

# Haarpy: a Python library for Weingarten calculus and integration over classical compact groups, their associated circular ensembles, the permutation and centered permutation groups, and quantum groups

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## Software

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## Summary

**Haarpy** is a Python library for the symbolic calculation of Weingarten functions and related integrals (also called moments or averages) of matrix ensembles under their Haar measure. There is a multiplicity of applications, both in physics and in mathematics, requiring the computation of averages of various polynomial functions over the classical compact groups (unitary, orthogonal and symplectic), their associated circular ensembles, the groups of permutation matrices and centered permutation matrices, and the quantum groups (free symmetric and free orthogonal) (Potters & Bouchaud, 2020). Rather than relying on Monte Carlo simulation, which provides approximate results (Mezzadri, 2006), Haarpy enables exact analytical computation of such moments.

Under the hood, Haarpy reduces the computation of integrals over the relevant ensembles to Weingarten calculus. This field has grown rapidly over the past few decades, following the foundational work of Collins, who introduced the terminology (Collins, 2003; Collins & Śniady, 2006). This machinery is exposed through functions such as `weingarten_unitary`, `weingarten_orthogonal`, and `weingarten_symplectic`, while the associated integrals are accessed through functions such as `haar_integral_unitary`, `haar_integral_orthogonal`, and `haar_integral_symplectic`. A full description of Haarpy's functionalities can be found in the library [documentation](#).

Built on top of the SymPy symbolic engine, Haarpy allows users to retain symbolic parameters and derive general formulas applicable across entire classes of problems (Meurer et al., 2017).

## Statement of need

The computation of averages over Haar-distributed random matrices is of growing importance in quantum information theory (Cardin & Quesada, 2024; Di Matteo, 2021; Martínez-Cifuentes et al., 2024; Mele, 2024; Turkeshi et al., 2025), random matrix theory (Bordenave & Collins, 2024; Daigle et al., 2025), and statistical physics (Cotler et al., 2017; Potters & Bouchaud, 2020). These averages describe the typical behavior of complex systems, including their quantum and classical correlations.

Standard approaches typically rely either on Monte Carlo simulation, which approximates integrals through sampling, or on manual analytical derivations, which often involve intricate combinatorics.

Haarpy is intended for researchers throughout physics, mathematicians working in representation

theory, and researchers studying random matrix models. By making Weingarten calculus readily accessible as a software package, Haarpy enhances reproducibility, reduces computational overhead, and lowers the barrier to performing rigorous analytical calculations.

## State of the field

Several software packages support symbolic Haar integration and related calculations. Early implementations were developed in the proprietary computer algebra systems Mathematica (de Guise, 2024; Puchała & Miszczak, 2017) and Maple (Ginory & Kim, 2021), while more recent projects such as RTNI2 (Fukuda, 2023) and IntegrateUnitary.jl (Pawela & Puchała, 2026) provide advanced functionality for tensor-network-based computations and symbolic integration within their respective ecosystems. In particular, RTNI2 offers a powerful framework for diagrammatic calculations based on tensor network methods, while IntegrateUnitary.jl provides symbolic tools for Haar integration in Julia.

Haarpy complements these efforts by providing an open-source, Python-native implementation tightly integrated with the scientific Python ecosystem and with SymPy. In addition to supporting integrations over the unitary and orthogonal groups, the package implements moment calculations for symplectic groups, circular ensembles, permutation and centered permutation matrices, and the free symmetric and free orthogonal groups within a unified interface. To the best of our knowledge, Haarpy is the first Python package to provide direct computation of Haar integrals within a unified symbolic framework, rather than requiring users to reconstruct integrals from intermediate outputs.

The package places a strong emphasis on verification, with an extensive test suite covering both symbolic and numerical computations. Rather than extending an existing codebase, the underlying Weingarten calculus machinery has been re-implemented in Python to provide a tool that integrates naturally with widely used Python scientific workflows. For classical compact groups, Haarpy additionally provides an alternative moment computation method based on the recursive algorithm of Gorin (Gorin & López, 2008), which does not rely on Weingarten calculus. In many settings, this latter approach can offer improved performance compared to Weingarten-based computations.

## Software design

Haarpy is designed with a focus on mathematical fidelity, symbolic flexibility, and usability. A key consideration is the trade-off between symbolic and numerical computation: while purely numerical methods can offer speed, they often obscure structure and reduce reproducibility, whereas symbolic computation preserves the exact algebraic form needed for derivations and verification. For this reason, the implementation is deliberately built around a lightweight symbolic stack centered on SymPy, avoiding heavier external dependencies in order to keep the system transparent, portable, and easier to validate. At the same time, efficiency is addressed through selective caching of intermediate combinatorial quantities, which reduces redundant work without relying on a precomputed database of results. In particular, moments are computed on demand rather than retrieved from tabulated Weingarten values.

Core Weingarten functionality is complemented by auxiliary tools, including:

- Implementation of the Murnaghan-Nakayama rule for characters of irreducible representations of the symmetric group (James, 1978),
- Functions for computing dimensions of irreducible representations of the symmetric and unitary groups (James, 1978),
- Generators yielding the permutations of the hyperoctahedral group and Young group (Macdonald, 1998),

- Generators yielding the sets of non-crossing partitions and non-crossing pairings (Nica & Speicher, 2006).

This modular design enables reuse in broader algebraic and combinatorial computations. Exposing group-theoretic primitives increases usability for diverse research applications.

## Research impact statement

Haarpy has been cited in an arXiv preprint, indicating external use beyond the original development context (Duschenes et al., 2025). The library is also used in ongoing collaborative research on Weingarten calculus and related random matrix theory problems. The project is publicly available on [GitHub](#) and is accompanied by comprehensive documentation hosted on [Read the Docs](#). The repository includes continuous integration with automated testing, with all tests passing and full test coverage reported via standard coverage tooling. These components provide a reproducible and verifiable implementation of algorithms for integration over a wide range of ensembles equipped with the Haar measure.

## AI usage disclosure

No AI tools were used in the development of Haarpy.

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