Artificial Intelligence formulated this projection for compatibility purposes from the original article published at Global Journals. However, this technology is currently in beta. *Therefore, kindly ignore odd layouts, missed formulae, text, tables, or figures.* 

# An Empirical Investigation of Using ANN Based N-State Sequential Machine and Chaotic Neural Network in the Field of Cryptography Abhinav Tiwari<sup>1</sup> <sup>1</sup> Shri Ram Institute of Technology Jabalpur *Received: 13 December 2011 Accepted: 3 January 2012 Published: 15 January 2012*

# 8 Abstract

Cryptography is the exchange of information among the users without leakage of information 9 to others. Many public key cryptography are available which are based on number theory but 10 it has the drawback of requirement of large computational power, complexity and time 11 consumption during generation of key [1]. To overcome these drawbacks, we analyzed neural 12 network is the best way to generate secret key. In this paper we proposed a very new 13 approach in the field of cryptography. We are using two artificial neural networks in the field 14 of cryptography. First One is ANN based n-state sequential machine and Other One is chaotic 15 neural network. For simulation MATLAB software is used. This paper also includes an 16 experimental results and complete demonstration that ANN based n-state sequential machine 17 and chaotic neural network is successfully perform the cryptography. 18

19

20 Index terms— ANN, n-state Sequential Machine, Chaotic Neural Network, Cryptography.

# <sup>21</sup> 1 Introduction

n Artificial Neural Network (ANN) is an information processing paradigm that is inspired by the way biological 22 nervous systems, such as the brain, process An ANN is configured for a specific application, such as pattern 23 recognition or data classification, through a learning process [12]. Learning in biological systems involves 24 adjustments to the synaptic connections that exist between the neurons. [12]Cryptosystems are commonly used 25 for protecting the integrity, confidentiality, and authenticity of information resources. In addition to meeting 26 standard specifications relating to encryption and decryption, such systems must meet increasingly stringent 27 specifications concerning information security. This is mostly due to the steady demand to protect data and 28 resources from disclosure, to guarantee the authenticity of data, and to protect systems from web based attacks. 29 30 For these reasons, the development and evaluation of cryptographic algorithms is a challenging task.

This paper is an investigation of using ANN based n-state sequential machine and chaotic neural network in the field of cryptography .the rest of the paper is organized as follows: section 2 discusses background and related work in the field of ANN based cryptography, section 3 proposed method related to nstate sequential machine and chaotic neural network section 4 discusses implementation section 5 discusses experimental report and test result and finally section 6 discusses conclusion.

# 36 **2** II.

Background and related work Jason L. Wright, Milos Manic Proposed a research paper on Neural Network
Approach to Locating Cryptography in Object Code. In this paper, artificial neural networks are used to classify
functional blocks from a disassembled program as being either cryptography related or not. The resulting system,
referred to as NNLC (Neural Net for Locating Cryptography) is presented and results of applying this system to

referred to as NNLC (Neural Net for Locatin
various libraries are described [2].

### 4 A) SEQUENTIAL MACHINE

John Justin M, Manimurugan S introduced A Survey on Various Encryption Techniques. This paper focuses mainly on the different kinds of encryption techniques that are existing, and framing all the techniques together as a literature survey. Aim an extensive experimental study of implementations of various available encryption techniques. Also focuses on image encryption techniques, information encryption techniques, double encryption and Chaos-based encryption techniques. This study extends to the performance parameters used in encryption processes and analysing on their security issues [3].

