
    prediction = X.dot(beta)
    return prediction
```

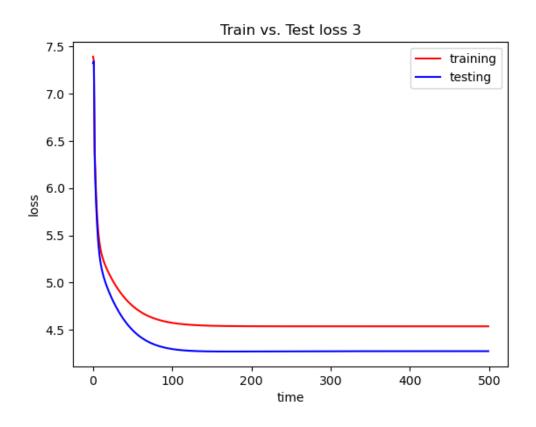
```
Training loss 1:
                                   Testing loss 1:
1 epoch training loss: 1638.8626
                                   1 epoch testing loss: 1561.5623
2 epoch training loss: 113.7888
                                   2 epoch testing loss: 85.7634
3 epoch training loss: 96.5312
                                   3 epoch testing loss: 72.8295
4 epoch training loss: 93.9689
                                   4 epoch testing loss: 71.6451
5 epoch training loss: 93.5849
                                   5 epoch testing loss: 71.7162
6 epoch training loss: 93.5273
                                   6 epoch testing loss: 71.8168
7 epoch training loss: 93.5187
                                   7 epoch testing loss: 71.8652
8 epoch training loss: 93.5174
                                   8 epoch testing loss: 71.8849
9 epoch training loss: 93.5172
                                   9 epoch testing loss: 71.8926
10 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8955
```



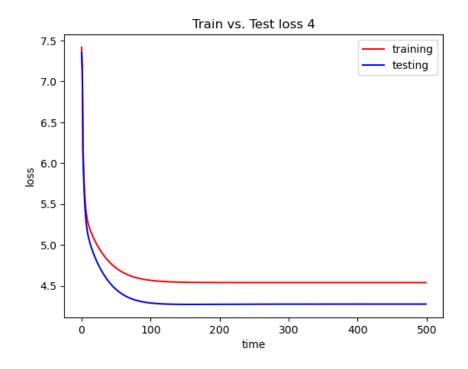
```
Testing loss 2:
Training loss 2:
1 epoch training loss: 1658.6368
                                   1 epoch testing loss: 1529.8699
                                   2 epoch testing loss: 87.9491
2 epoch training loss: 112.1784
                                   3 epoch testing loss: 73.2008
3 epoch training loss: 96.1909
                                   4 epoch testing loss: 71.6713
4 epoch training loss: 93.9032
                                   5 epoch testing loss: 71.6994
5 epoch training loss: 93.5730
6 epoch training loss: 93.5252
                                   6 epoch testing loss: 71.8039
7 epoch training loss: 93.5183
                                   7 epoch testing loss: 71.8585
8 epoch training loss: 93.5173
                                   8 epoch testing loss: 71.8819
                                   9 epoch testing loss: 71.8913
9 epoch training loss: 93.5172
10 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8950
```



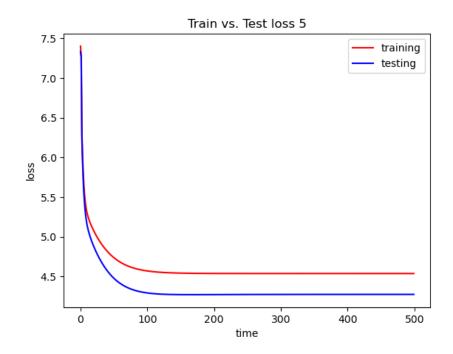
```
Testing loss 3:
Training loss 3:
                                   1 epoch testing loss: 1512.8913
1 epoch training loss: 1619.7421
                                   2 epoch testing loss: 89.6924
2 epoch training loss: 115.9243
                                   3 epoch testing loss: 73.5227
3 epoch training loss: 97.0052
                                   4 epoch testing loss: 71.7014
4 epoch training loss: 94.0647
                                   5 epoch testing loss: 71.6878
5 epoch training loss: 93.6032
                                   6 epoch testing loss: 71.7935
