<b>COMPUTING SUBJECT:</b>	Machine Learning
TYPE:	WORK ASSIGNMENT
<b>IDENTIFICATION:</b>	Regression Performance
COPYRIGHT:	Michael Claudius
DEGREE OF DIFFICULTY:	Easy
TIME CONSUMPTION:	1 hours
EXTENT:	< 60 lines
<b>OBJECTIVE:</b>	Basic understanding of RMSE regression

# **COMMANDS:**

# **IDENTIFICATION:** Regression Performance/MICL

### The Mission

To understand the idea behind linear regression and Root Square Mean Error (RSME). The context is limited to one variable, y, depending on the independent variable, x,

### Precondition

You must have done the exercise on Linear Regression.

The first 4 questions are similar to previous assignments on linear regression so a copy and paste is just fine!.

### The problem

Given a data list with values for y, and another data list with corresponding values for, x, you are to investigate the performance of linear regression:  $y = b^*x + a$ , as well as polynomial regression:  $y = A^*x^2 + B^*x + C$ . As an example we will use the data given in Appendix A and end up with



As performance measure for the regression, we use the Root Mean Square Error (RMSE):

Equation 2-1. Root Mean Square Error (RMSE)  
RMSE(
$$\mathbf{X}, h$$
) =  $\sqrt{\frac{1}{m} \sum_{i=1}^{m} (h(\mathbf{x}^{(i)}) - y^{(i)})^2}$ 

### Maybe Maybe Not Useful links

https://en.wikipedia.org/wiki/Root-mean-square\_deviation https://www.statisticshowto.datasciencecentral.com/rmse/

https://matplotlib.org/3.1.0/tutorials/introductory/pyplot.html

Assignment 1: Math behind Root Mean Square Error

Read the 1.5 pages (p. 39-41) in "Aurélien Géron Hands-on Machine Learning" Chapter 2 about "Performance measure". Discuss the formula for calculating RMSE:

Equation 2-1. Root Mean Square Error (RMSE) RMSE( $\mathbf{X}, h$ ) =  $\sqrt{\frac{1}{m} \sum_{i=1}^{m} (h(\mathbf{x}^{(i)}) - y^{(i)})^2}$ 

Before the serious calculations, we will play a little and try to guess the correct linear regression values.

<u>Assignment 2: Application program, define data and hypothesis</u> Start Jupyter and create a new file, *RegressionPerformance*. First, import libraries numpy, pandas and matplotlib.pyplot and math. In second cell, declare two lists x & y of same length

#Cost per click of individual keywords
x = [1.0, 2.1, 2.3, 2.5, 4.1, 4.5, 4.9, 5.9, 8.9]
#Total amount of clicks per day
y = [48.2, 63.0, 89.0, 71.0, 89.0, 82.2, 70.0, 80.0, 150.0]

In next cell declare two global values for slope and intersection:

and the hypothesis function, h:

def h(x): return  $b^*x + a$ 

Try to call and print h(2).

#### Assignment 3: Application plot of data and line

Use the plot library and plot the diagram and the data points like you have done before.

```
plt.axis([0, 10, 0, 200])
plt.scatter(x, y)
```

Then, use plt.title, plt.xlabel and plt.ylabel to apply text according to the plot on page 2.

BUT this time utilize the hypothesis function, h, to plot the regression line:

regression\_line = [h(item) for item in [0, 10]]

and hopefully you see:



Try to change the values of a and b and run the code again. Can you manually find a line that fits better by the look.

Now we move on to the RMSE

Assignment 4: Application program, the data We are still using the data:

> x: Cost per click of individual keywords x = [1.0, 2.1, 2.3, 2.5, 4.1, 4.5, 4.9, 5.9, 8.9] y: Total amount of clicks per day y = [48.2, 63.0, 89.0, 71.0, 89.0, 82.2, 70.0, 80.0, 150.0]

#### Assignment 5: Function sum of squares

Look at the formula and the inner part. First declare a function, Sum\_Of\_Squares, to calculate and return the sum of the squares:  $(h(x) - y)^2$  of elements in in two lists:

```
def Sum_Of_Squares(x, y, hFunc):
    . . .
    dif = hFunc(numX) - numY
    xy2.append(dif**2)
.....
```

Make the rest yourself....

Call the function with h as parameter:

result = Sum\_Of\_Squares(x, y, h)

and print the value. *Tip: Similar to xySum\_Prod from previous assignment.* 

#### Assignment 6: RMSE function

Declare a function for calculating and returning the value of RMSE. You just need to utilize *Sum\_Of\_Squares*, a square root and division by the number of data points:

```
def RMSE(x, y, hFunc):
    . . . . .
```

Print out the error for different values of a and b.

<u>Afterthoughts</u> Probably you already used your previous program to find the best fit ! BUT is linear regression best ? Let's investigate polynomial regression of degree 2.

### Assignment 7: Polynomial regression

We shall investigate a polynomial regression for the data set. Thus the hypothesis function is:

$$A^*x^2 + B^*x + C$$

Instead of

 $b^*x + a$ 

Step back to definition cell for h (second cell). Declare 3 variables A, B, C with values 2.0, 1.0, 60.0 And change the h(x)-function to return:

```
def h(x):
    # return b*x + a
    return A*x**2 + B*x + C
```

Run!

Can you find some values giving a lower RMSE than the linear regression ?

<u>Assignment 8: Discussion in the class</u> So what is best linear or polynomial regression ? Can we conclude? Ready to launch ? What shall we do ?

This ends your own mathematical programming, hopefully you got an idea of regression and understand some of the libraries to be used.

## Appendix A

x: Cost per click of individual keywords x = [1.0, 2.1, 2.3, 2.5, 4.1, 4.5, 4.9, 5.9, 8.9]

y: Total amount of clicks per day y = [48.2, 63.0, 89.0, 71.0, 89.0, 82.2, 70.0, 80.0, 150.0]