Ilker DALKIRAN, Kenan DANIS, MAN introduced a research paper on Artificial neural network based chaotic 48 generator for cryptology. In this paper, to overcome disadvantages of chaotic systems, the dynamics of Chua's 49 circuit namely x, y and z were modeled using Artificial Neural Network (ANN). ANNs have some distinctive 50 capabilities like learning from experiences, generalizing from a few data and nonlinear relationship between 51 inputs and outputs. The proposed ANN was trained in diffrent structures using different learning algorithms. 52 To train the ANN, 24 different sets including the initial conditions of Chua's circuit were used and each set 53 consisted of about 1800 input-output data. The experimental results showed that a feed-forward Multi Layer 54 Perceptron (MLP), trained with Bayesian Regulation back propagation algorithm, was found as the suitable 55 network structure. As a case study, a message was first encrypted and then decrypted by the chaotic dynamics 56 57 obtained from the proposed ANN and a comparison was made between the proposed ANN and the numerical 58 solution of Chua's circuit about encrypted and decrypted messages ??5].

Eva Volna ,Martin Kotyrba ,Vaclav Kocian, Michal Janosek developed a Cryptography Based on Neural Network. This paper deals with using neural network in cryptography, e.g. designing such neural network that would be practically used in the area of cryptography. This paper also includes an experimental demonstration [6].

Karam M. Z. Othman , Mohammed H. Al Jammas Introduced Implementation of Neural -Cryptographic
System Using Fpga. In this work, a Pseudo Random Number Generator (PRNG) based on artificial Neural
Networks (ANN) has been designed. This PRNG has been used to design stream cipher system with high
statistical randomness properties of its key sequence using ANN. Software simulation has been build using
MATLAB to firstly, ensure passing four wellknown statistical tests that guaranteed randomness characteristics.
Secondly, such stream cipher system is required to be implemented using FPGA technology, therefore, minimum

69 hardware requirements has to be considered [7].

T. Schmidt, H. Rahnama Developed A Review of Applications of Artificial Neural Networks In Cryptosystems.
 This paper presents a review of the literature on the use of artificial neutral networks in cryptography. Different

<sup>71</sup> runs paper presents a review of the interactive on the use of artificial neutral networks in cryptography. Different <sup>72</sup> neural network based approaches have been categorized based on their applications to different components

of cryptosystems such as secret key protocols, visual cryptography, design of random generators, digital
watermarking, and steganalysis [8].

Wenwu Yu, Jinde Cao introduced Cryptography based on delayed chaotic neural networks. In this Letter, a novel approach of encryption based on chaotic Hopfield neural networks with time varying delay is proposed. We use the chaotic neural network to generate binary sequences which will be used for masking plaintext. The plaintext is masked by switching of chaotic neural network maps and permutation of generated binary sequences. Simulation results were given to show the feasibility and effectiveness in the proposed scheme of this Letter. As

a result, chaotic cryptography becomes more practical in the secure transmission of large multi-media files over
 public data communication network [9] III.

# <sup>82</sup> 3 PROPOSED METHOD

A number of studies have already investigated different machine learning methodologies, specifically neural networks and their applications in cryptography, but It is uncommon technique to using Artificial Neural network based n-state sequential machine and Chaotic neural network in the field of cryptography.

# <sup>86</sup> 4 a) Sequential Machine

A sequential Machine output depends on state of the machine as well as the input given to the sequential machine. 87 Therefore Michel I. Jordan Network was designed because in which output are treated as input. We are used these 88 type of input as a state. As a sequential machine can be achieved by using a Michel I. Jordan neural network, 89 therefore data are successfully encrypted and decrypted. In this case the starting state of the n-state sequential 90 91 machine can act as a key. Data is used to train the neural network as it provides the way the machine moves from 92 one state to another. c) Cryptography Achieved by a chaotic neural network Cryptography scheme was done by 93 a chaotic neural network. A network is called chaotic neural network if its weights and biases are determined 94 by chaotic sequence. Specially encryption of digital signal we used chaotic neural network. The reason for using sequential machine for implementation is that the output and input can have any type of relationship and the 95 output depends on the starting state. The starting state is used as a key for encryption and decryption. If the 96 starting state is not known, it is not possible to retrieve the data by decryption even if the state table or the 97 working of the sequential state is known. For training of the neural network, any type of sequential machine can 98 be used with the key showing the complexity or the level of security obtained. 99

#### c) Cryptography Achieved Through chaotic neural network 5 100

A chaotic neural network in which weights and biases are determined by a chaotic sequence. g = digital signal101 of length M and g(n) 0? M-1, be the one-byte value of the signal g at position n. The decryption procedure is 102 the same as the above one except that the input signal to the decryption Chaotic neural network should be g'(n) 103 and its output signal should be g"(n). 104 V.

