# **Homework 1**

# **COMP9417, Machine Learning and Data Mining**

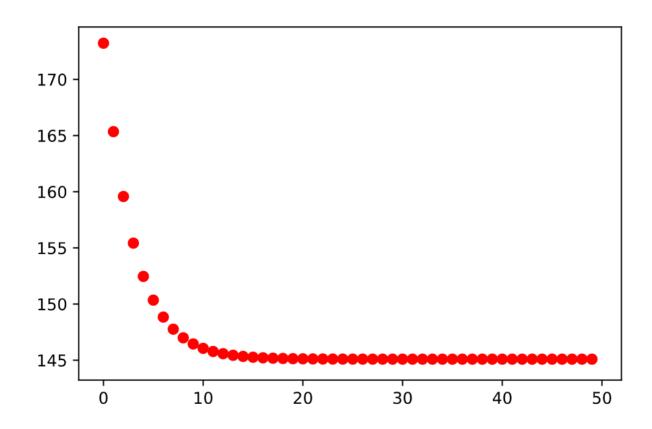
Author: Anant Krishna Mahale (z5277610)

1.The  $\theta$  parameters ( $\theta_0$ ,  $\theta_1$ ;) from step 3 when you are using house age feature.

 $\theta_0$  is 42.54078538346594

 $\theta_1$  is -10.319399022339129

2. A plot, which visualises the change in cost function  $J(\theta)$  at each iteration.



3. RMSE for your training set when you use house age feature.

RMSE for 'House\_age' training\_data 12.045510305912353

4. RMSE for test set, when you use house\_age feature.

RMSE for 'House\_age' testing\_data 16.58731450340051

5. RMSE for test set, when you use distance to the station feature.

RMSE for 'Distance\_to\_Stataion' testing data 12.652088009723935

6.RMSE for test set, when you use number of stores feature

RMSE value while using 'Number of Stores' testing data 14.731993508206784

# 7. Compare the performance of your three models and rank them accordingly.

These are the different RMSE Values obtained from different features.

## **House Age**

```
* Training: 12.045510305912353
* Testing: 16.58731450340051
```

### Distance to the nearest MRT station

```
* Training: 9.165754538401488
* Testing: 12.652088009723935
```

#### **Number of convenience stores**

```
* Training: 27.486676129636784
* Testing: 25.642117651334722
```

We can see that, RMSE value for training data of "Number of convenience stores"> "House Age"> "Distance to the nearest MRT station"

Since training RMSE value for feature "Distance to the nearest MRT station" is least amongst all and order of magnitude for test and train data is same. I would rate it 1.

In the same terms, I would rank 'House Age' 2, and 'Number of convenience stores' 3 as it's RMSE value is high among all the features.

