



## **Quality Control Procedures for Real-Time Water Level Data**

<b>Date</b>	<b>Revision Description</b>	<b>Notes</b>
July 2022	<ul style="list-style-type: none"><li>• Original Document</li></ul>	
November 2023	<ul style="list-style-type: none"><li>• Formal organization of Hohonu QC methodology</li><li>• Added comparison of raw and clean data</li><li>• Added comparison between clean Hohonu and NOAA data</li></ul>	<ul style="list-style-type: none"><li>• Identified future upgrades to QC methodology to increase precision and alignment with NOAA methods</li></ul>
July 2024	<ul style="list-style-type: none"><li>• Addition of overland station QC methodology</li></ul>	

### **Purpose and Scope**

Hohonu provides precision water level monitoring and forecasting to help communities prepare for and respond to flooding. The organization maintains a network of over 100 water level sensors located in 14 states, and it uses the latest technologies in hardware, software, and data science in order to deliver reliable and accessible water level data to its customers.

Adherence to rigorous quality assurance (QA) and quality control (QC) protocols is essential for providing high-quality data to stakeholders. The U.S. Integrated Ocean Observing System (IOOS) provides authoritative procedures through the Quality Assurance/Quality Control of Real-Time Oceanographic Data (QARTOD) program<sup>1</sup>. Additional resources are available specifically to QC flooding data on a real-time basis<sup>8</sup>. Hohonu implements these procedures within its own QA/QC program.

**This document provides an overview of the QC methodology used by Hohonu to ‘clean’ raw sensor data and ensure the provision of high-quality real-time water level measurements to its customers.** It is intended as technical support for users of Hohonu’s data, and covers both Hohonu’s tidal and overland sensor networks.. While procedures documenting data acquisition QA is outside the scope of this document, additional information on sensor specifications, calibration and maintenance, and leveling are referenced at the end<sup>2,3</sup>.

## Data Processing: Tidal Stations

The following section applies to Hohonu's tidal sensors which measure water levels that change continuously and periodically along coastlines. After water level data are acquired from a Hohonu tidal sensor, a set of internal and external QC steps are applied to convert 'raw' data into 'clean' data ready for storage, reporting, and forecasting (Figure 1). The below processes are implemented for all live Hohonu tidal sensors.

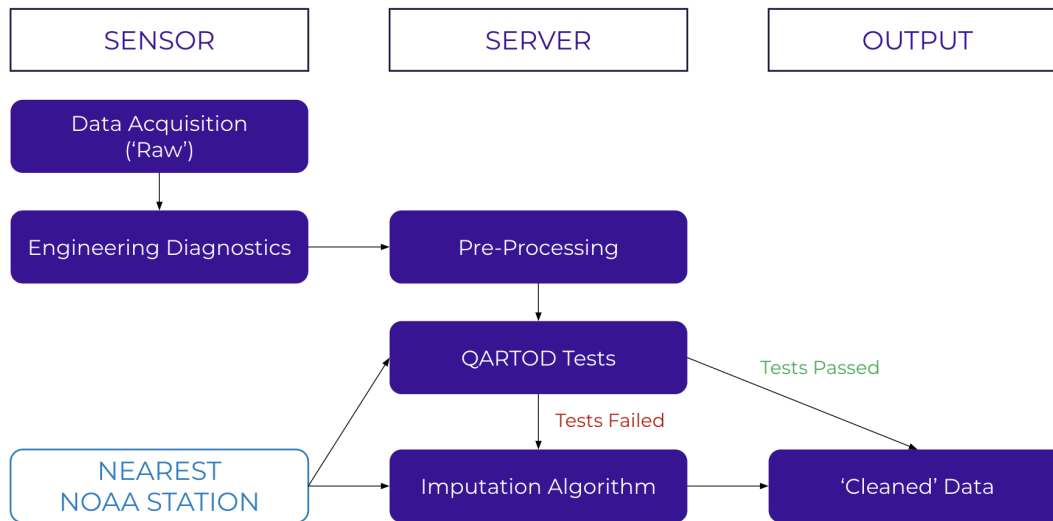


Figure 1. Overview of the Hohonu QC methodology for tidal sensors

### 1. Engineering Diagnostics

*Frequency: every sample*

Hohonu sensors sample at a rate of 1 Hz and internally calculate a 6-minute means from them. The first step of QC is performed onboard by removing all samples measured outside of 3 standard deviations from the mean. Data are then transmitted to the Hohonu server.

