

1 pvlib python: 2023 project update

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5 Summary

6 pvlib python is a community-developed, open-source software toolbox for simulating the
7 performance of solar photovoltaic (PV) energy systems. It provides peer-reviewed reference
8 implementations of over 100 empirical and physics-based models from the scientific literature,
9 including solar position algorithms, irradiance models, thermal models, and PV electrical
10 models. In addition to these individual low-level model implementations, pvlib python provides
11 high-level constructs that chain these models together like building blocks to form complete
12 “weather-to-power” photovoltaic system models. It also provides functions to fetch and import
13 a wide variety of weather datasets useful for PV modeling.

14 pvlib python has been developed since 2015 and follows modern best practices for open-source
15 python software, with comprehensive automated testing, standards-based packaging, and
16 semantic versioning. Its source code is developed openly on GitHub and releases are distributed
17 via the Python Package Index (PyPI) and the conda-forge repository. pvlib python’s source
18 code is made freely available under the permissive BSD-3 license.

19 Here we present an update on pvlib python, describing capability and community development
20 since our 2018 publication ([Holmgren, Hansen, & Mikofski, 2018](#)).

21 Statement of need

22 PV performance models are used throughout the field of solar photovoltaics. The rapid increase
23 in scale, technological diversity, and sophistication of the global solar energy industry demands
24 correspondingly more capable models. Per the United States Department of Energy, “the
25 importance of accurate modeling is hard to overstate” ([Solar Energy Technologies Office,
26 2022](#)).

27 Compared with other PV modeling tools, pvlib python stands out in several key aspects. One
28 is its reusable toolbox design, providing the user a level of flexibility beyond that of other tools.
29 Rather than organizing the user interface around pre-built modeling workflows, pvlib python
30 makes the individual “building blocks” of PV performance models accessible to the user. This
31 allows the user to assemble their own model workflows, including the ability of incorporating
32 their own custom modeling steps. This flexibility is essential for applications in both academia
33 and industry.

34 Another key aspect of pvlib python is that it is used via a general-purpose programming
35 language (Python). In addition to being more generally flexible, powerful, and scalable than a
36 traditional graphical user interface, pvlib python’s interface as a programming library allows it
37 to be combined with other Python packages. This enables integration with database query,
38 data manipulation, numerical optimization, plotting, and reporting packages, to name a few.

39 A final key aspect of pvlib python is its open peer review approach and foundation in published
40 scientific research, allowing it to be developed by a decentralized and diverse community of

41 PV researchers and practitioners without compromising its focus on transparent and reliable
42 model implementations.

43 These core tenets, along with sustained contributions from a passionate and committed
44 community, have lead to its widespread adoption across the PV field (Stein & Hansen, 2022).
45 In support of the claim that pvlib python provides meaningful value and addresses real needs,
46 we offer these quantitative metrics:

- 47 1. Its 2018 JOSS publication, at the time of this writing, ranks 14th by citation count out
48 of the 2000+ papers published by JOSS to date.
- 49 2. The Python Package Index (PyPI) classifies pvlib python as a “critical project” due to
50 being in the top 1% of the index’s packages by download count.
- 51 3. The project’s online documentation receives over 400,000 page views per year.
- 52 4. pvlib python was found to be the third most used python project in the broader open-
53 source sustainability software landscape, with the first two being netCDF4 utilities
54 applicable across many scientific fields (Augspurger et al., 2023).

55 Functionality additions

56 To meet new needs of the PV industry, substantial new functionality has been added in the
57 roughly five years since the 2018 JOSS publication.

58 First, several dozen new models have been implemented, expanding the package’s capability
59 in both existing and new modeling areas and prompting the creation of several new modules
60 within pvlib python. In response to the recent rapid increase in deployment of bifacial PV, a
61 capability enhancement of particular note is the inclusion of models for simulating irradiance
62 on the rear side of PV modules. Other notable additions include methods of fitting empirical
63 PV performance models to measurements and models for performance loss mechanisms like
64 soiling and snow coverage.

65 Figure 1 summarizes the number of models (or functions) per module for pvlib python versions
66 0.6.0 (released 2018-09-17) and 0.10.1 (released 2023-07-03), showing a substantial capability
67 expansion over the last five years.