### Summary of the Rank:

- 1. Distance to the nearest MRT station
- 2. House Age
- 3. Number of convenience stores

```
In [1]: #matplotlib inline
   import matplotlib.pyplot as plt
   import seaborn as sns
   import pandas as pd
   import numpy as np
   import math
   from sklearn import datasets, linear_model
   from sklearn.metrics import mean_squared_error, r2_score
   from sklearn.model_selection import train_test_split
```

```
In [2]: #fucntion to calculate the square
def square(x):
    return x*x
```

```
In [3]: def stocashtic_gradient_descent(X,y,theta0, theta1,learning_rate,iterations,col
    or):
        m = len(y)
        for i in range(iterations):
            h_x = theta0 + theta1 * X[j]
            theta0 = theta0 + learning_rate* (y[j] - h_x)
            theta1 = theta1 + learning_rate * (y[j] - h_x) * X[j]
        lossFn = sum([square(y[k] - theta0 - theta1*X[k]) for k in range(m)])/m
        #print(lossFn)
        plt.scatter(i, lossFn, c=color)
    return theta0, theta1
```

```
In [4]: #Method to find the RMSE.
def findRMSE (X,y, theta0, theta1):
    m = len(y)
    sumSoFar = 0
    for i in range(m):
        currentSum = square(y[i] - theta0 - theta1 * X[i])
        sumSoFar = sumSoFar+currentSum
    sumSoFar = (1/m)*sumSoFar
    RMSE = math.sqrt(sumSoFar)

return RMSE
```

```
In [5]: # loading the housing_price_dataset from the current directory.
    df=pd.read_csv('house_prices.csv')

#took each feature in the list.
    features = ["house age", "distance to the nearest MRT station", "number of convenience stores"]

#Normalization.
    for feature in features:
        df[feature]=((df[feature]-df[feature].min())/(df[feature].max()-df[feature].min()))
```

```
In [6]: #Splitting Training and Test Data
trainData = df.iloc[:300]
testData = df.iloc[300:]
```

```
In [7]: #decalring all the training data_set
    train_housevalue = trainData['house price of unit area'].values
    train_houseage = trainData['house age'].values
    train_dist_to_store = trainData['distance to the nearest MRT station'].values
    train_nbrOfStores = trainData['number of convenience stores'].values
```

```
In [8]: #decalring all the test data_set
    test_housevalue = testData['house price of unit area'].values
    test_houseage = testData['house age'].values
    test_dist_to_store = testData['distance to the nearest MRT station'].values
    test_nbrOfStores = testData['number of convenience stores'].values
```

```
#intializing the parameters for training the model on House Age Values.
In [9]:
        #Ouestion 1- 4
        init theta0 = -1
        init theta1 = -0.5
        learningRate = 0.01
        maxIteration = 50
        trained theta0 ha, trained theta1 ha = stocashtic gradient descent(train housea
        ge, train housevalue, init theta0, init theta1, learningRate, maxIteration, 'red')
        train RMSE ha = findRMSE(train houseage, train housevalue, trained theta0 ha, t
        rained thetal ha)
        test RMSE ha = findRMSE(test houseage, test housevalue, trained theta0 ha, trai
        ned thetal ha)
        print('----*Values obtained on House Age Feature*-----
        ----\n')
        print(f'Theta 0: {trained theta0 ha}')
        print(f'\nTheta 1: {trained_theta1_ha}')
        #plt.show()
        print('\nRMSE value for training data',train_RMSE_ha)
        print('\nRMSE value for testing data',test_RMSE_ha)
```

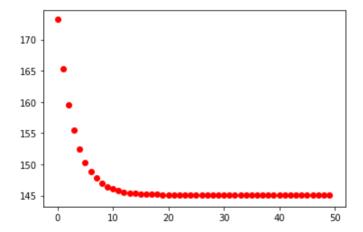
-----\*Values obtained on House Age Feature\*-----

Theta 0: 42.54078538346594

Theta\_1: -10.319399022339129

RMSE value for training data 12.045510305912353

RMSE value for testing data 16.58731450340051



```
In [10]:
         #Initializing the parameters for training the model based on the distance to t
         he station.
         #Question_5
         init theta0 = -1
         init theta1 = -0.5
         learningRate = 0.01
         maxIteration = 50
         trained theta0 ds, trained theta1 ds = stocashtic gradient descent(train dist t
         o store, train housevalue, init theta0, init theta1, learningRate, maxIteration, 'blu
         e')
         #plt.show()
         train_RMSE_ds = findRMSE(train_dist_to_store, train_housevalue, trained_theta0_
         ds, trained thetal ds)
         test RMSE ds = findRMSE(test dist to store, test housevalue, trained theta0 ds,
         trained thetal ds)
         print('----*Values obtained on \'Distance to the nearest M
         RT station Feature\'*----\n')
         print(f'Theta 0: {trained theta0 ds}')
         print(f'\nTheta_1: {trained_theta1_ds}')
         print('\nRMSE value for training data',train_RMSE_ds)
         print('\nRMSE value for testing data',test_RMSE_ds)
```

