

¹ Jaxion: A JAX package for Fuzzy Dark Matter

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14 jaxion is designed to be an open-source release of previous research code algorithms that have been used to investigate several aspects of FDM (Amin et al., 2022; Amin & Mocz, 2019; Church et al., 2019; Davies & Mocz, 2020; Dome et al., 2023, 2023; Foote et al., 2023; Lancaster et al., 2020; Luu et al., 2024, 2025; Mocz et al., 2017, 2018, 2019, 2020, 2023; Mocz & Szasz, 2021; Painter et al., 2024; Alvaro Pozo et al., 2024; A. Pozo et al., 2025).

⁵ Summary

⁶ We introduce jaxion, a Python library built on JAX for 3D numerical simulations of fuzzy dark matter (FDM), gas, and stars. Spectral, particle-mesh, and finite volume solvers are combined to model the various physics components, which are coupled through gravity. The code is scalable to multiple GPUs. JAX's automatic differentiation enables the simulations to be used with optimization and inference workflows. jaxion provides a flexible framework for rapid prototyping at scale and integration of simulations with inverse problems or hybrid physics-ML modeling.

¹³ Statement of need

¹⁴ jaxion is designed to be an open-source release of previous research code algorithms that have been used to investigate several aspects of FDM (Amin et al., 2022; Amin & Mocz, 2019; Church et al., 2019; Davies & Mocz, 2020; Dome et al., 2023, 2023; Foote et al., 2023; Lancaster et al., 2020; Luu et al., 2024, 2025; Mocz et al., 2017, 2018, 2019, 2020, 2023; Mocz & Szasz, 2021; Painter et al., 2024; Alvaro Pozo et al., 2024; A. Pozo et al., 2025).
¹⁹ This new release, written in JAX (Bradbury et al., 2018), has the added advantage of being differentiable and deployable on multiple GPUs.

²¹ Astrophysics research has long relied on sophisticated simulation codes. Established codes include: Athena++ (Stone et al., 2020), Arepo (Springel, 2010), FLASH (Fryxell et al., 2000), RAMSES (Teyssier, 2002), GAMER (Schive et al., 2018), PyUltraLight (Edwards et al., 2018). Such codes enable detailed studies of gas dynamics, star formation, cosmological structure formation, and galaxy evolution. These tools employ a combination of grid-based, particle-based, and spectral methods to solve the governing equations of hydrodynamics, gravity, and additional physics.

²⁸ Despite their successes, classical astrophysics codes are limited in their ability to interface with modern machine learning (ML) frameworks and support automatic differentiation. As ML and AI techniques are becoming more integrated with scientific fields, e.g. for parameter inference, model discovery, and hybrid physics-ML modeling, there is a growing need for simulation frameworks that are flexible and differentiable. jaxion fills this gap, by leveraging automatic differentiation, hardware acceleration, and seamless integration with ML workflows. Other recent developments of differentiable astrophysics code for various applications ranging from hydrodynamical simulations to modeling gravitational waves include (Horowitz & Lukic, 2025; Lanzieri et al., 2022; Wong et al., 2023)

³⁷ jaxion is a differentiable simulation library specifically designed for studying FDM coupled to baryons (stars and gas). FDM is a plausible dark matter candidate, modeled as a quantum wave-like field. It exhibits unique phenomena such as solitonic cores and granular interference patterns on kiloparsec scales (Hui et al., 2017). jaxion, with built-in automatic differentiability, is aimed to open new avenues for scientific discovery through gradient-based parameter

⁴² inference, optimization, and hybrid physics-ML modeling.

⁴³ Overview of functionality

⁴⁴ jaxion solves the following equations:

Component	Governing Equations	Numerical Method
Fuzzy Dark Matter	Schrodinger-Poisson	Spectral
Gas	Compressible Euler (isothermal)	Finite Volume
Stars	Collisionless N-body	Particle-Mesh
Gravity	Poisson equation	Spectral

⁴⁵ in a 3D periodic domain. It can solve equations in physical or comoving (cosmological)
⁴⁶ coordinates. Users can additionally optionally add an external potential. Features will continue
⁴⁷ to expand in future releases, including self-interactions, multiple axion fields, other fluid
⁴⁸ equations of state, and sink particles.

