

cases	doc_1		doc_2		decision	id
					DUPLICATES	326
			authors	<ul style="list-style-type: none">Xavier LagorceRyad Benosman		
	authors	<ul style="list-style-type: none">Xavier LagorceR. Benosman	title	STICK: Spike Time Interval Computational Kernel, A Framework for General Purpose Computation using Neurons, Precise Timing, Delays, and Synchrony		
	title	STICK: Spike Time Interval Computational Kernel, a Framework for General Purpose Computation Using Neurons, Precise Timing, Delays, and Synchrony	publication_date	2015-07-22 15:09:07+00:00		
	publication_date	2015-07-22 00:00:00	source	SupportedSources.ARXIV		
	source	SupportedSources.SEMANTIC_SCHOLAR	journal	None		
	journal	Neural Computation	volume			
	volume	27	doi			
	doi	10.1162/NECO_a_00783	urls	<ul style="list-style-type: none">http://arxiv.org/pdf/1507.06222v1http://arxiv.org/abs/1507.06222v1http://arxiv.org/pdf/1507.06222v1		
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	id	id368572185479275467	abstract	There has been significant research over the past two decades in developing new platforms for spiking neural computation. Current neural computers are primarily developed to mimic biology. They use neural networks, which can be trained to perform specific tasks to mainly solve pattern recognition problems. These machines can do more than simulate biology; they allow us to rethink our current paradigm of computation. The ultimate goal is to develop brain-inspired general purpose computation architectures that can breach the current bottleneck introduced by the von Neumann architecture. This work proposes a new framework for such a machine. We show that the use of neuron-like units with precise timing representation, synaptic diversity, and temporal delays allows us to set a complete, scalable compact computation framework. The framework provides both linear and nonlinear operations, allowing us to represent and solve any function. We show usability in solving real use cases from simple differential equations to sets of nonlinear differential equations leading to chaotic attractors.		
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