

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


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Summary

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

As the field becomes increasingly ambitious in the search for signs of even thin atmospheres on 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

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

While some forward models have been developed to simulate the impact of stellar heterogeneity 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).
79

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).

Future Developments

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

Similar Tools

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

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

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References

- Ahrer, E.-M., Radica, M., Piaulet-Ghorayeb, C., Raul, E., Wiser, L. S., Welbanks, L., Acuna, L., Allart, R., Coulombe, L.-P., Louca, A. J., MacDonald, R. J., Sidel, M., Evans-Soma, T. M., Benneke, B., Christie, D., Beatty, T. G., Cadieux, C., Cloutier, R., Doyon, R., ... Schlichting, H. E. (2025). Escaping helium and a highly muted spectrum suggest a metal-enriched atmosphere on sub-neptune GJ3090b from JWST transit spectroscopy. *arXiv e-Prints*, arXiv:2504.20428. <https://doi.org/10.48550/arXiv.2504.20428>
- Astropy Collaboration, Price-Whelan, A. M., Sipőcz, B. M., Günther, H. M., Lim, P. L., Crawford, S. M., Conseil, S., Shupe, D. L., Craig, M. W., Dencheva, N., Ginsburg, A., VanderPlas, J. T., Bradley, L. D., Pérez-Suárez, D., de Val-Borro, M., Aldcroft, T. L., Cruz, K. L., Robitaille, T. P., Tollerud, E. J., ... Astropy Contributors. (2018). The Astropy Project: Building an Open-science Project and Status of the v2.0 Core Package. *The Astronomical Journal*, 156(3), 123. <https://doi.org/10.3847/1538-3881/aabc4f>
- Astropy Collaboration, Robitaille, T. P., Tollerud, E. J., Greenfield, P., Droettboom, M., Bray, E., Aldcroft, T., Davis, M., Ginsburg, A., Price-Whelan, A. M., Kerzendorf, W. E., Conley, A., Crighton, N., Barbary, K., Muna, D., Ferguson, H., Grollier, F., Parikh, M. M., Nair, P. H., ... Streicher, O. (2013). Astropy: A community Python package for astronomy. *Astronomy & Astrophysics*, 558, A33. <https://doi.org/10.1051/0004-6361/201322068>

- 128 Bell, T., Ahner, E.-M., Brande, J., Carter, A., Feinstein, A. D., Caloca, G., Mansfield, M., Zieba,
129 S., Piaulet, C., Benneke, B., Filippazzo, J., May, E., Roy, P.-A., Kreidberg, L., & Stevenson,
130 K. (2022). Eureka!: An End-to-End Pipeline for JWST Time-Series Observations. *Journal*
131 *of Open Source Software*, 7(79), 4503. <https://doi.org/10.21105/joss.04503>
- 132 Foreman-Mackey, D. (2016). Corner.py: Scatterplot matrices in python. *The Journal of Open*
133 *Source Software*, 1(2), 24. <https://doi.org/10.21105/joss.00024>
- 134 Foreman-Mackey, D., Hogg, D. W., Lang, D., & Goodman, J. (2013). Emcee: The MCMC
135 hammer. *Publications of the Astronomical Society of the Pacific*, 125(925), 306. <https://doi.org/10.1086/670067>
- 136
- 137 Harris, C. R., Millman, K. J., Walt, S. J. van der, Gommers, R., Virtanen, P., Cournapeau, D.,
138 Wieser, E., Taylor, J., Berg, S., Smith, N. J., Kern, R., Picus, M., Hoyer, S., Kerkwijk,
139 M. H. van, Brett, M., Haldane, A., Río, J. F. del, Wiebe, M., Peterson, P., ... Oliphant,
140 T. E. (2020). Array programming with NumPy. *Nature*, 585(7825), 357–362. <https://doi.org/10.1038/s41586-020-2649-2>
- 141
- 142 Horne, K. (2013). *Pysynphot: Synthetic photometry software*. <https://ascl.net/1303.023>;
143 Astrophysics Source Code Library.
- 144 Hunter, J. D. (2007). Matplotlib: A 2D graphics environment. *Computing in Science &*
145 *Engineering*, 9(3), 90–95. <https://doi.org/10.1109/MCSE.2007.55>
- 146 Husser, T.-O., Wende-von Berg, S., Dreizler, S., Homeier, D., Reiners, A., Barman, T.,
147 & Hauschildt, P. H. (2013). A new extensive library of PHOENIX stellar atmospheres
148 and synthetic spectra. *Astronomy & Astrophysics*, 553, A6. <https://doi.org/10.1051/0004-6361/201219058>
- 149
- 150 Iyer, A. R., Line, M. R., Muirhead, P. S., Fortney, J. J., & Gharib-Nezhad, E. (2023).
151 The SPHINX m-dwarf spectral grid. I. Benchmarking new model atmospheres to derive
152 fundamental m-dwarf properties. *The Astrophysical Journal*, 944(1), 41. <https://doi.org/10.3847/1538-4357/acabc2>
- 153
- 154 Lim, O., Benneke, B., Doyon, R., MacDonald, R. J., Piaulet, C., Artigau, É., Coulombe, L.-P.,
155 Radica, M., L'Heureux, A., Albert, L., Rackham, B. V., de Wit, J., Salhi, S., Roy, P.-A.,
156 Flagg, L., Fournier-Tondreau, M., Taylor, J., Cook, N. J., Lafrenière, D., ... Darveau-Bernier,
157 A. (2023). Atmospheric Reconnaissance of TRAPPIST-1 b with JWST/NIRISS: Evidence
158 for Strong Stellar Contamination in the Transmission Spectra. *The Astrophysical Journal*
159 *Letters*, 955(1), L22. <https://doi.org/10.3847/2041-8213/acf7c4>
- 160 MacDonald, R. J., & Madhusudhan, N. (2024). *POSEIDON: Multidimensional atmospheric*
161 *retrieval of exoplanet spectra*. <https://doi.org/10.21105/joss.04873>
- 162 Piaulet-Ghorayeb, C., Benneke, B., Radica, M., Raul, E., Coulombe, L.-P., Ahner, E.-M.,
163 Kubyshkina, D., Howard, W. S., Krissansen-Totton, J., MacDonald, R. J., Roy, P.-A.,
164 Louca, A., Christie, D., Fournier-Tondreau, M., Allart, R., Miguel, Y., Schlichting, H. E.,
165 Welbanks, L., Cadieux, C., ... Knutson, H. A. (2024). JWST/NIRISS reveals the water-rich
166 “steam world” atmosphere of GJ 9827 d. *The Astrophysical Journal Letters*, 974(1), L10.
167 <https://doi.org/10.3847/2041-8213/ad6f00>
- 168 Piaulet-Ghorayeb, C., Benneke, B., Turbet, M., Moore, K., Roy, P.-A., Lim, O., Doyon,
169 R., Fauchez, T. J., Albert, L., Radica, M., Coulombe, L.-P., Lafrenière, D., Cowan, N.
170 B., Belzile, D., Musfirat, K., Kaur, M., L'Heureux, A., Johnstone, D., MacDonald, R.
171 J., ... Turner, J. D. (2025). Strict limits on potential secondary atmospheres on the
172 temperate rocky exo-earth TRAPPIST-1 d. *arXiv e-Prints*. <https://ui.adsabs.harvard.edu/abs/2024ApJ...974L..10P>
- 173
- 174 Rackham, B. V., Apai, D., & Giampapa, M. S. (2018). The transit light source effect: False
175 spectral features and incorrect densities for m-dwarf transiting planets. *The Astrophysical*
176 *Journal*, 853(2), 122. <https://doi.org/10.3847/1538-4357/aaa08c>

