

Wulfram Gerstner

List of Publications by Year in descending order

Source: <https://exaly.com/author-pdf/4475771/publications.pdf>

Version: 2024-02-01

129
papers

18,434
citations

26567

56
h-index

22102

113
g-index

150
all docs

150
docs citations

150
times ranked

9459
citing authors

#	ARTICLE	IF	CITATIONS
1	Brain signals of a Surprise-Actor-Critic model: Evidence for multiple learning modules in human decision making. <i>NeuroImage</i> , 2022, 246, 118780.	2.1	4
2	Learning in Volatile Environments With the Bayes Factor Surprise. <i>Neural Computation</i> , 2021, 33, 269-340.	1.3	17
3	Novelty is not surprise: Human exploratory and adaptive behavior in sequential decision-making. <i>PLoS Computational Biology</i> , 2021, 17, e1009070.	1.5	18
4	A functional model of adult dentate gyrus neurogenesis. <i>ELife</i> , 2021, 10, .	2.8	6
5	Rapid suppression and sustained activation of distinct cortical regions for a delayed sensory-triggered motor response. <i>Neuron</i> , 2021, 109, 2183-2201.e9.	3.8	46
6	When shared concept cells support associations: Theory of overlapping memory engrams. <i>PLoS Computational Biology</i> , 2021, 17, e1009691.	1.5	16
7	On the choice of metric in gradient-based theories of brain function. <i>PLoS Computational Biology</i> , 2020, 16, e1007640.	1.5	7
8	Mesoscopic population equations for spiking neural networks with synaptic short-term plasticity. <i>Journal of Mathematical Neuroscience</i> , 2020, 10, 5.	2.4	21
9	Dendritic Voltage Recordings Explain Paradoxical Synaptic Plasticity: A Modeling Study. <i>Frontiers in Synaptic Neuroscience</i> , 2020, 12, 585539.	1.3	7
10	Biologically plausible deep learning – But how far can we go with shallow networks?. <i>Neural Networks</i> , 2019, 118, 90-101.	3.3	71
11	How single neuron properties shape chaotic dynamics and signal transmission in random neural networks. <i>PLoS Computational Biology</i> , 2019, 15, e1007122.	1.5	29
12	Stability of working memory in continuous attractor networks under the control of short-term plasticity. <i>PLoS Computational Biology</i> , 2019, 15, e1006928.	1.5	57
13	Optimal Stimulation Protocol in a Bistable Synaptic Consolidation Model. <i>Frontiers in Computational Neuroscience</i> , 2019, 13, 78.	1.2	5
14	One-shot learning and behavioral eligibility traces in sequential decision making. <i>ELife</i> , 2019, 8, .	2.8	16
15	Multicontact Co-operativity in Spike-Timing-Dependent Structural Plasticity Stabilizes Networks. <i>Cerebral Cortex</i> , 2018, 28, 1396-1415.	1.6	21
16	Balancing New against Old Information: The Role of Puzzlement Surprise in Learning. <i>Neural Computation</i> , 2018, 30, 34-83.	1.3	56
17	Multi-Timescale Memory Dynamics Extend Task Repertoire in a Reinforcement Learning Network With Attention-Gated Memory. <i>Frontiers in Computational Neuroscience</i> , 2018, 12, 50.	1.2	3
18	Eligibility Traces and Plasticity on Behavioral Time Scales: Experimental Support of NeoHebbian Three-Factor Learning Rules. <i>Frontiers in Neural Circuits</i> , 2018, 12, 53.	1.4	174

