yieldplotlib: A unified library for exoplanet yield code visualizations

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Summary

NASA's next flagship observatory, as recommended by the Astro2020 decadal survey, is the Habitable World's Observatory (HWO) which has the ambitious goal to "search for biosignatures from a robust number of about 25 habitable zone planets and be a transformative facility for general astrophysics". In the phrasing of this science goal, the importance of expected exo-Earth yield (the total number of detected habitable zone planets) on the success of the mission is made apparent. As HWO is being developed and trade spaces are explored, yield codes such as the Altruistic Yield Optimization (AYO) and EXOSIMS that can calculate the expected number of detected and characterized planets for a given mission architecture are essential. While these yield codes have the same goal, they can be complex and have major differences in their inputs and outputs that has made comparing results difficult. The need for a unified library for visualizing the inputs and outputs of these yield codes in a complete, descriptive, and accessible way has therefore also become apparent. To this end we have developed yieldplotlib, an open-source python library to communicate the results of yield codes to the broader community and produce publication-quality plots. Currently, there are modules for analyzing AYO and EXOSIMS, but yieldplotlib is easily extensible and support for other yield codes can be easily added in the future.

Statement of need

Expected exoplanetary yield is an important metric when evaluating the success of proposed flagship space observatory architectures such as those currently being considered for the Habitable Worlds Observatory (HWO; (Feinberg et al., 2024)). As HWO is being developed and trade spaces are explored, yield codes such as the Altruistic Yield Optimizer (AYO; (Stark et al., 2014)) and EXOSIMS (Delacroix et al., 2016) that can calculate the expected number of detected and characterized planets for a given mission architecture are essential. While these yield codes have the same goal, they can be complex and have major differences in their inputs and outputs that has made comparing results difficult. The need for a unified library for visualizing these yield codes in a complete, descriptive, and accessible way has therefore also become apparent. This is non-trivial due to the differing syntaxes, structures, and assumptions that each of these codes make.

Despite the challenges, when these values are interrogated directly, new insights are achieved. Some of these insights are highlighted in detail in (Stark et al., 2025) where a comparison of just the internal exposure time calculations of AYO and EXOSIMS revealed sources of previously unknown discrepancy. `yieldplotlib` is in many ways a continuation of that initial work, but aimed instead at the higher level yield products which are of the most direct interest.

`yieldplotlib` is a Python package for visualizing the inputs and outputs of AYO and EXOSIMS through the use of a custom loading and parsing structure that allows for the easy access

42 of equivalent data across yield code outputs. `yieldplotlib` is designed to communicate the
43 results of yield codes to the broader community and produce publication-quality plots without
44 the need to understand the complex underlying yield codes that were used to generate the
45 data. Currently `yieldplotlib` contains modules for analyzing AYO and EXOSIMS, but is
46 easily extensible and support for other yield codes can be easily added in the future.

Methods and Functionality

Parsing and Getting Values

49 `yieldplotlib` uses a file node and directory structure to parse the yield output packages from
50 AYO and EXOSIMS. It then uses a user generated `key_map` to link the EXOSIMS and AYO
51 keys to a universal key in `yieldplotlib`. This `key_map` is generated from a CSV file in the
52 repository which is automatically updated on a daily basis daily from an active development
53 version hosted on Google Sheets for broader collaboration. An example few lines from the CSV
54 file can be found in the following table.

yieldplotlib key_map.csv sample

yieldplotlib name	description	EXOSIMS name	EXOSIMS file	EXOSIMS Class	AYO name	AYO file	AYO Class
star_dist	Distance to the star (in parsecs).	star_dist	reduce-star-target.csv	EXOSIMSCSVFile	dist (pc)	target_list.csv	AYOCSVFile
yield_earth	Yield of Earth-like exoplanet candidates	exoE_det_alt_mean	reduce-earth.csv	EXOSIMSCSVFile	exoEarth candidate yield	observations.csv	AYOCSVFile

Figure 1: Example portion of the yieldplotlib key map CSV file containing the mappings between AYO, EXOSIMS, and yieldplotlib parameters.