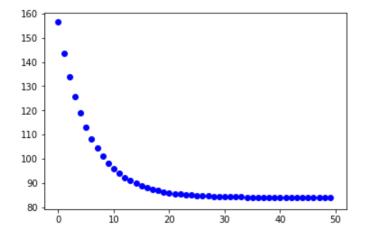
-----\*Values obtained on 'Distance to the nearest MRT stat ion Feature'\*-----

Theta\_0: 44.766087037899375

Theta\_1: -46.500633970906314

RMSE value for training data 9.165754538401488

RMSE value for testing data 12.652088009723935



```
In [11]:
         #Initializing the paramenters for the number of stores.
         #Question 6
         init theta0 = -1
         init theta1 = -0.5
         learningRate = 0.01
         maxIteration = 50
         trained theta0 ns, trained theta1 ns = stocashtic gradient descent(train nbr0fS
         tores, train housevalue, init theta0, init theta1, learningRate, maxIteration, 'gree
         n')
         #plt.show()
         train RMSE ns = findRMSE(train nbrOfStores, train housevalue, trained theta0 ns
         , trained thetal ns)
         test RMSE ns = findRMSE(test nbrOfStores, test housevalue, trained theta0 ns, t
         rained thetal ns)
         print('----*Values obtained on \'Number of convenience sto
         res\' Feature*----\n')
         print(f'Theta 0: {trained theta0 ns}')
         print(f'\nTheta 1: {trained theta1 ns}')
         print('\nRMSE value for training data',train_RMSE_ns)
         print('\nRMSE value for testing data',test_RMSE_ns)
```

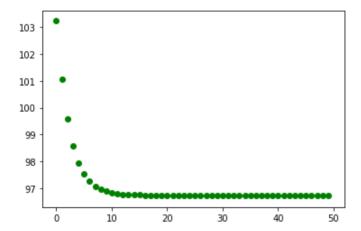
-----\*Values obtained on 'Number of convenience stores' Fe ature\*-----

Theta 0: 27.486676129636784

Theta 1: 25.642117651334722

RMSE value for training data 9.83487827563954

RMSE value for testing data 14.731993508206784



```
In [12]: print('Summary:')
        print('\n\n----*Values obtained on House Age Feature*---
         ----\n')
        print(f'Theta 0: {trained_theta0_ha}')
        print(f'\nTheta 1: {trained_theta1_ha}')
        print('\nRMSE value for training data',train_RMSE_ha)
        print('\nRMSE value for testing data',test_RMSE_ha)
        print('\n\n----*Values obtained on \'Distance to the neare
        st MRT station Feature\'*----\n')
        print(f'Theta 0: {trained_theta0_ds}')
        print(f'\nTheta 1: {trained theta1 ds}')
        print('\nRMSE value for training data',train RMSE ds)
        print('\nRMSE value for testing data',test RMSE ds)
        print('\n\n----*Values obtained on \'Number of convenience
        stores\' Feature*----\n')
        print(f'Theta 0: {trained_theta0_ns}')
        print(f'\nTheta_1: {trained_theta1_ns}')
        print('\nRMSE value for training data',train_RMSE_ns)
        print('\nRMSE value for testing data',test RMSE ns)
       Summary:
          -----*Values obtained on House Age Feature*-----
       Theta 0: 42.54078538346594
       Theta 1: -10.319399022339129
       RMSE value for training data 12.045510305912353
       RMSE value for testing data 16.58731450340051
        -----*Values obtained on 'Distance to the nearest MRT stat
        ion Feature'*-----
       Theta 0: 44.766087037899375
       Theta_1: -46.500633970906314
       RMSE value for training data 9.165754538401488
       RMSE value for testing data 12.652088009723935
        ature*-----
       Theta_0: 27.486676129636784
       Theta 1: 25.642117651334722
       RMSE value for training data 9.83487827563954
       RMSE value for testing data 14.731993508206784
```

