

PDR: The Planetary Data Reader

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Software

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Summary

The Planetary Data Reader, pdr, is an open-source Python-language package that reads data stored in planetary science formats and converts it into standard Python objects. It typically loads images as ndarrays, tables as Pandas DataFrames, and metadata and ancillary data as strings or dicts. pdr's interface is designed to be maximally accessible to the introductory Python user. To read a data file, a user must simply import pdr, then run pdr.read(fn), where fn is the data file or a detached 'label' (metadata) file associated with it. pdr will immediately load the product's metadata, then lazily load data objects when referenced.

pdr reads data products held by the Planetary Data System (PDS) that follow either PDS3 (Jet Propulsion Laboratory, 2009) or PDS4 (Jet Propulsion Laboratory, 2024) standards – meaning in practice that they have metadata labels that generally follow one of these two formats. (It also supports some common scientific interchange data formats that are not PDS, including FITS.) We knew from the outset that many products in the PDS were not fully standards-compliant, particularly those archived under PDS3; its holdings are extremely diverse and span over half a century. For this reason, we took a data-driven development approach rather than attempting to implement these standards to the letter. What this means in practice is that we built pdr around a core of extremely flexible heuristics that permit it to permissively accept and correctly handle many products that deviate from the standards.

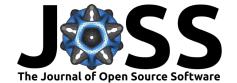
We have developed these heuristics through an iterative design process centered on examination of actually-existing data. We created manifests of the holdings of each of the PDS nodes, then used them to help identify 'types' of data and retrieve representative samples of each 'type' (the PDS holds hundreds of millions of files with a total data volume in the petabytes, so examining every file is impractical).

When we examine a new type, we verify the correctness of pdr's behavior across our sample of that type and make changes to pdr as necessary to support that type's characteristics. We then add one or two individual products of that type to a data corpus we use for regression testing. This has permitted us to design software that conforms to planetary data rather than planetary data standards. Our methods for adding dataset support are further described in Kaufman et al. (2022), and our testing toolchain can be found in Curtis et al. (2024).

pdr is an affliate package of planetarypy (PlanetaryPy Technical Committee, 2024). It is available on the Python Package Index and conda-forge.

Statement of need

- 37 Just accessing data is a major pain point for planetary scientists. Data archived under the
- PDS3 standards can be especially challenging due to inconsistent, specialized, or flatly incorrect
- ₃₉ formatting. While the newer PDS4 standards are significantly stricter, many of the holdings of
- 40 the PDS have not yet been migrated to this standard. pdr can remove months of preparatory
- 41 work, making it faster for scientists to get to core research tasks and making it much more



- practical for them to incorporate data sets they haven't worked with previously into their research.
- The simplicity and consistency of pdr's API, along with its speed and stability, make it ideal for use in automated data processing pipelines. pdr is currently used in a wide variety of planetary projects. These include the Perseverance rover's Mastcam-Z tactical pipeline (St. Clair, Million, et al., 2023) and PDS3 to PDS4 migration pipelines for data from Clementine (St. Clair et al., 2021), Chandrayaan-1 M3 (Pieters et al., 2021), and the Viking Orbiter cameras (St. Clair, Brown, et al., 2023). Its fast metadata parsing features make it especially appealing for converting metadata standards across tens of millions of products. For example, pdr is able to parse metadata and load the image arrays for 100 nominal-sized Mastcam-Z IOFs using their attached PDS3 labels in 1.5 seconds and in 0.75 seconds to parse the metadata alone. More complex labels such as the New Horizons ALICE calibrated Jupyter PDS3 labels took pdr 1.7 seconds to parse metadata for 100 files.

55 Other packages

There is a very wide variety of software intended to read data in planetary science formats.

pdr's most important distinctions are its emphasis on breadth, simplicity, and high compatibility
with other tools. pdr incorporates some of this software, including pds4_tools (Small Bodies
Node & Nagdimunov, 2021) and astropy.io.fits (Astropy Developers, 2024). pdr uses these
packages to read, respectively, PDS4 and FITS files, converting their outputs into standard
Python objects to provide users with a common interface regardless of file format.

lt is important to note that many pieces of software with narrower *format* scope than pdr have wider *application* scope. For instance, GDAL (Rouault et al., 2024) and rasterio (Mapbox, 2024) (which uses GDAL) read a narrower range of data and do not provide as consistent or straightforward an interface, but will deal with map projection transformations; plio (USGS Astrogeology, 2024) only reads data in a few formats, but is capable of applying instrument-specific metadata-parsing rules. Many of these tools also offer write capabilities, which pdr does not. Users who require write capabilities or subdomain-specific behaviors might find narrowly-focused tools more appropriate; they might also find pdr useful as a preprocessor for such tools.

Acknowledgements

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