

Qlunc: Quantification of lidar uncertainty

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Summary

Wind lidar is a flexible and versatile remote sensing device for wind energy applications (Beck & Kühn, 2017) which measures the wind vector remotely using laser light backscattered from aerosols. They are a key tool for wind energy and meteorology. Like any measurement method, it is essential to estimate their uncertainty. Qlunc, which stands for Quantification of lidar uncertainty, is an open-source python-based tool to create a digital twin of the lidar hardware, and estimate the uncertainty of wind lidar wind speed measurements.

Qlunc contains models of the uncertainty contributed by individual lidar components and modules (represented by python objects, which in turn represent physical lidar objects), that are then combined considering their different nature, to estimate the uncertainties in wind lidar measurements. The modules are based on the OpenLidar architecture (Clifton et al., 2019) and can be easily adapted for particular use cases thanks to the modularity of the code (see Figure 1). The terminology for the components and modules defined within Qlunc has also been aligned with a community-driven wind lidar ontology, which is in development (N. Vasiljevic, 2021; N. Vasiljevic & Clifton, 2021).

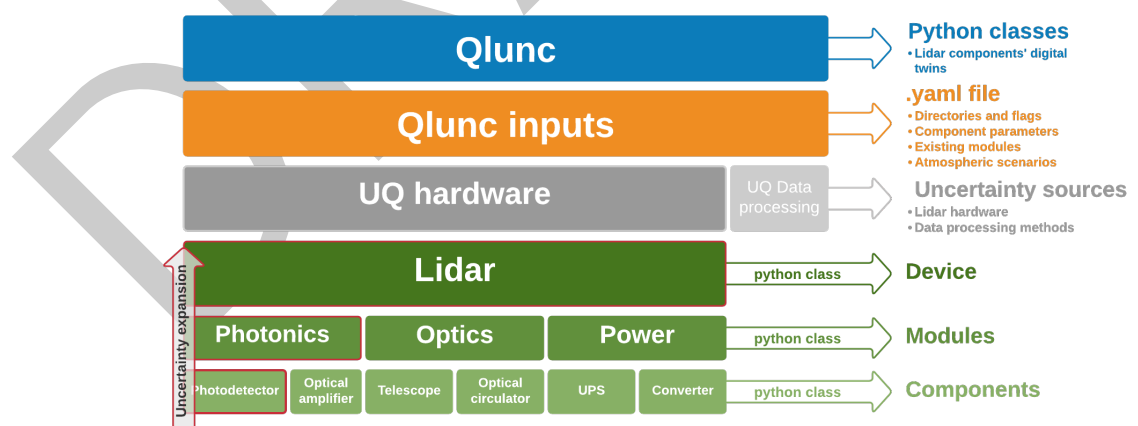


Figure 1: Qlunc basic structure.

The first release is focused on velocity azimuth display (VAD) (Browning & Wexler, 1968) scans and forward-looking nacelle-mounted measuring modes, which are common wind-energy-industry applications. Besides uncertainty estimations, Qlunc's functions could be extended for other applications, for example to compare different wind velocity vector calculation methods. This, combined with the underlying open-source code, defines an attractive scenario for sharing knowledge and fostering collaboration on wind lidars.

Statement of Need

Wind lidars are measuring devices, and as for any other measuring systems, their measurements have uncertainties (Borraccino & Courtney, 2016). Therefore, as already stated, it is crucial to assess their measurements uncertainty in order to increase confidence in lidar technology.

Measurement uncertainty means doubt about the validity of the result of a measurement (JCGM, 2008). It represents the dispersion of the values attributed to a measurand. The ability to simulate uncertainty through a model such as Qlunc, is important for judging measurement data but can also be useful for designing and setting up experiments and optimizing lidar design. Because wind lidar is important for wind energy applications (Clifton et al., 2018), better models for wind lidar hardware (e.g., Qlunc) and measurement processes (e.g., through MOCALUM (Nikola Vasiljevic, 2020) or YADDUM (Nikola Vasiljevic, 2019), with which Qlunc can feasibly combine) will directly contribute to the adoption of wind lidar for wind energy applications.

This project is influenced by fundamental open science principles (Bot et al., 2018). The scope is to create an open, standardized and collaborative framework to describe both generic and specific lidar architectures, characterize lidar uncertainties, and provide the tools for others to contribute within this framework.

Future development roadmap

Over the next year, we plan to implement further lidar hardware modules in the model and compute their combined uncertainties. In addition, we will identify main data processing methods and include those that we consider the highest contributors to uncertainty.

We also plan to further align the terminology used in Qlunc with the IEA Wind Task 32 controlled vocabulary for wind lidar (N. Vasiljevic et al., 2021). This will make it easier for users to understand what each of the modules and components do, and promotes interoperability.

All documentation from the project, tutorials, and raw code will be published through a web site, to enable users to dive into the numerical framework and get used to the Qlunc routines. We welcome contributions from the wind lidar community.

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