### 2. Data Pre-Processing

*Frequency: every 6 hours*

Prior to implementing QARTOD tests, Hohonu pre-processes water level datasets by first aligning samples to a 6-minute interval and identifying missing data points. Duplicate measurements and any samples measured at less than 0.5m D2W (distance from sensor to water) are removed. Flags are generated if sensor data indicate improper readings or when a sensor has become detached.

Finally, water level measurements are converted from the station datum (D2W) to a geodetic datum (i.e. NAVD88) or, if unavailable, a tidal datum (typically mean low low

water, or MLLW) using a pre-calculated conversion factor. The purpose of this step is to make comparable Hohonu data to the nearest NOAA station for certain QARTOD tests and missing data imputation. The tidal data datum for each station is calculated per the NOAA-aligned methodology referenced below<sup>1</sup>.

### 3. QARTOD Tests

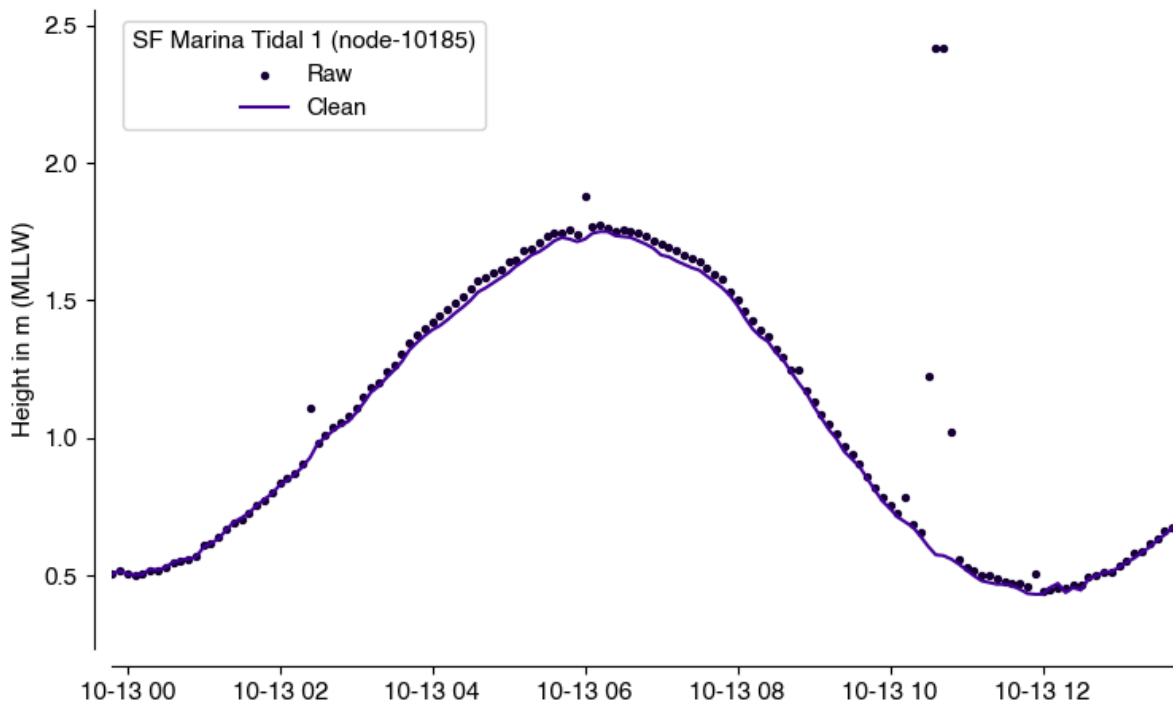
*Frequency: every 6 hours*

The IOOS QARTOD procedures provide a hierarchy of 11 QC tests for real-time water quality data. Hohonu implements 6 of these tests on its server using automated scripts modified from those provided on the IOOS Github repository<sup>4</sup>. These tests include:

