

¹ ESMBenchmarkViz: A Python Toolkit for Interactive Visualization of Earth System Model Evaluation and Benchmarking

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Software

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⁶ Summary

⁷ ESMBenchmarkViz is a Python toolkit designed to generate interactive graphs for visualizing ⁸ statistics and metrics from the evaluation of climate and Earth system models. This toolkit ⁹ enables researchers to perform more straightforward benchmarking and intercomparison of ¹⁰ models.

¹¹ Statement of Need

¹² Earth System Models (ESMs) are essential for understanding and predicting the complex ¹³ interactions within the Earth's climate system. These models integrate components such as the ¹⁴ atmosphere, oceans, land surface, and biosphere to address fundamental and applied questions ¹⁵ about Earth system processes, feedbacks, and the system's response to external forcing. As ¹⁶ ESMs grow in complexity, the need for effective evaluation and benchmarking methods to ¹⁷ ensure their reliability and accuracy becomes increasingly important.

¹⁸ Evaluating ESMs involves comparing model outputs against observational data and other models ¹⁹ to assess performance. This process is vital for identifying model strengths and weaknesses, ²⁰ guiding improvements, and deepening our understanding of climate processes. However, ESM ²¹ evaluations (particularly comprehensive evaluations) often generate large volumes of data (e.g., ²² [\(J. Lee et al., 2019\)](#), [\(Jiwoo Lee et al., 2021\)](#), [\(J. Lee et al., 2024\)](#); [\(Ahn et al., 2022\)](#); [\(Planton et al., 2021\)](#)), which can be challenging to interpret and communicate effectively. Wrangling ²³ this data to produce an effective, polished visualization is a cumbersome process that often ²⁴ requires constructing data and label lists, calculating plot coordinates, and making manual ²⁵ adjustments for final elements like color bar position or label orientation.

²⁷ To address these challenges, we have developed ESMBenchmarkViz, a modern Python library ²⁸ specifically designed for efficient interactive visualization of statistical performance metrics ²⁹ from ESM evaluation and intercomparison. This library leverages the power of Bokeh ([Bokeh Development Team, 2018](#)) to provide researchers and practitioners with user-friendly interactive ³⁰ tools for dynamic exploration of complex datasets. By enabling real-time manipulation of ³¹ visualizations, ESMBenchmarkViz facilitates deeper insights into model performance and ³² inter-model comparisons, making it easier to identify outliers and persistent biases.

³⁴ The development of ESMBenchmarkViz originated from the interactive visualization dashboard ³⁵ of the [Program for Climate Model Diagnosis and Intercomparison \(PCMDI\) Metrics Package](#) ³⁶ ([\(Jiwoo Lee et al., 2023\)](#) [\(J. Lee et al., 2024\)](#)), showing diverse evaluation metrics for ESMs ³⁷ along with diagnostic information (<https://pcmdi.llnl.gov/research/metrics/>). We refer to ³⁸ these diagnostics as "dive-down information," as they enable users to investigate metrics in ³⁹ greater detail.

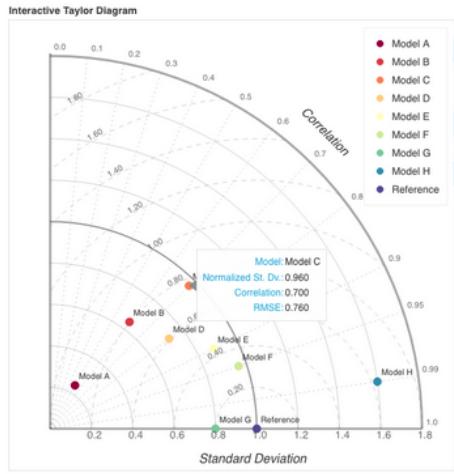
⁴⁰ In this document, we describe the core functionalities of ESMBenchmarkViz and demonstrate
⁴¹ example applications. Our aim is to empower climate scientists with enhanced visualization
⁴² capabilities, contributing to more robust ESM evaluations and benchmarking, and ultimately
⁴³ supporting informed decision-making for climate policy.

⁴⁴ **Key Features**

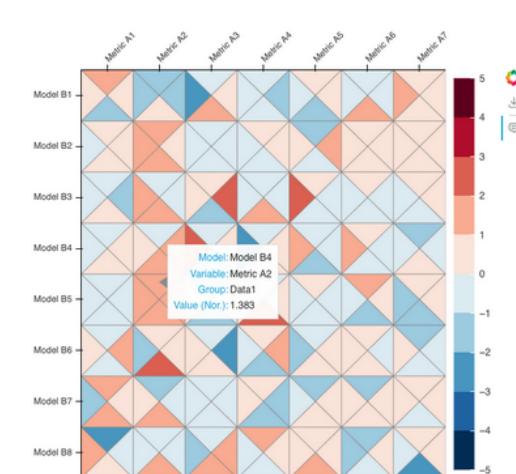
⁴⁵ ESMBenchmarkViz provides reusable functions to generate a suite of interactive plots cus-
⁴⁶ tomized for the evaluation, intercomparison, and benchmarking of ESMs. The toolkit is
⁴⁷ developed in Python 3 and built on top of the Bokeh library for interactive visualization. API
⁴⁸ reference documentation and interactive demo Jupyter notebooks are available for each type
⁴⁹ of plot.

⁵⁰ With singular functions for each plot, the library integrates seamlessly with existing data
⁵¹ analysis workflows and promotes reproducibility in climate research. Users can interact with
⁵² data by zooming, filtering, and hovering for detailed tooltips or displaying additional details as
⁵³ a sidenote, enhancing the communication of findings.

