

Eliu Huerta

List of Publications by Year in Descending Order

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The third column is the impact factor (IF) of the journal, and the fourth column is the number of citations of the article.

61 papers	3,660 citations	24 h-index	60 g-index
64 ext. papers	4,729 ext. citations	7.1 avg, IF	5.48 L-index

#	Paper	IF	Citations
61	Statistically-informed deep learning for gravitational wave parameter estimation. <i>Machine Learning: Science and Technology</i> , 2022 , 3, 015007	5.1	3
60	Interpretable AI forecasting for numerical relativity waveforms of quasicircular, spinning, nonprecessing binary black hole mergers. <i>Physical Review D</i> , 2022 , 105,	4.9	1
59	A FAIR and AI-ready Higgs boson decay dataset.. <i>Scientific Data</i> , 2022 , 9, 31	8.2	1
58	Inference-Optimized AI and High Performance Computing for Gravitational Wave Detection at Scale.. <i>Frontiers in Artificial Intelligence</i> , 2022 , 5, 828672	3	0
57	Advances in Machine and Deep Learning for Modeling and Real-Time Detection of Multi-messenger Sources 2022 , 1793-1819		
56	Deep Learning for Cardiologist-Level Myocardial Infarction Detection in Electrocardiograms. <i>IFMBE Proceedings</i> , 2021 , 341-355	0.2	5
55	A Gravitational-wave Measurement of the Hubble Constant Following the Second Observing Run of Advanced LIGO and Virgo. <i>Astrophysical Journal</i> , 2021 , 909, 218	4.7	46
54	Observation of eccentric binary black hole mergers with second and third generation gravitational wave detector networks. <i>Physical Review D</i> , 2021 , 103,	4.9	1
53	Deep learning for gravitational wave forecasting of neutron star mergers. <i>Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics</i> , 2021 , 816, 136185	4.2	14
52	Deep learning ensemble for real-time gravitational wave detection of spinning binary black hole mergers. <i>Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics</i> , 2021 , 812, 136029	4.2	15
51	Advances in Machine and Deep Learning for Modeling and Real-Time Detection of Multi-messenger Sources 2021 , 1-27		2
50	Deep Learning with Quantized Neural Networks for Gravitational-wave Forecasting of Eccentric Compact Binary Coalescence. <i>Astrophysical Journal</i> , 2021 , 919, 82	4.7	5
49	Artificial neural network subgrid models of 2D compressible magnetohydrodynamic turbulence. <i>Physical Review D</i> , 2020 , 101,	4.9	10
48	Deep transfer learning for star cluster classification: I. application to the PHANGS+ST survey. <i>Monthly Notices of the Royal Astronomical Society</i> , 2020 , 493, 3178-3193	4.3	16
47	The NANOGrav 11 yr Data Set: Evolution of Gravitational-wave Background Statistics. <i>Astrophysical Journal</i> , 2020 , 890, 108	4.7	13
46	The NANOGrav 11 yr Data Set: Limits on Gravitational Wave Memory. <i>Astrophysical Journal</i> , 2020 , 889, 38	4.7	22
45	Modeling the Uncertainties of Solar System Ephemerides for Robust Gravitational-wave Searches with Pulsar-timing Arrays. <i>Astrophysical Journal</i> , 2020 , 893, 112	4.7	23

44	Convergence of artificial intelligence and high performance computing on NSF-supported cyberinfrastructure. <i>Journal of Big Data</i> , 2020 , 7,	11.7	9
43	Gravitational wave denoising of binary black hole mergers with deep learning. <i>Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics</i> , 2020 , 800, 135081	4.2	39
42	Physics-inspired deep learning to characterize the signal manifold of quasi-circular, spinning, non-precessing binary black hole mergers. <i>Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics</i> , 2020 , 808, 135628	4.2	10
41	Prospects for observing and localizing gravitational-wave transients with Advanced LIGO, Advanced Virgo and KAGRA. <i>Living Reviews in Relativity</i> , 2020 , 23, 3	32.5	144
40	Enabling real-time multi-messenger astrophysics discoveries with deep learning. <i>Nature Reviews Physics</i> , 2019 , 1, 600-608	23.6	28
39	Deep learning at scale for the construction of galaxy catalogs in the Dark Energy Survey. <i>Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics</i> , 2019 , 795, 248-258	4.2	23
38	Probing neutron star structure via f-mode oscillations and damping in dynamical spacetime models. <i>Physical Review D</i> , 2019 , 99,	4.9	5
37	Supporting High-Performance and High-Throughput Computing for Experimental Science. <i>Computing and Software for Big Science</i> , 2019 , 3, 1	6	7
36	The NANOGrav 11 yr Data Set: Limits on Gravitational Waves from Individual Supermassive Black Hole Binaries. <i>Astrophysical Journal</i> , 2019 , 880, 116	4.7	58
35	Fusing numerical relativity and deep learning to detect higher-order multipole waveforms from eccentric binary black hole mergers. <i>Physical Review D</i> , 2019 , 100,	4.9	20
34	Characterization of numerical relativity waveforms of eccentric binary black hole mergers. <i>Physical Review D</i> , 2019 , 100,	4.9	10
33	Physics of eccentric binary black hole mergers: A numerical relativity perspective. <i>Physical Review D</i> , 2019 , 100,	4.9	12
32	Python Open source Waveform Extractor (POWER): an open source, Python package to monitor and post-process numerical relativity simulations. <i>Classical and Quantum Gravity</i> , 2018 , 35, 027002	3.3	6
31	Deep neural networks to enable real-time multimessenger astrophysics. <i>Physical Review D</i> , 2018 , 97,	4.9	100
30	The NANOGrav 11-year Data Set: High-precision Timing of 45 Millisecond Pulsars. <i>Astrophysical Journal, Supplement Series</i> , 2018 , 235, 37	8	295
29	Gravitational Waves from Accreting Neutron Stars Undergoing Common-envelope Inspiral. <i>Astrophysical Journal</i> , 2018 , 857, 38	4.7	10
28	Eccentric, nonspinning, inspiral, Gaussian-process merger approximant for the detection and characterization of eccentric binary black hole mergers. <i>Physical Review D</i> , 2018 , 97,	4.9	79
27	Deep Learning for real-time gravitational wave detection and parameter estimation: Results with Advanced LIGO data. <i>Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics</i> , 2018 , 778, 64-70	4.2	144

