

# PyMOCAT-MC: A Python Implementation of the MIT Orbital Capacity Assessment Toolbox Monte Carlo Module

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## Summary

The growth of satellite deployments and large-scale constellations has made orbital sustainability an urgent concern (Bastida Virgili et al., 2016; Lemmens et al., 2020; Lewis, 2020). Simulating the evolution of the space environment requires tools that are both computationally efficient and compatible with modern research workflows. The MIT Orbital Capacity Assessment Toolbox Monte Carlo module (MOCAT-MC) (Jang et al., 2025) is a leading framework for modeling space traffic and debris risk. Its MATLAB implementation, however, restricts accessibility due to licensing requirements and creates barriers to integrating with state-of-the-art machine learning algorithms and Python-based data science pipelines central to contemporary space sustainability research.

PyMOCAT-MC is a complete Python reimplement of MOCAT-MC, designed to maintain full functional compatibility with the original toolbox while improving performance, modularity, and accessibility. The translation involves converting more than 150 MATLAB functions into Python, preserving the core algorithms while restructuring them for clarity and efficiency within Python's scientific computing ecosystem. The resulting codebase comprises over 8,600 lines in Python and leverages open-source libraries including NumPy, SciPy, pandas, and Matplotlib.

Benchmarking against the MATLAB version shows that PyMOCAT-MC achieves a mean relative error of 0.96% and a maximum error of 2.38% across all tested scenarios. Performance tests demonstrate speed gains of up to four times, with the "Realistic Launch" scenario completing in under 30 seconds in Python compared to over 75 seconds in MATLAB. This combination of high accuracy and faster execution enables more extensive simulation campaigns and facilitates integration into a wider range of research and policy workflows.

## Statement of need

As the orbital environment becomes more congested (Kukreja et al., 2025), the ability to model collisions, debris evolution, and the long-term sustainability of satellite operations has significant implications for both policy-making and engineering design. The original MOCAT-MC toolbox (ARCLab-MIT, 2025) has been used by researchers to perform Monte Carlo-based assessments of orbital capacity, yet its MATLAB implementation limits accessibility for researchers and organizations that rely on open-source tools. It also presents barriers to integrating with state-of-the-art machine learning frameworks and Python-based data analysis workflows, which are increasingly central to advancing space situational awareness modeling. These constraints can hinder collaboration, slow down simulation work, and restrict the adoption of the software

41 in the broader space sustainability community.

42 PyMOCAT-MC addresses these barriers by delivering a functionally equivalent, open-source  
43 Python version that is free to use and modify. The Python implementation not only replicates  
44 the original capabilities but also improves runtime performance and introduces a modular  
45 design that makes it easier to integrate with other orbital mechanics packages such as Astropy  
46 or poliastro, as well as related tools like MOCAT-pySSEM (Brownhall et al., 2025). The  
47 open-source nature of the project supports reproducibility, encourages community contributions,  
48 and creates opportunities for extending the toolbox to new modeling approaches, such as agent-  
49 based simulations of satellite behavior. In the broader landscape, PyMOCAT-MC complements  
50 established environment models such as ESA's MASTER and related frameworks (Flegel et al.,  
51 2012).

## 52 Methodology

53 The development of PyMOCAT-MC began with a careful analysis of the MATLAB source  
54 code to understand the data structures, algorithms, and dependencies used in MOCAT-MC.  
55 Each function was translated individually into Python, maintaining the mathematical and  
56 algorithmic logic while adopting Pythonic conventions for readability and maintainability. Where  
57 appropriate, vectorized operations and optimized data handling were introduced to enhance  
58 computational performance without altering the results. Atmospheric modeling considerations,  
59 critical for accurate orbit propagation (Ding et al., 2023; Emmert, 2015), were preserved in  
60 the translation.