55 Once the yield packages are loaded and parsed, a getter can be called on the directories to
56 return the corresponding value from the respective yield code.

```
from yieldplotlib.load.ayo_directory import AYODirectory
from yieldplotlib.load.exosims_directory import EXOSIMSDirectory

ayo = AYODirectory(Path("path/to/my/ayo_data"))
exosims = EXOSIMSDirectory(Path("path/to/my/exosims_data"))

ayo.get("yield_earth")
exosims.get("yield_earth")
```

57 Yield input packages (YIPs) specifying input coronagraph parameters can also be read in and
58 accessed using the same file node and directory structure. This allows users to access key
59 coronagraph performance metrics that serve as critical inputs to these yield codes. In order to
60 process the YIPs, `yieldplotlib` uses `yippy` as a backend, though the user interface is identical
61 to generating the AYO and EXOSIMS directories.

```
from yieldplotlib.load.yip_directory import YIPDirectory

yip = YIPDirectory(Path("path/to/my/yip_data"))
```

Plotting

Generic and Comparison Plots

yieldplotlib extends the commonly used Python plotting package matplotlib to take advantage of the wide variety of customization options that matplotlib offers, as well as the extensive knowledge base many users of yieldplotlib will have with that package. The yieldplotlib generic plots are designed to be used for single yield run visualizations and can make scatter plots, standard plots, and histograms.

The comparison plots are designed for plotting multiple yield runs in either the same multi-panel figure, or on the same set of axes for more complex comparisons.

Plotting Scripts

yieldplotlib contains scripts for generating common plots used in yield code visualizations to provide instant usability for comparing AYO and EXOSIMS as motivated by the rapid pace of the ongoing architecture trade studies for HWO. This also serves to provide examples on how the package can be used for those who want to adapt the generic yieldplotlib parsing structure and plotting methods to generate their own bespoke visualizations.

Figure Figure 2 and Figure Figure 3 show two different types of yield outputs. Figure Figure 2 shows the fraction of a star's habitable zone that cen be sampled by during the lifetime of a mission known as the "habitable zone completeness" with the two yield codes in side by side axes and using the same color bar for ease of comparison. Figure Figure 3 shows histograms of the total number of detected planets found for each yield code as a function of planet type.
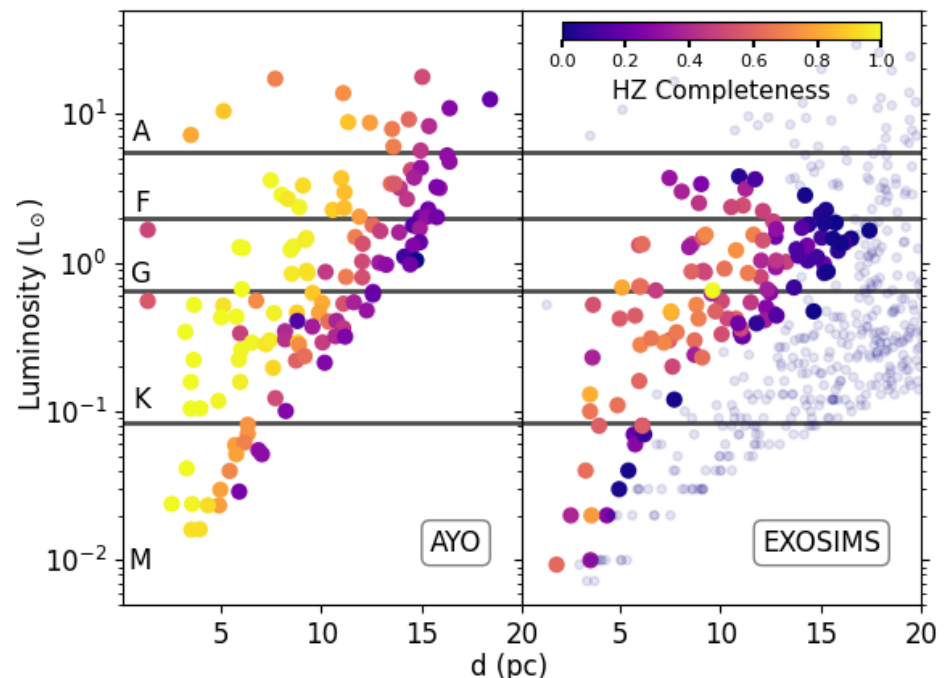