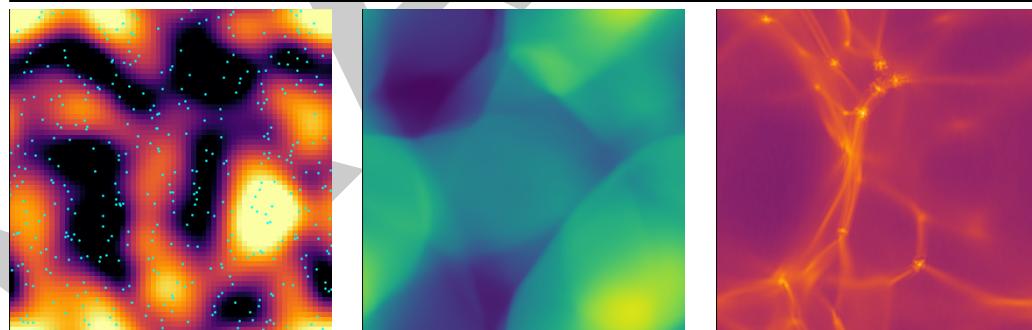
⁴⁹ The code generates checkpoints (for restart and analysis) and images.

⁵⁰ Documentation is found at: <https://jaxion.readthedocs.io/>

⁵¹ The Github Repository is at: <https://github.com/JaxionProject/jaxion>

⁵² Examples of simulation setups are found in the examples/ directory, including inverse problems
⁵³ (optimization).

⁵⁴ Below are snapshots from some of the examples:



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69 References

- 70 Amin, M. A., Jain, M., Karur, R., & Mocz, P. (2022). Small-scale structure in vector
71 dark matter. *Journal of Cosmology and Astroparticle Physics*, 2022(8), 014. <https://doi.org/10.1088/1475-7516/2022/08/014>
- 73 Amin, M. A., & Mocz, P. (2019). Formation, gravitational clustering, and interactions of
74 nonrelativistic solitons in an expanding universe. *Physical Review D*, 100(6), 063507.
75 <https://doi.org/10.1103/PhysRevD.100.063507>
- 76 Bradbury, J., Frostig, R., Hawkins, P., Johnson, M. J., Leary, C., Maclaurin, D., Necula, G.,
77 Paszke, A., VanderPlas, J., Wanderman-Milne, S., & Zhang, Q. (2018). JAX: Composable
78 transformations of Python+NumPy programs (Version 0.3.13). <http://github.com/jax-ml/jax>
- 80 Church, B. V., Mocz, P., & Ostriker, J. P. (2019). Heating of Milky Way disc stars by dark
81 matter fluctuations in cold dark matter and fuzzy dark matter paradigms. *Machine*, 485(2),
82 2861–2876. <https://doi.org/10.1093/mnras/stz534>
- 83 Davies, E. Y., & Mocz, P. (2020). Fuzzy dark matter soliton cores around supermassive
84 black holes. *Monthly Notices of the Royal Astronomical Society*, 492(4), 5721–5729.
85 <https://doi.org/10.1093/mnras/staa202>
- 86 Dome, T., Fialkov, A., Sartorio, N., & Mocz, P. (2023). Cosmic web dissection in fuzzy dark
87 matter cosmologies. *Monthly Notices of the Royal Astronomical Society*, 525(1), 348–363.
88 <https://doi.org/10.1093/mnras/stad2276>
- 89 Edwards, F., Kendall, E., Hotchkiss, S., & Easther, R. (2018). PyUltraLight: a pseudo-spectral