#	ARTICLE	IF	CITATIONS
19	Excitable neuronal assemblies with adaptation as a building block of brain circuits for velocity-controlled signal propagation. <i>PLoS Computational Biology</i> , 2018, 14, e1006216.	1.5	12
20	Hebbian plasticity requires compensatory processes on multiple timescales. <i>Philosophical Transactions of the Royal Society B: Biological Sciences</i> , 2017, 372, 20160259.	1.8	151
21	The temporal paradox of Hebbian learning and homeostatic plasticity. <i>Current Opinion in Neurobiology</i> , 2017, 43, 166-176.	2.0	138
22	Exponentially Long Orbits in Hopfield Neural Networks. <i>Neural Computation</i> , 2017, 29, 458-484.	1.3	3
23	Cortical Dynamics in Presence of Assemblies of Densely Connected Weight-Hub Neurons. <i>Frontiers in Computational Neuroscience</i> , 2017, 11, 52.	1.2	22
24	Towards a theory of cortical columns: From spiking neurons to interacting neural populations of finite size. <i>PLoS Computational Biology</i> , 2017, 13, e1005507.	1.5	112
25	Predicting non-linear dynamics by stable local learning in a recurrent spiking neural network. <i>ELife</i> , 2017, 6, .	2.8	58
26	A Model of Synaptic Reconsolidation. <i>Frontiers in Neuroscience</i> , 2016, 10, 206.	1.4	9
27	Enhanced Sensitivity to Rapid Input Fluctuations by Nonlinear Threshold Dynamics in Neocortical Pyramidal Neurons. <i>PLoS Computational Biology</i> , 2016, 12, e1004761.	1.5	32
28	Does computational neuroscience need new synaptic learning paradigms?. <i>Current Opinion in Behavioral Sciences</i> , 2016, 11, 61-66.	2.0	28
29	Nonlinear Hebbian Learning as a Unifying Principle in Receptive Field Formation. <i>PLoS Computational Biology</i> , 2016, 12, e1005070.	1.5	46
30	Automated High-Throughput Characterization of Single Neurons by Means of Simplified Spiking Models. <i>PLoS Computational Biology</i> , 2015, 11, e1004275.	1.5	68
31	Synaptic Consolidation: From Synapses to Behavioral Modeling. <i>Journal of Neuroscience</i> , 2015, 35, 1319-1334.	1.7	42
32	Diverse synaptic plasticity mechanisms orchestrated to form and retrieve memories in spiking neural networks. <i>Nature Communications</i> , 2015, 6, 6922.	5.8	268
33	Neuromodulated Spike-Timing-Dependent Plasticity, and Theory of Three-Factor Learning Rules. <i>Frontiers in Neural Circuits</i> , 2015, 9, 85.	1.4	233
34	Stochastic variational learning in recurrent spiking networks. <i>Frontiers in Computational Neuroscience</i> , 2014, 8, 38.	1.2	56
35	Spike-timing prediction in cortical neurons with active dendrites. <i>Frontiers in Computational Neuroscience</i> , 2014, 8, 90.	1.2	30
36	Limits to high-speed simulations of spiking neural networks using general-purpose computers. <i>Frontiers in Neuroinformatics</i> , 2014, 8, 76.	1.3	55

#	ARTICLE	IF	CITATIONS
37	Fluctuations and information filtering in coupled populations of spiking neurons with adaptation. <i>Physical Review E</i> , 2014, 90, 062704.	0.8	32
38	Connection-type-specific biases make uniform random network models consistent with cortical recordings. <i>Journal of Neurophysiology</i> , 2014, 112, 1801-1814.	0.9	12
39	Optimal Control of Transient Dynamics in Balanced Networks Supports Generation of Complex Movements. <i>Neuron</i> , 2014, 82, 1394-1406.	3.8	259
40	Temporal whitening by power-law adaptation in neocortical neurons. <i>Nature Neuroscience</i> , 2013, 16, 942-948.	7.1	164
41	Synaptic Plasticity in Neural Networks Needs Homeostasis with a Fast Rate Detector. <i>PLoS Computational Biology</i> , 2013, 9, e1003330.	1.5	144
42	Reinforcement Learning Using a Continuous Time Actor-Critic Framework with Spiking Neurons. <i>PLoS Computational Biology</i> , 2013, 9, e1003024.	1.5	121
43	Inference of neuronal network spike dynamics and topology from calcium imaging data. <i>Frontiers in Neural Circuits</i> , 2013, 7, 201.	1.4	82
44	The Silent Period of Evidence Integration in Fast Decision Making. <i>PLoS ONE</i> , 2013, 8, e46525.	1.1	9
45	Reward-based learning under hardware constraints using a RISC processor embedded in a neuromorphic substrate. <i>Frontiers in Neuroscience</i> , 2013, 7, 160.	1.4	27
46	Changing the responses of cortical neurons from sub- to suprathreshold using single spikes in vivo. <i>ELife</i> , 2013, 2, e00012.	2.8	26
47	Paradoxical Evidence Integration in Rapid Decision Processes. <i>PLoS Computational Biology</i> , 2012, 8, e1002382.	1.5	17
48	Coding and Decoding with Adapting Neurons: A Population Approach to the Peri-Stimulus Time Histogram. <i>PLoS Computational Biology</i> , 2012, 8, e1002711.	1.5	42
49	The Performance (and Limits) of Simple Neuron Models: Generalizations of the Leaky Integrate-and-Fire Model. , 2012, , 163-192.		7
50	Microcircuits of excitatory and inhibitory neurons in layer 2/3 of mouse barrel cortex. <i>Journal of Neurophysiology</i> , 2012, 107, 3116-3134.	0.9	207
51	Parameter extraction and classification of three cortical neuron types reveals two distinct adaptation mechanisms. <i>Journal of Neurophysiology</i> , 2012, 107, 1756-1775.	0.9	91
52	Theory and Simulation in Neuroscience. <i>Science</i> , 2012, 338, 60-65.	6.0	141
53	Perceptual learning, roving and the unsupervised bias. <i>Vision Research</i> , 2012, 61, 95-99.	0.7	28
54	Improved Similarity Measures for Small Sets of Spike Trains. <i>Neural Computation</i> , 2011, 23, 3016-3069.	1.3	37

