

¹ STCTM: a forward modeling and retrieval framework ² for stellar contamination and stellar spectra

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Summary

¹⁶ Transmission spectroscopy is a key avenue for the near-term study of small-planet atmospheres
¹⁷ and the most promising method when it comes to searching for atmospheres on temperate
¹⁸ rocky worlds, which are often too cold for planetary emission to be detectable. At the same
¹⁹ time, the small planets that are most amenable for such atmospheric probes orbit small M dwarf
²⁰ stars. This “M-dwarf opportunity” has encountered a major challenge because of late-type
²¹ stars’ magnetic activity, which lead to the formation of spots and faculae at their surface.
²² If inhomogeneously distributed throughout the photosphere, this phenomenon can give rise
²³ to “stellar contamination”, or the transit light source effect (TLSE). Specifically, the TLSE
²⁴ describes the fact that spectral contrasts between bright and dark spots at the stellar surface
²⁵ outside of the transit chord can leave wavelength-dependent imprints in transmission spectra
²⁶ that may be mistaken for planetary atmosphere absorption.

²⁷ As the field becomes increasingly ambitious in the search for signs of even thin atmospheres on
²⁸ small exoplanets, the TLSE is becoming a limiting factor, and it becomes imperative to develop
robust inference methods to disentangle planetary and stellar contributions to the observed
spectra. Here, I present stctm, the STellar ConTamination Modeling framework, a flexible
Bayesian retrieval framework to model the impact of the TLSE on any exoplanet transmission
spectrum, and infer the range of stellar surface parameters that are compatible with the
observations in the absence of any planetary contribution. With the exotune sub-module,
users can also perform retrievals directly on out-of-transit stellar spectra in order to place
data-driven priors on the extent to which the TLSE can impact any planet’s transmission
spectrum. The input data formats, stellar models, and fitted parameters are easily tunable
using human-readable files and the code is fully parallelized to enable fast inferences.

Statement of need

³⁰ The interpretation of high-precision exoplanet transmission spectra from facilities such as the
³¹ Hubble Space Telescope (HST) and the James Webb Space Telescope (JWST) is increasingly
³² dependent on a robust accounting for the effects of stellar contamination, particularly for
³³ small planets orbiting small stars. Despite a growing awareness of the Transit Light Source
³⁴ Effect (TLSE; ([Rackham et al., 2018](#); [TRAPPIST-1 JWST Community Initiative et al., 2024](#))),
³⁵ the community currently lacks flexible, open-source tools that allow for robust modeling and
³⁶ retrieval of stellar contamination signatures. Further, uncertainties in stellar models motivate
³⁷ flexible implementations with reproducible model setups supporting any user-specified stellar
³⁸ model source, such as PHOENIX or SPHINX model grids ([Husser et al., 2013](#); [Iyer et al.,](#)
³⁹ 2023).

⁴⁰ While some forward models have been developed to simulate the impact of stellar heterogeneity
⁴¹ on transmission spectra, these tools are either not publicly available, computationally intractable

⁴² due to their serial-mode-only implementation, part of much larger codes that require more
⁴³ advanced user training (e.g. atmospheric retrievals), or not designed for inference. Further,
⁴⁴ the community lacks frameworks that enable to retrieve stellar surface properties from both
⁴⁵ observed planetary transmission spectra and out-of-transit stellar spectra in a Bayesian context.
⁴⁶ This gap limits our ability to quantify uncertainties in exoplanet atmospheric properties and to
⁴⁷ test the robustness of atmospheric detections.

⁴⁸ `stctm` addresses this need by providing an open-source, modular, and user-friendly framework.
⁴⁹ It allows users to model a wide range of stellar surface configurations leveraging any spectral
⁵⁰ models, and to infer which stellar parameters could explain observations without invoking
⁵¹ planetary absorption. It also supports retrievals on out-of-transit stellar spectra to independently
⁵² assess the extent of potential stellar contamination by the host star. By enabling flexible, fast,
⁵³ and reproducible inference of the TLSE, `stctm` empowers the community to critically assess
⁵⁴ the reliability of exoplanet atmosphere detections.