90 solver for ultralight dark matter dynamics. *Journal of Cosmology and Astroparticle Physics*,
91 2018(10), 027. <https://doi.org/10.1088/1475-7516/2018/10/027>
- 92 Foote, H. R., Besla, G., Mocz, P., Garavito-Camargo, N., Lancaster, L., Sparre, M., Cunningham,
93 E. C., Vogelsberger, M., Gómez, F. A., & Laporte, C. F. P. (2023). Structure,
94 Kinematics, and Observability of the Large Magellanic Cloud's Dynamical Friction
95 Wake in Cold versus Fuzzy Dark Matter. *Astrophysical Journal*, 954(2), 163.
96 <https://doi.org/10.3847/1538-4357/ace533>
- 97 Fryxell, B., Olson, K., Ricker, P., Timmes, F. X., Zingale, M., Lamb, D. Q., MacNeice, P.,
98 Rosner, R., Truran, J. W., & Tufo, H. (2000). FLASH: An Adaptive Mesh Hydrodynamics
99 Code for Modeling Astrophysical Thermonuclear Flashes. *Astrophysical Journal Supplement
100 Series*, 131(1), 273–334. <https://doi.org/10.1086/317361>
- 101 Horowitz, B., & Lukic, Z. (2025). Differentiable Cosmological Hydrodynamics for Field-Level
102 Inference and High Dimensional Parameter Constraints. *arXiv e-Prints*, arXiv:2502.02294.
103 <https://doi.org/10.48550/arXiv.2502.02294>
- 104 Hui, L., Ostriker, J. P., Tremaine, S., & Witten, E. (2017). Ultralight scalars as cosmological
105 dark matter. *Physical Review D*, 95(4), 043541. <https://doi.org/10.1103/PhysRevD.95.043541>
- 107 Lancaster, L., Giovanetti, C., Mocz, P., Kahn, Y., Lisanti, M., & Spergel, D. N. (2020).
108 Dynamical friction in a Fuzzy Dark Matter universe. *Journal of Cosmology and Astroparticle
109 Physics*, 2020(1), 001. <https://doi.org/10.1088/1475-7516/2020/01/001>
- 110 Lanzieri, D., Lanusse, F., & Starck, J.-L. (2022). Hybrid Physical-Neural ODEs for Fast
111 N-body Simulations. *Machine Learning for Astrophysics*, 60. <https://doi.org/10.48550/arXiv.2207.05509>
- 113 Luu, H. N., Mocz, P., Vogelsberger, M., May, S., Borrow, J., Tye, S.-H. H., & Broadhurst,
114 T. (2024). Nested solitons in two-field fuzzy dark matter. *Monthly Notices of the Royal
115 Astronomical Society*, 527(2), 4162–4172. <https://doi.org/10.1093/mnras/stad3482>