#	ARTICLE	IF	CITATIONS
55	A history of spike-timing-dependent plasticity. <i>Frontiers in Synaptic Neuroscience</i> , 2011, 3, 4.	1.3	311
56	Extraction of Network Topology From Multi-Electrode Recordings: Is there a Small-World Effect?. <i>Frontiers in Computational Neuroscience</i> , 2011, 5, 4.	1.2	93
57	Synaptic tagging and capture: a bridge from molecular to behaviour. <i>BMC Neuroscience</i> , 2011, 12, .	0.8	0
58	Connectivity reflects coding: a model of voltage-based STDP with homeostasis. <i>Nature Neuroscience</i> , 2010, 13, 344-352.	7.1	517
59	STDP in Adaptive Neurons Gives Close-To-Optimal Information Transmission. <i>Frontiers in Computational Neuroscience</i> , 2010, 4, 143.	1.2	23
60	Voltage and spike timing interact in STDP - a unified model. <i>Frontiers in Synaptic Neuroscience</i> , 2010, 2, 25.	1.3	72
61	From Hebb Rules to Spike-Timing-Dependent Plasticity: A Personal Account. <i>Frontiers in Synaptic Neuroscience</i> , 2010, 2, 151.	1.3	7
62	Functional Requirements for Reward-Modulated Spike-Timing-Dependent Plasticity. <i>Journal of Neuroscience</i> , 2010, 30, 13326-13337.	1.7	121
63	Spike-Based Reinforcement Learning in Continuous State and Action Space: When Policy Gradient Methods Fail. <i>PLoS Computational Biology</i> , 2009, 5, e1000586.	1.5	82
64	How Good Are Neuron Models?. <i>Science</i> , 2009, 326, 379-380.	6.0	220
65	Stress, genotype and norepinephrine in the prediction of mouse behavior using reinforcement learning. <i>Nature Neuroscience</i> , 2009, 12, 1180-1186.	7.1	68
66	Is there a geometric module for spatial orientation? Insights from a rodent navigation model.. <i>Psychological Review</i> , 2009, 116, 540-566.	2.7	100
67	Phenomenological models of synaptic plasticity based on spike timing. <i>Biological Cybernetics</i> , 2008, 98, 459-478.	0.6	455
68	Extracting non-linear integrate-and-fire models from experimental data using dynamic I-V curves. <i>Biological Cybernetics</i> , 2008, 99, 361-370.	0.6	65
69	The quantitative single-neuron modeling competition. <i>Biological Cybernetics</i> , 2008, 99, 417-426.	0.6	103
70	Firing patterns in the adaptive exponential integrate-and-fire model. <i>Biological Cybernetics</i> , 2008, 99, 335-347.	0.6	250
71	Special issue on quantitative neuron modeling. <i>Biological Cybernetics</i> , 2008, 99, 237-239.	0.6	12
72	A benchmark test for a quantitative assessment of simple neuron models. <i>Journal of Neuroscience Methods</i> , 2008, 169, 417-424.	1.3	121

