9417 Assignment1

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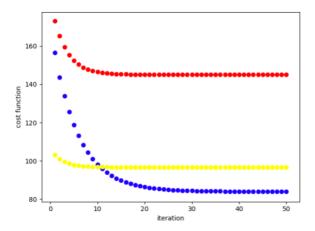
Q1

theta0 for house age: 42.54078538346594 theta1 for house age: -10.319399022339129

Q2

The red curve: house age

The blue curve: distance to the nearest MRT station The yellow curve: number of convenience stores



Q3

RMSE for training set of house age: 12.045510305912353

Q4

RMSE for test set of house age: 16.58731450340051

Q5

RMSE for test set of distance to the nearest MRT station: 12.652088009723935

Q6

RMSE for test set of number of convenience stores: 14.731993508206784

Q7

From the graph of Q2, we can see that as iteration grows to 50, the value of cost function for distance to the nearest MRT station is smallest, and number of convenience stores is the second smallest, house age is the largest. We can also observe the same conclusion when it comes to the RMSE for test sets from Q4, Q5 and Q6. Therefore, the rank for these three features would be distance to the nearest MRT station, followed by number of convenience stores, then house age.

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The whole code is shown as follow:
import matplotlib.pyplot as plt
import numpy as numpy
import pandas as pd
from math import sqrt
def square(x):
       return x*x
#read house_prices.csv and read all the values
df = pd.read_csv('house_prices.csv')
target_name = "house price of unit area"
target = df[target_name].values
house_age_all = df["house age"].values
dis_to_n_MRT_station_all = df["distance to the nearest MRT station"].values
num_of_con_stores_all = df["number of convenience stores"].values
 #normalization
\label{linear_model} \mbox{dis\_to\_n\_MRT\_station\_norm} = \mbox{[(i-min(dis\_to\_n\_MRT\_station\_all))/(max(dis\_to\_n\_MRT\_station\_all)-min(dis\_to\_n\_MRT\_station\_all))/ \mbox{for i } \mbox{In} \mbox{\cite{them.ps}} \mbox{\cite{them.ps}} = \mbox{\cite{them.ps}} \mbox{\cit
dis_to_n_MRT_station_all]
num\_of\_con\_stores\_nll))/(max(num\_of\_con\_stores\_all))/(max(num\_of\_con\_stores\_all)) \\
num_of_con_stores_all]
#stochatic gradient descent
learning_rate = 0.01
theta_0 = -1
theta_1 = -0.5
iterate = 0
while iterate < 50:
      for j in range(300):
              h_fun = theta_0 + theta_1 * house_age_norm[j]
             theta_0 = theta_0 + learning_rate * (target[j] - h_fun)
              theta_1 = theta_1 + learning_rate * (target[j] - h_fun) * house_age_norm[j]
       iterate = iterate + 1:
      Loss = sum([square(target[i]-theta\_0-theta\_1*house\_age\_norm[i]) \ \textit{for} \ i \ \textit{in} \ range(300)])/300.0
       plt.scatter(iterate,Loss,color = "red")
 #plt.show()
print ("theta0 for house age: ",theta_0)
print ("theta1 for house age: ",theta_1)
#RMSE for houseage, trainging set and test set respectively
RMSE_train_houseage = sqrt(sum([square(target[i]-theta_0-theta_1*house_age_norm[i]) for i in range(300)])/300.0)
print ("RMSE for training set of house age: ",RMSE_train_houseage)
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RMSE\_test\_houseage = sqrt(sum([square(target[i]-theta\_0-theta\_1*house\_age\_norm[i]) \\ \textit{for} i \\ \textit{in} range(300,400)])/100.0)
print ("RMSE for test set of house age: ",RMSE_test_houseage)
#reset the thetas and iterate and do the same procedure
iterate = 0
theta_0 = -1
theta_1 = -0.5
while iterate < 50:
         for j in range(300):
                  h_fun = theta_0 + theta_1 *dis_to_n_MRT_station_norm[j]
                  theta_0 = theta_0 + learning_rate * (target[j] - h_fun)
                  theta_1 = theta_1 + learning_rate * (target[j] - h_fun) * dis_to_n_MRT_station_norm[j]
         iterate = iterate + 1:
         Loss = sum([square(target[i]-theta\_0-theta\_1*dis\_to\_n\_MRT\_station\_norm[i]) \ \textbf{for} \ i \ \textbf{in} \ range(300)])/300.0
         plt.scatter(iterate,Loss,color = "blue")
print ("theta0 for distance to the nearest MRT station: ",theta_0)
print ("theta1 for distance to the nearest MRT station: ",theta_1)
RMSE\_train\_dis\_to\_n\_MRT\_station\_norm = sqrt(sum([square(target[ii]-theta\_0-theta\_1*dis\_to\_n\_MRT\_station\_norm[i]) \\ \textit{for } in the target in target in the target in the target in ta
range(300)])/300.0)
RMSE\_test\_dis\_to\_n\_MRT\_station\_norm = sqrt(sum([square(target[i]-theta\_0-theta\_1^*dis\_to\_n\_MRT\_station\_norm[i]) \begin{tabular}{l} for infinity of the context of the con
range(300,400)])/100.0)
print ("RMSE for training set of distance to the nearest MRT station: ",RMSE_train_dis_to_n_MRT_station_norm)
print ("RMSE for test set of distance to the nearest MRT station: ",RMSE_test_dis_to_n_MRT_station_norm)
iterate = 0
theta_0 = -1
theta_1 = -0.5
while iterate < 50:
         for j in range(300):
                  h_fun = theta_0 + theta_1 * num_of_con_stores_norm[j]
                  theta_0 = theta_0 + learning_rate * (target[j] - h_fun)
                  theta_1 = theta_1 + learning_rate * (target[j] - h_fun) * num_of_con_stores_norm[j]
         iterate = iterate + 1:
         Loss = sum([square(target[i]-theta\_0-theta\_1*num\_of\_con\_stores\_norm[i]) \ \textit{for} \ i \ \textit{in} \ range(300)])/300.0
         plt.scatter(iterate,Loss,color = "yellow")
print ("theta0 for number of convenience stores: ",theta_0)
print ("theta1 for number of convenience stores: ",theta_1)
RMSE\_train\_num\_of\_con\_stores\_norm = sqrt(sum([square(target[i]-theta\_0-theta\_1*num\_of\_con\_stores\_norm[i]) \\ \textit{for} i \\ \textit{in} range(300)])/300.0)
RMSE\_test\_num\_of\_con\_stores\_norm = sqrt(sum([square(target[i]-theta\_0-theta\_1*num\_of\_con\_stores\_norm[i]) \\ \textit{for} i in range(300,400)])/100.0)
print ("RMSE for training set of number of convenience stores: ",RMSE_train_num_of_con_stores_norm)
```

 $\textbf{print} \; (\texttt{"RMSE} \; \textbf{for test set of number of convenience stores: "}, \texttt{RMSE_test_num_of_con_stores_norm})$

#show the graph
plt.xlabel("iteration")
plt.ylabel("cost function")
plt.show()