- *Time/Gap Test*: Determines that the most recent data point has been measured and received within the expected time window
- *Gross Range Test*: Checks that values are within reasonable range bounds. This range is established using 3 times the standard deviation
- *Spike Test*: Determines if there is a spike in the data given the previous point. Spikes are assessed using a cutoff of 3 times the standard deviations
- *Rate of Change Test*: Checks the first order difference of a series of values to see if there are any values exceeding a threshold. This threshold is established using 3 times the standard deviation.
- *Flat Line Test*: Checks for consecutively repeated values within a tolerance. The tolerance is set as 0.00001m.
- *Neighbor Test*: Compares Hohonu data with nearest NOAA station to detect anomalies. Flags occur when the stations' difference in subsequent samples is greater than 1m.

All test thresholds are determined automatically using the data. If any of these tests fail, the data point is flagged and its value is removed. Hohonu only uses the GOOD, FAILED, and MISSING flag types provided in the QARTOD procedures.

Of the remaining QARTOD tests, two are implemented onboard the sensor (Syntax Test and Location Test), one is performed manually upon daily station review (Climatology Test), and one is not applicable given the sensor type (Multivariate Test). Finally, Hohonu is presently assessing incorporating the Attenuated Signal Test into its automated procedures.



*Figure 2. Comparison between raw and cleaned tidal data collected in San Francisco, CA. In this example, the raw data failed the spike, rate of change, and neighbor tests in multiple points, and were accordingly flagged, removed, and imputed.*

#### 4. Imputation Algorithm

*Frequency: every 6 hours*

Any missing data or data that failed QARTOD tests are filled using an imputation algorithm. The algorithm is a multivariate imputer that estimates the missing values by utilizing available data in similar features<sup>5</sup>. In this case, the similar feature refers to data collected from the nearest NOAA station. Data are considered 'clean' upon passing all QARTOD tests or after imputation (Figure 2). Finally, clean data are converted back into the original station datum.

