

ParquetDB: A Lightweight Python Parquet-Based Database

Logan L. Lang¹, Eduardo R. Hernandez², Kamal Choudhary³, and Aldo H. Romero¹

¹ Department of Physics, West Virginia University, Morgantown, United States ² Instituto de Ciencia de Materiales de Madrid, Madrid, Spain ³ National Institute of Standards and Technology, Gaithersburg, United States

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

Software

- [Review](#)
- [Repository](#)
- [Archive](#)

Editor: [Open Journals](#)

Reviewers:

- [@openjournals](#)

Submitted: 01 January 1970

Published: unpublished

License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License ([CC BY 4.0](#)).

Summary

ParquetDB is a Python library serving as a “medium ware” solution, bridging the gap between file-based storage and full database systems. Leveraging Apache Parquet (“Parquet,” n.d.; [Welcome to The Apache Software Foundation, n.d.](#)), it combines file storage portability with advanced querying capabilities, enabling efficient compression and read performance without dedicated server overhead. ParquetDB addresses limitations in both traditional approaches by seamlessly handling complex data types (arrays, nested structures, Python objects), simplifying data interaction compared to direct file manipulation or manual serialization. Performance benchmarks show competitive read/write speeds and effective query performance via predicate pushdown, demonstrating its utility for managing medium-to-large datasets where database complexity is unwarranted but basic file I/O is insufficient.

Statement of need

The demand for efficient, scalable, and adaptable data storage solutions is critical across research domains. Traditional file formats (e.g., CSV, JSON, TXT) offer simplicity but suffer from inefficiencies, particularly with numerical data due to ASCII/UTF encoding overhead, leading to larger files and slower I/O. While binary formats like HDF5 ([HDF5 for Python — H5py 3.13.0 Documentation, n.d.](#)) improve efficiency for large numerical datasets, they function primarily as structured file containers, lacking the rich querying APIs and transactional integrity features common in databases. These file-based approaches often require manual data relationship management and lack built-in indexing, hindering agility as projects scale or require rapid iteration.

Database systems like SQLite (Allen & Owens, 2010) or MongoDB (Guo, 2017) provide robust encoding, indexing, and querying. Relational databases ensure integrity via structured schemas but can be rigid when data models evolve (Pascal, 2000). NoSQL options offer flexibility but may introduce consistency challenges or require complex optimization (Pivert, 2018). Furthermore, many databases involve server configurations or lack transparent file-based portability, adding overhead unsuitable for lightweight experimentation or simpler deployment scenarios. While SQLite is serverless and ubiquitous, its row-based nature can be less performant for analytical queries scanning wide datasets compared to columnar formats, and managing complex nested data can be cumbersome.

Directly using libraries like Apache Arrow (PyArrow) to work with Parquet files offers access to columnar efficiency and querying primitives like predicate pushdown. However, this still requires developers to build abstractions for database-like operations (CRUD), manage schema

41 consistency across multiple files, handle serialization of complex Python objects, and orchestrate
42 data updates or deletions manually.

43 ParquetDB addresses this gap, providing a “medium ware” layer built upon Python and the
44 Parquet format. It offers a familiar database-like interface (CRUD operations) while leveraging
45 columnar storage for compression and read performance benefits. Crucially, ParquetDB adds
46 value beyond direct Parquet file manipulation by automating schema management (including
47 evolution), simplifying the storage/retrieval of complex Python objects, and providing a unified
48 API to manage collections of Parquet files as a single logical datastore. It supports predicate
49 and column pushdown for optimization within a lightweight, serverless architecture, offering a
50 pragmatic balance for scenarios demanding more than basic files but less than a full database
51 system, particularly where schema flexibility and ease of use are paramount. For a comprehensive
52 feature list, visit our documentation (<https://parquetdb.readthedocs.io/en/latest/>).

53 Benchmarks

54 We evaluated ParquetDB's performance against SQLite and MongoDB using synthetic datasets
55 (100 integer columns, varying record counts). Our first experiment compared write and read
56 performance. ParquetDB's create times are competitive, performing second best behind
57 SQLite as dataset size increases. For bulk read operations, ParquetDB initially lags slightly but
58 significantly outperforms both competitors on larger datasets (beyond several hundred/thousand
59 rows), benefiting from Parquet's columnar efficiency. (See Figure 1)

