

¹ RustSASA: A Rust Crate for Accelerated Solvent Accessible Surface Area Calculations

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Summary

⁹ Solvent accessible surface area (SASA) calculations are fundamental for understanding protein
¹⁰ structure, function, and dynamics in computational biology. These calculations quantify the
¹¹ surface area of biomolecules accessible to solvent molecules, providing insights into protein
¹² folding, stability, and intermolecular interactions. The Shrake-Rupley algorithm has served as
¹³ the standard for SASA calculations since 1973, but existing implementations often become
¹⁴ computational bottlenecks when analyzing large protein datasets. As proteomics datasets
¹⁵ continue to grow with initiatives like AlphaFold producing hundreds of millions of predicted
¹⁶ protein structures the need for efficient SASA calculation tools has increased dramatically.
¹⁷ RustSASA addresses this challenge by providing a high-performance implementation of the
¹⁸ Shrake-Rupley algorithm written in pure Rust, delivering a 5× speed improvement over Freesasa
¹⁹ while maintaining calculation accuracy and providing interfaces for multiple programming
²⁰ languages and frameworks (i.e: MDAnalysis).

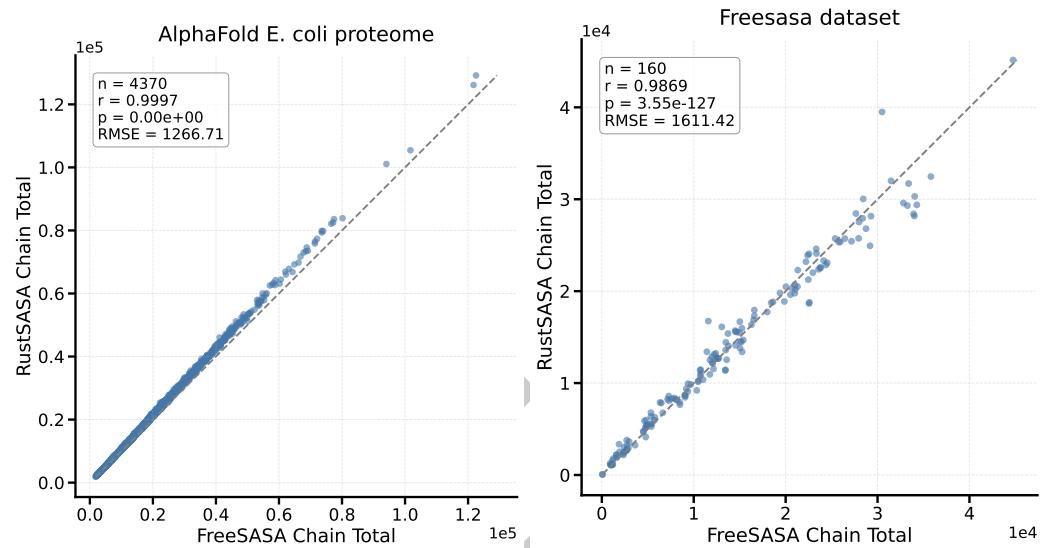
Statement of need

²¹ Current SASA calculation tools represent a significant computational bottleneck in structural
²² biology workflows, particularly for molecular dynamics simulations and high-throughput analyses.
²³ Popular implementations such as those in Biopython and Freesasa, while accurate, become
²⁴ prohibitively slow when processing large protein datasets. RustSASA addresses this performance
²⁵ gap by leveraging Rust's efficient parallelization abstractions (via Rayon) and readily available
SIMD instructions (via Pulp). These optimizations enable RustSASA's performance advantage
over the simpler C implementation of the same algorithm in Freesasa.

²⁶ Benchmarking on representative protein structures demonstrates that RustSASA achieves a
²⁷ 5× improvement over Freesasa, and a 63× performance improvement over Biopython. This
²⁸ performance advantage reduces computational costs for high-throughput structural analyses
²⁹ and makes large-scale comparative studies feasible. Furthermore, RustSASA's multi-language
³⁰ support (Rust and Python), command-line interface, and MDAnalysis package ensure broad
³¹ accessibility across the computational biology community.

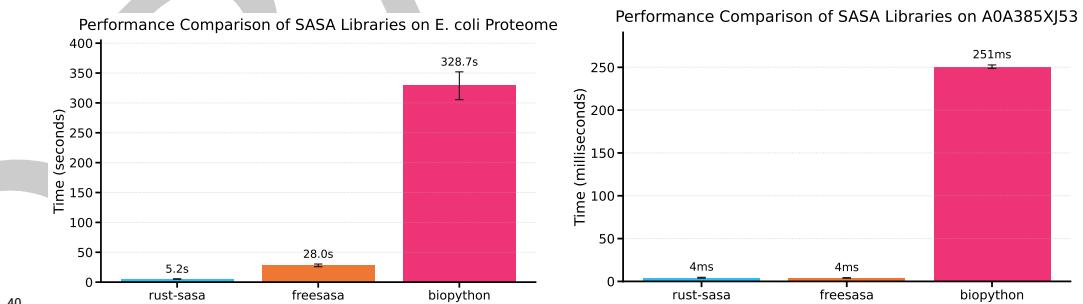
32 Results

33 Calculation Quality



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

39 Performance



40
 41 We evaluated the performance of Freesasa, RustSASA, and Biopython (Cock et al., 2009)
 42 in two common use cases. First, we performed SASA calculations for all proteins in the E.
 43 coli proteome. Second, we evaluated the performance of these methods on a single randomly
 44 selected protein (A0A385XJ53) from the AlphaFold E. coli proteome.

45 For the full proteome benchmark we used Hyphen (Peter, 2023) with 3 runs and 3 warmup
 46 iterations. All methods utilized parallel processing across eight cores. GNU parallel (Tange,
 47 2011) was used to parallelize Freesasa and Biopython, while RustSASA utilized its internal
 48 parallelization. RustSASA processed the entire proteome in ~5 seconds compared to ~28 seconds
 49 for Freesasa and ~368 seconds for Biopython, representing 5× and 63× speed improvements,
 50 respectively.

51 For the single protein benchmark, we used Hyphen with 3 warmup iterations and 25 runs.
 52 RustSASA processed the protein in 4.3ms (± 0.5), Freesasa processed in 4.0ms
 53 (± 0.2), and Biopython processed the protein in 250.8ms (± 2.0).

54 All runs were conducted on a 2024 Apple MacBook Air with an M3 processor and 24GB of
55 unified memory.

56 Conclusion

57 RustSASA provides a significant advancement in SASA calculation performance while main-
58 taining accuracy, addressing a bottleneck in computational structural biology. The 5× speed
59 improvement over current standards enables previously intractable analyses of large protein
60 datasets and molecular dynamics simulations. By providing interfaces for multiple programming
61 languages alongside a command-line tool and MDAnalysis package, RustSASA ensures broad
62 accessibility across the research community. As structural biology datasets continue to expand,
63 efficient computational tools like RustSASA become essential for advancing our understanding
64 of protein structure and function.

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67 to this project. We would also like to thank the reviewers for their valuable comments as well
68 as Tsai et al. for creating the ProtOr radii ([Tsai et al., 1999](#)).

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