





floatCSEP: An application to deploy and conduct reproducible and prospective earthquake forecasting

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Software

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Summary

floatCSEP is a Python application that standardizes and orchestrates the workflow of earthquake forecasting experiments. Based on principles established by the Collaboratory for the Study of Earthquake Predictability (CSEP, <https://cseptest.org>), it enables reproducible, transparent, and reusable evaluations of earthquake forecasts (both prospective or pseudo-prospective). floatCSEP builds on the existing pyCSEP toolkit for core evaluation routines and adds the functionality needed to deploy and conduct entire experiments, including catalog handling, forecast generation, evaluation, visualization, and reporting. Accompanying tutorials illustrate experiment use cases, which users can extend to incorporate new models, alternative evaluation metrics, or different regions and timeframes. Ultimately, floatCSEP will support new official CSEP experiments, and also encourage and empower independent researchers to validate their own models.

Background

Earthquake forecasts are probabilistic statements about future earthquake occurrence (Jordan et al., 2011), used for informing building codes, emergency response planning, and risk reduction strategies. Because earthquake occurrence is driven by complex and highly non-linear processes (e.g., Geller, 1997; Y. Y. Kagan, 1994), forecasts should be expressed and evaluated in a probabilistic framework designed to describe their fundamental uncertainties (Y. Kagan & Jackson, 1994). To assess their reliability, further challenges are the large time scales required to collect sufficient observations for evaluation (especially of large earthquakes) and the multiple subjective biases from any post-hoc adjustments in modeling or evaluation environments (e.g., Schorlemmer & Gerstenberger, 2007). To address such challenges, the Collaboratory for the Study of Earthquake Predictability (CSEP) was established to facilitate rigorous, prospective forecasting experiments where all forecasting models, data sources, evaluation metrics, and related software are defined prior to the evaluation time period (e.g., Jordan, 2006; Schorlemmer et al., 2018). CSEP experiments were carried out in so-called Testing Centers, i.e., hardware and software infrastructure designed to ensure (i) controlled access, (ii) reproducible environments for automated forecast generation, and (iii) long-term archiving of input data, metadata and results (Zecher et al., 2010). With this framework, CSEP has successfully hosted and published experiments across diverse geographic regions, such as California, New Zealand, Italy, Japan, and globally (Bayona et al., 2021, 2022, 2023; Eberhard et al., 2012; Field, 2007; Gerstenberger

43 & Rhoades, 2010; Iturrieta et al., 2024; Nanjo et al., 2011; Strader et al., 2018; Taroni et al.,
44 2014, 2018; Tsuruoka et al., 2012; Werner et al., 2010; Zechar et al., 2013). These efforts
45 have substantially advanced scientific rigor and established community standards in earthquake
46 forecasting research, thereby contributing to better forecasts and seismic hazard assessments
47 (e.g., Michael & Werner, 2018; Schorlemmer et al., 2018).

48 Statement of Need

49 Despite their achievements, the original CSEP Testing Centers were centralized, rigid, and
50 with data management tightly coupled to local hardware, significantly limiting software
51 reusability, scalability, and broader community engagement (e.g., Savran, Bayona, et al., 2022;
52 Schorlemmer et al., 2018; Zechar et al., 2010). As also noted by Mizrahi et al. (2024)
53 Mizrahi et al. (2024), there is broad consensus in the earthquake forecasting community that
54 transparency and reproducibility are essential in forecast testing. However, due to the complexity
55 of Testing Centers, independent researchers often require advanced technical expertise to
56 access, reproduce, and analyze CSEP experiments. To overcome these limitations, the Python
57 package pyCSEP (Graham et al., 2024; Savran, Bayona, et al., 2022; Savran, Werner, et
58 al., 2022) was developed to provide core forecast evaluation routines, which can be directly
59 integrated into modelers' workflows.

60 However, pyCSEP alone lacks key features required to deploy and conduct entire forecasting
61 experiments, such as interfacing with external model source code, automating catalog access,
62 managing data, standardizing workflow execution, and generating summary reports. This
63 lack highlights the need for comprehensive software that provides Testing Center capabilities
64 while remaining decoupled from specific hosting hardware. The solution should manage the
65 entire experiment lifecycle, from model integration and initial deployment, to the incremental
66 updating of input data, forecasts, results, and reports as new observations become available.

67 Software Overview

68 The primary objective of floatCSEP is to provide a portable, automated, and reproducible
69 Testing Center environment that can run on any computer with sufficient computational
70 resources. Experiments are defined through human-readable YAML (yaml.org) configuration
71 files, which are processed through a simple command-line interface to ensure ease of use even
72 for users without extensive computational expertise. This declarative approach simplifies the
73 experiment setup, standardizes its workflow (Figure 1) and enhances its reproducibility.

