




pygwb: a Python-based library for gravitational-wave background searches

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

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Submitted: 31 March 2023

Published: unpublished

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Introduction

A gravitational-wave background (GWB) is expected from the superposition of all gravitational waves (GWs) too faint to be detected individually, or by the incoherent overlap of a large number of signals in the same band ([A. I. Renzini et al., 2022](#)). A GWB is primarily characterized by its spectral emission, usually parameterized by the GW fractional energy density spectrum $\Omega_{\text{GW}}(f)$, which is the target for stochastic GW searches ([Allen & Romano, 1999](#)),

$$\Omega_{\text{GW}}(f) = \frac{1}{\rho_c} \frac{d\rho_{\text{GW}}(f)}{d \ln f},$$

where $d\rho_{\text{GW}}$ is the energy density of GWs in the frequency band f to $f + df$, and ρ_c is the critical energy density of the Universe. Different categories of GW sources may be identified by the unique spectral shape of their background emission; hence, the detection of a GWB will provide invaluable information about the evolution of the Universe and the population of GW sources within it.

Statement of need

Due to the considerable amount of data to analyze, and the vast panorama of GWB models to test, the detection and characterization of a GWB requires a community effort. Furthermore,

This article and software are linked

with Research article (2023) pygwb: a Python-based library for gravitational-wave background searches. *Journal of Open Source Software*, 0(0), 5454. [10.5547/joss.00000](https://doi.org/10.5547/joss.00000)

[Update draft](#) with the DOI from AAS once you

know it., published in the *Astrophysical Journal* <- The

name of the AAS journal..

data handling and model building entail a number of different choices, depending on specific analysis purposes. Up until the previous LIGO-Virgo-KAGRA Collaboration (LVK) observing run, O3, the collaboration has relied on an internal MATLAB-based pipeline available at (LVK, 2020) to perform stochastic analyses. This pipeline lacks the ability to perform parameter estimation, as well as modularity and flexibility. This exemplifies the need for an accessible, flexible, and user-friendly open-source codebase for the current and upcoming LVK runs: pygwb. To fully cater to user needs, pygwb is modular and extensively customizable, and is accompanied by exhaustive documentation.

Method

The GWB spectrum estimation implemented in pygwb is based on the unbiased minimum variance cross-correlation estimator (Romano & Cornish, 2017),

$$\hat{\Omega}_{\text{GW},f} = \frac{\text{Re}[C_{IJ,f}]}{\gamma_{IJ}(f)S_0(f)}.$$

Here, $C_{IJ,f}$ is the cross-correlation spectral density between two detectors I and J , γ_{IJ} is the overlap reduction function (Allen & Romano, 1999), and $S_0(f) = \frac{3H_0^2}{10\pi^2} \frac{1}{f^3}$, where H_0 is the Hubble constant today (Aghanim & others, 2020). The variance of the estimator is given by

$$\sigma_{\text{GW},f}^2 = \frac{1}{2T\Delta f} \frac{P_{I,f}P_{J,f}}{\gamma_{IJ}^2(f)S_0^2(f)},$$

where $P_{I,f}$ is the power spectral density from detector I and T is the duration of data used to produce the above spectral densities. This estimator is optimal and unbiased under the assumption that the signal is Gaussian, isotropic, and continuous. Details on how the estimation is carried out, as well as the implementation of the estimator on large datasets and with many potentially overlapping data segments can be found in our companion methods paper (Arianna I. Renzini & others, 2023).

Model testing in pygwb is performed through Bayesian inference on a select set of parameters, given a parametric GWB model and a likelihood p of observing the data given the model. Concretely, the above cross-correlation estimator is input data to a Gaussian residual likelihood,

$$p(\hat{\Omega}_{\text{GW},f}^{IJ}|\lambda) \propto \exp \left[-\frac{1}{2} \sum_{IJ} \sum_f \left(\frac{\hat{\Omega}_{\text{GW},f}^{IJ} - \Omega_{\text{M}}(f|\lambda)}{\hat{\sigma}_{\text{GW},f}^{IJ}} \right)^2 \right],$$

where $\Omega_{\text{M}}(f|\lambda)$ is the GWB model and λ are its parameters. pygwb currently admits a variety of GWB models, compatible with the Gaussian likelihood above. More information about the parameter estimation and the implemented models can be found in our companion methods paper (Arianna I. Renzini & others, 2023).

pygwb

pygwb is a Python-based, open-source stochastic GW analysis package specifically tailored to searches for isotropic GWBs with current ground-based interferometers, namely the Laser Interferometer Gravitational-wave Observatory (LIGO), the Virgo observatory, and the KAGRA detector.

