

Chapter V

Preprocessing Perceptrons and Multivariate Decision Limits

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ABSTRACT

Classification networks, consisting of preprocessing layers combined with well-known classification networks, are well suited for medical data analysis. Additionally, by adjusting network complexity to corresponding complexity of data, the parameters in the preprocessing network can, in comparison with networks of higher complexity, be more precisely understood and also effectively utilised as decision limits. Further, a multivariate approach to preprocessing is shown in many cases to increase correctness rates in classification tasks. Handling network complexity in this way thus leads to efficient parameter estimations as well as useful parameter interpretations.

INTRODUCTION

Decision limits, in the sense of discrimination values used for interpretative purposes (Solberg, 1987), are arrived at through traditional bio-statistical approaches to medical data analysis (see e.g. (Armitage & Berry, 1994)). In discrimination (or classification), such as typically used in differential diagnosis, involving several attributes, this analysis should not be performed on univariate level. Attributes cannot be analysed one by one, since synthesization of these decision limits must then be considered separately. Such a discriminator is then like a hybrid, with isolated univariate analysis composed with a classification network, where network parameters are trained independently from the level of univariate decision limits. This would provide a separation of univariate decision limits and discrimination, which

is unfavourable for correctness rates. Furthermore, efforts to increase correctness rates are typically based on increase of network complexities, leading to other complications such as overtraining and difficulties to interpret network parameters. An analytical presentation of the usefulness to correlate data complexity with network complexity is outside the scope of this chapter. We refer e.g. to (Duda et. al., 2001) for further reading.

In this chapter, we propose to use preprocessing perceptrons which include a preprocessing layer allowing establishing optimal multivariate decision limits. Classification networks coupled with preprocessing can be kept small, and, indeed, with a network complexity in correlation with data complexity. The classification network coupled with preprocessing can be ordinary logistic regression, i.e., a perceptron using neural network terminology. Several other classification networks can be used but it has been shown (Eklund, 1994) difficult to establish significantly higher correctness rates with classifiers of higher complexities.

Hybrid networks, used either in classification or closed-loop problems, are often used in order to increase network complexity to cope with higher data complexity. Typical examples include moving from character to voice recognition or applications of intelligent control. The hybrid described in this chapter is intended to enhance decision limit related preprocessing within the classifier itself. Careful preprocessing with parameter estimations, also for the preprocessing layer, enables the use of classifier networks with low complexities. We support our claims by using some typical case studies in binary classification. Examples are drawn from chromosomal anomalies, endocrinology and virology.

This chapter is organised as follows; First, a description of the proposed preprocessing perceptron is given, including the mathematical formulas behind it. In the next section, three case studies, where the preprocessing perceptron has been used, are described together with the methods used and the results. The results are compared with the results of a multilayer perceptron run on the same data. Next, univariate versus multivariate decision limits are discussed and the final section in this chapter contains a conclusion of the chapter.

THE PREPROCESSING PERCEPTRON

A linear regression is given by the weighted sum:

$$y = \gamma + \sum_{i=1}^n w_i x_i$$

where x_i are inputs and w_i and γ are parameters of the linear function. A logistic regression performs a sigmoidal activation of the weighted sum, i.e.:

$$y = \frac{1}{1 + e^{(-\gamma - \sum_{i=1}^n w_i x_i)}}$$

Note that a logistic regression function is precisely a (single-layer) perceptron in the terminology of neural networks (Duda et. al., 2001). The preprocessing perceptron consists similarly of a weighted sum but includes preprocessing functions for each input variable. Suitable preprocessing functions are sigmoids (sigmoidal functions):

$$g[\alpha, \beta](x) = \frac{1}{1 + e^{-\beta(x-\alpha)}}, \beta \neq 0 \tag{1}$$

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