

# Samuel Gershman

## List of Publications by Year in descending order

Source: <https://exaly.com/author-pdf/5528427/publications.pdf>

Version: 2024-02-01

130  
papers

11,970  
citations

43973

48  
h-index

39575

94  
g-index

181  
all docs

181  
docs citations

181  
times ranked

7191  
citing authors

#	ARTICLE	IF	CITATIONS
1	Model-Based Influences on Humans' Choices and Striatal Prediction Errors. <i>Neuron</i> , 2011, 69, 1204-1215.	3.8	1,388
2	Building machines that learn and think like people. <i>Behavioral and Brain Sciences</i> , 2017, 40, e253.	0.4	978
3	The hippocampus as a predictive map. <i>Nature Neuroscience</i> , 2017, 20, 1643-1653.	7.1	593
4	A tutorial on Bayesian nonparametric models. <i>Journal of Mathematical Psychology</i> , 2012, 56, 1-12.	1.0	430
5	Computational rationality: A converging paradigm for intelligence in brains, minds, and machines. <i>Science</i> , 2015, 349, 273-278.	6.0	380
6	The Curse of Planning. <i>Psychological Science</i> , 2013, 24, 751-761.	1.8	308
7	Reinforcement Learning in Multidimensional Environments Relies on Attention Mechanisms. <i>Journal of Neuroscience</i> , 2015, 35, 8145-8157.	1.7	284
8	Reinforcement Learning and Episodic Memory in Humans and Animals: An Integrative Framework. <i>Annual Review of Psychology</i> , 2017, 68, 101-128.	9.9	280
9	Context, learning, and extinction.. <i>Psychological Review</i> , 2010, 117, 197-209.	2.7	275
10	The successor representation in human reinforcement learning. <i>Nature Human Behaviour</i> , 2017, 1, 680-692.	6.2	250
11	Learning latent structure: carving nature at its joints. <i>Current Opinion in Neurobiology</i> , 2010, 20, 251-256.	2.0	242
12	Predictive representations can link model-based reinforcement learning to model-free mechanisms. <i>PLoS Computational Biology</i> , 2017, 13, e1005768.	1.5	203
13	Retrospective revaluation in sequential decision making: A tale of two systems.. <i>Journal of Experimental Psychology: General</i> , 2014, 143, 182-194.	1.5	192
14	Toward a universal decoder of linguistic meaning from brain activation. <i>Nature Communications</i> , 2018, 9, 963.	5.8	178
15	A Unified Framework for Dopamine Signals across Timescales. <i>Cell</i> , 2020, 183, 1600-1616.e25.	13.5	161
16	Believing in dopamine. <i>Nature Reviews Neuroscience</i> , 2019, 20, 703-714.	4.9	156
17	Dopamine reward prediction errors reflect hidden-state inference across time. <i>Nature Neuroscience</i> , 2017, 20, 581-589.	7.1	152
18	Cost-Benefit Arbitration Between Multiple Reinforcement-Learning Systems. <i>Psychological Science</i> , 2017, 28, 1321-1333.	1.8	150

#	ARTICLE	IF	CITATIONS
19	Deconstructing the human algorithms for exploration. <i>Cognition</i> , 2018, 173, 34-42.	1.1	148
20	Interplay of approximate planning strategies. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2015, 112, 3098-3103.	3.3	145
21	When Does Model-Based Control Pay Off?. <i>PLoS Computational Biology</i> , 2016, 12, e1005090.	1.5	142
22	Multistability and Perceptual Inference. <i>Neural Computation</i> , 2012, 24, 1-24.	1.3	131
23	Human Reinforcement Learning Subdivides Structured Action Spaces by Learning Effector-Specific Values. <i>Journal of Neuroscience</i> , 2009, 29, 13524-13531.	1.7	112
24	Rethinking dopamine as generalized prediction error. <i>Proceedings of the Royal Society B: Biological Sciences</i> , 2018, 285, 20181645.	1.2	111
25	The Successor Representation: Its Computational Logic and Neural Substrates. <i>Journal of Neuroscience</i> , 2018, 38, 7193-7200.	1.7	106
26	The algorithmic architecture of exploration in the human brain. <i>Current Opinion in Neurobiology</i> , 2019, 55, 7-14.	2.0	106
27	Gradual extinction prevents the return of fear: implications for the discovery of state. <i>Frontiers in Behavioral Neuroscience</i> , 2013, 7, 164.	1.0	105
28	Discovering latent causes in reinforcement learning. <i>Current Opinion in Behavioral Sciences</i> , 2015, 5, 43-50.	2.0	104
29	Exploring a latent cause theory of classical conditioning. <i>Learning and Behavior</i> , 2012, 40, 255-268.	0.5	102
30	The Medial Prefrontal Cortex Shapes Dopamine Reward Prediction Errors under State Uncertainty. <i>Neuron</i> , 2018, 98, 616-629.e6.	3.8	100
31	A Unifying Probabilistic View of Associative Learning. <i>PLoS Computational Biology</i> , 2015, 11, e1004567.	1.5	100
32	Neural and Psychological Maturation of Decision-making in Adolescence and Young Adulthood. <i>Journal of Cognitive Neuroscience</i> , 2013, 25, 1807-1823.	1.1	98
33	Structured Event Memory: A neuro-symbolic model of event cognition.. <i>Psychological Review</i> , 2020, 127, 327-361.	2.7	98
34	Moderate levels of activation lead to forgetting in the think/no-think paradigm. <i>Neuropsychologia</i> , 2013, 51, 2371-2388.	0.7	95
35	Human memory reconsolidation can be explained using the temporal context model. <i>Psychonomic Bulletin and Review</i> , 2011, 18, 455-468.	1.4	94
36	Empirical priors for reinforcement learning models. <i>Journal of Mathematical Psychology</i> , 2016, 71, 1-6.	1.0	92

