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1 2	Backpropagation in HL7 in Medical Informatics to Analysis Speed of Sending Data
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7 Abstract

6

In this paper, analysis the speed of sending message in Healthcare standard 7 with the use of 8 back propagation in neural network. Various algorithms are define in backpropagation in neural 9 network we can use trainlm algorithm for sending message purpose. This algorithm appears to 10 be fastest method for training moderate sized feedforward neural network. It has a very 11 efficient matlab implementation. The need of trainlm algorithm are used for analysis, increase 12 the speed of sending message faster and accurately and more efficiently. The proposed work is 13 used in healthcare medical data. With the use of backpropagation in health care standard 14 seven (HL7) sending message between two systems. To increase the speed of the healthcare 15 sending data we can use Train LM algorithm. Train LM algorithm is more fastest algorithm it 16 can be increase efficiency and improve accuracy of the system and also provide real time 17 application. To increase speed of sending message these algorithm used. With the use of this 18 algorithm it can be decreasing time of sending message to the other system. 19

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21 Index terms— medical informatics, HL7, backpropagation.

²² 1 Introduction

edical informatics is the sub-discipline of health informatics that directly impacts the patientphysician relation-23 ship. It focuses on the information technology that enables the effective collection of data using technology 24 tools to develop medical knowledge and to facilitate the delivery of patient medical care. The goal of medical 25 informatics is to ensure access to critical patient medical information at the precise time and place it is needed 26 to make medical decisions. Medical informatics also focuses on the management of medical data for research and 27 education. a) Healthcare Standards support clinical practice and the management, delivery, and evaluation of 28 health services, and are recognized as the most commonly used in the world. Healthcare provides seven standards 29 to perform various functionalities. The latest standard implement in Healthcare is Health Level Seven (HL1-7) 30 31 is a standard series of predefined logical formats for packaging healthcare data into messages to be transmitted 32 among computer system.

³³ 2 b) Neural Networks

Are originally modelled as a computational model to mimic the way the brain works. Brain is made from small functional units called neurons. A neural has a cell body, several short dendrites and single long axon. By the dendrites and axon several neurons connected. Dendrites take various signals and pass to the other neurons as a input signal. These input increase or decrease to the electrical potential of the cell body and if it is reaches a threshold, a electric pulse is sent to the axon and the output occurs.

39 **3** II.

Types of Neural Network a) Biological Neural Network Are made up of real biological neurons that are connected or functionally related in a nervous system. In the field of neuroscience they often identified groups of neurons that perform a specific physiological function in laboratory analysis.

43 4 b) Artificial Neural Network

44 Are composed of interconnecting artificial neurons (programming constructs that mimic the properties of 45 biological neurons).it is used for solving artificial intelligence problems without necessary creating a model of

 $_{\rm 46}~$ a real biological system. 3 layers in neural network I/P, hidden layer and O/P.

47 5 Input

- Hidden layer Output ii. Simulating Annealing It is a global minimum can guarantee of optimal solution but it is
 slower than gradient decent and also much more complicated implementation.
- 50 iii. Genetic Algorithm Faster than simulated annealing and also less like to get stuck in local minima but it 51 is slower than gradient descent and also memory intensive for large network.

⁵² 6 iv. Simplex Algorithm

⁵³ It is similar to gradient decent but faster and easy to implement but does not gurantee a global minima.

⁵⁴ 7 v. Train LM Algorithm

It is much faster than all algorithm and also used to calculate performance easily implement in matlab. It used to solve the fitting problem and also provide fastest many mode sizes feed forward network.

57 8 III.

⁵⁸ 9 Methodology a) Levenberg-Marquardt Algorithm

59 In order to make sure that the approximated Hessian matrix JTJ is invertible.

⁶⁰ 10 Levenberg-Marquardt algorithm introduces another approx-⁶¹ imation to Hessian matrix:

62 (1.1) where ? is always positive, called combination coefficient.

63 I is the identity matrix.

From Equation ??.1, one may notice that the elements on the main diagonal of the Hessian matrix will be larger than zero. Therefore, with this approximation (Equation ??.1), it can be sure that matrix H is always invertible.

(1.2) By combining Equations 1.1 and 1.2, the update rule of Levenberg-Marquardt algorithm can be presented
 as ??1.3) As the combination of the steepest descent algorithm and the Gauss-Newton algorithm, the Levenberg Marquardt algorithm switches between the two algorithms during the training process. When the combination

70 coefficient ? is very small (nearly zero).

Equation (1.1) approaching to Equation (1.2) and Gauss-Newton algorithm is used. When combination coefficient ? is very large, Equation ??.1 approximates to ??1.4) and the steepest descent method is used. If the combination coefficient ? in Equation ??2.25 is very big, it can be interpreted as the learning coefficient in the steepest descent method (1.4).

The training process using Levenberg-Marquardt algorithm could be designed as follows:

i. With the initial weights (randomly generated), evaluate the total error (SSE). ii. Do an update as directed 76 by Equation ??.1 to adjust weights. iii. With the new weights, evaluate the total error. iv. If the current total 77 error is increased as a result of the update, then retract the step (such as reset the weight vector to the precious 78 value) and increase combination coefficient? by a factor of 10 or by some other factors. Then go to step ii and 79 try an update again. v. If the current total error is decreased as a result of the update, then accept the step 80 (such as keep the new weight vector as the current one) and decrease the combination coefficient? by a factor of 81 10 or by the same factor as step iv. vi. Go to step ii with the new weights until the current total error is smaller 82 83 than the required value.

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⁸⁶ 12 Conclusion

To analyzing the speed of sending messages between the systems. Improving the quality and accuracy of the message sending in HL7 standard. Less time require exchanging data between systems. It can be based on real time application. Provide efficient and accurate data. Train LM algorithm easily implement in matlab and provide better result as compare to all other backpropagation algorithms. Fastest method for training moderate sized feed forward neural network. In the future we can also work on Dicom images to increase the speed of sending image fastly and best quality with use of this algorithm.

93 V.

94 13 Future Scope

- 95 In future work, also more improve the speed of sending message with some another network and also more
- distortion measures and feature domains will be used as the image samples. Also, the relationship between the
- 97 metrics adopted for the combination will be further investigated to find the best combination among them. More 98 experiments are needed to validate properties of the network such as it optimum number of neurons in hidden
- layers, validation etc. Performance comparison of LMBP with other networks should also be discussed.



Figure 1: M © 2013 D



Figure 2: D

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13 FUTURE SCOPE

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