105

#### Experiment and test result a) Sequential Machine 6 106

A general n-state Sequential Machine was implemented. As an example, the serial adder was implemented using 107 this machine. There is initial state is informed to the user for the input bits to be added. The output is the sum 108 and the carry bit. After that the execution of program has been completed it is automatically jumps to the new 109 carry state. This output is considered as previous carry state. 110

Following Graph illustrate Mean Square Error when we Enter input 0,1 and 2 .we apply this input in two feed 111 forward Adder One is of Multilayer single output feed forward Adder and other is Multilayer multiple output 112 feed forward Adder. 113

First we apply Input Number 0,1 and 2 on Multilayer single output feed forward and find MSE on Linear 114 scale. Year and generated encrypted letter similarly is the starting state is 1 the letter is shifted by 2. The state 115

is automatically move to next state .If the next input is again A the output will be C as the current state now 116

is 1. For H, state 0 will flip the letter to A while state 1 will flip the output to B. This method can be used to 117 1 2 encrypt a word containing only the letters A to H.



Figure 1: Fig. 3 . 1 :

118

<sup>&</sup>lt;sup>1</sup>© 2012 Global Journals Inc. (US) Global Journal of Computer Science and Technology  $^{2}$ © 2012 Global Journals Inc. (US)



Figure 2:

$$\begin{split} \Delta_{p}\omega_{jk} &= \gamma \delta_{k}^{p} y_{j}^{p} \begin{pmatrix} 4 - \lambda \\ k \text{ receiving the input and the outp} \\ k \text{ receiving the input and the outp} \\ \text{sending this signal along the constraints} \\ \delta_{0}^{p} &= (d_{0}^{p} - y_{0}^{p}) F'(s_{0}^{p}) \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} \delta_{0} \text{ error signal is given} \\ \hline y^{p} &= F(s^{p}) = \frac{1}{1 + e^{-s^{2}}} \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} F \text{ activation function} \\ \hline F'(s^{p}) &= y^{p}(1 - y^{p}) \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} Derivative is equal \\ \hline \delta_{0}^{p} &= (d_{0}^{p} - y_{0}^{p}) y_{0}^{p} (1 - y_{0}^{p}) \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} Derivative is equal \\ \hline \delta_{0}^{p} &= (d_{0}^{p} - y_{0}^{p}) y_{0}^{p} (1 - y_{0}^{p}) \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} Error signal for an outp \\ \hline \delta_{0}^{h} &= y_{h}^{p}(1 - y_{h}^{p}) \sum_{\alpha=1}^{N_{e}} \delta_{0}^{p} \omega_{h\alpha}) \\ \hline \delta_{0}^{h} &= y_{h}^{p}(1 - y_{h}^{p}) \sum_{\alpha=1}^{N_{e}} \delta_{0}^{p} \omega_{h\alpha}) \\ \hline \delta_{0}^{\mu} &= \gamma \delta_{k}^{p} y_{j}^{p} \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} \text{ Sigmoid activation function} \\ \hline \delta_{0}^{\mu} &= y_{h}^{p}(1 - y_{h}^{p}) \sum_{\alpha=1}^{N_{e}} \delta_{0}^{p} \omega_{h\alpha}) \\ \hline \delta_{0}^{\mu} &= \gamma \delta_{k}^{p} y_{j}^{p} \begin{pmatrix} 4 - \lambda \\ k - \nu \end{pmatrix} \text{ Gradient descent on the error signal}$$

Figure 3:



Figure 4: Fig. 5. 3 :

