





1 QhX: A Python package for periodicity detection in red 2 noise

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10 Summary

11 QhX is a Python-based package developed for periodicity detection in red noise time series data,
12 with a focus on the Vera C. Rubin Observatory Legacy Survey of Space and Time (LSST,
13 [Ivezić et al., 2019](#)). Traditional methods, such as those based on the Fourier Transform, often
14 struggle with red noise signals, which are prevalent in real-world datasets. The core of QhX is a
15 2D Hybrid method (cross-correlation of wavelet transforms of light curves, [Kovačević et al.,](#)
16 [2018](#)) that operates in a “period-period” phase space, capable of detecting oscillations in one
17 or more light curves. The QhX pipeline first transforms input data into a time-period plane
18 via wavelets and then (auto)correlates the resulting wavelet matrices to obtain a correlation
19 density in the period-period plane. After integrating the correlation density, the final decision
20 on detected periods is made by a statistical robovetter based on the significance, upper and
21 lower errors of detected periods, and the Intersection over Union (IoU) metric for measuring
22 the proximity and overlap of periods across bands. Beyond compiling the numerical catalog of
23 vetted periods, QhX offers visualization across photometric bands.

24 **Statement of need**

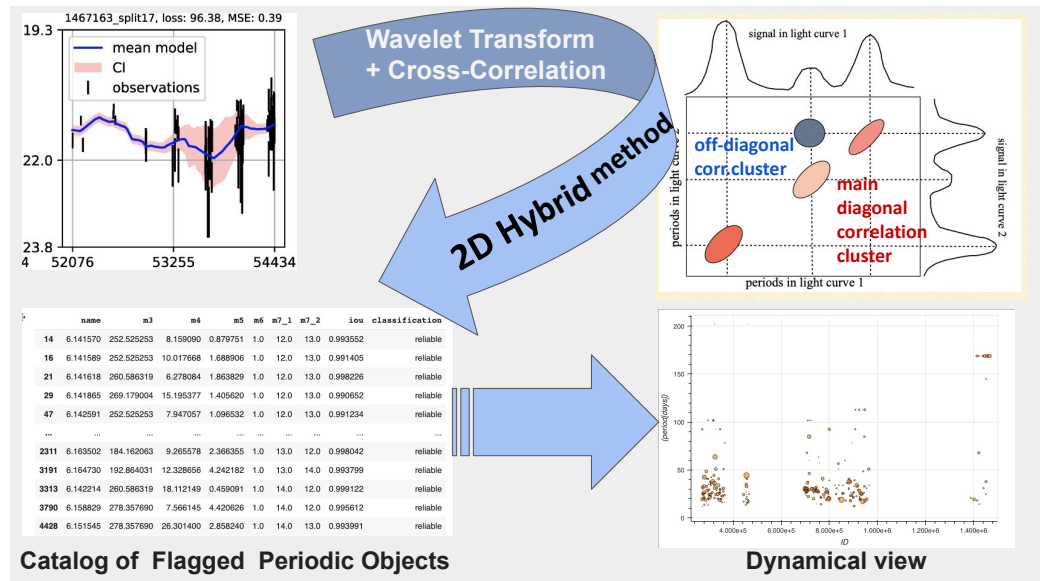


Figure 1: The left panel shows a 1D light curve with observational data (black error bars) and a model (blue line). QhX applies wavelet transforms to convert the time-series data (either observed or modeled) into the time-frequency domain, then cross-correlates wavelet transform matrices generating a correlation density map (right) in 2D period-period phase space where clusters reveal consistent periodic signals. After applying statistical procedures, the package generates a numerical catalog of flagged periodic objects (bottom left) and a dynamical view of detected periods across objects (bottom right).

25 Periodic variability can be encountered across a wide range of astronomical objects, from
 26 asteroids and stars to quasars. However, identifying meaningful signals in these variable sources
 27 is often complicated by red noise (see, e.g., Figure 1 in Gaia Collaboration et al., 2019; Kasliwal
 28 et al., 2015; Kovačević, Radović, et al., 2022), which shows fractal-like patterns across time
 29 scales (Belete et al., 2018; Vio et al., 1991). This, along with the non-stationary nature of
 30 signals and often unfavorable sampling (Brandt et al., 2018; D’Orazio & Charisi, 2023), makes
 31 identifying coherent signals challenging. Traditional time-frequency analysis, limited by the
 32 Fourier uncertainty principle (i.e., Gabor limit, Gabor, 1947), struggles with these complex
 33 signals, highlighting the need for a nonlinear approach (Abry et al., 1995; Cohen, 1995) for
 34 analyzing astronomical signals like quasar light curves.