6 epoch training loss: 93.5307
                                   7 epoch testing loss: 71.8529
7 epoch training loss: 93.5193
                                   8 epoch testing loss: 71.8792
8 epoch training loss: 93.5175
9 epoch training loss: 93.5172
                                   9 epoch testing loss: 71.8901
                                   10 epoch testing loss: 71.8944
10 epoch training loss: 93.5172
```



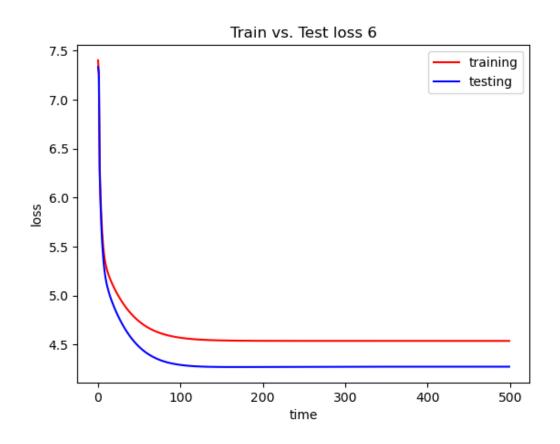
```
Training loss 4:
                                   Testing loss 4:
1 epoch training loss: 1664.6028
                                   1 epoch testing loss: 1557.9476
2 epoch training loss: 111.7846
                                   2 epoch testing loss: 85.9667
3 epoch training loss: 96.1102
                                   3 epoch testing loss: 72.8624
4 epoch training loss: 93.8880
                                   4 epoch testing loss: 71.6470
5 epoch training loss: 93.5703
                                   5 epoch testing loss: 71.7146
6 epoch training loss: 93.5248
                                   6 epoch testing loss: 71.8156
7 epoch training loss: 93.5183
                                   7 epoch testing loss: 71.8646
8 epoch training loss: 93.5173
                                   8 epoch testing loss: 71.8847
9 epoch training loss: 93.5172
                                   9 epoch testing loss: 71.8925
10 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8955
```



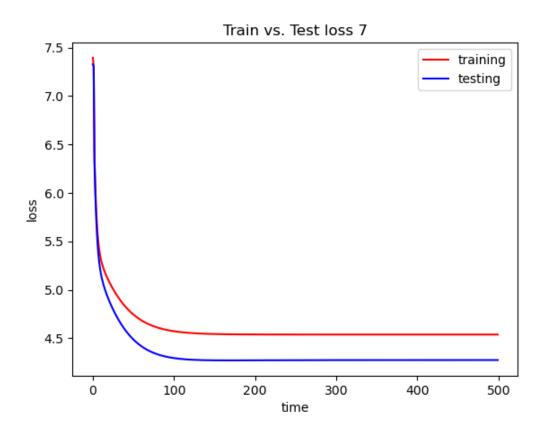
```
Testing loss 5:
Training loss 5:
                                    1 epoch testing loss: 1529.5974
1 epoch training loss: 1636.3971
                                    2 epoch testing loss: 88.0297
2 epoch training loss: 114.0815
                                    3 epoch testing loss: 73.2153
3 epoch training loss: 96.5947
4 epoch training loss: 93.9814
                                    4 epoch testing loss: 71.6726
                                    5 epoch testing loss: 71.6988
5 epoch training loss: 93.5873
6 epoch training loss: 93.5278
                                    6 epoch testing loss: 71.8034
7 epoch training loss: 93.5188
                                    7 epoch testing loss: 71.8583
8 epoch training loss: 93.5174
                                    8 epoch testing loss: 71.8818
                                    9 epoch testing loss: 71.8912
9 epoch training loss: 93.5172
                                    10 epoch testing loss: 71.8949
10 epoch training loss: 93.5172
```



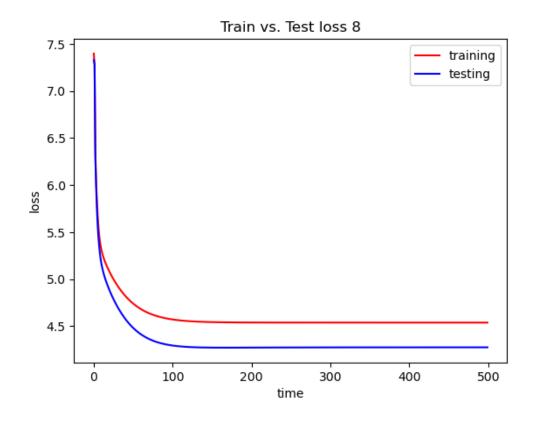
```
Training loss 6:
                                   Testing loss 6:
1 epoch training loss: 1636.9329
                                   1 epoch testing loss: 1530.1608
2 epoch training loss: 113.9849