# Command Window

Enter	The Number Of INPUT 2
Enter	The Number Of OUTPUT1
Enter	The Number Of State 2
Enter	INPUT And STATE[0 0 0]
Enter	OUTPUT And STATE[0 0]
Enter	INPUT And STATE[0 1 0]
Enter	OUTPUT And STATE[1 0]
Enter	INPUT And STATE[1 0 0]
Enter	OUTPUT And STATE[1 0]
Enter	INPUT And STATE[1 1 0]
Enter	OUTPUT And STATE[0 1]
Enter	INPUT And STATE[0 1 1]
Enter	OUTPUT And STATE[0 1]
Enter	INPUT And STATE[1 0 1]
Enter	OUTPUT And STATE[0 1]
Enter	INPUT And STATE[1 1 1]
Enter	OUTPUT And STATE[1 1]
Enter	INPUT And STATE[0 0 1]
Enter	OUTPUT And STATE[0 1]

4

Figure 5: Fig. 5. 4 :



Figure 6:

 $\mathbf{5}$ 

Figure 7: Table 5 .

Conditions (Values of $x(0)$ and $\mu()$ )
Same Input Encrypted with Different Initial Conditions
(Values of $x(0)$ and $\mu()$ )

		Output with	Output with	output with
	ASCII	x(0) = 0.75	x(0) = 0.85	x(0) = 0.90
INPUT	CODE	µ=3.9	$\mu = 3.5$	$\mu = 3.2$
А	97	199	233	204
В	98	195	112	98
$\mathbf{C}$	99	200	239	11
D	100	253	108	31
E	101	220	226	1
F	102	17	115	25
G	103	187	234	56
Η	104	101	109	235
Ι	105	6	236	49
J	106	138	115	226
Κ	107	107	229	36
$\mathbf{L}$	108	32	110	225
Μ	109	180	238	42
Ν	110	119	112	254
0	111	225	224	45
Р	112	184	113	225
Q	113	63	243	49
R	114	168	83	227
S	115	103	252	51
Т	116	245	116	229
U	117	160	244	53
V	118	83	84	231
W	119	209	248	55
Х	120	219	120	233
Υ	121	209	248	57
Ζ	122	231	88	235

Figure 8: Table 1 :

Output Obtained Using

( D D D D ) D

# $\mathbf{2}$

# Decrypted Using Same and Different Initial Conditions Encrypted Data of Table 1 (Column 2) Decrypted Using Same and Different Initial Conditions

Output Obtained Using Same Initial Condition

Same Initial Condition			Different Initial Condition	
		output with	output with	output with
	ASCII	x(0) = 0.75	x(0) = 0.85	x(0) = 0.90
INPUT	CODE	$\mu = 3.9$	$\mu = 3.5$	$\mu = 3.2$
А	199	97	79	106
В	195	98	209	195
С	200	99	68	160
D	253	100	245	134
Ε	220	101	91	184
F	17	102	4	110
G	187	103	54	228
Н	101	104	96	230
Ι	6	105	131	94
J	138	106	147	2
Κ	107	107	229	36
L	32	108	34	173
М	180	109	55	243
Ν	119	110	105	231
0	225	111	110	163
Р	184	112	185	41
Q	63	113	189	127
R	168	114	137	57
S	103	115	232	39
ΤU	$245 \ 160$	$116 \ 117$	245 33	$100 \ 224$
V	83	118	113	194
W	209	119	94	145
Х	219	120	219	74
Υ	209	121	80	145
Z	231	122	197	118

Figure 9: Table 2 :

# Decrypted Using Same and Different Initial Conditions Encrypted Data of Table 1 (Column 3) Decrypted Using Same and Different Initial Conditions

