New Artificial Neural Network Models for Bio Medical Image Compression:
Bio Medical Image Compression

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ABSTRACT

This article presents an image compression method using feed-forward back-propagation neural networks (NNs). Marked progress has been made in the area of image compression in the last decade. Image compression removing redundant information in image data is a solution for storage and data transmission problems for huge amounts of data. NNs offer the potential for providing a novel solution to the problem of image compression by its ability to generate an internal data representation. A comparison among various feed-forward back-propagation training algorithms was presented with different compression ratios and different block sizes. The learning methods, the Levenberg Marquardt (LM) algorithm and the Gradient Descent (GD) have been used to perform the training of the network elements and finally, the performance is evaluated in terms of MSE and PSNR using medical images. The decompressed results obtained using these two algorithms are computed in terms of PSNR and MSE along with performance plots and regression plots from which it can be observed that the LM algorithm gives more accurate results than the GD algorithm.

KEYWORDS
Artificial Neural Network, Backpropagation Neural Network, Gradient Descent Algorithm (GD), Image Compression, Levenberg Marquardt Algorithm (LM)

1. INTRODUCTION

Artificial neural networks (ANNs) are archetypes of the biological neuron system and thus have been drawn from the abilities of a human brain. The architecture of ANN being drawn from the concept of brain functioning, a neural network is a hugely reticulated network of a huge number of neurons which are processing elements. ANNs are employed to summarize and prototype some of the functional aspects of the human brain system in an effort so as to acquire some of its computational strengths. A NN consists of eight components: neurons, signal function, activation state vector, activity aggregation rule, pattern of connectivity, learning rule, activation rule, and environment. Recently, ANNs are applied in areas in which high rates of computation are essential and considered as probable solutions to problems of image compression. Generally, two different categories have been put forward for enhancing the performance of compression methods. Firstly, a method for compression by using ANN technology has to be developed to improve the design. Secondly, neural networks have to be applied to develop compression methods. Backpropagation algorithm is extensively used learning algorithms in ANNs. With generalization ability and high accuracy, the feedforward neural network

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architecture is capable of approximating most problems. This architecture is based on the learning rule of error-correction. Error propagation comprises of two passes, a forward pass and a backward pass through different layers of network. The effect, of input vector’s application to the sensory nodes of the network, transmits through the network layer by layer in the forward pass. In the end, a set of outputs are produced as an actual response of this process. All the synaptic weights of the networks are fixed during the forward pass only and adjusted according to the need of error-correction during the back pass. The error signal is produced when the actual output of the network is subtracted from the expected output. This error signal is then propagated backward against the direction of synaptic conditions through the network. Until the actual output of the network so produced is nearer to the expected output, the synaptic weights are adjusted. To produce a complex output, the backpropagation neural network is essentially made of a network of simple processing elements working together. From the above knowledge of back propagation neural networks, image compression, and decompression can be achieved.

2. RELATED WORK

A new self-organization algorithm, which is based on the centroid learning rule and frequency-sensitive cost function in order to construct the codebooks. The results include a good adaptivity for varied statistics of source data (Chen, Sheu & Fang, 1994). A neural network data compression method which involves a new training method called as the Nested Training Algorithm (NTA), results in maintaining low distortion and high compression ratio (Chin & Arozullah, 1996). Image compression is achieved by appropriate image thresholding and these thresholds are obtained with a principle of moment preserving and was proposed by (Yang & Tsai, 1998). Block adaptive prediction based neural network scheme is used for lossless data compression. The results involve that the adaptations of the improvised method increases performance of the classical predictors evaluated (Logeswaran, 2002). Image compression can also be performed by a non-uniform thresholding and observed the effects of thresholding on reconstructed image quality (Sansgiry & Mihaila, 2003). The design of optimized codebooks by using the vector quantisation (VQ) that included the strategy of reinforced learning (RL). The results have shown that RL is insensitive to the selection of the initial codebook and an additional parameter known as learning rate parameter introduced by RL learning rate control parameter and was described by (Xu, Nandi, & Zhang, 2003). The method of Compression and encryption/decryption using neural networks was proposed and attained good quality decryption and reconstruction of 3D objects (Shortt, Naughton, & Javidi, 2006). A technique for noise removal and image compression in wavelet domain thresholding which is based on Partial Differential Equation (PDE) and it takes the advantage of variations in framework (Chan, & Zhou, 2007). Image compression method which consumes less time and follows a strategy where thresholds are optimized with optimization techniques for which objective function is distortion (Kaur, Gupta, Chauhan, & Saxena, 2007). An evolutionary fuzzy particle swarm optimization (FPSO) learning algorithm to self-extract the nearest optimum codebook of vector quantization (VQ) to carry out image compression (Feng, Chen, & Ye, 2007). Image compression can also be performed with Multistage Lattice Vector Quantization (MLVQ) and by thresholding of DWT coefficients. Proposed combination tries to minimize the quantization error and its computational complexity is less compared to ordinary VQ (Salleh & Soraghan, 2007). Electrocardiography (ECG) signals are compressed by transforming the signal with the help of discrete wavelet transform (Mohammadpour & Mollaei, 2009). Another kind of image compression where image to be compressed is transformed to frequency domain with the help of bandlet and required bandlet coefficients are obtained with type II Fuzzy thresholding and results are compared with the ordinary thresholding (Rajeswari & Rajesh, 2012). Image compression is achieved by the neural network, the results are proved that the Training algorithm and the back propagation neural network can increase the performance and decrease the convergence time (Patel & Agrawal, 2013). In order to overcome the problem of haziness in decompressed image, various artificial neural networks
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