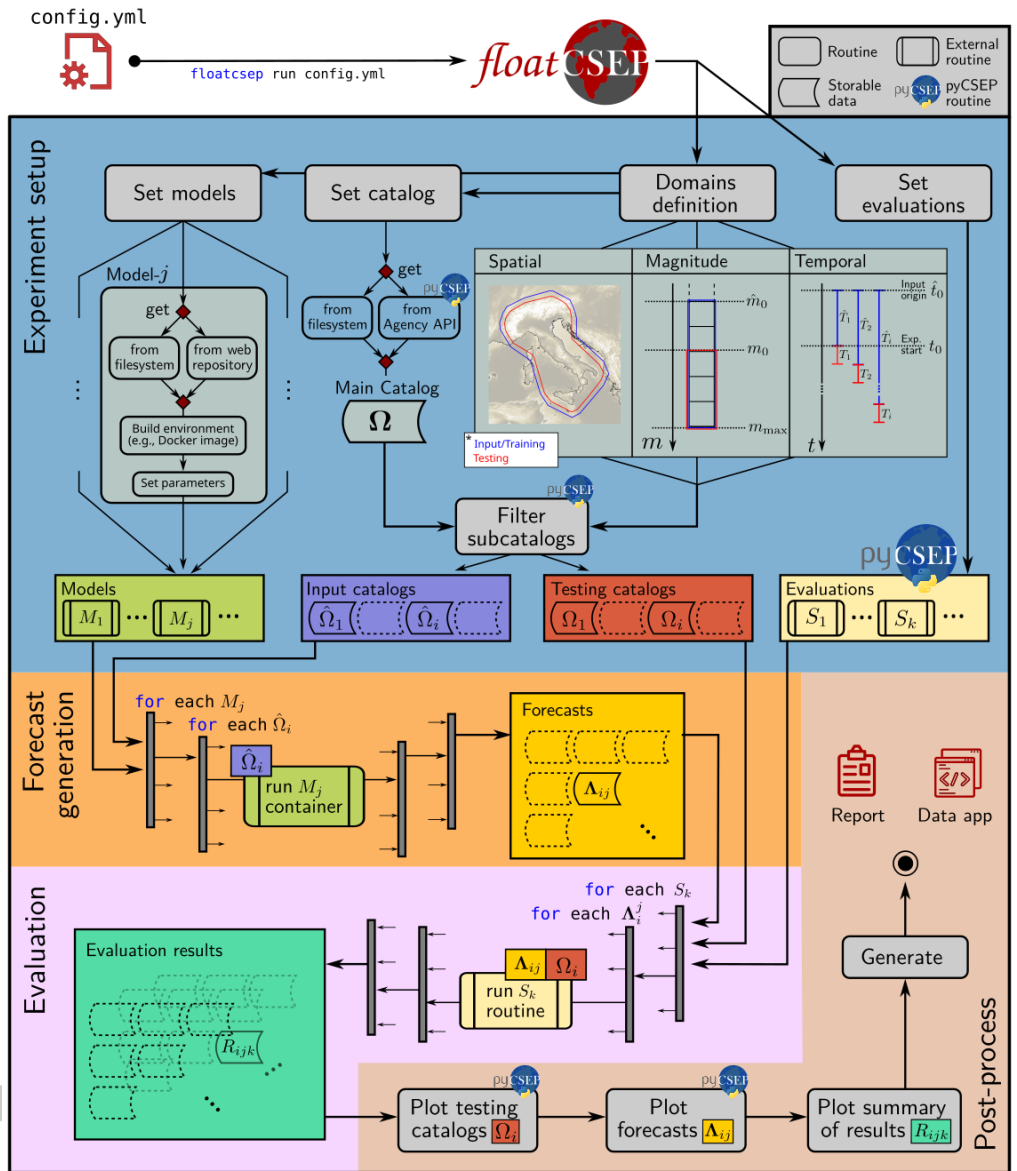


Figure 1: Workflow diagram of floatCSEP for a time-dependent experiment, which roughly consists of: 1) Defining time-space-magnitude ranges and discretizations for forecast generation and evaluation; 2) Querying and filtering earthquake catalogs (both for model input and evaluation); 3) Building the source code of external models, configuring its parameters and input data; 5) Generating forecasts by running each model source code in a containerized environment; 6) Performing forecast evaluations and comparisons using pyCSEP's or user-implemented testing metrics; and 7) Producing reports including test results and visual representations.

floatCSEP uses pyCSEP as a dependency, incorporating its core functionality (forecast and catalog classes, and evaluation routines) alongside additional Testing Center operations, such as data management and computational containerization. The application supports multiple forecast formats and accommodates both time-invariant and time-variant experiments. It handles forecasts produced either by models managed directly by floatCSEP or provided externally through raw files. Representative use cases are included as tutorials, which users can extend by incorporating new models, adding alternative evaluations, or by replicating in different regions and timeframes. The software integrates seamlessly with pyCSEP's existing

82 testing routines, but also provides custom hooks for user-defined tests, visualizations, and
83 reports.