⁵⁵ Main features of the code

⁵⁶ The user inputs are communicated to the code via an input `.toml` file for both TLSE retrievals
⁵⁷ and inferences from out-of-transit stellar spectra. The code follows similar phases for both
⁵⁸ types of retrievals:

- ⁵⁹ ▪ Reading in and parsing of the inputs (data file, stellar models file, saving options, MCMC
⁶⁰ fit setup)
- ⁶¹ ▪ Running the MCMC fit
- ⁶² ▪ Post-processing to create diagnostic plots, record model comparison and goodness-of-fit
⁶³ metrics, produce publication-ready figures, and store sample spectra and parameters for
⁶⁴ post-processing and publication support and reproducibility (e.g. Zenodo)

⁶⁵ exotune retrievals on out-of-transit stellar spectra have an additional (optional) pre-processing
⁶⁶ step, allowing users to:

- ⁶⁷ ▪ start from a full time-series of spectra as the input (e.g. the output from Stage 3 of the
⁶⁸ Eureka! ([Bell et al., 2022](#)) pipeline which is widely used in the community), rather than
⁶⁹ simply a pre-computed stellar spectrum
- ⁷⁰ ▪ exclude certain time intervals/exposures (e.g. containing the transit) and wavelength
⁷¹ ranges (e.g. saturated regions) when computing the median spectrum to perform the
⁷² retrieval over

⁷³ For exotune retrievals, an error inflation parameter can be leveraged to account for the
⁷⁴ often-large mismatch between the data and stellar models.

⁷⁵ Documentation

⁷⁶ The full documentation for `stctm` with installation, testing instructions, and real-data example
⁷⁷ retrievals on transmission spectra and out-of-transit stellar spectra are available at <https://stctm.readthedocs.io/>. A description of `stctm` can be found in several of the early papers
⁷⁸ that employed it (see next section).

⁸⁰ Uses of STCTM in the literature

⁸¹ `stctm` has been applied widely to the interpretation of transmission spectra of rocky planets
⁸² and small sub-Neptunes, including in ([Ahrer et al., 2025](#); [Lim et al., 2023](#); [Piaulet-Ghorayeb et](#)
⁸³ [al., 2024, 2025](#); [Radica et al., 2025](#); [Roy et al., 2023](#)).

84 Future Developments

85 The latest version of stctm at the time of writing (v2.1.1) supports MCMC retrievals on
86 transmission spectra and on out-of-transit stellar spectra (exotune), and provides model
87 comparison statistics, model and parameter samples as well as publication-ready figures.
88 Future versions will expand on these functionalities to include user-friendly scripts tailored to
89 post-processing only for an already-run retrieval (creating custom plots), as well as a Nested
90 Sampling alternative for the retrievals. Users are encouraged to propose or contribute any
91 other features.

92 Similar Tools

93 Here are a few open-source codes that offer functionalities focused on retrievals of the TLSE
94 or on out-of-transit stellar spectra:

- 95 ▪ Generic atmospheric retrievals (including TLSE-only retrievals on transmission spectra):
96 POSEIDON ([MacDonald & Madhusudhan, 2024](#))
- 97 ▪ Retrievals on out-of-transit stellar spectra (Nested Sampling, serial run mode only):
98 StellarFit ([Radica et al., 2025](#))

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101 and ease of set-up for the user.stctm relies heavily on other Python libraries which include
102 numpy ([Harris et al., 2020](#)), scipy ([Virtanen et al., 2020](#)), astropy ([Astropy Collaboration et
103 al., 2013, 2018](#)), matplotlib ([Hunter, 2007](#)), pandas([team, 2020](#)), emcee([Foreman-Mackey
104 et al., 2013](#)), corner ([Foreman-Mackey, 2016](#)), and pysynphot ([Horne, 2013](#)). Users are also
105 strongly encouraged to use msg([Townsend & Lopez, 2023](#)) to obtain the grids of stellar models
106 used in the inference step.

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