#	ARTICLE	IF	CITATIONS
73	Dynamic $I-V$ Curves Are Reliable Predictors of Naturalistic Pyramidal-Neuron Voltage Traces. <i>Journal of Neurophysiology</i> , 2008, 99, 656-666.	0.9	183
74	Tag-Trigger-Consolidation: A Model of Early and Late Long-Term-Potential and Depression. <i>PLoS Computational Biology</i> , 2008, 4, e1000248.	1.5	110
75	Gamma Oscillations in a Nonlinear Regime: A Minimal Model Approach Using Heterogeneous Integrate-and-Fire Networks. <i>Neural Computation</i> , 2008, 20, 2973-3002.	1.3	25
76	Spike-triggered averages for passive and resonant neurons receiving filtered excitatory and inhibitory synaptic drive. <i>Physical Review E</i> , 2008, 78, 011914.	0.8	15
77	Modeling spatial and temporal aspects of visual backward masking. <i>Psychological Review</i> , 2008, 115, 83-100.	2.7	38
78	Optimality Model of Unsupervised Spike-Timing-Dependent Plasticity: Synaptic Memory and Weight Distribution. <i>Neural Computation</i> , 2007, 19, 639-671.	1.3	41
79	Predicting neuronal activity with simple models of the threshold type: Adaptive Exponential Integrate-and-Fire model with two compartments. <i>Neurocomputing</i> , 2007, 70, 1668-1673.	3.5	53
80	Optimal Spike-Timing-Dependent Plasticity for Precise Action Potential Firing in Supervised Learning. <i>Neural Computation</i> , 2006, 18, 1318-1348.	1.3	208
81	Predicting spike timing of neocortical pyramidal neurons by simple threshold models. <i>Journal of Computational Neuroscience</i> , 2006, 21, 35-49.	0.6	246
82	Dependence of the spike-triggered average voltage on membrane response properties. <i>Neurocomputing</i> , 2006, 69, 1062-1065.	3.5	15
83	Adaptive sensory processing for efficient place coding. <i>Neurocomputing</i> , 2006, 69, 1211-1214.	3.5	0
84	From spiking neurons to rate models: A cascade model as an approximation to spiking neuron models with refractoriness. <i>Physical Review E</i> , 2006, 73, 051908.	0.8	21
85	Triplets of Spikes in a Model of Spike Timing-Dependent Plasticity. <i>Journal of Neuroscience</i> , 2006, 26, 9673-9682.	1.7	515
86	Lecturers. <i>Les Houches Summer School Proceedings</i> , 2005, 80, ix.	0.2	0
87	Noise-enhanced computation in a model of a cortical column. <i>NeuroReport</i> , 2005, 16, 1237-1240.	0.6	4
88	Robust self-localisation and navigation based on hippocampal place cells. <i>Neural Networks</i> , 2005, 18, 1125-1140.	3.3	66
89	A Computational Model of Parallel Navigation Systems in Rodents. <i>Neuroinformatics</i> , 2005, 3, 223-242.	1.5	44
90	Short-Term Synaptic Plasticity Orchestrates the Response of Pyramidal Cells and Interneurons to Population Bursts. <i>Journal of Computational Neuroscience</i> , 2005, 18, 323-331.	0.6	25

#	ARTICLE	IF	CITATIONS
91	Generalized Bienenstock-Cooper-Munro rule for spiking neurons that maximizes information transmission. Proceedings of the National Academy of Sciences of the United States of America, 2005, 102, 5239-5244.	3.3	97
92	Competition between cue response and place response: a model of rat navigation behaviour. Connection Science, 2005, 17, 167-183.	1.8	5
93	Adaptive Exponential Integrate-and-Fire Model as an Effective Description of Neuronal Activity. Journal of Neurophysiology, 2005, 94, 3637-3642.	0.9	960
94	Synaptic Shot Noise and Conductance Fluctuations Affect the Membrane Voltage with Equal Significance. Neural Computation, 2005, 17, 923-947.	1.3	94
95	Generalized Integrate-and-Fire Models of Neuronal Activity Approximate Spike Trains of a Detailed Model to a High Degree of Accuracy. Journal of Neurophysiology, 2004, 92, 959-976.	0.9	233
96	Predicting spike times of a detailed conductance-based neuron model driven by stochastic spike arrival. Journal of Physiology (Paris), 2004, 98, 442-451.	2.1	13
97	Noninvasive Brain-Actuated Control of a Mobile Robot by Human EEG. IEEE Transactions on Biomedical Engineering, 2004, 51, 1026-1033.	2.5	562
98	Cognitive Navigation Based on Nonuniform Gabor Space Sampling, Unsupervised Growing Networks, and Reinforcement Learning. IEEE Transactions on Neural Networks, 2004, 15, 639-652.	4.8	78
99	Coding and learning of behavioral sequences. Trends in Neurosciences, 2004, 27, 11-14.	4.2	55
100	Optimal Hebbian Learning: A Probabilistic Point of View. Lecture Notes in Computer Science, 2003, , 92-98.	1.0	11
101	Stable Propagation of Activity Pulses in Populations of Spiking Neurons. Neural Computation, 2002, 14, 987-997.	1.3	56
102	Mathematical formulations of Hebbian learning. Biological Cybernetics, 2002, 87, 404-415.	0.6	289
103	Noise and the PSTH response to current transients: II. Integrate-and-fire model with slow recovery and application to motoneuron data. Journal of Computational Neuroscience, 2002, 12, 83-95.	0.6	16
104	Intrinsic Stabilization of Output Rates by Spike-Based Hebbian Learning. Neural Computation, 2001, 13, 2709-2741.	1.3	147
105	Coding properties of spiking neurons: reverse and cross-correlations. Neural Networks, 2001, 14, 599-610.	3.3	27
106	Noise and the PSTH response to current transients: I. General theory and application to the integrate-and-fire neuron. Journal of Computational Neuroscience, 2001, 11, 135-151.	0.6	35
107	Spatial orientation in navigating agents: Modeling head-direction cells. Neurocomputing, 2001, 38-40, 1059-1065.	3.5	21
108	Spatial cognition and neuro-mimetic navigation: a model of hippocampal place cell activity. Biological Cybernetics, 2000, 83, 287-299.	0.6	240

