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The Production Model of Fuzzy Neural Network in Information Security Systems

Zyryanov S. I.¹, Berezhnov V. P.², Perepelkina Yu. V.³, Shevtsov A. I.⁴, Shevchenko K. K.⁵, Kozlov V. V.⁶

¹Department of Higher mathematics and natural science, synergy University, Moscow, Russia

²Department of Higher mathematics and natural science, synergy University, Moscow, Russia

³Department of Higher mathematics and natural science, synergy University, Moscow, Russia

⁴Department of Higher mathematics and natural science, synergy University, Moscow, Russia ⁵Department of Higher mathematics and natural science, synergy University, Moscow, Russia

⁶Department of Higher mathematics and natural science, synergy University, Moscow, Russia

ABSTRACT

In this article, a model for evaluating the level of information security based on a fuzzy neural production network is proposed. The methodology of qualitative assessment of the level of information security in the system is based on the results of measurements and expert assessments, which may be unclear and insufficiently expressed in order to be described by mathematical dependencies. The functional status of such systems can be described using fuzzy rule constructs. Fuzzy production networks are identical in their structure to multilayered neural networks, and this property was applied by the authors to construct a neural network model for evaluating the level of information security. The concept of technological security profiles is introduced as a set of security States that correspond to the identified technical channels of information leakage at a certain time. It is proposed to rank these channels by importance before processing in the neuro marketing system.

The results obtained allow us to formulate directions for further research on the development of new effective information protection systems using intelligent technologies.

Key words : information security, fuzzy production models, neuron, neuronal network, technological portrait of security, technical channel of information leakage.

1. INTRODUCTION

A set of security States that correspond to the technical channels of information leakage detected on the OID at a certain time can be represented as dynamic systems. These States are called events and are represented as technological portraits of the protected environment [1-4]. This term, which covers the corresponding concepts of a physical, informational and secure nature, refers to the occupation of a dynamic distributed network of a certain technological state or configuration of States [5]. The evolutionary changes that

take place between stages are not taken into account and it is assumed that the dynamics of the system develops discretely, from event to event. Such a system is discrete-mode [5].

The environment for a neural network system (NMS) to assess the level of information security can be represented as a set of discrete-mode systems with associated discrete technological States of security [6-9].

Getting the necessary number of instrumental measurements and special studies for each of the technical channels of information leakage received on the OID, neo needs to develop such a procedure for processing measurements, which allows you to automatically receive information about the technological state of security of each of the channels according to the approved threat model [10].

2. MATERIALS AND METHODS

The results of instrumental measurements and special ossiers for each of the technical channels of information leakage are perceived by the sensor matrix in the form of a set of observations [11-13]:

X =(X X1, 2,..., Xi,..., Xm), and = 1,2,...,n.

In a separate sensor channel, a reduction of the sample space X occurs, y resulting in a sequence of discrete variables Uk, k = 0, 1, ..., n-1, which take the values Z1, Z2,..., Zr.

It is necessary to synthesize the structure of a neuropod-like classifier that implements the crucial function $\gamma(U)$ on the reduced sample space U [6].

The sequence of discrete variables Ur, k = 0,1,..., n-1, which take the value z, a = 1,2,..., r, can be approximated by vectors Ξ , $f(0)\mu$ and Ξ ,

F (k)µ.

The structure of the simplest neural network-like security assessment system is a set of M+1 ensembles of first-layer neural networks. The ensemble consists of n-neurons, whose excitation level is defined as n [14]

 $Y\mu(k)=\sum \Xi a(k)Fa(k)\mu a=1$,

where $\Xi,\,f(0)\mu$ and $\Xi,\,f(k)\mu.$ - vectors that can be used to approximate a sequence of discrete variables Ur, k=0,1,...,

n-1, takes the values za, a = 1, 2, ..., r [5].

Each neuron carries out encoding process, which is determined by the so-called method of labeled lines, wherein the particular value the process provided in accordance specific (labeled) line Z1,Z2,...,Z,a,...,Zk , and hence the specified value of the process parameter corresponds to one maximum excited synaptics the connection a (k) =1 [15].



Figure 1: Fuzzy logic neural network-based in information security system

Figure 1 shows us fuzzy logic neural network. In contrast to a typical neuron, whose synaptic connections are equivalent, a neuron that encodes by the method of labeled lines has priority synaptic connections. A synaptic input with a larger number corresponds to a larger value of the process parameter. Another difference is that k only one synaptic link is triggered at the k-th instant of time, and thus the task of entering and controlling the threshold Θ , using the weight function w, is greatly simplified w [16-18].

For each technical channel in the set {ml}TKVI it is necessary to determine its importance [8].

Despite the difference in the number of threats in accordance with [5], which define each possible technical channel, this number does not determine the degree of danger of the technical channel itself. Therefore, to determine the importance coefficients, it is rational to use the ratio between the total number of threats of a particular channel and the number of threats that took the value "1" according to the results of an expert survey [19].

Then the formula for determining the importance coefficients of each technical channel in the list looks like this:

M1 ll = Mll , l = 1..L

where Ml1 - the number of threats of the l-th technical channel, which took the value "1" according to the results of an expert survey; Ml-the total number of threats that characterize the l-th technical channel [20].

Thus, a set of values for the importance of each possible technical channel is formed information leaks - $\{\lambda l\}$ TKVI, l =1,L. The synaptic input with a high number corresponds to the value of the parameter of the technical channel with a high rating.

It is necessary to conduct an inventory of technical channels of importance and form a final list of possible technical channels for information leakage to the OID [21-24]:

 $\{ml\}TKVI, l=1, L where L \leq 8.$

The proposed approaches allow us to formalize, for example, the results of further research on the development of new effective information security systems using intelligent technologies.



Figure 2: Training of neural network for security dimension

Figure 2 shows us training of neural network and algorithms. The difference between it and existing systems for assessing the level of security that the neural network structure is focused on solving a specific task – creating

and certifying an object of information activity or creating a comprehensive information security system in its. The requirement of problem orientation of the neural network (NM) leads to the implementation of the principle of adequacy of its structure and external environment, i.e. the possibility of flexible structural and functional adjustment [25]. This determines the most important property of NM-adaptability to changes in the environment of its functioning, which is very relevant for the complexityof the structure of modern its. The initial structuring of the neural network should be carried out by formal synthesis methods, which determine the optimal structure, including the number of neural layers and neural ensembles, the number of neural-like elements in each layer, and the presence of deterministic elements links between them and the initial weight coefficients [22].

3. CONCLUSION

The current novelty of the work is to expand the understanding of cloud computing, review the requirements for both methods and forms of organizing scientific research, preserving scientific information and sharing it, and in the role of a scientist and his professional ICT competencein organizing scientific work.

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