26	Prospects for observing and localizing gravitational-wave transients with Advanced LIGO, Advanced Virgo and KAGRA. <i>Living Reviews in Relativity</i> , 2018 , 21, 3	32.5	543
25	Prospects for observing and localizing gravitational-wave transients with Advanced LIGO, Advanced Virgo and KAGRA 2018 , 21, 1		2
24	The NANOGrav 11 Year Data Set: Pulsar-timing Constraints on the Stochastic Gravitational-wave Background. <i>Astrophysical Journal</i> , 2018 , 859, 47	4.7	209
23	Classification and unsupervised clustering of LIGO data with Deep Transfer Learning. <i>Physical Review D</i> , 2018 , 97,	4.9	65
22	The basic physics of the binary black hole merger GW150914. <i>Annalen Der Physik</i> , 2017 , 529, 1600209	2.6	45
21	BOSS-LDG: A Novel Computational Framework That Brings Together Blue Waters, Open Science Grid, Shifter and the LIGO Data Grid to Accelerate Gravitational Wave Discovery 2017 ,		5
20	Search for Gravitational Waves Associated with Gamma-Ray Bursts during the First Advanced LIGO Observing Run and Implications for the Origin of GRB 150906B. <i>Astrophysical Journal</i> , 2017 , 841, 89	4.7	42
19	Complete waveform model for compact binaries on eccentric orbits. <i>Physical Review D</i> , 2017 , 95,	4.9	73
18	DETECTING ECCENTRIC SUPERMASSIVE BLACK HOLE BINARIES WITH PULSAR TIMING ARRAYS: RESOLVABLE SOURCE STRATEGIES. <i>Astrophysical Journal</i> , 2016 , 817, 70	4.7	23
17	Proposed search for the detection of gravitational waves from eccentric binary black holes. <i>Physical Review D</i> , 2016 , 93,	4.9	39
16	Characterization of transient noise in Advanced LIGO relevant to gravitational wave signal GW150914. <i>Classical and Quantum Gravity</i> , 2016 , 33,	3.3	155
15	Prospects for Observing and Localizing Gravitational-Wave Transients with Advanced LIGO and Advanced Virgo. <i>Living Reviews in Relativity</i> , 2016 , 19, 1	32.5	393
14	Detection of eccentric supermassive black hole binaries with pulsar timing arrays: Signal-to-noise ratio calculations. <i>Physical Review D</i> , 2015 , 92,	4.9	35
13	Accurate and efficient waveforms for compact binaries on eccentric orbits. <i>Physical Review D</i> , 2014 , 90,	4.9	67
12	Self-forced evolutions of an implicit rotating source: A natural framework to model comparable and intermediate mass-ratio systems from inspiral through ringdown. <i>Physical Review D</i> , 2014 , 90,	4.9	5
11	Effect of eccentricity on binary neutron star searches in advanced LIGO. <i>Physical Review D</i> , 2013 , 87,	4.9	54
10	Enhanced sensitivity of the LIGO gravitational wave detector by using squeezed states of light. <i>Nature Photonics</i> , 2013 , 7, 613-619	33.9	572
9	Importance of including small body spin effects in the modelling of intermediate mass-ratio inspirals. II. Accurate parameter extraction of strong sources using higher-order spin effects. <i>Physical Review D</i> , 2012 , 85,	4.9	13

8	Accurate modeling of intermediate-mass-ratio inspirals: Exploring the form of the self-force in the intermediate-mass-ratio regime. <i>Physical Review D</i> , 2012 , 86,	4.9	10
7	Importance of including small body spin effects in the modelling of extreme and intermediate mass-ratio inspirals. <i>Physical Review D</i> , 2011 , 84,	4.9	26
6	Intermediate-mass-ratio inspirals in the Einstein Telescope. I. Signal-to-noise ratio calculations. <i>Physical Review D</i> , 2011 , 83,	4.9	22
5	Intermediate-mass-ratio inspirals in the Einstein Telescope. II. Parameter estimation errors. <i>Physical Review D</i> , 2011 , 83,	4.9	19
4	Influence of conservative corrections on parameter estimation for extreme-mass-ratio inspirals. <i>Physical Review D</i> , 2009 , 79,	4.9	40
3	Accelerated, scalable and reproducible AI-driven gravitational wave detection. <i>Nature Astronomy</i> ,	12.1	9
2	Star cluster classification in the PHANGS-BIST survey: Comparison between human and machine learning approaches. <i>Monthly Notices of the Royal Astronomical Society</i> ,	4.3	7
1	Confluence of Artificial Intelligence and High Performance Computing for Accelerated, Scalable and Reproducible Gravitational Wave Detection		3