- 177 Radica, M., Piaulet-Ghorayeb, C., Taylor, J., Coulombe, L.-P., Benneke, B., Albert, L., Artigau,
178 É., Cowan, N. B., Doyon, R., Lafrenière, D., L'Heureux, A., & Lim, O. (2025). Promise and
179 peril: Stellar contamination and strict limits on the atmosphere composition of TRAPPIST-
180 1 c from JWST NIRISS transmission spectra. *The Astrophysical Journal Letters*, 979(1),
181 L5. <https://doi.org/10.3847/2041-8213/ada381>
- 182 Roy, P.-A., Benneke, B., Piaulet, C., Gully-Santiago, M. A., Crossfield, I. J. M., Morley, C. V.,
183 Kreidberg, L., Mikal-Evans, T., Brande, J., Delisle, S., Greene, T. P., Hardegree-Ullman, K.
184 K., Barman, T., Christiansen, J. L., Dragomir, D., Fortney, J. J., Howard, A. W., Kosiarek,
185 M. R., & Lothringer, J. D. (2023). Water absorption in the transmission spectrum of
186 the water world candidate GJ 9827 d. *The Astrophysical Journal Letters*, 954(2), L52.
187 <https://doi.org/10.3847/2041-8213/acebf0>
- 188 team, T. pandas development. (2020). Pandas-dev/pandas: pandas. *Zenodo*. <https://doi.org/10.5281/zenodo.3509134>
- 189
- 190 Townsend, R., & Lopez, A. (2023). MSG: A software package for interpolating stellar
191 spectra in pre-calculated grids. *The Journal of Open Source Software*, 8(81), 4602.
192 <https://doi.org/10.21105/joss.04602>
- 193 TRAPPIST-1 JWST Community Initiative, Wit, J. de, Doyon, R., Rackham, B. V., Lim, O.,
194 Ducrot, E., Kreidberg, L., Benneke, B., Ribas, I., Berardo, D., Niraula, P., Iyer, A., Shapiro,
195 A., Kostogryz, N., Witzke, V., Gillon, M., Agol, E., Meadows, V., Burgasser, A. J., ... Way,
196 M. J. (2024). A roadmap for the atmospheric characterization of terrestrial exoplanets with
197 JWST. *Nature Astronomy*, 8, 810–818. <https://doi.org/10.1038/s41550-024-02298-5>
- 198 Virtanen, P., Gommers, R., Oliphant, T. E., Haberland, M., Reddy, T., Cournapeau, D.,
199 Burovski, E., Peterson, P., Weckesser, W., Bright, J., van der Walt, S. J., Brett, M., Wilson,
200 J., Millman, K. J., Mayorov, N., Nelson, A. R. J., Jones, E., Kern, R., Larson, E., ... SciPy
201 1.0 Contributors. (2020). SciPy 1.0: Fundamental Algorithms for Scientific Computing in
202 Python. *Nature Methods*, 17, 261–272. <https://doi.org/10.1038/s41592-019-0686-2>