Chapter 6 Training Algorithms

ABSTRACT

The process of assigning the weight to each connection is called training. A network can be subject to supervised or unsupervised training. In this chapter, supervised and unsupervised learning are explained and then various training algorithms such as multilayer perceptron (MLP) and Back Propagation (BP) as supervised training algorithms are introduced. The unsupervised training algorithm, namely Kohonen's self-organizing map (SOM), is introduced as one of most popular neural network models. SOMs convert high-dimensional, non-linear statistical relationships into simple geometric relationships in an n-dimensional array.

1 INTRODUCTION

A network can be subject to supervised or unsupervised learning. The learning would be supervised if external criteria are used and matched by the network output, and if not, the learning is unsupervised. Unsupervised approaches are also termed selforganizing. Typically more feedbacks and intralayer connections between neurons cause more interaction between neurons.

Supervised networks are a little more straightforward to conceptualize than unsupervised networks. In supervised learning, the inputs to the supervised network are applied along with an expected response. The network is trained by pairs of stimulus-response records. For example, a stock market forecaster may present

DOI: 10.4018/978-1-4666-6146-2.ch006

Training Algorithms

economic data (the stimulus) along with metrics of stock market performance (the response) to the neural network to the present and attempt to predict the future once training is complete.

The unsupervised network is provided with only stimulus. For example, if we want the unsupervised network to correctly classify parts from a conveyor belt into part numbers, providing an image of each part to do the classification (the stimulus). The unsupervised network in this case would act like a look-up memory that is indexed by its contents, or a Content-Addressable-Memory (CAM).

2 SUPERVISED LEARNING

Synapses interconnect the individual neurons through to make a network. As information is processed, the connections allow the neurons to signal each other. One weight is assigned to each connection and connections are not equal. The connection weight is zero, if there is no connection between two neurons. These weights are what determine the output of the neural network. Therefore, it can be said that the connection weights form the memory of the neural network.

The process of assigning the weight to each connection is called training. The beginning of the most training algorithms is by assigning random numbers to the weight matrix. Then the validity of the neural network is examined. Next, based on the valid performance of the network the weights are adjusted. This process is repeated until the validation error is within an acceptable limit. There are many ways to train neural networks. Neural network training methods generally fall into the categories of supervised, unsupervised and various hybrid approaches.

Supervised training is accomplished by giving the neural network a set of sample data along with the anticipated outputs from each of these samples. Supervised training is the most common form of neural network training. As supervised training proceeds, the neural network is taken through several iterations, or epochs, until the actual output of the neural network matches the anticipated output, with a reasonably small error. Each epoch is one pass through the training samples.

Supervised learning or Associative learning in which the network is trained by providing it with input and matching output patterns. These input-output pairs can be provided by an external teacher, or by the system, which contains the neural network (self-supervised).

It is very important to understand how to properly train a neural network. This book explores several methods of neural network training, including Back Propagation, Particle Swarm Optimization (PSO), and genetic algorithms. Once the neural network is trained, it must be validated to see if it is ready for use. 15 more pages are available in the full version of this document, which may be purchased using the "Add to Cart"

button on the publisher's webpage: www.igi-

global.com/chapter/training-algorithms/111001

Related Content

Modelling Analysis and Simulation for Reliability Prediction for Thermal Power System

Vikram Kumar Kamboj, Kamalpreet Sandhuand Shamik Chatterjee (2020). *Al Techniques for Reliability Prediction for Electronic Components (pp. 136-163).* www.irma-international.org/chapter/modelling-analysis-and-simulation-for-reliability-prediction-for-thermal-power-system/240495

Application of ANN and PSO Swarm Optimization for Optimization in Advanced Manufacturing: A Case With CNC Lathe

Nehal Dash, Sanghamitra Debtaand Kaushik Kumar (2022). *Research Anthology on Artificial Neural Network Applications (pp. 804-823).*

www.irma-international.org/chapter/application-of-ann-and-pso-swarm-optimization-foroptimization-in-advanced-manufacturing/288987

A Novel Prediction Perspective to the Bending Over Sheave Fatigue Lifetime of Steel Wire Ropes by Means of Artificial Neural Networks

Tuba Özge Onurand Yusuf Aytaç Onur (2020). *Artificial Intelligence and Machine Learning Applications in Civil, Mechanical, and Industrial Engineering (pp. 39-58).* www.irma-international.org/chapter/a-novel-prediction-perspective-to-the-bending-over-sheave-fatigue-lifetime-of-steel-wire-ropes-by-means-of-artificial-neural-networks/238138

Hybrid-Learning Methods for Stock Index Modeling

Yuehui Chenand Ajith Abraham (2006). Artificial Neural Networks in Finance and Manufacturing (pp. 64-79).

www.irma-international.org/chapter/hybrid-learning-methods-stock-index/5349

Disease Identification in Plant Leaf Using Deep Convolutional Neural Networks

Venu K., Natesan Palanisamy, Krishnakumar B.and Sasipriyaa N. (2020). Handbook of Research on Applications and Implementations of Machine Learning Techniques (pp. 46-62).

www.irma-international.org/chapter/disease-identification-in-plant-leaf-using-deep-convolutionalneural-networks/234117