

popclass: A Python Package for Classifying Microlensing Events

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DOI: 10.21105/joss.07769

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Submitted: 07 October 2024 Published: 24 May 2025

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Summary

popclass is a Python package that provides a flexible, probabilistic framework for classifying the lens of a gravitational microlensing event. Gravitational microlensing occurs when a massive foreground object (e.g., a star, white dwarf or black hole) passes in front of and lenses the light from a distant background source. This causes an apparent brightening, and shift in position, of the background source. In most cases, characteristics of the microlensing signal do not contain enough information to definitively identify the lens type. Different lens types lie in different but overlapping regions of the characteristics of the microlensing signal. For example, black holes tend to be more massive than stars and therefore cause microlensing signals that are longer. Current Galactic simulations enable us to predict where different lens types lie in the observational space and can therefore be used to classify events (e.g., Lam et al., 2020).

popclass allows the user to match the characteristics of a microlensing signal with a simulation of the Galaxy to calculate lens type probabilities for the event (see Figure 1). Constraints on any microlensing signal properties and any Galactic model can be used. popclass comes with an interface to ArviZ (Kumar et al., 2019) and PyMultiNest (Buchner et al., 2014) for microlensing signal constraints, as well as pre-loaded Galactic models, plotting functionality, and methods to quantify the classification uncertainty. The probabilistic framework for popclass was developed in Perkins et al. (2024), used in Fardeen et al. (2024) and has been applied to classifying events in Kaczmarek et al. (2025).

Statement of need

The advent of the Vera C. Rubin Observatory (Ivezić et al., 2019) and the Nancy Grace Roman Space Telescope (Spergel et al., 2015) will provide tens of thousands of microlensing events per year (e.g., Abrams et al., 2025; Penny et al., 2019). To maximize the scientific output from this event stream, it is critical to identify events that have a high probability of being caused by a certain lens type such as a black hole (Lam et al., 2022; Sahu et al., 2022), before we allocate expensive follow-up observations such as space-based astrometry (e.g., Sahu et al., 2022) or ground-based adaptive optics imaging (e.g., Terry et al., 2022) to confirm their nature.



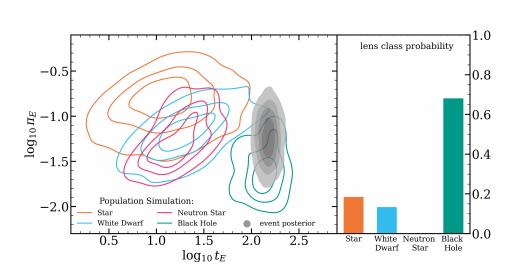


Figure 1: Left: posterior distribution of an event in log10(timescale)-log10(parallax) space, overlaid on 'star', 'white dwarf', 'neutron star' and 'black hole' contours. Right: bars showing probabilities of that event belonging to each of the lens populations.

Current microlensing software packages such as DarkLensCode (Howil et al., 2025) or pyLIMASS (Bachelet et al., 2024) estimate lens mass and distance constraints using the light curves of microlensing events and auxiliary information (e.g., source proper motions, distances, color, or finite source effects). Using auxiliary information makes these current methods powerful but limits their application to only be effective for events with the available auxiliary data. Moreover, no current software tools explicitly predict the lens type, and they always assume a fixed Galactic model. popclass fills the need for a flexible microlensing classification software package that can be broadly applied to classify all events from the Vera C. Rubin Observatory and Nancy Grace Roman Space Telescope and can be used with any Galactic model in the form of a simulation.

Method

popclass is based on the Bayesian classification framework in Perkins et al. (2024). Consider the data from a single microlensing event d, and a model of the Galaxy \mathcal{G} . popclass calculates the probability that the lens of the event belongs to each lens class, class, where class, d classes and, for example, classes = {star, neutron star, white dwarf, black hole}. Namely, popclass calculates

$$p(\mathsf{class}_L|\boldsymbol{d},\mathcal{G}) \text{ for } \mathsf{class}_L \in \mathsf{classes}.$$

Using Bayes' theorem,

$$p(\mathsf{class}_L|\boldsymbol{d},\mathcal{G}) = \frac{p(\mathsf{class}_L|\mathcal{G})p(\boldsymbol{d}|\mathsf{class}_L,\mathcal{G})}{p(\boldsymbol{d}|\mathcal{G})}.$$

Assuming that the set of considered lens classes is complete, $p(\boldsymbol{d}|\mathcal{G})$ is a normalization factor such that all lens class probabilities sum to unity. Using importance sampling (e.g., Hogg et al., 2010) with S independent posterior samples $\theta_c \sim p(\theta|\boldsymbol{d})$ drawn under some prior, $\pi(\theta)$, obtained from fitting some set of microlensing signal parameters, θ ,



$$p(\mathsf{class}_L|\boldsymbol{d},\mathcal{G}) = \frac{p(\mathsf{class}_L|\mathcal{G})}{p(\boldsymbol{d}|\mathcal{G})} \times \frac{1}{S} \sum_{c=0}^S \frac{p(\theta_c|\mathsf{class}_L,\mathcal{G})}{\pi(\theta_c)}.$$

This allows the use of previously calculated posterior samples to perform lens classification for a single event in the context of a Galactic model. The term $p(\theta_c|{\rm class}_L,\mathcal{G})$ is calculated using kernel density estimation in popclass over θ with a simulated catalog of microlensing events from $\mathcal{G}.$ $p({\rm class}_L|\mathcal{G})$ is the prior probability that an event belongs to each class before any data is seen, which is simply set by the relative number of expected events predicted by the Galactic model $\mathcal{G}.$

Acknowledgements

popclass depends on NumPy (Harris et al., 2020), SciPy (Virtanen et al., 2020), ASDF (Greenfield et al., 2015), Matplotlib (Hunter, 2007), and scikit-learn (Pedregosa et al., 2011). This work was performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory (LLNL) under Contract DE-AC52-07NA27344. The document number is LLNL-JRNL-870290 and the code number is LLNL-CODE-2000456. The theoretical foundation of this work was established under support from LLNL's Laboratory Directed Research and Development Program under project 22-ERD-037. The software implementation for this project was funded under the LLNL Space Science Institute's Institutional Scientific Capability Portfolio funds in partnership with LLNL's Academic Engagement Office. ZK acknowledges support from the 2024 LLNL Data Science Summer Institute and is a Fellow of the International Max Planck Research School for Astronomy and Cosmic Physics at the University of Heidelberg (IMPRS-HD).

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