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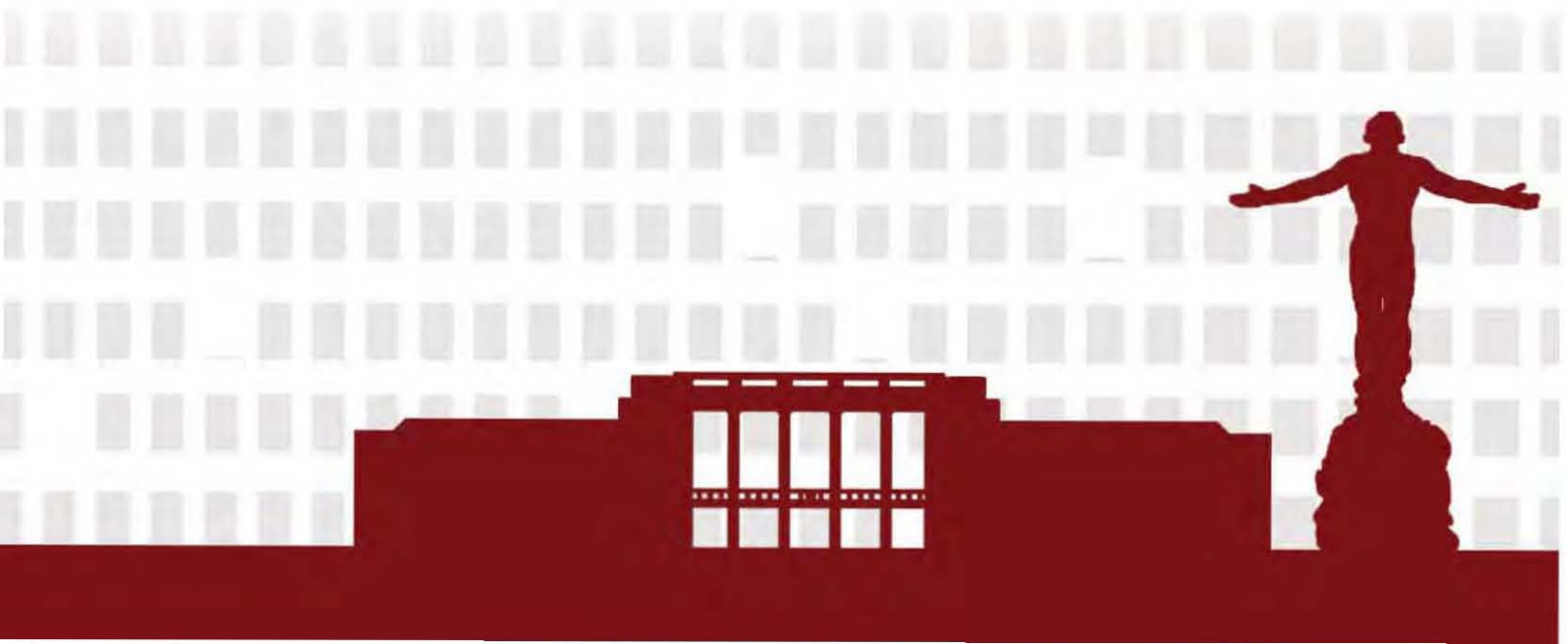

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SOLUTION TO THE *SUBSET SUM* PROBLEM USING THE FRAMEWORK OF SPIKING NEURAL P SYSTEMS WITH STRUCTURAL PLASTICITY

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ABSTRACT

Spiking neural P systems (in short, SNP systems) are parallel, distributed models of computations based on the structure and function of neural cells or *neurons*. Neurons process only a single type of signal or object known as the *spike*. The neurons are placed on vertices of a directed graph, where each edge in the graph is called a *synapse*. Information cannot be discerned from the spikes, as spikes are *indistinct signals*. Instead, information is obtained from the time intervals between spikes, or the presence (absence) of spikes at certain time steps. Time therefore is a means to *encode information*, rather than simply being a background of the computations. It is known that SNP systems and their variants are *Turing universal*, i.e. they can simulate any Turing machine, and thus can carry out any effective computation that we know of.

Since the introduction of SNP systems in 2006 (see [3]), many theoretical and practical problems have been solved using SNP systems. See e.g. [4] and the SNP systems chapter in [5]. In this extended abstract we use the variant known as *SNP systems with structural plasticity* (in short, SNPSP systems). SNPSP systems were introduced in [6] to include the neuroscience feature of strutural plasticity in the SNP systems framework. In SNPSP systems, *plasticity rules* allow neurons to create or delete synapses.

We use SNPSP systems in this work to provide a constant time, nondeterministic solution to the *Subset sum* problem. This problem is a well known computationally hard problem with important use in cryptography. The hardness of the *Subset sum* problem is applied to practical use in order to secure many systems requiring encryption, see e.g. [1,2]. Briefly, the *Subset sum* problem has as its inputs a set of natural numbers $V = \{v1, v2, \ldots, vn\}$ and a natural number S. The task is to find a subset B of V where the elements of B sum exactly to S, see e.g. [7].

An SNPSP system solving *Subset sum* is given in graphical form in Figure 1. Using plasticity rules (see Figure 1), we are able to reduce the number of neurons in our system by a linear amount (with respect to problem input size n) compared to the number of neurons in the SNP system given in [8].

Keywords: Spiking neural P systems, Structural plasticity, Subset sum

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