The pygwb package is class-based and modular to facilitate the evolution of the code and to increase flexibility of the analysis pipeline. The advantage of the Python language lies in

73 rapid code execution, while maintaining a certain level of user-friendliness, which results in
74 a shallow learning curve and will encourage future contributions to the code from the whole
75 GW community. A summary of all pygwb modules and its main external dependencies can be
76 found in the pygwb schema [Figure 1](#).

77 The package is compatible with GW frame files in a variety of formats, relying on the I/O
78 functionality of gwpy ([Macleod et al., 2021](#)). NumPy ([Harris et al., 2020](#)) is heavily used
79 within the pygwb code, as well as matplotlib ([Hunter, 2007](#)) for plotting purposes. Some
80 of the frequency-related computations rely on functionalities of the scipy ([Virtanen et al.,
81 2020](#)) package. The astropy ([Astropy Collaboration et al., 2022](#)) package is employed for
82 cosmology-related computations. The parameter estimation module included in pygwb is based
83 on Bilby ([Ashton et al., 2019](#)) and the dynesty ([Speagle, 2020](#)) sampler package.

84 A customizable pipeline script, pygwb_pipe, is provided with the package and can be run in
85 default mode, which reproduces the methodology of the LVK isotropic analysis implemented
86 on the most recent observation run ([Abbott et al., 2021](#)). On the other hand, the modularity
87 of the package allows users to develop custom pygwb pipelines to fit their needs. A set
88 of simple statistical checks can be performed on the data after a pygwb run by using the
89 statistical_checks module. In addition, a parameter estimation script, pygwb_pe, is also
90 included and allows to test a subset of default models with user-defined parameters. pygwb_pe
91 is based on the pygwb parameter estimation module, pe, which allows the user to test both
92 predefined and user-defined models and obtain posterior distributions on the parameters of
93 interest. Users are encouraged to develop and test their own models within the pe module.
94 The pygwb package also contains built-in support for running on HTCondor-supported servers
95 using dag files to parallelize the analysis of long stretches of data. Using the dedicated
96 pygwb_combine script, the output can be combined into an overall estimation of the GWB for
97 the whole data set.