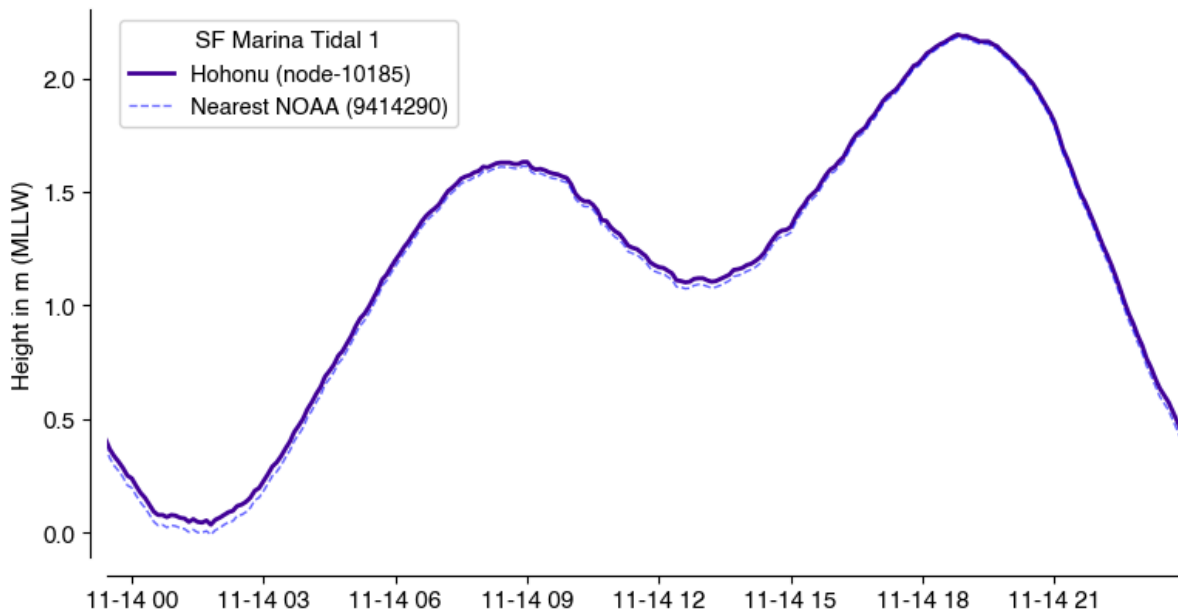


Figure 3. Comparison between cleaned Hohonu data and data from the nearest NOAA station (located approximately 1.4 miles away). Data at these stations are strongly correlated (Pearson correlation of 0.999 for the plotted time series and 0.996 over the preceding 5 months).

### Data Processing: Overland Stations

The following section applies to Hohonu’s overland sensors which measure either distance to a static surface or inundation in the area as it is occurring. After water level data are acquired from a Hohonu overland sensor, a set of internal and server-side QC steps are applied to convert ‘raw’ data into various tiers of ‘clean’ data ready for storage, reporting, and forecasting. Hohonu’s QC system for overland stations utilizes many of the same tests employed for tidal stations. The following processes are implemented for all live Hohonu overland sensors.

#### 1. Data Pre-Processing

*Frequency: every sample transmission*

Hohonu sensors sample at a rate of 1 Hz and internally calculate a mean value approximately every 30 seconds. The first step of QC is performed onboard by removing all samples measured outside of 3 standard deviations from the mean and then recalculating the mean. Flags are generated if sensor data indicate improper readings or when a sensor has become detached. Data are then transmitted to the Hohonu server where they are deduplicated by timestamp and saved to the

database. These data can be accessed via the Hohonu API as 'QC Level 0' data quality, but are presently unavailable via the Hohonu Dashboard.

Prior to the server-side QC steps, overland measurements are converted from the station datum to the geodetic datum (i.e. NAVD88) using surveyed information about the sensor's position relative to the ground.

## 2. Basic QC Testing and Filtering

*Frequency: every sample transmission*

The next step of the overland QC process is a basic filtering step which yields what Hohonu labels as 'Raw' data on its Dashboard and 'QC Level 1' data through the API. First, a Gross Range test is performed by flagging all points measured belowground, within the sensor's noise floor (i.e. ½" above the ground) and noise ceiling (2' below the sensor). Flagged points are replaced with the ground height, and filtered data are saved in the database. In testing, this basic filtering step typically leads to the removal of > 90% of non-inundation noise in the signal.