Figure 2: Plot of the Habitable Zone (HZ) completeness as a function of host star luminosity (in units of Solar luminosity) and distance (in parsecs). Here the AYO results are on the left and the EXOSIMS results are on the right.
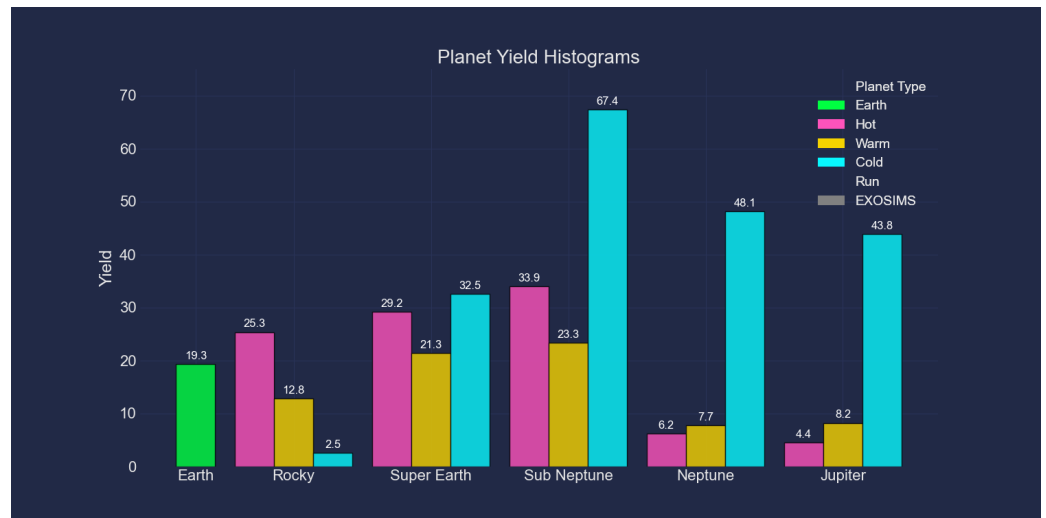
Figure 3: Bar chart showing expected EXOSIMS planet yields for hot (pink), warm (yellow), and cold (blue) Rocky planets, Super Earths, Sub-Neptunes, Neptunes and Jupiters. Earth-like planets which are of the most interest for HWO are shown in green. This plot uses the "cyberpunk" theme from matplotlib which is supported as a keyword argument to yieldplotlib as a dark mode alternative to the standard plotting color schemes.

Yield code inputs can also have a profound impact on their results and so plotting these values is important to ensure consistency. Figure Figure 4 shows the throughput for a key series of optics in the observatory known as a coronagraph. Smaller throughputs mean less planet light makes it onto the detector and can have a profound impact on final yields.

Figure 4: Core throughput vs. separation (in lambda/D) for the coronagraph assumed by AYO (dotted purple), EXOSIMS (dashed pink) and pulled directly from the yield input package using yippy (solid blue). Slight differences between the codes can be attributed to how the "core" is defined. EXOSIMS and yippy adopt a fixed radius circular aperture whereas AYO defines an aperture based on pixels having more than 30% of the peak flux. Additional sources of difference can also lie in the interpolation methods used by all of the codes. This highlights the types of insights that tools like yieldplotlib can help to uncover.

Pipeline and Command Line Interface

In order to generate summary plots quickly, `yieldplotlib` comes packaged with a command line interface and plotting pipeline to create a suite of commonly used yield plots. This is accessed by running `run_ypl` and providing a path to a single folder containing the outputs for a single AYO or EXOSIMS run, or to a directory containing subdirectories of many AYO and EXOSIMS runs.

Acknowledgements

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