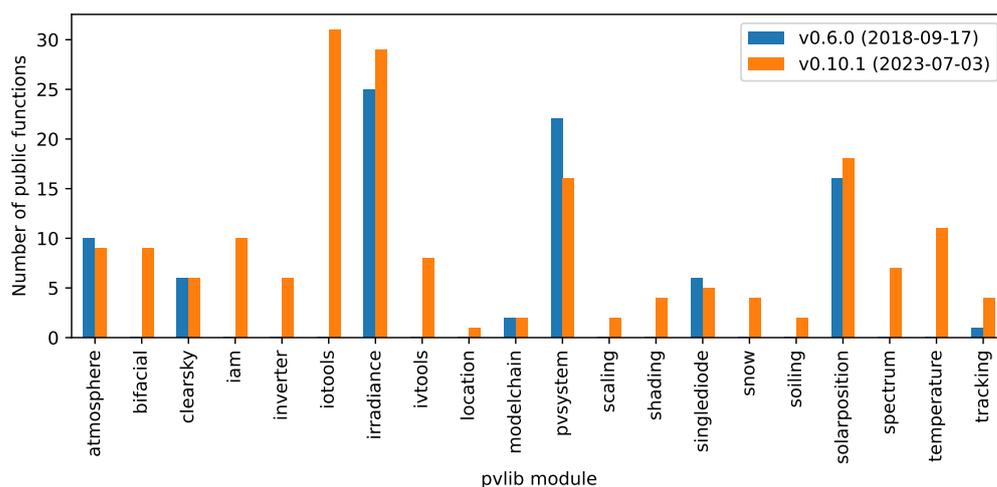


Figure 1: Comparison of public function counts for selected pvlib modules for v0.6.0 and v0.10.1. Some modules are smaller in v0.10.1 due to moving functions to new modules (e.g. from pvsystem to iam).

68 Second, in addition to the new function-level model implementations, the package’s high-
69 level classes have also been expanded to support the complexity of emerging system designs,

70 including heterogeneous systems whose subsystems differ in mounting or electrical configuration
71 and systems that require custom orientation/steering models.

72 Third, the creation of `pvlb.iotools`, a sub-package for fetching and importing datasets
73 relevant to PV modeling. These functions provide a standardized interface for reading data
74 files in various complex data formats, offering conveniences like optionally standardizing the
75 dataset variable names and units to `pvlb`'s conventions (Jensen et al., submitted). As of
76 version 0.10.1, `pvlb.iotools` contains functions to download data from over ten online data
77 providers, plus file reading/parsing functions for a dozen solar resource file formats.

78 These additions are discussed in more detail in (Hansen et al., 2023) and (Anderson et al.,
79 2022). Complete descriptions of the changes in each release can be found in the project's
80 documentation.

81 Community growth

82 It is difficult to comprehensively describe the community around open-source projects like `pvlb`
83 python, but some aspects of it are more easily quantifiable than others. Here we examine
84 the community from a few convenient perspectives, but we emphasize that this section is
85 necessarily a limited view of the community as a whole.

86 First, we examine contributors to `pvlb python`'s code repository. The project's use of version
87 control software enables easy quantification of repository additions (to code, documentation,
88 tests, etc) over time. The project's repository currently comprises contributions from over 100
89 people spanning industry, academia, and government research institutions. Figure 2 shows the
90 number of unique repository contributors over time, demonstrating continued and generally
91 accelerating attraction of new contributors.

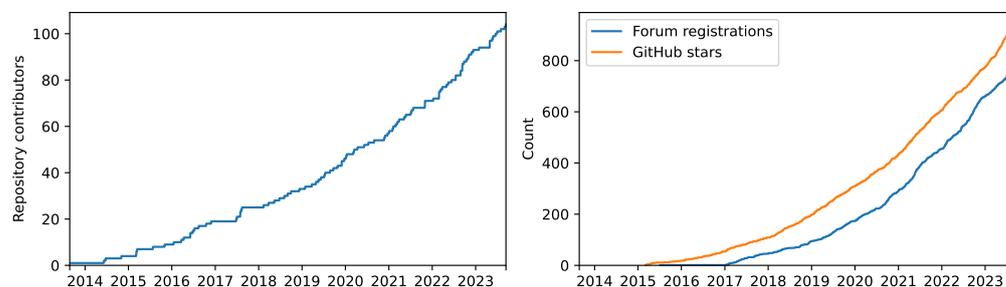


Figure 2: Total repository contributor count over time (left) and other community size statistics (right).