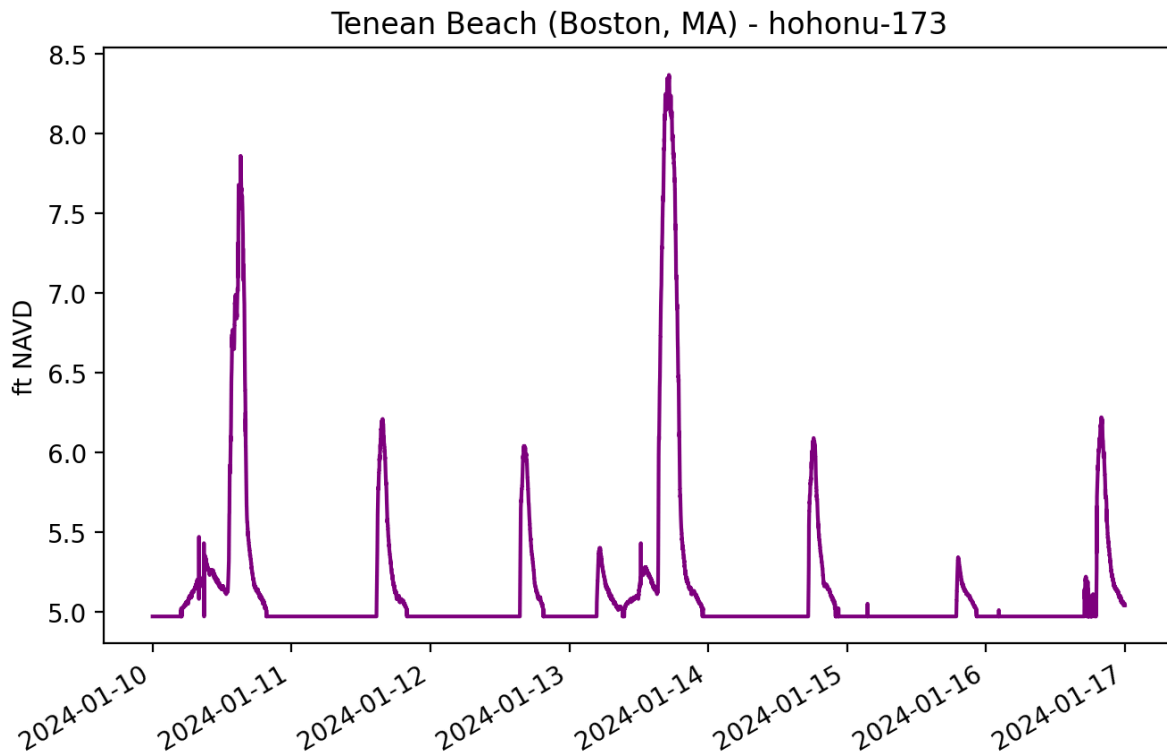


Figure 4. Overland data after passing through the Basic QC step

## 3. Advanced QC Testing and Filtering

*Frequency: every hour*

The final QC step is a more advanced process that is based off of those used in the QARTOD process<sup>1</sup>, though thresholds for each of these tests are set based on testing and previously observed inundation data. The result data are labeled as 'Clean' on the Dashboard and 'QC Level 2' via the API.

To initiate this process, data that have previously been passed through the Basic QC step are pulled from the database and resampled to 1-min mean values. Note that any missing data points in the 1-min time series are presently not imputed. The resampled data then go through a series of advanced QC tests which include a Spike Test, a Rate of Change Test, and a Climatology Test that flags above ground measurements collected outside the window of extreme tidal level. If a data point is flagged for a spike or extreme rate of change, it is replaced by an interpolation of surrounding points; if a data point is flagged based on a climatology test, it is replaced with the ground height. Next, the signal is scanned one final time for small inundation events, defined as jumps in the signal with duration less than 30 minutes. Small inundations are flagged and replaced with the ground height. The small inundation filter helps capture any remaining noise not covered by the rate of change, spike, and climatology filters, though some noise may remain even after this step.