                                   2 epoch testing loss: 87.9429
3 epoch training loss: 96.5737
                                   3 epoch testing loss: 73.1997
4 epoch training loss: 93.9773
                                   4 epoch testing loss: 71.6713
5 epoch training loss: 93.5865
                                   5 epoch testing loss: 71.6995
6 epoch training loss: 93.5276
                                   6 epoch testing loss: 71.8039
7 epoch training loss: 93.5188
                                   7 epoch testing loss: 71.8586
8 epoch training loss: 93.5174
                                   8 epoch testing loss: 71.8819
9 epoch training loss: 93.5172
                                   9 epoch testing loss: 71.8913
10 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8950
```



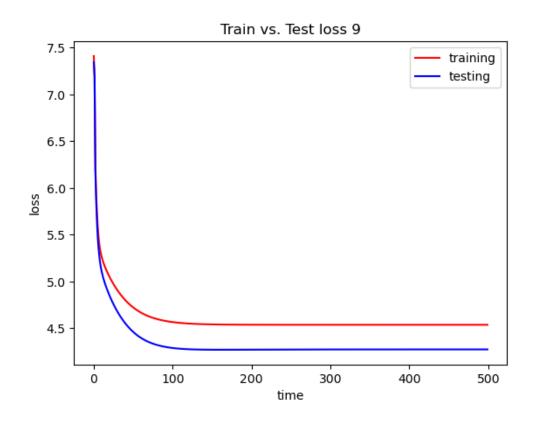
```
Testing loss 7:
Training loss 7:
                                    1 epoch testing loss: 1520.5462
1 epoch training loss: 1627.3549
                                    2 epoch testing loss: 88.8309
2 epoch training loss: 114.9701
                                    3 epoch testing loss: 73.3609
3 epoch training loss: 96.7904
                                    4 epoch testing loss: 71.6855
4 epoch training loss: 94.0207
                                    5 epoch testing loss: 71.6933
5 epoch training loss: 93.5947
                                    6 epoch testing loss: 71.7986
6 epoch training loss: 93.5291
                                   7 epoch testing loss: 71.8557
7 epoch training loss: 93.5190
                                    8 epoch testing loss: 71.8806
8 epoch training loss: 93.5175
                                    9 epoch testing loss: 71.8907
9 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8947
10 epoch training loss: 93.5172
```



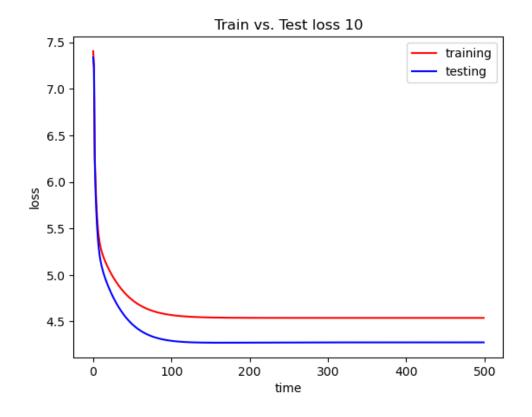
```
Training loss 8:
                                   Testing loss 8:
1 epoch training loss: 1632.8604
                                   1 epoch testing loss: 1526.0503
2 epoch training loss: 114.4062
                                   2 epoch testing loss: 88.3224
3 epoch training loss: 96.6659
                                   3 epoch testing loss: 73.2680
4 epoch training loss: 93.9956
                                   4 epoch testing loss: 71.6771
5 epoch training loss: 93.5900
                                   5 epoch testing loss: 71.6968
6 epoch training loss: 93.5282
                                   6 epoch testing loss: 71.8016
7 epoch training loss: 93.5189
                                   7 epoch testing loss: 71.8573
8 epoch training loss: 93.5174
                                   8 epoch testing loss: 71.8813
9 epoch training loss: 93.5172