#	ARTICLE	IF	CITATIONS
37	The computational nature of memory modification. <i>ELife</i> , 2017, 6, .	2.8	92
38	Do learning rates adapt to the distribution of rewards?. <i>Psychonomic Bulletin and Review</i> , 2015, 22, 1320-1327.	1.4	91
39	The Successor Representation and Temporal Context. <i>Neural Computation</i> , 2012, 24, 1553-1568.	1.3	88
40	A comparative evaluation of off-the-shelf distributed semantic representations for modelling behavioural data. <i>Cognitive Neuropsychology</i> , 2016, 33, 175-190.	0.4	87
41	Learning the Structure of Social Influence. <i>Cognitive Science</i> , 2017, 41, 545-575.	0.8	86
42	Toward the neural implementation of structure learning. <i>Current Opinion in Neurobiology</i> , 2016, 37, 99-105.	2.0	84
43	Neural Context Reinstatement Predicts Memory Misattribution. <i>Journal of Neuroscience</i> , 2013, 33, 8590-8595.	1.7	81
44	Hippocampal remapping as hidden state inference. <i>ELife</i> , 2020, 9, .	2.8	76
45	Belief state representation in the dopamine system. <i>Nature Communications</i> , 2018, 9, 1891.	5.8	75
46	Statistical Computations Underlying the Dynamics of Memory Updating. <i>PLoS Computational Biology</i> , 2014, 10, e1003939.	1.5	70
47	Pure correlates of exploration and exploitation in the human brain. <i>Cognitive, Affective and Behavioral Neuroscience</i> , 2018, 18, 117-126.	1.0	70
48	Incentives Boost Model-Based Control Across a Range of Severity on Several Psychiatric Constructs. <i>Biological Psychiatry</i> , 2019, 85, 425-433.	0.7	66
49	Time representation in reinforcement learning models of the basal ganglia. <i>Frontiers in Computational Neuroscience</i> , 2014, 7, 194.	1.2	64
50	Novelty and Inductive Generalization in Human Reinforcement Learning. <i>Topics in Cognitive Science</i> , 2015, 7, 391-415.	1.1	64
51	Decision by sampling implements efficient coding of psychoeconomic functions.. <i>Psychological Review</i> , 2018, 125, 985-1001.	2.7	64
52	Uncertainty and exploration.. <i>Decision</i> , 2019, 6, 277-286.	0.4	63
53	How to never be wrong. <i>Psychonomic Bulletin and Review</i> , 2019, 26, 13-28.	1.4	61
54	Explaining compound generalization in associative and causal learning through rational principles of dimensional generalization.. <i>Psychological Review</i> , 2014, 121, 526-558.	2.7	60