(a) Taylor Diagram



(b) Portrait Plot



(c) Scatter plot (with enabling the side dive-down-image viewer)

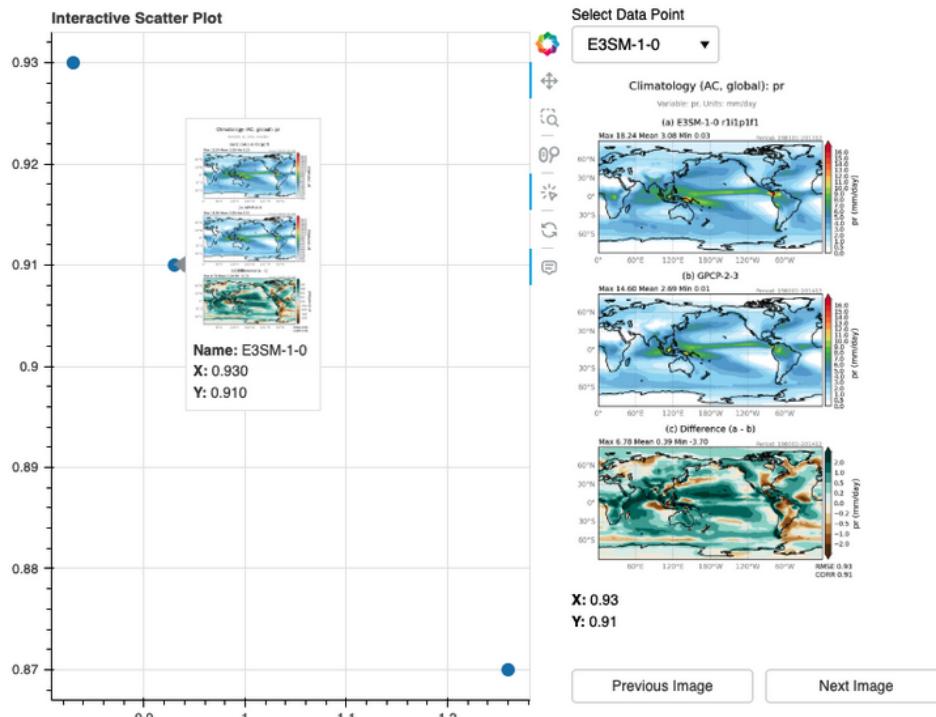


Figure 1: Demonstration of the core features: (a) Taylor Diagram (Taylor, 2001), (b) Portrait Plot ((Gleckler et al., 2008)), and (c) scatter plot with the side dive-down image viewer option activated. Users' mouse cursor hovering over for a specific data point (i.e., a specific ESM model and for its metrics) interactively shows a tooltip that includes detailed information, with the capability of clicking it to open the associated "dive-down" image. Images can be also included to the tooltips or to the side viewer.

54 The toolkit offers convenient APIs to generate and save the following types of graphs in both
55 static and interactive modes: two specialized plots used in ESM evaluation—Taylor Diagram
56 ((Taylor, 2001); Fig. 1a) and Portrait Plot ((Gleckler et al., 2008); Fig. 1b)—as well as the
57 widely used scatter plot (Fig. 1c). These graph types were selected for their utility in ESM
58 evaluation and benchmarking, and because few tools currently provide such capabilities.

59 Taylor Diagram

60 The Taylor Diagram (([Taylor, 2001](#)); Fig. 1a) provides a concise graphical summary of how
61 well patterns simulated by a model match observations. It simultaneously displays three
62 statistics—spatial pattern correlation, standard deviation, and root-mean-square error—making
63 it especially useful for comparing multiple models or datasets against a reference.

64 Portrait Plot

65 The Portrait Plot (([Gleckler et al., 2008](#)); Fig. 1b) presents a matrix-like visualization that
66 summarizes model performance across multiple variables, metrics, or regions. It enables quick
67 identification of patterns, strengths, and weaknesses by displaying performance scores as colored
68 cells, facilitating comprehensive intercomparison among models. This type of plot has been
69 actively used for various climate model evaluation studies (e.g., ([J. Lee et al., 2019](#)), ([Jiwoo Lee
et al., 2021](#)), ([Ahn et al., 2022](#))). The PCMDI Metrics Package (([Jiwoo Lee et al., 2023](#)), ([J.
Lee et al., 2024](#))) Team had developed a precursor version of this package to present evaluation
72 output from hundreds of simulations in an efficient way (<https://pcmdi.llnl.gov/metrics/>).

73 Scatter Plot

74 The Scatter Plot (Fig. 1c) displays the relationship between two variables, allowing users
75 to visually assess correlations, trends, and outliers in model evaluation data. It is a flexible
76 and widely used tool for exploring and communicating the distribution and association of key
77 metrics. Although it is a very widely applied type of plot interdisciplinary, we have included it
78 to the package for its synergy with tooltips and images accompanying together, as shown in
79 Fig. 1c.

80 There are more types of plots planned for the future advancement of the package, for example,
81 Parallel Coordinate Plots in the way it has been used for ESM evaluations (([J. Lee et al.,
2024](#))).

83 Documentation

84 The ESMBenchmarkViz [documentation](#) includes the [public API list](#) and a Jupyter Notebook
85 [Gallery](#) that demonstrates usage of the package.

86 Distribution

87 ESMBenchmarkViz is available for Linux and MacOS, following the [installation instructions](#).
88 We host all development activity at the [GitHub Repository](#). We plan to set up a conda-forge
89 channel on Anaconda for an easier installation.

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97 project.

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