

Decision Support Systems

Team 3

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Chapter 1: Introduction

Chapter 2: Regression

This chapter details the work of LAB exercise 3.6.2, 3.6.3, 4.6.1 and 4.6.2 from "An Introduction to Statistical Learning". It starts by recapitulating the theory behind linear regressions, both simple and multiple, then proceeds to describe the accompanied LAB exercises and conclude on their findings.

2.1 Multiple Linear Regression

2.1.1 Theory

Basic theory for simple and multiple lin regs here. From the slides or book.

Simple Linear Regression is used to make linear models of data. It has a response Y on the basis of a single predictor variable X . We can write it as $Y = \beta_0 + \beta_1 X_1 + \epsilon_i$. $\beta_0 + \beta_1$ are unknown and to get a response, we must use data to estimate the coefficients. $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$ represent n observation pairs, each of which consists of a measurement of X and a measurement of Y . The drawback of this method is that only a single predictor variable is used and often have more. In cases where we want examined the relationship between multiple predictor variables we use Multiple Linear Regression. The model takes the following form $Y = \beta_0 + \beta_1 X_1 + \dots + \beta_n X_n + \epsilon_i$

To obtain the Coefficients in the model we use the least squares method to minimize the sum of squared residuals. We pick $\beta_0, \beta_1, \dots, \beta_p$ to to minimize the sum of squared residuals.

$$RSS = \sum (y - \hat{y})^2 = \sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_{i1} + \hat{\beta}_2 x_{i2} - \dots - \hat{\beta}_p x_{ip})^2$$

2.1.2 Results

LAB 3.6.2 + 3.6.3

2.1.3 Conclusion

2.1.4 Logistic Regression

2.1.5 Theory

Basic theory for logistic lin regs here. From the slides or book.

2.1.6 Results

LAB 4.6.1 + 4.6.2

2.1.7 Conclusion

Chapter 3: Linear Discriminant Analysis

Chapter 4: Cross Validation

Chapter 5: Subset Selection

Chapter 6: Shrinkage Methods

Chapter 7: Clustering Methods

Chapter 8: Discussion

Chapter 9: Conclusion

Chapter 10: Perspectives

Chapter 11: References