35 QhX provides a robust framework with features specifically designed to address these challenges.
 36 The first feature of QhX is its core 2D Hybrid method (see Figure 1), detailed in (Kovačević
 37 et al., 2018), with an analogy to 2D Correlation Spectroscopy (Noda, 2018) discussed in
 38 Kovačević (2024). 2D Hybrid method enables a nonlinear approach (via cross-correlation),
 39 expanding detection into a two-dimensional period-period phase space. By applying wavelet
 40 transforms, QhX maps time-series data into a time-frequency domain, then (auto)correlates
 41 it to produce a correlation density in the period-period plane, enhancing detection in red
 42 noise-dominated data.

43 To further ensure robust detection, QhX uses an innovative Intersection over Union (IoU)
 44 metric alongside standard statistical measures (significance and upper and lower error bounds)
 45 to assess the overlap of detected periods across optical bands and objects. In addition to
 46 correlation density maps, each period is visualized as the center of an “IoU ball,” with a
 47 radius that represents relative error, calculated as the mean of the upper and lower error
 48 bounds—similar to a circular aperture in photometry (Saxena et al., 2024).

49 The third feature of QhX introduces a novel approach beyond traditional periodogram and
 50 wavelet transform plots by generating both numerical and interactive visual catalogs. These
 51 catalogs rank periodicity candidates by reliability, allowing for interactive inspection of signal
 52 consistency—a level of interpretability that static plots cannot achieve.

53 **QhX structure**

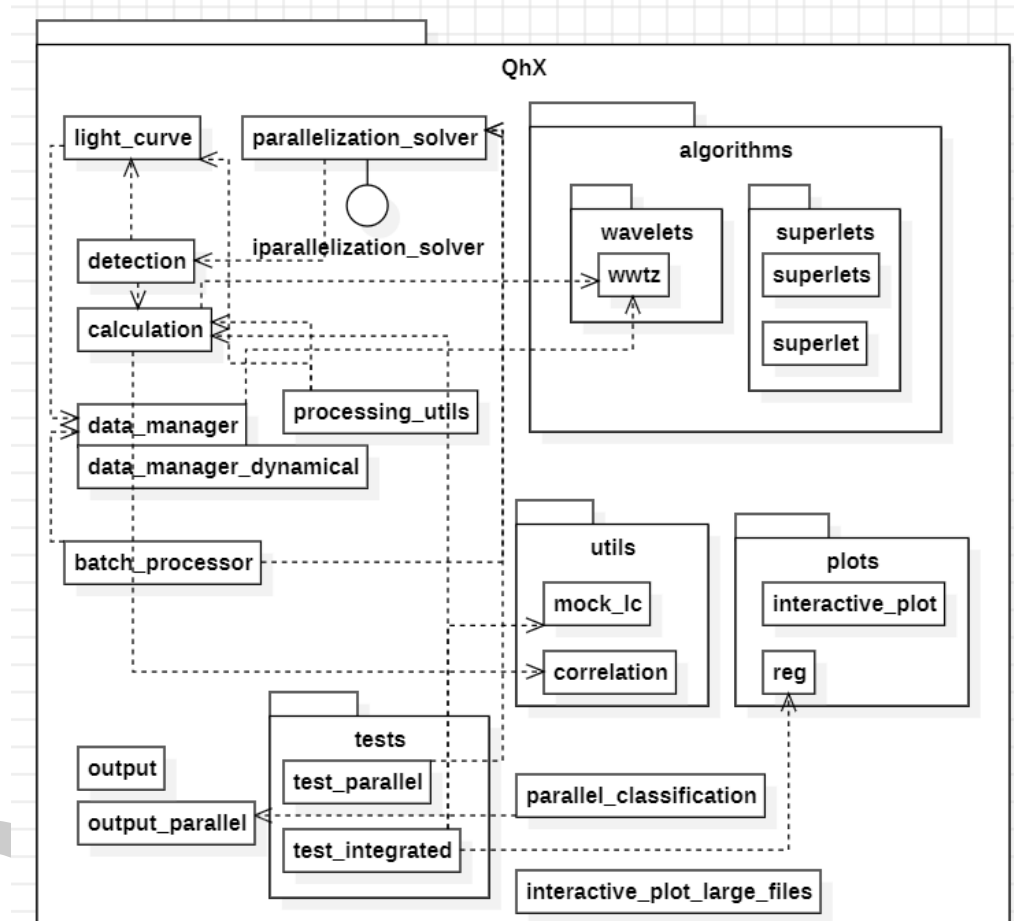


Figure 2: Schematic representation of the QhX package architecture.