84 Example Use

85 An example configuration file (config.yml) for a time-invariant, grid-based experiment in Italy
86 with two models is shown in Figure 2.

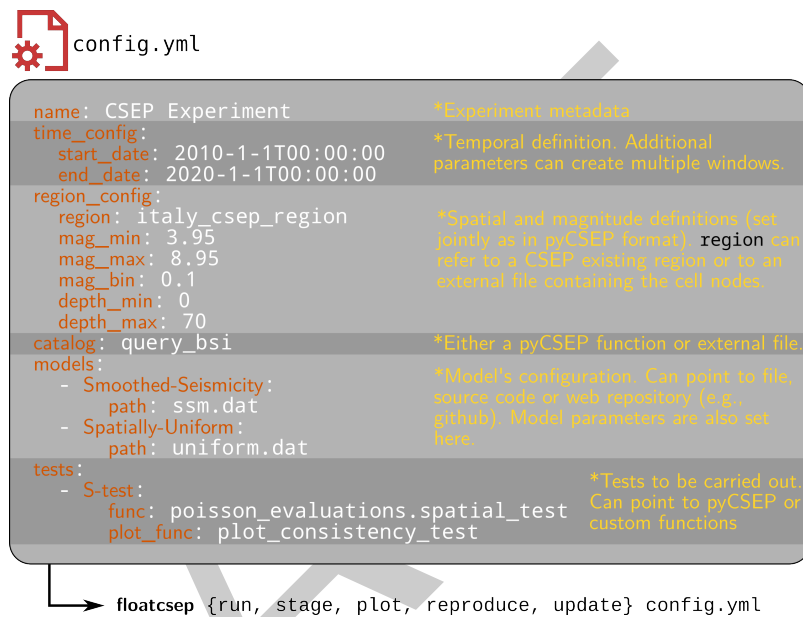


Figure 2: Simplest example of a configuration file for a time-invariant, grid-based experiment with two models. The run command executes an experiment end-to-end. The stage command accesses and builds the models' source code and prepares input/testing catalogs. The reproduce command re-executes an experiment and compares it with existing results using statistical and computational metrics. plot executes the post-process and visualization of the experiment; The update command generates and tests all forecasts missing since the last execution up to today.

87 Applications

88 floatCSEP is designed to support the following applications:

- 89 ■ Deploy and conduct new prospective experiments that incrementally incorporate new
90 data, update forecasts, and provide evaluation results. While the CSEP community
91 plans to use floatCSEP for new (official) experiments, we also encourage independent
92 researchers to adopt floatCSEP for prospectively evaluating their models
- 93 ■ Reproduce the results of completed prospective CSEP experiments within a containerized
94 computational environment (e.g., Iturrieta et al., 2024).
- 95 ■ Create new retrospective or pseudo-prospective experiments for their easy reproduction
96 and shareability.
- 97 ■ Plug in new models into a completed or ongoing (float)CSEP experiment. Since CSEP
98 experiments are clearly defined, they can be effectively used as benchmarks for comparing
99 and developing new forecasting models (e.g., Serafini et al., 2025)
- 100 ■ Support continuous evaluation of Operational Earthquake Forecasting systems that
101 provide authoritative, near-real-time forecasts (e.g., Jordan et al., 2011; Mizrahi et
102 al., 2024). However, most systems generate forecasts for overlapping windows (e.g.,
103 weekly forecasts updated daily) and evaluating the overall performance of such forecast

collections remains an open methodological challenge (e.g., Brehmer et al., 2025). floatCSEP contributes to a growing CSEP software ecosystem that, together with reproducibility packages (e.g., Allison et al., 2018; Bayona et al., 2022, 2023; Graham et al., 2024; Iturrieta et al., 2024; Savran, Bayona, et al., 2022), open-source forecasting models (e.g., Mizrahi et al., 2023), and long-term open-science repositories, could lay the foundation for building robust, collaborative benchmarks in earthquake forecasting.

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