CS589 Homework3

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1 Question 1

1. Base on the program, the $\bar{x}=48$, and $\bar{y}=48.3$. Calculating by the β function, the value of $\beta_1=1$. If the point(48, 50) was changed to (48, 60), the $\bar{y}=48.3+(10\ /\ 10)=49.3$, the $_1$ will not be change since the function: $\beta_1=\frac{\sum_{1}^{10}(x_i-\bar{x})(y_i-(\bar{y}+1))}{\sum_{1}^{10}(x_i-\bar{x})^2}=1, \text{ since when we factor out the 1 from the function, we got } \frac{\sum_{1}^{10}(x_i-\bar{x})(1)}{\sum_{1}^{10}(x_i-\bar{x})^2}=0, \text{ where } x_i=48. \text{ Since } \beta_1=1.0, \ \beta_0=1.0+0.3=1.3$

2 Question 2

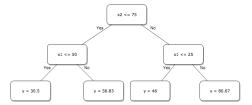


Figure 1: Graph for question 2

1. Please look at the code for with comment "For question 2", the result of the program is ==> 75 y-axis, 50 x-axis, 25 x-axis.

3 Question 3

 If all the points are in the shaded area, the loss function will not be change, which is true. However, the outside points will affect the loss and have different loss if the boundaries of the shaded region. So, that's possible the loss function will be changed. $2. \ \ I2$ is included in the loss function, so even slightly different slope will make the result changed.

4 Question 4

The answer of question 4 is: 0.287619686546, please look at the program for detail.

```
import math
import scipy.stats
import numpy as np
import sklearn.metrics as met
def question_one(x):
   y = x + 0.3
    array = np.array([(x_i, y_i) for (x_i, y_i) in zip(x, y)])
    print("the average of x = ", np.average(array[:, 0]), "; the average of y = ", np.average(array[:, 1]))
    return beta_function(array[:, 0], array[:, 1])
def question_two(dimensions, maxD):
    values = split helper(dimensions, maxD)
    print(values)
    return values
def split_helper(dimensions, maxD):
    maes = []
    split_points_x = [0, 25, 50, 75, 100]
    if maxD != 0:
        for x in range(0, 2):
            for i in split_points_x:
                d1 = \{\}
                d2 = \{\}
                for k, v in dimensions.items():
                    if k[x] <= i:
                        d1.update({k: v})
                    else:
                        d2.update({k: v})
                mae_value = mean_absolute_error(d1, d2)
                maes.append([i, x, mae_value])
        minimum = sorted((el for el in maes), key=lambda L: L[2], reverse = False)[0]
        print(minimum)
        cut(minimum[0], minimum[1], dimensions, maxD)
    else:
        return
def cut(split_value, axis, dimensions, maxD):
    d1 = \{\}
    d2 = \{\}
    if axis == 0:
        print(split_value, "x")
    else:
        print(split_value, "y")
    for k, v in dimensions.items():
        if k[axis] <= split_value:</pre>
            d1.update({k: v})
        else:
            d2.update({k: v})
    split_helper(d1, maxD - 1)
    split_helper(d2, maxD - 1)
def mean absolute error(part1, part2):
    if len(part1) == 0 or len(part2) == 0:
        return float("inf")
    else:
        part1_ave = [sum(part1.values()) / len(part1)] * len(part1)
        part2_ave = [sum(part2.values()) / len(part2)] * len(part2)
        mae_part1 = met.mean_absolute_error(list(part1.values()), part1_ave)
        mae_part2 = met.mean_absolute_error(list(part2.values()), part2_ave)
        return mae part1 + mae part2
    # if len(part1) == 0 and len(part2) != 0:
          part2_ave = [sum(part2.values()) / len(part2)] * len(part2)
          mae part2 = met.mean absolute error(list(part2.values()), part2 ave)
          return mae part2
    # elif len(part2) == 0 and len(part1) != 0:
          part1_ave = [sum(part1.values()) / len(part1)] * len(part1)
          mae part1 = met.mean absolute error(list(part1.values()), part1 ave)
          return mae part1
    # elif len(part2) != 0 and len(part1) != 0:
          part1 ave = [sum(part1.values()) / len(part1)] * len(part1)
          part2 ave = [sum(part2.values()) / len(part2)] * len(part2)
          mae part1 = met.mean absolute error(list(part1.values()), part1 ave)
          mae part2 = met.mean absolute error(list(part2.values()), part2 ave)
          return mae part1 + mae part2
    # else:
          return float(0)
def question four(x values, y values):
    beta 1 = beta function(x values, y values)
   x ave = np.average(x values)
   y ave = np.average(y values)
```

beta_0 = y_ave - beta_1 * x_ave

```
temp = 54 * beta_1 + beta_0
    y_hat = x_values * beta_1 + beta_0
    sig square = 0
    for a, b in zip(y_values, y_hat):
        sig_square = sig_square + (((a - b) ** 2) / len(y_values))
    p = scipy.stats.norm(temp, math.sqrt(sig_square)).cdf(temp - (65 - temp))
    return p
def beta function(x, y):
    return sum([(x_i - np.average(x)) * (y_i - np.average(y)) for (x_i, y_i) in zip(x, y)]) / sum([(x_i - np.average(x)) ** 2 for x_i in x])
if __name__ == '__main__':
    x_{values_1} = np.array([8, 21, 33, 35, 37, 48, 58, 77, 79, 84])
                                                                                   #for question 1
    x_{values_4} = np.array([88, 98, 31, 15, 66, 49, 9, 69, 55, 60])
                                                                                   #for question 4
    y_values_4 = np.array([93, 90, 41, 17, 75, 51, 18, 85, 78, 50])
                                                                                   #for question 4
    points = {(12, 12): 12, (12, 37): 24, (12, 62): 42, (12, 87): 46,
              (37, 12): 18, (37, 37): 40, (37, 62): 47, (37, 87): 71,
              (62, 12): 37, (62, 37): 53, (62, 62): 58, (62, 87): 80,
              (87, 12): 49, (87, 37): 63, (87, 62): 81, (87, 87): 91}
    question_two(points, 3)
    answer1 = question_one(x_values_1)
    print("without the value change, the value of beta_1 is: ", answer1)
    answer4 = question_four(x_values_4, y_values_4)
    print("the answer of question 4 is: ", answer4)
```