98 The source code can be found at <https://git.ligo.org/pygwb/pygwb> and <https://github.com/a-renzini/pygwb>, and can be installed from PyPi via `pip install pygwb`. The online
99 documentation, tutorials and examples are hosted at [https://pygwb.docs.ligo.org/pygwb/
100 index.html](https://pygwb.docs.ligo.org/pygwb/index.html). The package includes a unit test suite which currently covers 80% of the modules.
101 pygwb is released under a OSI Approved :: MIT License.
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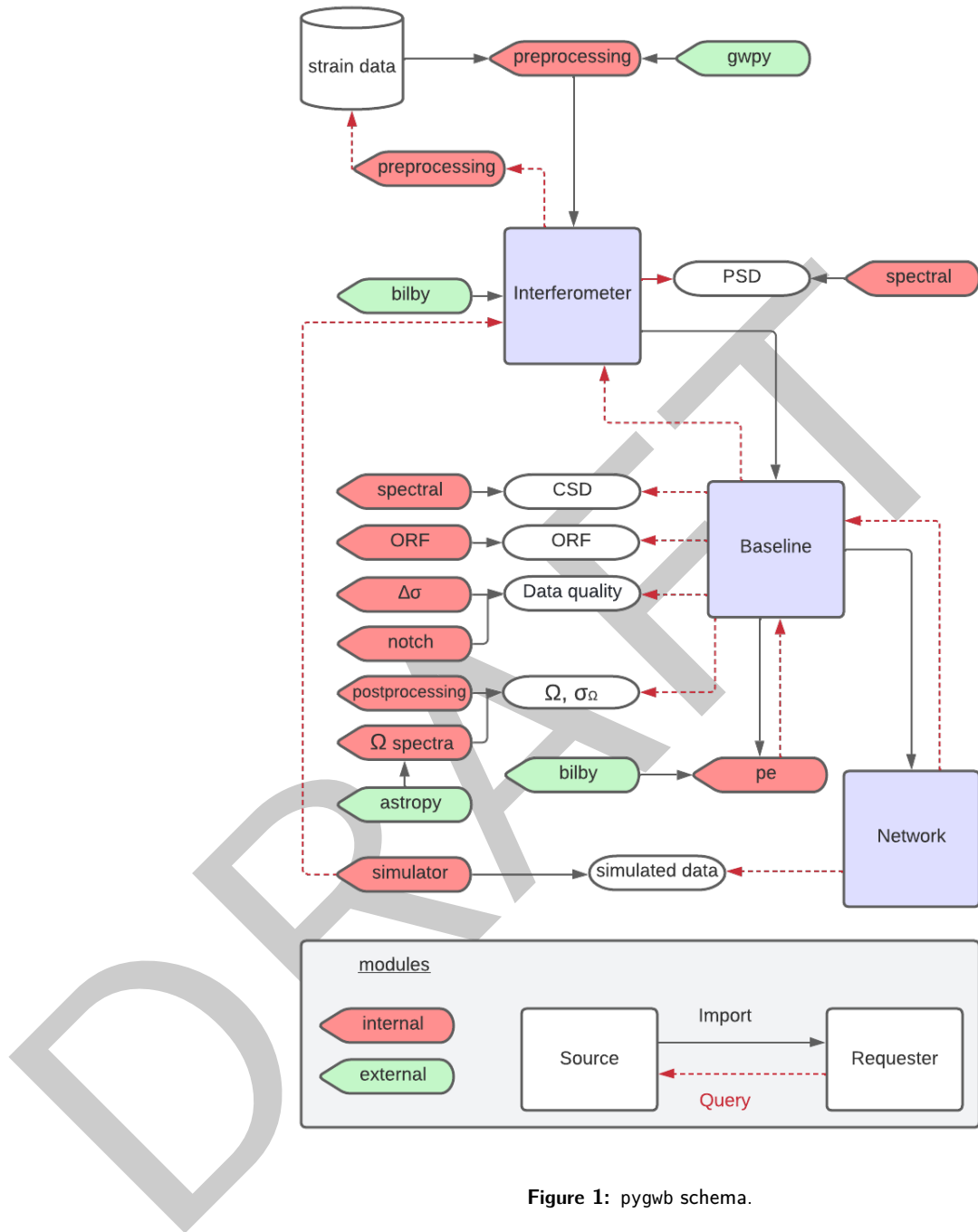


Figure 1: pygwb schema.

Acknowledgements

We thank S. Banagiri, S. Bose, T. Callister, F. De Lillo, L. D’Onofrio, F. Garufi, G. Harry, J. Lawrence, V. Mandic, A. Macquet, I. Michaloliakos, S. Mitra, K. Pham, R. Poggiani, T. Regimbau, J. Romano, N. van Remortel, and H. Zhong for contributing to the review and tests of the code.

We thank the LVK stochastic group for its support. AIR is supported by the NSF award 1912594. ARR is supported in part by the Strategic Research Program “High-Energy Physics” of the Research Council of the Vrije Universiteit Brussel and by the iBOF “Unlocking the

111 Dark Universe with Gravitational Wave Observations: from Quantum Optics to Quantum
 112 Gravity” of the Vlaamse Interuniversitaire Raad and by the FWO IRI grant I002123N “Essential
 113 Technologies for the Einstein Telescope”. KT is supported by FWO-Vlaanderen through grant
 114 number 1179522N. PMM is supported by the NANOGrav Physics Frontiers Center, National
 115 Science Foundation (NSF), award number 2020265. LT is supported by the National Science
 116 Foundation through OAC-2103662 and PHY-2011865. KJ is supported by FWO-Vlaanderen
 117 via grant number 11C5720N. DD is supported by the NSF as a part of the LIGO Laboratory.

118 This material is based upon work supported by NSF’s LIGO Laboratory which is a major
 119 facility fully funded by the National Science Foundation. LIGO was constructed by the
 120 California Institute of Technology and Massachusetts Institute of Technology with funding
 121 from the National Science Foundation, and operates under cooperative agreement PHY-
 122 1764464. Advanced LIGO was built under award PHY-0823459. The authors are grateful
 123 for computational resources provided by the LIGO Laboratory and supported by NSF Grants
 124 PHY-0757058 and PHY-0823459.

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