                                   9 epoch testing loss: 71.8910
10 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8949
```



```
Training loss 9:
                                    Testing loss 9:
1 epoch training loss: 1653.8247
                                    1 epoch testing loss: 1547.1189
2 epoch training loss: 112.5328
                                    2 epoch testing loss: 86.6375
3 epoch training loss: 96.2645
                                    3 epoch testing loss: 72.9734
4 epoch training loss: 93.9172
                                    4 epoch testing loss: 71.6541
5 epoch training loss: 93.5755
                                    5 epoch testing loss: 71.7092
6 epoch training loss: 93.5257
                                    6 epoch testing loss: 71.8116
7 epoch training loss: 93.5184
                                    7 epoch testing loss: 71.8626
8 epoch training loss: 93.5174
                                    8 epoch testing loss: 71.8838
9 epoch training loss: 93.5172
                                    9 epoch testing loss: 71.8921
10 epoch training loss: 93.5172
                                    10 epoch testing loss: 71.8953
```



```
Training loss 10:
                                   Testing loss 10:
1 epoch training loss: 1644.1486
                                   1 epoch testing loss: 1537.4148
2 epoch training loss: 113.2993
                                   2 epoch testing loss: 87.3260
3 epoch training loss: 96.4261
                                   3 epoch testing loss: 73.0911
4 epoch training loss: 93.9483
                                   4 epoch testing loss: 71.6626
5 epoch training loss: 93.5811
                                   5 epoch testing loss: 71.7040
6 epoch training loss: 93.5267
                                   6 epoch testing loss: 71.8076
7 epoch training loss: 93.5186
                                   7 epoch testing loss: 71.8605
8 epoch training loss: 93.5174
                                   8 epoch testing loss: 71.8828
9 epoch training loss: 93.5172
                                   9 epoch testing loss: 71.8917
10 epoch training loss: 93.5172
                                   10 epoch testing loss: 71.8951
```



Least Square Method

```
Model 1:
```

betas: [[4.36577429e+01 -2.65825847e-01 -5.43776072e-03 1.26079838e+00]], Train loss: 93.5172, Test loss: 71.8973, R Square: 0.5218

Model 2:

betas: [[4.86562639e+01 -1.12230729e+00 2.15915059e-02 -4.76640083e-03 1.24039850e+00]],

Train loss: 86.6068, Test loss: 67.9734, R Square: 0.5571

Model 3:

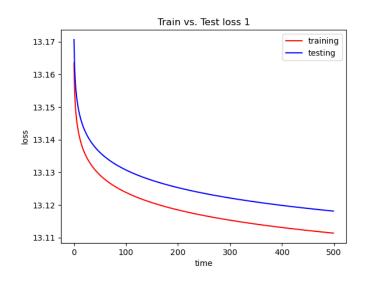
betas: [[5.11693952e+01 -2.83577957e-01 -1.40409553e-02 1.63456916e-06 6.73212945e-01]],

Train loss: 82.9494, Test loss: 66.2943, R Square: 0.5758

Disscusion:

1.

