

# MLAP Open Assessment A

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## 1 Linear regression and Logistic regression

### 1.1 Task 1

In order to experiment with linear regression I have chosen to use the following 8 features, referred to in equations as  $a$  through  $h$  for brevity:

- $a$  stock volume of previous day
- $b$  difference between the previous two days' stock volumes
- $c$  mean of stock volumes from previous ten days
- $d$  standard deviation of stock volumes from previous ten days
- $e$  stock price of previous day
- $f$  difference between the previous two days' stock prices
- $g$  mean of stock prices from previous ten days
- $h$  standard deviation of stock prices from previous ten days

Further, the elements of a vector  $\theta$  represent the coefficients of a regression function, with  $\theta_0$  always representing the constant term. Hence, a regression function might look as follows:

$$f(\theta) = \theta_0 + \theta_1 a + \theta_2 a^2$$

Figure 1 shows the Mean Squared Errors (MSEs) obtained in my initial phase of experimentation with the chosen features, shown to three significant figures. In this phase of experimentation, I evaluated the performance of each feature used on its own in first- and second-order polynomials.

It is clear from these results that the best performances come when using the features that relate to stock price as opposed to stock value ( $e - h$ ). However, feature  $f$ , the difference between the last two days' stock prices does not appear to perform very well. Feature  $e$  does not perform well as a first-order polynomial, but is exceptionally good as a second-order polynomial.

My next phase of experimentation is to take the high-performing features and try using them on their own in third-order polynomial functions. Following this, I will experiment with combining the better performing features to see what improvements can be made. The results of the initial third-order polynomial experiments are shown in figure 2. Out of interest, I have chosen to try feature  $c$  as a third-order polynomial as it was the best performing of the stock volume-related features.

As seen by the results, each feature tested in third-order polynomials had very similar results to the second-order polynomial tests. Going forward, I have chosen to try a function combining  $e$  and  $g$  both as second-order polynomials to see if they perform well as a pair. I am also interested to see if the recent change in stock volume ( $b$ ) combined with the mean of the last ten days' stock prices ( $g$ ) gives an indication of the next stock price. Further, combining  $a$ ,  $b$ ,  $d$  and  $g$  all together may give good results. The MSEs obtained for these tests are shown in figure 3 (note the change of scale for this chart).

From these results we can see

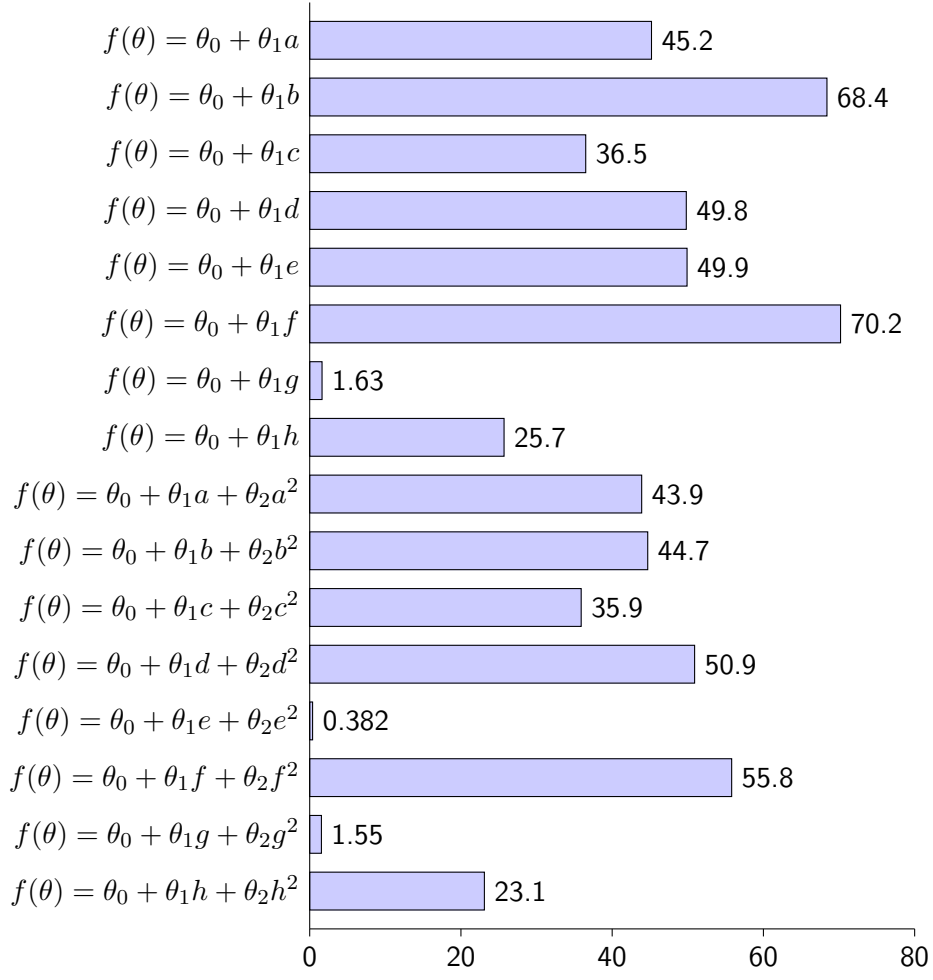


Figure 1: MSEs obtained using each of the features on their own in first- and second-order polynomials

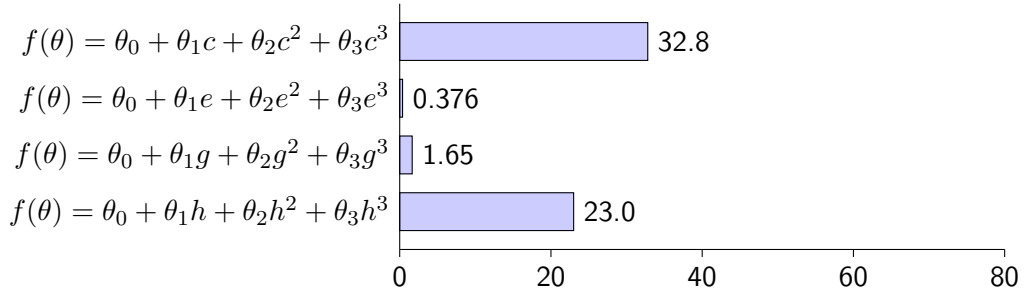


Figure 2: MSEs obtained using selected features on their own in three-order polynomials

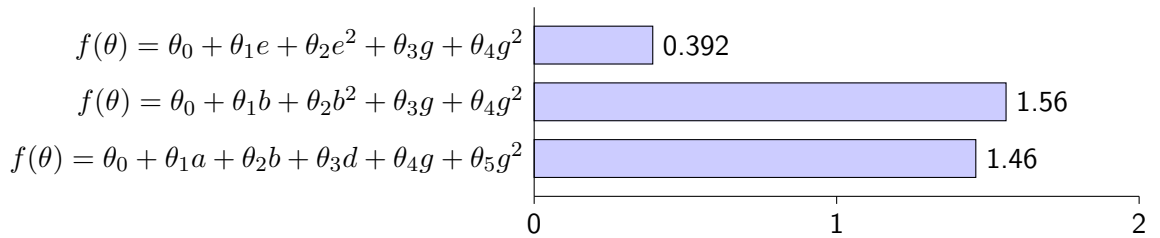


Figure 3: MSEs obtained when combining features into more complex polynomials

1.2 Task 2

1.3 Task 3

## 2 Bayesian networks

2.1 Task 4

2.2 Task 5