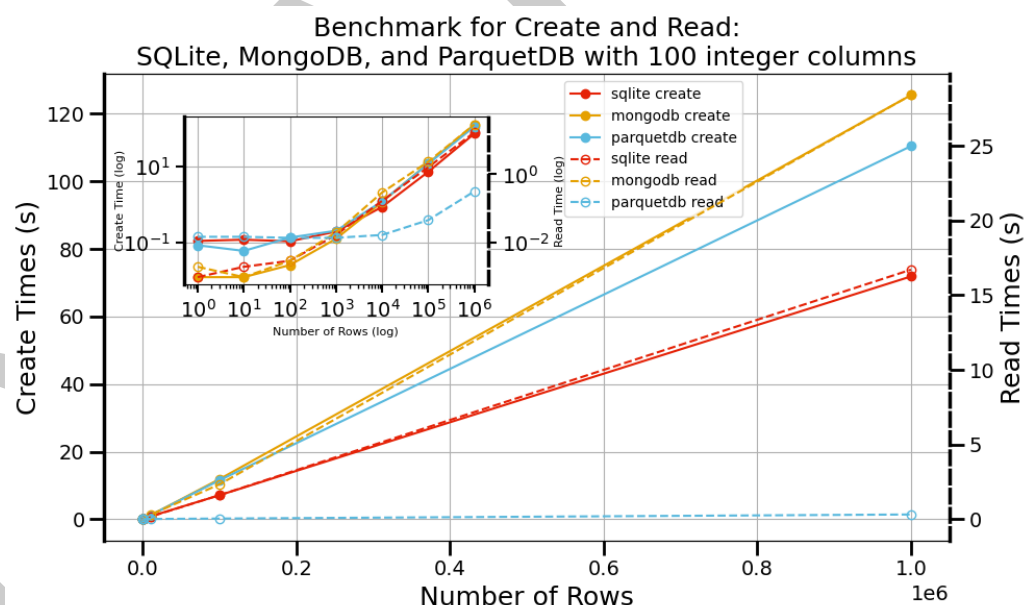


Figure 1: Benchmark Create and Read Times for Different Databases. Create time is plotted on the left y-axis, read time on the right y-axis, and the number of rows on the x-axis. A log plot is shown in the inset.

60 A “needle-in-a-haystack” benchmark assessed specific record retrieval. While lacking traditional
61 B-tree indexes, ParquetDB uses predicate pushdown leveraging Parquet's field-level statistics for
62 efficient filtering without full scans. It is important to note that performance advantages depend
63 on the workload; for instance, complex analytical queries involving aggregations or returning
64 small, highly filtered results might favor the mature query engine and indexing of systems like
65 SQLite. ParquetDB excels when querying or returning substantial portions of wide datasets.
66 Detailed benchmarks are in our extended paper (<https://arxiv.org/abs/2502.05311>)(Lang et

67 [al., 2025](#)).

68 Installation

69 For installation, please use pip:

```
pip install parquetdb
```

70 For more details, please visit the [GitHub repository](#). The repository contains additional
71 examples, API documentation, and guidelines for contributing to the project.

72 Acknowledgements

73 We thank the Pittsburgh Supercomputer Center (Bridges2) and San Diego Supercomputer
74 Center (Expanse) through allocation DMR140031 from the Advanced Cyberinfrastructure Co-
75 ordination Ecosystem: Services & Support (ACCESS) program, which is supported by National
76 Science Foundation grants #2138259, #2138286, #2138307, #2137603, and #2138296. We
77 gratefully acknowledge the computational resources provided by the WVU Research Computing
78 Dolly Sods HPC cluster, partially funded by NSF OAC-2117575. Additionally, we recognize the
79 support from the West Virginia Higher Education Policy Commission through the Research
80 Challenge Grant Program 2022 (Award RCG 23-007), as well as NASA EPSCoR (Award
81 80NSSC22M0173), for their contributions to this work. The work of E.R.H. is supported by
82 MCIN/AEI/ 10.13039/501100011033/FEDER, UE through projects PID2022-139776NB-C66.
83 K.C. thanks funding from the CHIPS Metrology Program, part of CHIPS for America, National
84 Institute of Standards and Technology, U.S. Department of Commerce. Certain commercial
85 equipment, instruments, software, or materials are identified in this paper in order to specify
86 the experimental procedure adequately. Such identifications are not intended to imply rec-
87 ommendation or endorsement by NIST, nor are they intended to imply that the materials or
88 equipment identified are necessarily the best available for the purpose.

89 References

- 90 Allen, G., & Owens, M. (2010). *The Definitive Guide to SQLite*. Apress. <https://doi.org/10.1007/978-1-4302-3226-1>
- 91
- 92 Guo, R. (2017). MongoDB's JavaScript fuzzer. *Commun. ACM*, 60(5), 43–47. <https://doi.org/10.1145/3052937>
- 93
- 94 *HDF5 for Python — H5py 3.13.0 documentation*. (n.d.). <https://docs.h5py.org/en/stable/index.html>.
- 95
- 96 Lang, L., Hernandez, E., Choudhary, K., & Romero, A. H. (2025). *ParquetDB: A Lightweight*
97 *Python Parquet-Based Database* (No. arXiv:2502.05311). arXiv. <https://doi.org/10.48550/arXiv.2502.05311>
- 98
- 99 Parquet. (n.d.). In *Apache Parquet*. <https://parquet.apache.org/>.
- 100 Pascal, F. (2000). *Practical Issues in Database Management: A Reference for the Thinking*
101 *Practitioner* (1st edition). Addison-Wesley Professional. ISBN: 978-0-201-48555-4
- 102 Pivert, O. (Ed.). (2018). *NoSQL Data Models: Trends and Challenges* (1st edition). Wiley-
103 ISTE. ISBN: 978-1-78630-364-6
- 104 *Welcome to The Apache Software Foundation*. (n.d.). <https://www.apache.org/>.