Output Obtained Using			Output Obtained Using	
Same Initial Condition			Different Initial Con	dition
		output	output with	output with
		with		
	ASCII	x(0) = 0.75	x(0) = 0.85	x(0) = 0.90
INPUT	CODE	$\mu = 3.9$	$\mu = 3.5$	$\mu = 3.2$
А	233	79	97	68
В	112	209	98	112
С	239	68	99	135
D	108	245	100	23
Ε	226	91	101	134
F	115	4	102	12
G	234	54	103	181
Н	109	96	104	238
Ι	236	131	105	180
J	115	147	106	251
К	229	229	107	170
L	110	34	108	227
М	238	55	109	169
Ν	112	105	110	224
0	224	110	111	162
Р	113	185	112	224
Q	243	189	113	179
R	83	137	114	194
S	252	232	115	188
Т	116	245	116	229
U	244	33	117	180
V	84	113	118	197
W	248	94	119	184
Х	120	219	120	233
Υ	248	80	121	184
Ζ	88	197	122	201

Figure 10: Table 3 :

# 3

 $\mathbf{4}$ 

## Decrypted Using Same and Different Initial Conditions Encrypted Data of Table 1 (Column 4) Decrypted Using Same and Different Initial Conditions

Output Obtained Using Same Initial Condition	output with		Output Obtained Using Different Initial Condition output with	
	ASCII	x(0) = 0.75	x(0) = 0.85	$\operatorname{output}$ with
	0055			x(0) =
INPUT	CODE	$\mu = 3.9$	$\mu = 3.5$	0.90
A	204	100	69	$\mu = 3.2$
A	204	100	08	97
В	98	195	112	98
	11	160	135	99
D	31	134	23	100
E		184	134	101
F'	25	110	12	102
G	56	228	181	103
Н	235	230	238	104
Ι	49	94	180	105
J	226	2	251	106
Κ	36	36	170	107
L	225	173	227	108
М	42	243	169	109
Ν	254	231	224	110
0	45	163	162	111
Р	225	41	224	112
Q	49	127	179	113
R	227	57	194	114
S	51	39	188	115
Т	229	100	229	116
U	53	224	180	117
V	231	194	197	118
W	55	145	184	110
X	233	74	233	120
V	57	145	184	191
7	01 935	118	201	199
			201	144

It is clear from table 2, 3 and 4 that we can

decrypt an encrypted data correctly by knowing the exact values of  $$\mathbf{x}$$ 

VI.

x (0) and  $\mu$  otherwise we get the

wrong data as shown in table 2,3 and 4.

Conclusion

Figure 11: Table 4 :

- 119 [Volna et al.], Eva Volna, Martin Kotyrba, Vaclav Kocian, Michal Janosek -Cryptography Based On Neural
- Network . Ostrava; Czech Republic. 702 p. 0. Department of Informatics and Computers University of Ostrava
   Dvorakova 7
- [Karam M et al. ()], Z Karam M, Mohammed H Othman, Jammas -Implementation Of Neural -Cryptographic
   System Using Al, Fpga. journal of Engineering Science and Technology 2011. 6 (4) p. . © School of
- Engineering, Taylor's University
- [Yen and Guo (1999)] 'A New Image Encryption Algorithm and Its VLSI Architecture'. J C Yen , J I Guo .
   *Grand Hotel*, (Taipei, Taiwan) 1999. Oct. 18-22. 1999. p. .
- ISrividya and Nandakumar (2011)] 'A Triple-Key chaotic image encryption method'. G Srividya, P Nandakumar
   Communications and Signal Processing 2011. Feb. 2011. p. .
- [Stuart J. Russell and Peter Norvig (ed.)] Artificial Intelligence A Modern Approach, Stuart J. Russell and Peter
   Norvig (ed.)
- IIker and Kenan ()] 'Artificial neural network based chaotic generator for cryptology'. Dalkiran Ilker, Danis, man
   Kenan. Turk J Elec Eng & Comp Sci 2010. 18 (2) pp. © T"UB?ITAK.
- [Daniel C. Biederman and Esther Ososanya (ed.)] Capacity of Several Neural Networks With Respect to Digital
   Adder and Multiplier, Daniel C. Biederman and Esther Ososanya (ed.)
- [Michie et al. (1994)] 'Ciletti 21'. D Michie , D J Spiegelhalter , C C Taylor . Machine Learning, Neural and
   Statistical Classification, M, Morris Mano, D Michael (ed.) February 17. 1994. (Digital design)
- 137 [Godhavari (2005)] 'Cryptography using neural network'. T Godhavari . IEEE Indicon 2005 Conference,
   138 (Chennai, India) Dec. 2005. p. .
- 139 [Shweta et al. (2011)] 'Design and Realization of A New Chaotic Neural Encryption/Decryption Network'. B
- Shweta, Suryawanshi, Devesh D Nawgaje. *IJCA Proceedings on 2nd National Conference on Information and Communication Technology NCICT*, Scott Su, Alvin Lin, Jui-Cheng Yen (ed.) (New York, USA) November