- 116 Luu, H. N., Mocz, P., Vogelsberger, M., Pozo, A., Broadhurst, T., Tye, S.-H. H., Liu,
 117 T., Fung, L. W. H., Smoot, G. F., Emami, R., & Hernquist, L. (2025). Diverse dark
 118 matter haloes in two-field fuzzy dark matter. *Phys. Rev. D*, 111(12), L121302. <https://doi.org/10.1103/w9j1-k7b3>
- 120 Mocz, P., Fialkov, A., Vogelsberger, M., Becerra, F., Amin, M. A., Bose, S., Boylan-Kolchin,
 121 M., Chavanis, P.-H., Hernquist, L., Lancaster, L., Marinacci, F., Robles, V. H., & Zavala, J.
 122 (2019). First Star-Forming Structures in Fuzzy Cosmic Filaments. *Physical Review Letters*,
 123 123(14), 141301. <https://doi.org/10.1103/PhysRevLett.123.141301>
- 124 Mocz, P., Fialkov, A., Vogelsberger, M., Becerra, F., Shen, X., Robles, V. H., Amin, M.
 125 A., Zavala, J., Boylan-Kolchin, M., Bose, S., Marinacci, F., Chavanis, P.-H., Lancaster,
 126 L., & Hernquist, L. (2020). Galaxy formation with BECDM - II. Cosmic filaments and
 127 first galaxies. *Monthly Notices of the Royal Astronomical Society*, 494(2), 2027–2044.
 128 <https://doi.org/10.1093/mnras/staa738>
- 129 Mocz, P., Fialkov, A., Vogelsberger, M., Boylan-Kolchin, M., Chavanis, P.-H., Amin, M. A.,
 130 Bose, S., Dome, T., Hernquist, L., Lancaster, L., Notis, M., Painter, C., Robles, V. H.,
 131 & Zavala, J. (2023). Cosmological structure formation and soliton phase transition in
 132 fuzzy dark matter with axion self-interactions. *Monthly Notices of the Royal Astronomical
 133 Society*, 521(2), 2608–2615. <https://doi.org/10.1093/mnras/stad694>
- 134 Mocz, P., Lancaster, L., Fialkov, A., Becerra, F., & Chavanis, P.-H. (2018). Schrödinger-
 135 Poisson-Vlasov-Poisson correspondence. *Physical Review D*, 97(8), 083519. <https://doi.org/10.1103/PhysRevD.97.083519>
- 137 Mocz, P., & Szasz, A. (2021). Toward Cosmological Simulations of Dark Matter on Quantum
 138 Computers. *Astrophysical Journal*, 910(1), 29. <https://doi.org/10.3847/1538-4357/abe6ac>
- 139 Mocz, P., Vogelsberger, M., Robles, V. H., Zavala, J., Boylan-Kolchin, M., Fialkov, A., &
 140 Hernquist, L. (2017). Galaxy formation with BECDM - I. Turbulence and relaxation of
 141 idealized haloes. *Monthly Notices of the Royal Astronomical Society*, 471(4), 4559–4570.
 142 <https://doi.org/10.1093/mnras/stx1887>
- 143 Painter, C. A., Boylan-Kolchin, M., Mocz, P., & Vogelsberger, M. (2024). An attractive model:
 144 simulating fuzzy dark matter with attractive self-interactions. *Monthly Notices of the Royal
 145 Astronomical Society*, 533(2), 2454–2472. <https://doi.org/10.1093/mnras/stae1912>
- 146 Pozo, Alvaro, Broadhurst, T., Smoot, G. F., Chiueh, T., Luu, H. N., Vogelsberger, M., &
 147 Mocz, P. (2024). Dwarf galaxies united by dark bosons. *Phys. Rev. D*, 109(8), 083532.
 148 <https://doi.org/10.1103/PhysRevD.109.083532>
- 149 Pozo, A., Emami, R., Mocz, P., Broadhurst, T., Hernquist, L., Vogelsberger, M., Smith, R.,
 150 Tremblay, G., Narayan, R., Steiner, J., Grindlay, J., & Smoot, G. (2025). Galaxy formation
 151 with wave/fuzzy dark matter: The core-halo structure and the solitonic imprint. *A&A*, 699,
 152 A308. <https://doi.org/10.1051/0004-6361/202450443>
- 153 Schive, H.-Y., ZuHone, J. A., Goldbaum, N. J., Turk, M. J., Gaspari, M., & Cheng, C.-
 154 Y. (2018). GAMER-2: a GPU-accelerated adaptive mesh refinement code - accuracy,
 155 performance, and scalability. *Monthly Notices of the Royal Astronomical Society*, 481(4),
 156 4815–4840. <https://doi.org/10.1093/mnras/sty2586>
- 157 Springel, V. (2010). E pur si muove: Galilean-invariant cosmological hydrodynamical simu-
 158 lations on a moving mesh. *Monthly Notices of the Royal Astronomical Society*, 401(2),
 159 791–851. <https://doi.org/10.1111/j.1365-2966.2009.15715.x>
- 160 Stone, J. M., Tomida, K., White, C. J., & Felker, K. G. (2020). The Athena++ Adaptive
 161 Mesh Refinement Framework: Design and Magnetohydrodynamic Solvers. *Astrophysical
 162 Journal Supplement Series*, 249(1), 4. <https://doi.org/10.3847/1538-4365/ab929b>
- 163 Teyssier, R. (2002). Cosmological hydrodynamics with adaptive mesh refinement. A new

¹⁶⁴ high resolution code called RAMSES. *Astronomy and Astrophysics*, 385, 337–364. <https://doi.org/10.1051/0004-6361:20011817>

¹⁶⁵ Wong, K. W. K., Isi, M., & Edwards, T. D. P. (2023). Fast Gravitational-wave Parameter
¹⁶⁷ Estimation without Compromises. *Astrophysical Journal*, 958(2), 129. <https://doi.org/10.3847/1538-4357/acf5cd>

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