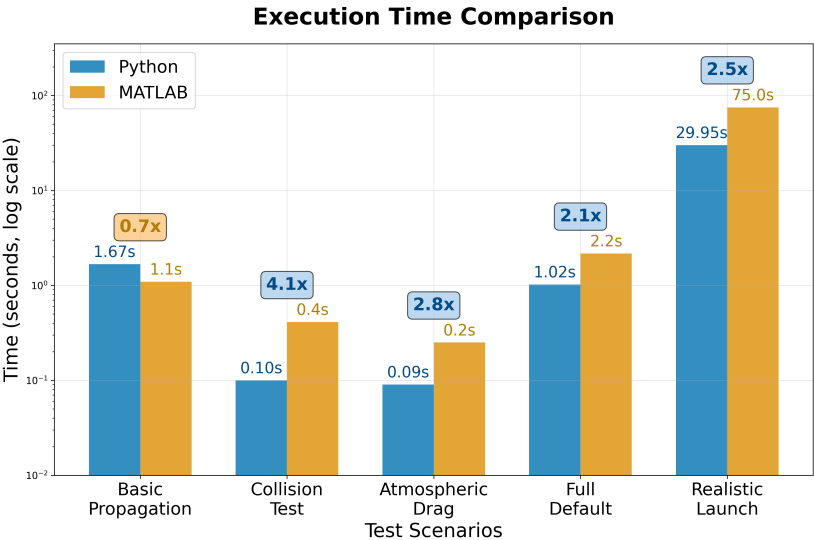
61 Validation was an integral part of the translation process. Unit tests were created to verify the  
62 correctness of individual functions, while side-by-side comparisons of MATLAB and Python  
63 simulation outputs were performed for each standard scenario. These comparisons ensure that  
64 differences in results remained within acceptable numerical tolerances. Manual code reviews  
65 were also conducted to confirm fidelity to the original design.

66 The repository is organized to facilitate both research use and further development. The  
67 `mocat_mc.py` module contains the main Monte Carlo simulation engine, supported by additional  
68 modules for orbital propagation, collision detection, atmospheric drag modeling, and debris  
69 generation. Continuous integration workflows were implemented to automatically run tests and  
70 verify that code changes preserve numerical fidelity across all benchmark scenarios. Example  
71 scripts demonstrate common simulation scenarios, including baseline, no-launch, and realistic-  
72 launch cases. Comparison scripts and performance analysis tools are included to reproduce the  
73 accuracy and speed results reported here. Supporting data, such as historical two-line element  
74 (TLE) sets, the JB2008 atmospheric density model, and launch schedules for megaconstellations,  
75 are bundled with the software to enable immediate use.

## 76 Results

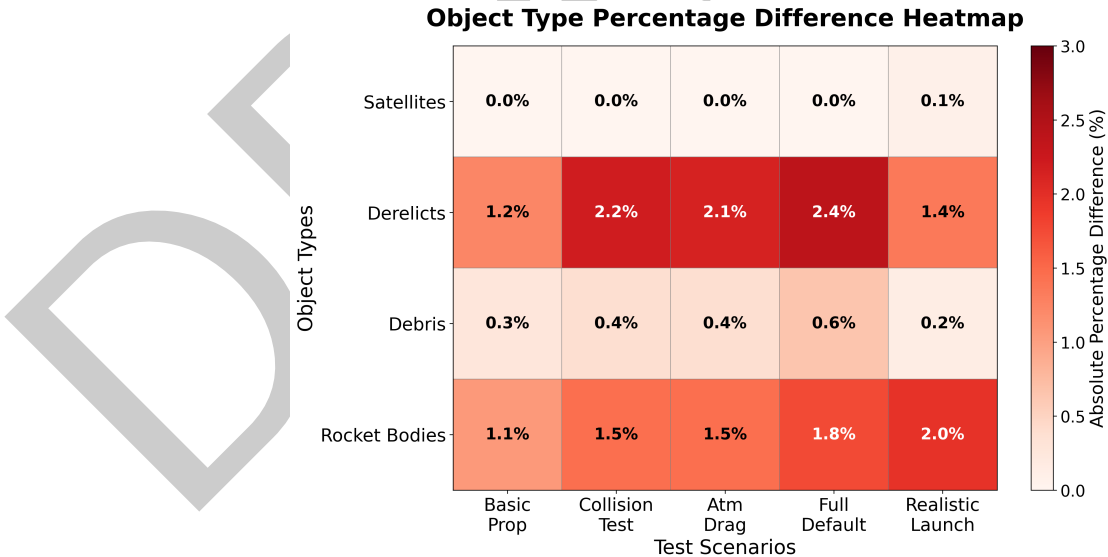
77 Across all tested scenarios, PyMOCAT-MC reproduces the results of the MATLAB implemen-  
78 tation with high fidelity. Differences in total object counts between the two implementations  
79 are small, with a maximum deviation of 150 objects out of approximately 13,700.

