

RustSASA: A Rust Crate for Accelerated Solvent Accessible Surface Area Calculations

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Summary

Solvent accessible surface area (SASA) calculations are fundamental for understanding protein structure, function, and dynamics in computational biology. These calculations quantify the surface area of biomolecules accessible to solvent molecules, providing insights into protein folding, stability, and intermolecular interactions. The Shrake-Rupley algorithm has served as the standard for SASA calculations since 1973, but existing implementations often become computational bottlenecks when analyzing large protein datasets. As proteomics datasets continue to grow with initiatives like AlphaFold producing hundreds of millions of predicted protein structures the need for efficient SASA calculation tools has increased dramatically. RustSASA addresses this challenge by providing a high-performance implementation of the Shrake-Rupley algorithm written in pure Rust, delivering a 7 \times speed improvement over Freesasa while maintaining calculation accuracy and providing interfaces for multiple programming languages and frameworks (i.e: MDAnalysis).

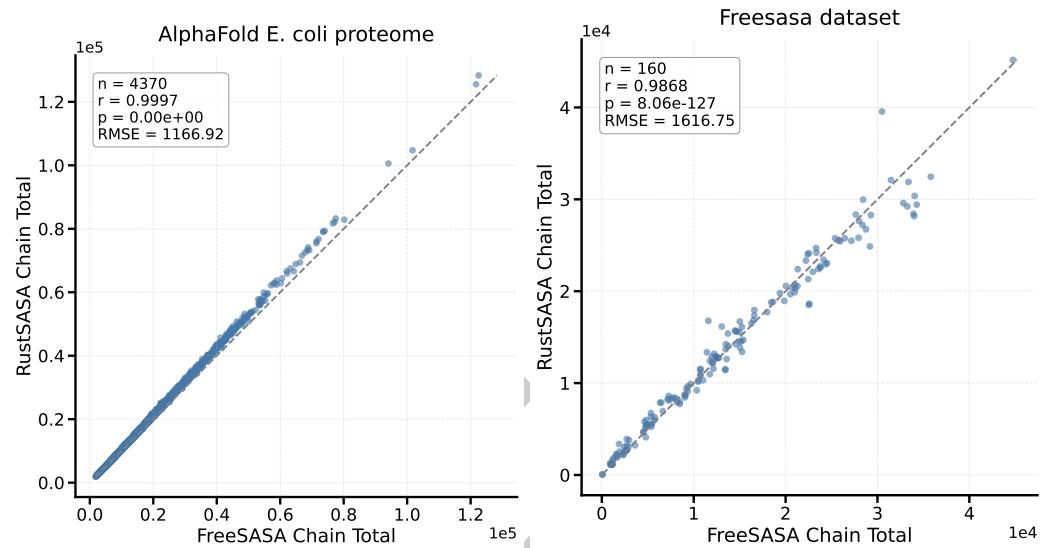
Statement of need

Current SASA calculation tools represent a significant computational bottleneck in structural biology workflows, particularly for molecular dynamics simulations and high-throughput analyses. Popular implementations such as those in Biopython and Freesasa, while accurate, become prohibitively slow when processing large protein datasets.

RustSASA addresses this performance gap by leveraging Rust's zero-cost abstractions and memory safety guarantees to create a SASA calculation crate that is significantly faster than Freesasa and Biopython. Benchmarking on representative protein structures demonstrates that RustSASA achieves a 7 \times improvement over Freesasa, and a 46 \times performance improvement over Biopython. This performance advantage reduces computational costs for high-throughput structural analyses and makes large-scale comparative studies feasible. Furthermore, Rust-SASA's multi-language support (Rust and Python), command-line interface, and MDAnalysis package ensure broad accessibility across the computational biology community.

31 Results

32 Calculation Quality



33 To evaluate the accuracy of RustSASA calculations, we compared results to Freesasa (Mitter-
 34 nacht, 2016) on both the predicted E. coli proteome from AlphaFold (Jumper et al., 2021)
 35 and the Freesasa evaluation dataset. RustSASA produces SASA values that closely match
 36 those from Freesasa, achieving Pearson correlation coefficients > 0.98 on both datasets.
 37

