

Laurent U Perrinet

List of Publications by Year in descending order

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Version: 2024-02-01

90
papers

2,008
citations

331670

21
h-index

276875

41
g-index

107
all docs

107
docs citations

107
times ranked

1918
citing authors

#	ARTICLE	IF	CITATIONS
1	PyNN: a common interface for neuronal network simulators. <i>Frontiers in Neuroinformatics</i> , 2008, 2, 11.	2.5	409
2	Perceptions as Hypotheses: Saccades as Experiments. <i>Frontiers in Psychology</i> , 2012, 3, 151.	2.1	290
3	Self-Invertible 2D Log-Gabor Wavelets. <i>International Journal of Computer Vision</i> , 2007, 75, 231-246.	15.6	136
4	Networks of integrate-and-fire neurons using Rank Order Coding B: Spike timing dependent plasticity and emergence of orientation selectivity. <i>Neurocomputing</i> , 2001, 38-40, 539-545.	5.9	86
5	Smooth Pursuit and Visual Occlusion: Active Inference and Oculomotor Control in Schizophrenia. <i>PLoS ONE</i> , 2012, 7, e47502.	2.5	78
6	Functional consequences of correlated excitatory and inhibitory conductances in cortical networks. <i>Journal of Computational Neuroscience</i> , 2010, 28, 579-594.	1.0	71
7	More is not always better: adaptive gain control explains dissociation between perception and action. <i>Nature Neuroscience</i> , 2012, 15, 1596-1603.	14.8	60
8	The behavioral receptive field underlying motion integration for primate tracking eye movements. <i>Neuroscience and Biobehavioral Reviews</i> , 2012, 36, 1-25.	6.1	51
9	Coding Static Natural Images Using Spiking Event Times: Do Neurons Cooperate?. <i>IEEE Transactions on Neural Networks</i> , 2004, 15, 1164-1175.	4.2	49
10	Active inference, eye movements and oculomotor delays. <i>Biological Cybernetics</i> , 2014, 108, 777-801.	1.3	44
11	Bayesian modeling of dynamic motion integration. <i>Journal of Physiology (Paris)</i> , 2007, 101, 64-77.	2.1	42
12	The Flash-Lag Effect as a Motion-Based Predictive Shift. <i>PLoS Computational Biology</i> , 2017, 13, e1005068.	3.2	40
13	Complex dynamics in recurrent cortical networks based on spatially realistic connectivities. <i>Frontiers in Computational Neuroscience</i> , 2012, 6, 41.	2.1	37
14	Suppressive Traveling Waves Shape Representations of Illusory Motion in Primary Visual Cortex of Awake Primate. <i>Journal of Neuroscience</i> , 2019, 39, 4282-4298.	3.6	36
15	Role of Homeostasis in Learning Sparse Representations. <i>Neural Computation</i> , 2010, 22, 1812-1836.	2.2	35
16	Push-Pull Receptive Field Organization and Synaptic Depression: Mechanisms for Reliably Encoding Naturalistic Stimuli in V1. <i>Frontiers in Neural Circuits</i> , 2016, 10, 37.	2.8	35
17	Sparse spike coding in an asynchronous feed-forward multi-layer neural network using matching pursuit. <i>Neurocomputing</i> , 2004, 57, 125-134.	5.9	32
18	Motion clouds: model-based stimulus synthesis of natural-like random textures for the study of motion perception. <i>Journal of Neurophysiology</i> , 2012, 107, 3217-3226.	1.8	32

