



Backpropagation of error modelling applied to the HFC Bird Creek Data Set

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Neurocomputing is a set of emergent technologies concerned with information processing systems that develop operational capabilities in an autonomous manner as an adaptive response to an information environment. The principal data processing structure of interest is the artificial neural network; although other classes of adaptive learning mechanism are also included. The most popular 'default' training algorithm is 'backpropagation of error' (Rumelhart *et al.*, 1986; Tvetter, 2004). This optimisation procedure provides an efficient computational mechanism for evaluating the derivatives of the network performance function with respect to a given set of network parameters. It corresponds to a propagation of errors backwards through the network. This explains its peculiar sounding title. The term has also been adopted to describe feed forward multi-layered networks trained with the 'backpropagation of error' algorithm: producing 'backprop' or 'backpropagation' neural networks. Networks of this type have emerged as major workhorses in various areas of business and commerce; it is also the most common type of neural network that has been used to perform hydrological modelling operations (Maier & Dandy, 2000). Indeed, for most explorations, the standard backpropagation neural network is the first point-of-call and will often produce acceptable results. As such, the use of more complex solutions will seldom, if ever, need to be investigated. Thus fresh developments in neural network modelling should always be compared to standard backpropagation models in order to establish the potential advantages that are on offer. Following the neural network procedures described in Dawson & Wilby (2001), this paper will report the construction and application of a standard backpropagation neural network solution developed

on the competition dataset. The selection of modelling inputs was based on a statistical consideration of raw inputs and moving averages. Trial and error was used to determine an optimal number of hidden units. Early stopping based on the use of a cross validation dataset was required to prevent over fitting. The results are compared with multiple linear regression outputs developed on identical datasets.

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