38 Performance

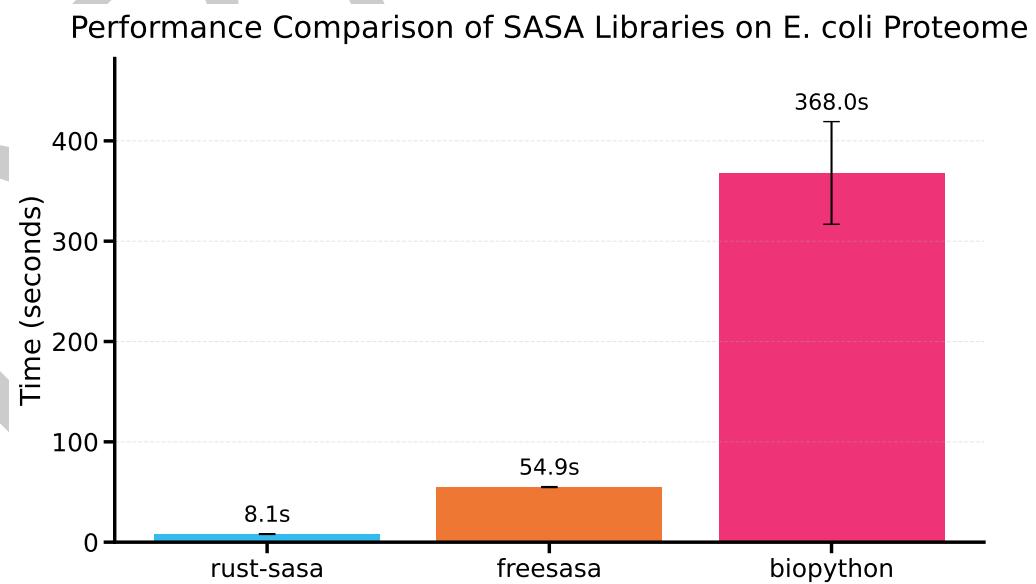


Figure 1: Comparing Freesasa, RustSasa, and Biopython performance on E. coli proteome

39 We evaluated the performance of Freesasa, RustSASA, and Biopython (Cock et al., 2009)
 40 on the predicted E. coli proteome using Hyperfine (Peter, 2023) with three runs and three
 41 warmup iterations on a 2024 Apple MacBook Air with an M3 processor and 24GB of unified

42 memory. All methods utilized parallel processing across eight cores. GNU parallel ([Tange, 43 2011](#)) was used to parallelize Freesasa and Biopython, while RustSASA's utilized its internal 44 parallelization.

45 RustSASA processed the entire proteome in ~8 seconds compared to ~55 seconds for Freesasa 46 and ~368 seconds for Biopython, representing 7× and 46× speed improvements, respectively.

47 Conclusion

48 RustSASA provides a significant advancement in SASA calculation performance while main- 49 taining accuracy, addressing a bottleneck in computational structural biology. The 7× speed 50 improvement over current standards enables previously intractable analyses of large protein 51 datasets and molecular dynamics simulations. By providing interfaces for multiple programming 52 languages alongside a command-line tool and MDAnalysis package, RustSASA ensures broad 53 accessibility across the research community. As structural biology datasets continue to expand, 54 efficient computational tools like RustSASA become essential for advancing our understanding 55 of protein structure and function.

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59 References

- 60 Cock, P. J., Antao, T., Chang, J. T., Chapman, B. A., Cox, C. J., Dalke, A., Friedberg, I.,
61 Hamelryck, T., Kauff, F., Wilczynski, B., & others. (2009). Biopython: Freely available
62 python tools for computational molecular biology and bioinformatics. *Bioinformatics*,
63 25(11), 1422–1423. <https://doi.org/10.1093/bioinformatics/btp163>
- 64 Jumper, J., Evans, R., Pritzel, A., Green, T., Figurnov, M., Ronneberger, O., Tunyasuvunakool,
65 K., Bates, R., Žídek, A., Potapenko, A., Bridgland, A., Meyer, C., Kohl, S. A. A., Ballard,
66 A. J., Cowie, A., Romera-Paredes, B., Nikolov, S., Jain, R., Adler, J., ... Hassabis, D.
67 (2021). Highly accurate protein structure prediction with AlphaFold. *Nature*, 596(7873),
68 583–589. <https://doi.org/10.1038/s41586-021-03819-2>
- 69 Mitternacht, S. (2016). FreeSASA: An open source c library for solvent accessible surface area
70 calculations. *F1000Research*, 5, 189. <https://doi.org/10.12688/f1000research.7931.1>
- 71 Peter, D. (2023). *Hyperfine*. <https://github.com/sharkdp/hyperfine>
- 72 Tange, O. (2011). GNU parallel - the command-line power tool; *Login: The USENIX Magazine*,
73 36(1), 42–47. <http://www.gnu.org/s/parallel>