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19	Dynamics of distributed 1D and 2D motion representations for short-latency ocular following. <i>Vision Research</i> , 2008, 48, 501-522.	1.4	30
20	Edge co-occurrences can account for rapid categorization of natural versus animal images. <i>Scientific Reports</i> , 2015, 5, 11400.	3.3	25
21	Motion-based prediction explains the role of tracking in motion extrapolation. <i>Journal of Physiology (Paris)</i> , 2013, 107, 409-420.	2.1	23
22	Networks of integrate-and-fire neuron using rank order coding A: How to implement spike time dependent Hebbian plasticity. <i>Neurocomputing</i> , 2001, 38-40, 817-822.	5.9	22
23	Pursuing motion illusions: A realistic oculomotor framework for Bayesian inference. <i>Vision Research</i> , 2011, 51, 867-880.	1.4	22
24	Saccadic Foveation of a Moving Visual Target in the Rhesus Monkey. <i>Journal of Neurophysiology</i> , 2011, 105, 883-895.	1.8	20
25	Motion-Based Prediction Is Sufficient to Solve the Aperture Problem. <i>Neural Computation</i> , 2012, 24, 2726-2750.	2.2	19
26	Eye tracking a self-moved target with complex hand-target dynamics. <i>Journal of Neurophysiology</i> , 2016, 116, 1859-1870.	1.8	17
27	Sparse deep predictive coding captures contour integration capabilities of the early visual system. <i>PLoS Computational Biology</i> , 2021, 17, e1008629.	3.2	16
28	Reinforcement effects in anticipatory smooth eye movements. <i>Journal of Vision</i> , 2018, 18, 14.	0.3	15
29	Anisotropic connectivity implements motion-based prediction in a spiking neural network. <i>Frontiers in Computational Neuroscience</i> , 2013, 7, 112.	2.1	13
30	Sparse Approximation of Images Inspired from the Functional Architecture of the Primary Visual Areas. <i>Eurasip Journal on Advances in Signal Processing</i> , 2006, 2007, 1.	1.7	12
31	Revisiting horizontal connectivity rules in V1: from like-to-like towards like-to-all. <i>Brain Structure and Function</i> , 2022, 227, 1279-1295.	2.3	12
32	Feature detection using spikes: The greedy approach. <i>Journal of Physiology (Paris)</i> , 2004, 98, 530-539.	2.1	11
33	Phase space analysis of networks based on biologically realistic parameters. <i>Journal of Physiology (Paris)</i> , 2010, 104, 51-60.	2.1	11
34	Speed-Selectivity in Retinal Ganglion Cells is Sharpened by Broad Spatial Frequency, Naturalistic Stimuli. <i>Scientific Reports</i> , 2019, 9, 456.	3.3	11
35	Modeling spatial integration in the ocular following response using a probabilistic framework. <i>Journal of Physiology (Paris)</i> , 2007, 101, 46-55.	2.1	10
36	Humans adapt their anticipatory eye movements to the volatility of visual motion properties. <i>PLoS Computational Biology</i> , 2020, 16, e1007438.	3.2	10

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37	Effect of Top-Down Connections in Hierarchical Sparse Coding. <i>Neural Computation</i> , 2020, 32, 2279-2309.	2.2	10
38	Testing the odds of inherent vs. observed overdispersion in neural spike counts. <i>Journal of Neurophysiology</i> , 2016, 115, 434-444.	1.8	9
39	Finding independent components using spikes: A natural result of hebbian learning in a sparse spike coding scheme. <i>Natural Computing</i> , 2004, 3, 159-175.	3.0	8
40	A dual foveal-peripheral visual processing model implements efficient saccade selection. <i>Journal of Vision</i> , 2020, 20, 22.	0.3	8
41	PyNN: towards a universal neural simulator API in Python. <i>BMC Neuroscience</i> , 2007, 8, .	1.9	7
42	A novel bio-inspired static image compression scheme for noisy data transmission over low-bandwidth channels. , 2010, , .		6
43	Sparse spike coding : applications of neuroscience to the processing of natural images. <i>Proceedings of SPIE</i> , 2008, , .	0.8	5
44	Bayesian Modeling of Motion Perception Using Dynamical Stochastic Textures. <i>Neural Computation</i> , 2018, 30, 3355-3392.	2.2	5
45	An Adaptive Homeostatic Algorithm for the Unsupervised Learning of Visual Features. <i>Vision (Switzerland)</i> , 2019, 3, 47.	1.2	5
46	Visual Search as Active Inference. <i>Communications in Computer and Information Science</i> , 2020, , 165-178.	0.5	5
47	Emergence of filters from natural scenes in a sparse spike coding scheme. <i>Neurocomputing</i> , 2004, 58-60, 821-826.	5.9	4
48	Dynamical neural networks: Modeling low-level vision at short latencies. <i>European Physical Journal: Special Topics</i> , 2007, 142, 163-225.	2.6	4
49	A homeostatic gain control mechanism to improve event-driven object recognition. , 2021, , .		4
50	Coherence detection in a spiking neuron via Hebbian learning. <i>Neurocomputing</i> , 2002, 44-46, 133-139.	5.9	3
51	Sparse Gabor wavelets by local operations. , 2005, , .		3
52	Signature of an anticipatory response in area VI as modeled by a probabilistic model and a spiking neural network. , 2014, , .		3
53	M ² APix: A Bio-Inspired Auto-Adaptive Visual Sensor for Robust Ground Height Estimation. , 2018, , .		3
54	NeuralEnsemble.Org: Unifying neural simulators in Python to ease the model complexity bottleneck. <i>Frontiers in Neuroinformatics</i> , 0, 3, .	2.5	3