142 2011. 17. (Chaotic Neural Network for Cryptography in Image Processing)

- [Wright] Milos Manic Neural Network Approach to Locating Cryptography in Object Code. Emerging Technologies
   and Factory Automation INL Laboratory, Jason L Wright .
- <sup>145</sup> [Shweta et al. (2012)] 'Nawgaje-a triple-key chaotic neural network for cryptography in image processing'. B
- Shweta , Suryawanshi , D Devesh . International Journal of Engineering Sciences & Emerging Technologies
   2231 -6604. April 2012. 2 (1) p. .
- [Shweta et al. (2012)] 'Nawgaje-a triple-key chaotic neural network for cryptography in image processing'. B
   Shweta , Suryawanshi , D Devesh . International Journal of Engineering Sciences & Emerging Technologies
   2231 -6604. April 2012. 2 (1) p. .
- [Wasserman ()] Neural Computing, Theory and Practice, Philip D Wasserman . 1989. New York: Van Nordstrand
   Reinhold.
- [Schmidt and Dept] of computer science, ryerson university, canada -a review of applications of artificial neural
   networks in cryptosystems, T Schmidt , Dept .
- [Boyd ; C. Alexopoulos ()] 'Picture Data Encryption Using SC4N Pattern'. C Boyd ; C. Alexopoulos . *Electronics & Communication Journal* Oct. 1993. 131. 1992. 25 (6) p. . (Pattern Recognition)
- [Kumar et al. ()] 'Ratan Singh -Modeling and Simulation of Backpropogation Algorithm Using VHDL'. Jay
   Kumar , Ankit Sinha , Guncha Goswami , Manisha Kumari . International Journal of Computer Applications
   in Engineering Sciences 2231-4946. JUNE 2011. II.
- [Laskari et al. ()] 'Studying the performance of artificial neural networks on problems related to cryptography'.
   E C Laskari , G C Meletiou , D K Tasoulis , M N Vrahatis . Nonlinear Analysis: Real World Applications 2006. 7 p. .
- <sup>163</sup> [Justin and Manimurugan (2011)] 'Survey on Various Encryption Techniques'. John Justin , M Manimurugan ,
- S -A. Tripatjot Singh Panag CRYPTOGRAPHY USING CHAOTIC NEURAL NETWORK International Journal of Information Technology 2231-2307. March 2012 5. Management July-December 2011. 4 (2) p. .
   (International Journal of Soft Computing and Engineering (IJSCE))
- [Miles et al. ()] 'The Data Encryption Standard: Past and Future'. E Miles , Dennis K Smid , Branstad .
   *proceedings of the ieee* 1988. 76 (5) p. .
- [Smid and Branstad ()] 'The Data Encryption Standard: Past and Future'. M E Smid , D K Branstad .
   Proceedings of The IEEE 1988. 76 (5) p. .
- [Yu and Bishop (ed.) (2006)] Wenwu Yu . Jinde Cao -Cryptography based on delayed chaotic neural networks
   Department of Mathematics, A R Bishop (ed.) (Nanjing) 210096. 1 February 2006. 10 March 2006. March
   2006 Available online 17 April 2006Communicated. Southeast University (received in revised)