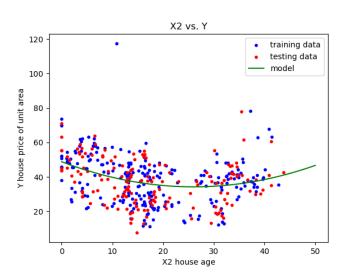
根據以上的結果,可以判斷(β 0, β 1, β 2, β 3)和(β 0', β 1', β 2', β 3')算出來的結果相同。由於題目沒 有規定 random 取值(weighting)的範圍,因此我取 0-0.001 較符合我的 regression model。比如當我 超出以上的 random 範圍時, train 和 test 的 loss 不一定能夠找到最好的值。如下圖我們可以發現它 在經過一段時間便趨近平緩,並未達到最好的結果。由此我理解到 learning rate 的設定和 weighting factor 很有關係,當 weight-ing factor 的初始值距離 minium 很遠的話,learing rate 的初始值也要跟 著改變。

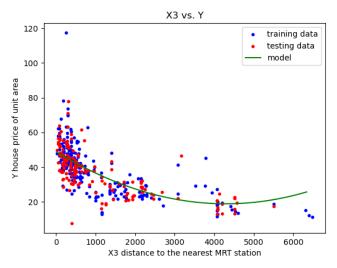


而換句話說,當初使條件設定不良的話,gradient descend 不一定能找到一個最小的值。而 Least Square Method 的目標是最小化預測值和實際值之間差的平方和,可以直接通過解矩陣方程式 來找到係數,也就代表可以直接通過計算得到最優解,不需要迭代過程。以上便是兩者的差異。

以下三張圖為我所找到的每個 feature 的 best fitting model。首先我在程式中跑了九張圖,也就是每個 feature 對每個 model 的圖,然後透過觀察去判斷。X2、X3 分別選擇含有 2 次項的 model,(X2 為 model 2, X3 為 model 3),因為我從圖上觀察到它們的分布比較像 2 次曲線。而 X4 選擇 model 1 的原因是根據 model 的公式,由於 model 2 和 model 3 都含有 2 次項,因此 X4 在那個 model 的權重也會因此改變,也就會讓它相比 model 1 來的差

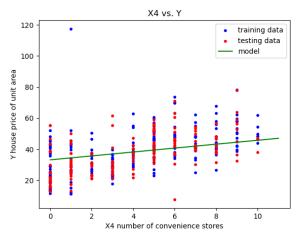
從以下的圖觀察,首先我發現房價和屋齡的關係呈現U型,如預想的一樣,屋齡越小的房子價格越高,但比較特別的是在屋齡 30~40 的區段,房價有回升的跡象,我猜測是因為能維持如此久的房子本身用料就不差,所以這個區段的房子本來就屬於高價房。而房價與距離捷運站遠近也有一定的關聯性,可以看到距離捷運越近的房價很高,且有許多房屋集中於接近捷運站的地方,代表交通是房價的一個重要考量之一。而便利商店的數量和房價的關係相較之下比較不是最主要的原因,但也可以觀察到便利商店的數量和房價也是呈正比關係,也就代表便利商店的數量越多也會影響房價,便利商店的數量多寡可能代表都市化的程度,畢竟都市的便利商店數量一定大於鄉村,而都市的房價也大於鄉村。





$$Y = \beta_4 + \beta_5 X_2 + \beta_6 X_2^2 + \beta_7 X_3 + \beta_8 X_4$$

$$Y = \beta_9 + \beta_{10}X_2 + \beta_{11}X_3 + \beta_{12}X_3^2 + \beta_{13}X_4$$



$$Y = {\beta_0}' + {\beta_1}' X_2 + {\beta_2}' X_3 + {\beta_3}' X_4$$