92 However, the project as a whole is the product of not only of those who contribute code but
93 also those who submit bug reports, propose ideas for new features, participate in online fora,
94 and support the project in other ways. Along those lines, two easily tracked metrics are the
95 number of people registered in the `pvlb python` online discussion forum and the number of
96 GitHub "stars" (an indicator of an individual's interest, akin to a browser bookmark) on the
97 `pvlb python` code repository. Figure 2 shows these counts over time. Although these numbers
98 almost certainly substantially underestimate the true size of the `pvlb python` community, their increase
99 over time indicates continued and accelerating community growth.

100 In addition to continuous interaction online, community members sometimes meet in person
101 at user's group and tutorial sessions run by `pvlb python` maintainers and community members
102 alike. To date, these meetings have been held at the IEEE Photovoltaics Specialists Conference
103 (PVSC), the PVPWC Workshops, and the PyData Global conference. Figure 3 shows a timeline
104 of these meetings, along with other notable events in the project's history.

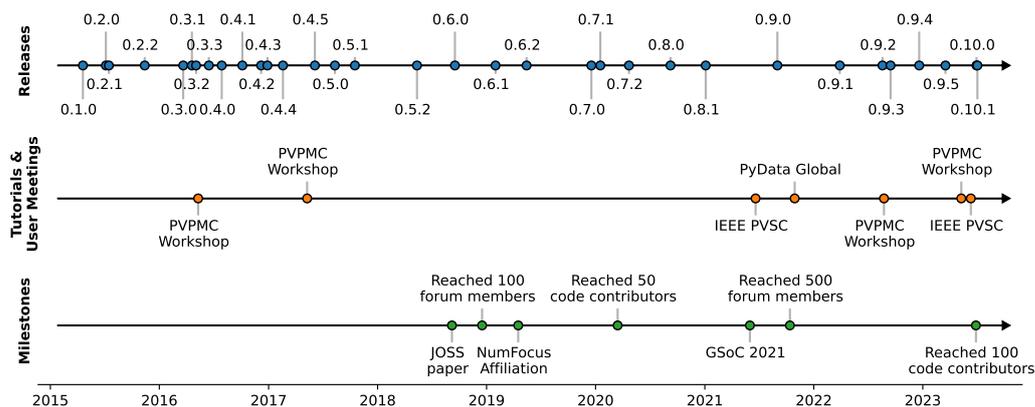


Figure 3: pvlib python major event timeline: releases (top), community events (middle), and other project milestones (bottom).

105 Finally, it is worth pointing out that pvlib python contributors and users are part of a broader
 106 community around not just pvlib python itself but also other members of the pvlib software
 107 “family”: pvanalytics, a package for PV data quality assurance and feature recognition
 108 algorithms (Perry et al., 2022), and twoaxistracking, a package for simulating self-shading in
 109 arrays of two-axis solar trackers (Jensen et al., 2022). Moreover, looking beyond pvlib and its
 110 affiliated packages, we see that Python is proving to be the most common programming language
 111 in general for open-source PV modeling and analysis software. The packages mentioned here
 112 make up one portion of a growing landscape of Python-for-PV projects (Holmgren, Hansen,
 113 Stein, et al., 2018).

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 118 Performance Modeling Collaborative (PVPWC) and other projects
- 119 ■ The Danish Energy Agency through grant nos. 64020-1082 and 134232-510237
- 120 ■ NumFOCUS’s Small Development Grant program
- 121 ■ Google’s Summer of Code program

122 pvlib python benefits enormously from building on top of various high-quality packages that
 123 have become de facto standards in the python ecosystem: numpy (Harris et al., 2020), pandas
 124 (McKinney, 2010), scipy (Virtanen et al., 2020), and numba (Lam et al., 2015) for numerics,
 125 matplotlib (Hunter, 2007) for plotting, sphinx (Komiya et al., 2023) for documentation, and
 126 pytest (Krekel et al., 2004) for automated testing. The project also benefits from online
 127 infrastructure generously provided free of charge, including GitHub (code development and
 128 automated testing) and ReadTheDocs.org (documentation building and hosting).

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