54 The QhX package is a modular and extensible API (see Figure 2) designed for efficient
 55 detection and analysis of periodic signals in astronomical time series data, particularly for
 56 projects like LSST. Given that astronomical data analysis often requires rapid prototyping and
 57 experimentation with various algorithms, the modular design of QhX enables users to easily
 58 swap or modify functions, supporting diverse research needs without the constraints of a fixed
 59 class structure.

60 The package is organized into interconnected modules, each fulfilling a specialized role:

61 **1. Core Algorithms:**

- 62 ■ The algorithms module provides essential signal-processing techniques, including
 63 the Weighted Wavelet Z-Transform (wwtz) and prototype superlet transforms, which
 64 enable robust analysis of continuous time series data.

- 65
- 66 ■ The correlation function within the `utils` module supports the 2D Hybrid method
 - 67 by converting light curve data into wavelet matrices and performing (auto)correlation,
 - 68 creating correlation density maps that highlight periodicity as main diagonal
 - 69 clusters in the period-period phase space.
- 70 **2. Signal Detection and Validation:**
- 71 ■ The detection module identifies candidate periodic signals and assesses their validity
 - 72 using statistical measures, including significance testing and error calculations, informed
 - 73 by methods developed with the LSST community (Johnson et al., 2019). The Intersection
 - 74 over Union (IoU) metric is applied to ensure robust detection across bands and objects.
 - 75 ■ Robovetters, or statistical validation tools, finalize the reliability of detected periods,
 - 76 adding an additional layer of quality control.
- 77 **3. Data Management:**
- 78 ■ The `data_manager` and `data_manager_dynamical` modules manage data flow, handling
 - 79 tasks like data loading, outlier removal, and format compatibility. They also support
 - 80 custom data loaders to process various data formats.
 - 81 ■ The `batch_processor` and `parallelization_solver` modules optimize task distribution
 - 82 across multiple processors, boosting computational efficiency for large datasets.
 - 83
- 84 **4. Visualization and Output:**
- 85 ■ The `plots` module includes tools for creating interactive visualizations, such as
 - 86 `interactive_plot`, which allows for exploring detected periodicities across bands
 - 87 and objects. For large datasets, `interactive_plot_large_files` enables in-depth
 - 88 inspection of signal consistency.
 - 89 ■ The `output` and `output_parallel` modules handle result storage, supporting both
 - 90 single-threaded and parallelized workflows.
- 91 **5. Testing and Validation:**
- 92 ■ The `tests` module, with functions like `test_parallel` and `test_integrated`, validates
 - 93 the functionality of different components, ensuring robustness across various
 - 94 processing setups.

95 QhX (Version 0.2.0) is a standalone, open-source package designed for handling datasets with

96 varying numbers of filters across surveys. It offers both dynamic and fixed modes, along

97 with parallel processing capabilities for large datasets. The `dynamical_mode.py` module offers

98 optional inclusion of observational errors, enhancing the accuracy of periodic signal detection

99 across multiple bands.

100 Representative Applications

101 The QhX method has been applied to various studies, including:

- 102 ■ **Quasar Periodicity:** Investigating periodic signals in various quasars (Fatović et al., 2023;
- 103 Kovačević et al., 2018, 2019; Kovačević, Popović, et al., 2020).
- 104 ■ **Quasi-Periodic Oscillations:** Analyzing oscillations in quasars (Kovačević, Yi, et al.,
- 105 2020).
- 106 ■ **Very Low-Frequency Signals:** Detecting VLF signals variability before, during and after
- 107 earthquakes (Kovačević, Nina, et al., 2022).

108 Additionally, the QhX pipeline is listed as a [directable software in-kind contribution](#) to the LSST

109 project, highlighting its role in the LSST.

110 Documentation and Tutorials

111 Comprehensive documentation for QhX, available at [Github Pages](#), includes several example

112 notebooks:

- 113 ▪ **Basic Tutorial:** Introduces the fundamentals of QhX using a mock light curve, helping
- 114 new users get started quickly.
- 115 ▪ **Parallel Processing Example:** Demonstrates how to perform parallel processing with
- 116 quasar light curves from the [LSST AGN Data Challenge database](#), showcasing the
- 117 software's capability to handle large datasets efficiently.
- 118 ▪ **Task Distribution:** Showcases how to distribute tasks across multiple processors using,
- 119 enhancing computational performance.
- 120 ▪ **Merging Large Files:** Provides guidance on handling extensive datasets by merging large
- 121 files, which is essential for high-volume data analysis.
- 122 ▪ **Visualization of Large Datasets:** Illustrates how to visualize large files obtained from
- 123 High-Performance Computing (HPC) environments, enabling effective interpretation of
- 124 results.

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