#	ARTICLE	IF	CITATIONS
55	Resource-rational decision making. <i>Current Opinion in Behavioral Sciences</i> , 2021, 41, 15-21.	2.0	59
56	Reconsidering the evidence for learning in single cells. <i>ELife</i> , 2021, 10, .	2.8	58
57	Structured, uncertainty-driven exploration in real-world consumer choice. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2019, 116, 13903-13908.	3.3	57
58	Dopamine Ramps Are a Consequence of Reward Prediction Errors. <i>Neural Computation</i> , 2014, 26, 467-471.	1.3	56
59	Compositional inductive biases in function learning. <i>Cognitive Psychology</i> , 2017, 99, 44-79.	0.9	55
60	Controllability governs the balance between Pavlovian and instrumental action selection. <i>Nature Communications</i> , 2019, 10, 5826.	5.8	52
61	Dissociable neural correlates of uncertainty underlie different exploration strategies. <i>Nature Communications</i> , 2020, 11, 2371.	5.8	49
62	Where do hypotheses come from?. <i>Cognitive Psychology</i> , 2017, 96, 1-25.	0.9	48
63	Neural Computations Underlying Causal Structure Learning. <i>Journal of Neuroscience</i> , 2018, 38, 7143-7157.	1.7	46
64	Integrating Models of Interval Timing and Reinforcement Learning. <i>Trends in Cognitive Sciences</i> , 2018, 22, 911-922.	4.0	45
65	Discovery of hierarchical representations for efficient planning. <i>PLoS Computational Biology</i> , 2020, 16, e1007594.	1.5	44
66	Individual differences in learning predict the return of fear. <i>Learning and Behavior</i> , 2015, 43, 243-250.	0.5	42
67	Planning Complexity Registers as a Cost in Metacontrol. <i>Journal of Cognitive Neuroscience</i> , 2018, 30, 1391-1404.	1.1	41
68	The Generative Adversarial Brain. <i>Frontiers in Artificial Intelligence</i> , 2019, 2, 18.	2.0	40
69	Dopamine neuron ensembles signal the content of sensory prediction errors. <i>ELife</i> , 2019, 8, .	2.8	39
70	Causal Inference About Good and Bad Outcomes. <i>Psychological Science</i> , 2019, 30, 516-525.	1.8	38
71	Flexible modulation of sequence generation in the entorhinalâ€“hippocampal system. <i>Nature Neuroscience</i> , 2021, 24, 851-862.	7.1	38
72	Dopamine, Inference, and Uncertainty. <i>Neural Computation</i> , 2017, 29, 3311-3326.	1.3	36

#	ARTICLE	IF	CITATIONS
73	Dopaminergic genes are associated with both directed and random exploration. <i>Neuropsychologia</i> , 2018, 120, 97-104.	0.7	36
74	A theory of learning to infer.. <i>Psychological Review</i> , 2020, 127, 412-441.	2.7	36
75	Perceptual estimation obeys Occam's razor. <i>Frontiers in Psychology</i> , 2013, 4, 623.	1.1	35
76	Model-free and model-based learning processes in the updating of explicit and implicit evaluations. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2019, 116, 6035-6044.	3.3	35
77	Discovering hierarchical motion structure. <i>Vision Research</i> , 2016, 126, 232-241.	0.7	33
78	Context-dependent learning and causal structure. <i>Psychonomic Bulletin and Review</i> , 2017, 24, 557-565.	1.4	33
79	Competition and Cooperation Between Multiple Reinforcement Learning Systems. , 2018, , 153-178.		33
80	Pavlovian Control of Escape and Avoidance. <i>Journal of Cognitive Neuroscience</i> , 2018, 30, 1379-1390.	1.1	32
81	The transdiagnostic structure of mental effort avoidance. <i>Scientific Reports</i> , 2019, 9, 1689.	1.6	32
82	Social structure learning in human anterior insula. <i>ELife</i> , 2020, 9, .	2.8	31
83	Predicting the past, remembering the future. <i>Current Opinion in Behavioral Sciences</i> , 2017, 17, 7-13.	2.0	30
84	Origin of perseveration in the trade-off between reward and complexity. <i>Cognition</i> , 2020, 204, 104394.	1.1	30
85	Discovering social groups via latent structure learning.. <i>Journal of Experimental Psychology: General</i> , 2018, 147, 1881-1891.	1.5	30
86	A topographic latent source model for fMRI data. <i>NeuroImage</i> , 2011, 57, 89-100.	2.1	29
87	Rationally inattentive intertemporal choice. <i>Nature Communications</i> , 2020, 11, 3365.	5.8	29
88	Memory as a Computational Resource. <i>Trends in Cognitive Sciences</i> , 2021, 25, 240-251.	4.0	29
89	The role of state uncertainty in the dynamics of dopamine. <i>Current Biology</i> , 2022, 32, 1077-1087.e9.	1.8	29
90	Adapting the flow of time with dopamine. <i>Journal of Neurophysiology</i> , 2019, 121, 1748-1760.	0.9	28