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55	On efficient sparse spike coding schemes for learning natural scenes in the primary visual cortex. BMC Neuroscience, 2007, 8, .	1.9	2
56	Control of the temporal interplay between excitation and inhibition by the statistics of visual input. BMC Neuroscience, 2009, 10, .	1.9	2
57	Computational neuroscience, from multiple levels to multi-level. Journal of Physiology (Paris), 2010, 104, 1-4.	2.1	2
58	Speed uncertainty and motion perception with naturalistic random textures. Journal of Vision, 2018, 18, 345.	0.3	2
59	Synchrony in thalamic inputs enhances propagation of activity through cortical layers. BMC Neuroscience, 2007, 8, .	1.9	1
60	Different pooling of motion information for perceptual speed discrimination and behavioral speed estimation. Journal of Vision, 2010, 10, 834-834.	0.3	1
61	A Behavioral Receptive Field for Ocular Following in Monkeys: Spatial Summation and Its Spatial Frequency Tuning. ENeuro, 2022, 9, ENEURO.0374-21.2022.	1.9	1
62	Dynamics of non-linear cortico-cortical interactions during motion integration in early visual cortex: a spiking neural network model of an optical imaging study in the awake monkey. BMC Neuroscience, 2009, 10, .	1.9	0
63	Decoding the population dynamics underlying ocular following response using a probabilistic framework. BMC Neuroscience, 2009, 10, .	1.9	0
64	Motion-based predictive coding is sufficient to solve the aperture problem. BMC Neuroscience, 2011, 12, .	1.9	0
65	The relationship between cortical network structure and the corresponding state space dynamics. BMC Neuroscience, 2011, 12, .	1.9	0
66	Active inference, eye movements and oculomotor delays. BMC Neuroscience, 2013, 14, .	1.9	0
67	Motion based prediction and development of response to an "on the way" stimulus. BMC Neuroscience, 2013, 14, .	1.9	0
68	Sparse coding of natural images using a prior on edge co-occurrences. , 2015, , .		0
69	Biologically-inspired characterization of sparseness in natural images. , 2016, , .		0
70	Dynamical state spaces of cortical networks representing various horizontal connectivities. Frontiers in Systems Neuroscience, 0, 3, .	2.5	0
71	Decoding spatial information in population of neurons for the ocular following response. Frontiers in Neuroinformatics, 0, 3, .	2.5	0
72	A recurrent Bayesian model of dynamic motion integration for smooth pursuit. Journal of Vision, 2010, 10, 545-545.	0.3	0

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73	Variations of horizontal cortical network structures and their corresponding state space dynamics. <i>Frontiers in Computational Neuroscience</i> , 0, 5, .	2.1	0
74	Pattern discrimination for moving random textures: Richer stimuli are more difficult to recognize. <i>Journal of Vision</i> , 2011, 11, 749-749.	0.3	0
75	Effect of image statistics on fixational eye movements. <i>Journal of Vision</i> , 2012, 12, 1014-1014.	0.3	0
76	How and why do image frequency properties influence perceived speed?. <i>Journal of Vision</i> , 2013, 13, 354-354.	0.3	0
77	Different temporal integration for ocular following and speed perception. <i>Journal of Vision</i> , 2013, 13, 385-385.	0.3	0
78	Beyond simply faster and slower: exploring paradoxes in speed perception. <i>Journal of Vision</i> , 2014, 14, 491-491.	0.3	0
79	Edge co-occurrences are sufficient to categorize natural versus animal images. <i>Journal of Vision</i> , 2014, 14, 1310-1310.	0.3	0
80	Motion-based prediction model for flash lag effect. <i>Journal of Vision</i> , 2014, 14, 471-471.	0.3	0
81	The characteristics of microsaccadic eye movements varied with the change of strategy in a match-to-sample task.. <i>Journal of Vision</i> , 2014, 14, 110-110.	0.3	0
82	Anticipatory smooth eye movements and reinforcement. <i>Journal of Vision</i> , 2015, 15, 1019.	0.3	0
83	A dynamic model for decoding direction and orientation in macaque primary visual cortex. <i>Journal of Vision</i> , 2015, 15, 484.	0.3	0
84	Operant reinforcement versus reward expectancy: effects on anticipatory eye movements. <i>Journal of Vision</i> , 2016, 16, 1356.	0.3	0
85	Dynamic modulation of volatility by reward contingencies: effects on anticipatory smooth eye movement. <i>Journal of Vision</i> , 2017, 17, 273.	0.3	0
86	AB009. Learning dynamics in a neural network model of the primary visual cortex. <i>Annals of Eye Science</i> , 0, 4, AB009-AB009.	2.1	0
87	Humans adapt their anticipatory eye movements to the volatility of visual motion properties. , 2020, 16, e1007438.		0
88	Humans adapt their anticipatory eye movements to the volatility of visual motion properties. , 2020, 16, e1007438.		0
89	Humans adapt their anticipatory eye movements to the volatility of visual motion properties. , 2020, 16, e1007438.		0
90	Humans adapt their anticipatory eye movements to the volatility of visual motion properties. , 2020, 16, e1007438.		0