#	ARTICLE	IF	CITATIONS
109	Noise in Integrate-and-Fire Neurons: From Stochastic Input to Escape Rates. <i>Neural Computation</i> , 2000, 12, 367-384.	1.3	183
110	Population Dynamics of Spiking Neurons: Fast Transients, Asynchronous States, and Locking. <i>Neural Computation</i> , 2000, 12, 43-89.	1.3	372
111	Hebbian learning and spiking neurons. <i>Physical Review E</i> , 1999, 59, 4498-4514.	0.8	526
112	How the threshold of a neuron determines its capacity for coincidence detection. <i>BioSystems</i> , 1998, 48, 105-112.	0.9	29
113	Extracting Oscillations: Neuronal Coincidence Detection with Noisy Periodic Spike Input. <i>Neural Computation</i> , 1998, 10, 1987-2017.	1.3	92
114	Reduction of the Hodgkin-Huxley Equations to a Single-Variable Threshold Model. <i>Neural Computation</i> , 1997, 9, 1015-1045.	1.3	273
115	Learning navigational maps through potentiation and modulation of hippocampal place cells. , 1997, 4, 79-94.		100
116	What Matters in Neuronal Locking?. <i>Neural Computation</i> , 1996, 8, 1653-1676.	1.3	224
117	Vertical signal flow and oscillations in a three-layer model of the cortex. <i>Journal of Computational Neuroscience</i> , 1996, 3, 125-136.	0.6	4
118	A neuronal learning rule for sub-millisecond temporal coding. <i>Nature</i> , 1996, 383, 76-78.	13.7	1,038
119	Rapid Phase Locking in Systems of Pulse-Coupled Oscillators with Delays. <i>Physical Review Letters</i> , 1996, 76, 1755-1758.	2.9	83
120	Spontaneous Excitations in the Visual Cortex: Stripes, Spirals, Rings, and Collective Bursts. <i>Neural Computation</i> , 1995, 7, 905-914.	1.3	51
121	Time structure of the activity in neural network models. <i>Physical Review E</i> , 1995, 51, 738-758.	0.8	411
122	Emergence of spatiotemporal receptive fields and its application to motion detection. <i>Biological Cybernetics</i> , 1994, 72, 81-92.	0.6	12
123	A biologically motivated and analytically soluble model of collective oscillations in the cortex. <i>Biological Cybernetics</i> , 1994, 71, 349-358.	0.6	4
124	Why spikes? Hebbian learning and retrieval of time-resolved excitation patterns. <i>Biological Cybernetics</i> , 1993, 69, 503-515.	0.6	225
125	A biologically motivated and analytically soluble model of collective oscillations in the cortex. <i>Biological Cybernetics</i> , 1993, 68, 363-374.	0.6	116
126	Coherence and incoherence in a globally coupled ensemble of pulse-emitting units. <i>Physical Review Letters</i> , 1993, 71, 312-315.	2.9	107

#	ARTICLE	IF	CITATIONS
127	Associative memory in a network of "spiking" neurons. Network: Computation in Neural Systems, 1992, 3, 139-164.	2.2	143
128	Universality in neural networks: the importance of the "mean firing rate". Biological Cybernetics, 1992, 67, 195-205.	0.6	50
129	Associative memory in a network of "spiking" neurons. , 0, .		102