#	ARTICLE	IF	CITATIONS
91	Computational Phenotyping: Using Models to Understand Individual Differences in Personality, Development, and Mental Illness. <i>Personality Neuroscience</i> , 2018, 1, e18.	1.3	27
92	Finding structure in multi-armed bandits. <i>Cognitive Psychology</i> , 2020, 119, 101261.	0.9	26
93	Policy compression: An information bottleneck in action selection. <i>Psychology of Learning and Motivation - Advances in Research and Theory</i> , 2021, , 195-232.	0.5	24
94	Multi-task reinforcement learning in humans. <i>Nature Human Behaviour</i> , 2021, 5, 764-773.	6.2	23
95	Remembrance of inferences past: Amortization in human hypothesis generation. <i>Cognition</i> , 2018, 178, 67-81.	1.1	22
96	Plans, Habits, and Theory of Mind. <i>PLoS ONE</i> , 2016, 11, e0162246.	1.1	21
97	Online learning of symbolic concepts. <i>Journal of Mathematical Psychology</i> , 2017, 77, 10-20.	1.0	20
98	Social-Structure Learning. <i>Current Directions in Psychological Science</i> , 2020, 29, 460-466.	2.8	20
99	Confidence and central tendency in perceptual judgment. <i>Attention, Perception, and Psychophysics</i> , 2021, 83, 3024-3034.	0.7	20
100	A linear threshold model for optimal stopping behavior. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2020, 117, 12750-12755.	3.3	19
101	Rational inattention and tonic dopamine. <i>PLoS Computational Biology</i> , 2021, 17, e1008659.	1.5	18
102	Using computational theory to constrain statistical models of neural data. <i>Current Opinion in Neurobiology</i> , 2017, 46, 14-24.	2.0	16
103	Hierarchical structure is employed by humans during visual motion perception. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2020, 117, 24581-24589.	3.3	15
104	The rational use of causal inference to guide reinforcement learning strengthens with age. <i>Npj Science of Learning</i> , 2020, 5, 16.	1.5	14
105	Neural signatures of arbitration between Pavlovian and instrumental action selection. <i>PLoS Computational Biology</i> , 2021, 17, e1008553.	1.5	13
106	Estimating Scale-Invariant Future in Continuous Time. <i>Neural Computation</i> , 2019, 31, 681-709.	1.3	12
107	Ingredients of intelligence: From classic debates to an engineering roadmap. <i>Behavioral and Brain Sciences</i> , 2017, 40, e281.	0.4	11
108	Human visual motion perception shows hallmarks of Bayesian structural inference. <i>Scientific Reports</i> , 2021, 11, 3714.	1.6	11

#	ARTICLE	IF	CITATIONS
109	Editors' Introduction: Computational Approaches to Social Cognition. Topics in Cognitive Science, 2019, 11, 281-298.	1.1	10
110	On the Blessing of Abstraction. Quarterly Journal of Experimental Psychology, 2017, 70, 361-365.	0.6	9
111	The neurobiology of deep reinforcement learning. Current Biology, 2020, 30, R629-R632.	1.8	9
112	What Is the Model in Model-Based Planning?. Cognitive Science, 2021, 45, e12928.	0.8	9
113	Just looking: The innocent eye in neuroscience. Neuron, 2021, 109, 2220-2223.	3.8	9
114	Heuristics from bounded meta-learned inference.. Psychological Review, 2022, 129, 1042-1077.	2.7	7
115	Inference and Search on Graph-Structured Spaces. Computational Brain & Behavior, 2021, 4, 125-147.	0.9	6
116	Causal Inference Gates Corticostriatal Learning. Journal of Neuroscience, 2021, 41, 6892-6904.	1.7	6
117	Medial Prefrontal Cortex Updates Its Status. Neuron, 2016, 92, 937-939.	3.8	5
118	Moral dynamics: Grounding moral judgment in intuitive physics and intuitive psychology. Cognition, 2021, 217, 104890.	1.1	5
119	The penumbra of learning: A statistical theory of synaptic tagging and capture. Network: Computation in Neural Systems, 2014, 25, 97-115.	2.2	4
120	Impulsivity and risk-seeking as Bayesian inference under dopaminergic control. Neuropsychopharmacology, 2022, 47, 465-476.	2.8	3
121	Human-in-the-Loop Interpretability Prior. Advances in Neural Information Processing Systems, 2018, 31, .	2.8	3
122	The Reward-Complexity Trade-off in Schizophrenia. Computational Psychiatry, 2021, 5, 38-53.	1.1	2
123	Hierarchical motion structure is employed by humans during visual perception. Journal of Vision, 2019, 19, 282.	0.1	2
124	Bayesian belief updating after a replication experiment. Behavioral and Brain Sciences, 2018, 41, e134.	0.4	1
125	Analyzing Machine-Learned Representations: A Natural Language Case Study. Cognitive Science, 2020, 44, e12925.	0.8	1
126	Discovery of hierarchical representations for efficient planning. , 2020, 16, e1007594.		0



#	ARTICLE	IF	CITATIONS
127	Discovery of hierarchical representations for efficient planning. , 2020, 16, e1007594.		0
128	Discovery of hierarchical representations for efficient planning. , 2020, 16, e1007594.		0
129	Discovery of hierarchical representations for efficient planning. , 2020, 16, e1007594.		0
130	Discovery of hierarchical representations for efficient planning. , 2020, 16, e1007594.		0