80 In addition to matching the accuracy of MATLAB, the Python version delivers substantial  
81 performance gains. The most computationally demanding scenario, which includes realistic  
82 launch patterns for megaconstellations, runs in less than 29.95 seconds with PyMOCAT-MC  
83 compared to 75.02 seconds in MATLAB, representing a speed-up larger than a factor of two.

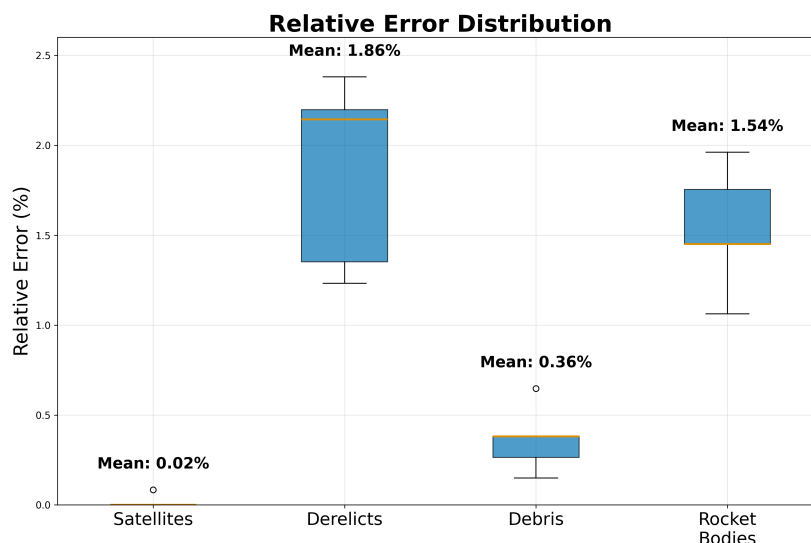


**Figure 1:** Runtime comparison between MATLAB and Python implementations for multiple scenarios, showing up to a 4× improvement in computational speed for PyMOCAT-MC.

84 Error analysis confirms that the differences between MATLAB and Python remain minimal  
85 across simulation years and object types. The heatmap and box plots highlights that relative  
86 errors are uniformly low; the mean relative error is 0.96%, and the maximum error across  
87 object types and time horizons is 2.38%. Specifically among satellites and debris, the error is  
88 consistently lower than 0.61%.



**Figure 2:** Heatmap of relative error (%) across simulation years and object types, illustrating consistent accuracy of the Python implementation compared to MATLAB.



**Figure 3:** Relative error distribution box plots across object classes, showing PyMOCAT-MC achieves near-zero error for satellites and consistently low error across all categories.

89 These improvements reduce the time required for large simulation batches, enabling exploration  
 90 of a wider range of parameters and more detailed sensitivity analyses. PyMOCAT-MC can be  
 91 applied to test how different launch strategies influence orbital sustainability, evaluate debris  
 92 mitigation policies, and generate scenario libraries that inform both engineering design and  
 93 regulatory decision-making.

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