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Atmospheric Retrievals with petitRADTRANS

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Summary

The `petitRADTRANS` (pRT) codebase has undergone significant updates since its initial publication in Mollière et al. (2019). A retrieval module combining the pRT spectrum calculations with the `MultiNest` (F. Feroz et al., 2009; Farhan Feroz et al., 2019; F. Feroz & Hobson, 2008) and `Ultranest` (Buchner et al., 2014) samplers has been included to streamline retrievals of exoplanet atmospheres in emission and transmission.

SUMMARY HERE

Statement of need

Retrievals are important. Faster sampling using nested sampling. Unique in ability to do c-k, line by line, transmission and emission, or any combination thereof. Widely used in the community (e.g. ...)

petitRADTANS Retrieval Module

Brief Intro

Multiple datasets can be included into a single retrieval, with each dataset receiving its own `RadTrans` object used for the radiative transfer calculation, allowing for highly flexible retrievals where multiple spectral resolutions, wavelength ranges and even atmospheric models can be combined in a single retrieval. Each dataset can also receive scaling factors (for the flux, uncertainties or both), error inflation factors and offsets. Several atmospheric models are built into the `models` module, allowing for a wide range of P-T, cloud and chemistry parameterizations. These models are used to compute a spectrum $\vec{S}$, which is convolved to the instrumental resolution and binned to the wavelength bins of the data using a custom binning function to account for non-uniform bin sizes. The resulting spectrum compared to the data with flux $\vec{F}$ and covariance $\mathbf{C}$ in the likelihood function:

$$-2\log\mathcal{L} = \left(\vec{S}-\vec{F}\right)^T \mathbf{C}^{-1}\left(\vec{S}-\vec{F}\right) + \log\left(2\pi\det\left(\mathbf{C}\right)\right). \quad (1)$$

The second term is included in the likelihood to allow for uncertainties to vary as a free parameter during the retrieval, and penalizes overly large uncertainties.

pRT can compute spectra either using line-by-line calculations, or using correlated-k tables for defining the opacities of molecular species. We include up-to-date correlated-k line lists from Exomol [Tennyson & Yurchenko (2012); mckemmish2016; polyansky2018; chubb2020] and HITEMP (Rothman et al., 2010), with the full set of available opacities listed in the online

documentation. The `exo-k` package is used to resample the the correlated-k opacity tables to a lower spectral resolution in order to reduce the computation time (Leconte, 2021).

Included in pRT is an option to use an adaptive pressure grid with a higher resolution around the location of the cloud base, and a lower resolution elsewhere. The higher resolution grid is 10 times as fine as the remaining grid, and replaces one grid cell above and below the cloud base layer, as well as the cloud base layer cell itself. This allows for more precise positioning of the cloud layers within the atmosphere. Including this adaptive mesh, our pressure grid contains a total of 154 layers when two cloud species are used, which is the standard grid used in this work.

Finally, photometric data are fully incorporated into the retrieval process. As with spectroscopic data, a model is computed using a user-defined function. This model spectrum is then multiplied by a filter transmission profile from the SVO database using the `species` package (Stolker et al., 2020). This results in accurate synthetic photometry, which can be compared to the values specied by the user with the `add_photometry` function.

Combining the c-k opacities of multiple species requires mixing the distributions in g space. Previously, this was accomplished by taking 1000 samples of each distribution. This sampling process resulted in non-deterministic spectral calculations, resulting in unexpected behaviour from the nested sampling process, as the same set of parameters could result in varying log-likelihood. This has been updated to fully mix the c-k distributions. Considering the first species, the second species is added in, and the resulting grid is sorted. The cumulative opacity grid is then mixed with the next species, a process which iterates until every species with significant opacity contributions ($>0.1\%$ of the current opacity in any bin) is mixed in to the opacity grid. Once complete, the resulting grid is linearly interpolated back to the 16 g points at each pressure and frequency bin as required by pRT. This fully deterministic process stabilized the log-likelihood calculations in the retrievals, and resulted in a $5\times$ improvement in the speed of the c-k mixing function.

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