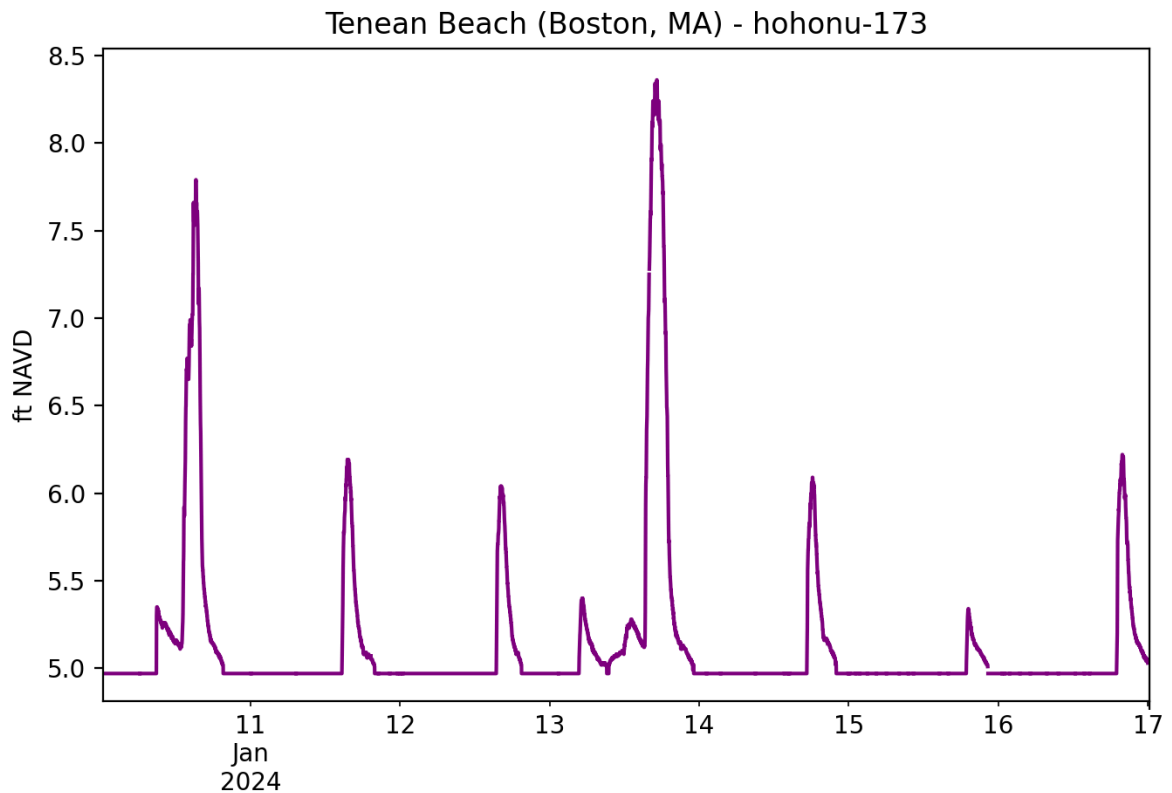


Figure 5. Overland data after passing through the Advanced QC step

## Data Storage and Reporting

All raw and cleaned sensor data are stored on the Hohonu server and are available for further QC, reporting, manipulation, and forecasting. Data may also be compared against NOAA control stations (Figure 3). Users can access these data via the user interface on the Hohonu Dashboard, via a mobile iOS app, or through the Hohonu Public API<sup>6</sup>. Present functionality also allows for water level data to be converted between different datums including the Flood Depth datum for overland stations. References for data access are provided below.

Any flags generated during the QC process are also stored on the Hohonu server and are examined daily by manually inspecting an aggregated station report. This inspection allows for response to issues with data stations. Hohonu is constantly working to update its protocols to align with the best available methodologies. Future updates will include more robust training modules for users and automatic thresholding to differentiate between bad data and acute events (e.g. storm surge or tsunamis).

## References

1. National Ocean Service, Manual for real-time quality control of water level data: a guide to quality control and quality assurance for water level observations, 2021. <https://repository.library.noaa.gov/view/noaa/29274>
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3. Hohonu FAQ: <https://www.hohonu.io/faqs>
4. IOOS QC: QARTOD and other Quality Control tests implemented in Python. [https://ioos.github.io/ioos\\_qc/](https://ioos.github.io/ioos_qc/)
5. Scikit-learn IterativeImputer: <https://scikit-learn.org/stable/modules/generated/sklearn.impute.IterativeImputer.html>
6. Hohonu Public API. <https://hohonu.readme.io/reference/getting-started-with-your-api>
7. FloodNet Methodology. New York University. <https://www.floodnet.nyc/methodology/>