

# Big Data Analytics

## Exercise Sheet 2

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### Exercise 1: Gradient Descent (5 Points)

- a) Apply gradient descent on the function  $f(x) = \frac{1}{4}x^4 + \frac{1}{3}x^3 - \frac{1}{2}x^2$  with following configurations:
- a) Use step length  $\alpha = 0.3$  and starting point  $x_0 = -1$  and show the first four iterations. What is your minimum?
- b) Use step length  $\alpha = 2$  and starting point  $x_0 = -1$  and show the first four iterations. What has happened and why?
- c) Use step length  $\alpha = 0.3$  and starting point  $x_0 = 0$  and show the first two iterations. What has happened and why?  
Do the same again with  $\alpha = 0.8$  and starting point  $x_0 = 0.5$  and show the first four iterations. Where is your minimum now?  
What would be a possible solution to overcome the problem just identified?

### Exercise 2: Linear Regression using Gradient Descent(10 Points)

A restaurant chain wants to predict profit given the population of the city, in which they want to open a new franchise. Attached is the restaurant data set with two columns, Population of the city and second column is Profit earned in that city.

- a) To visualize this data, you are required to plot this data by putting “Profits in 10,000s” as label of Profit on y-axis and “Population of city in 10,000s” as label of Population on x-axis.
- b) Program a Gradient Descent algorithm that minimizes cost function of Uni-variate Linear Regression by learning optimal values for  $\beta$ . The objective of linear regression is to minimize the cost function

$$J(\beta) = \frac{1}{2m} \sum_{i=1}^m (h_{\beta}(x^i) - y^i)^2$$

Where the hypothesis  $h_{\beta}(x)$  is given by linear model

$$h_{\beta}(x) = \beta^T x = \beta_0 + \beta_1 x$$

### Exercise 3: Logistic Regression (5 Points)

The table below presents the test-firing results for 25 surface-to-air anti aircraft missiles at targets of varying speed. The result of each test is either a hit ( $t = 1$ ) or a miss ( $t = 0$ ). The explanatory variable  $x$  gives the speed of the target in knots.

Target speed			Target speed		
Test	( $x$ ) in knots	$t$	Test	( $x$ ) in knots	$t$
1	400	0	14	330	1
2	220	1	15	280	1
3	490	0	16	210	1
4	210	1	17	300	1
5	500	0	18	470	1
6	270	0	19	230	0
7	200	1	20	430	0
8	470	0	21	460	0
9	480	0	22	220	1
10	310	1	23	250	1
11	240	1	24	200	1
12	490	0	25	390	0
13	420	0			

Logistic Regression model is given by:

$$\hat{y}(\mathbf{x}; \beta) = p(t = 1|x) = \frac{\exp(\beta_0 + \beta_1 x)}{1 + \exp(\beta_0 + \beta_1 x)}$$

Using maximum likelihood, we estimated the following values of parameters.

$$\beta_0 = 6.071 \text{ and } \beta_1 = -0.018$$

- $\beta_1$  is negative. What can we infer from that about the relation between target speed and the probability of a hit?
- What is the fitted probability of a hit for a target speed of 350 knots?
- Give the classification rule corresponding to the fitted model, assuming we always allocate to the class with highest fitted probability for a given target speed.