

ABSESpy: An agent-based modeling framework for social-ecological systems

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DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

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Submitted: 22 November 2023

Published: unpublished

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Summary

ABSESpy is a novel agent-based modeling (ABM) framework that enhances socio-ecological systems (SES) research fidelity. Addressing critical needs in SES study, such as complex decision-making, scaling, and data integration, it features a Branch-Leaf architecture for clear separation and integration of human and natural subsystems, promoting replicability and model coupling. ABSESpy also supports modeling human behavior through well-recognized workflows of perception, decision-making definitions, and responses. Moreover, it advances real-world modeling with multiple time operating modes, accommodating the diverse temporal scales of SES phenomena and integrating time-sensitive event simulations. These innovations position ABSESpy as a crucial tool in addressing current gaps in SES research, fostering more ABMs for real-world SES issues.

Statement of need

Social-ecological systems (SES) represent an integrated concept that recognizes the complex and interdependent dynamics between human societies and ecological systems (Folke et al., 2010). Consisting of decision-making agents (representing people, communities, organizations, and environmental components) capable of following heterogeneous objectives (Levin et al., 2013), SES has specific needs for research support from agent-based modeling.

However, ABMs' potential is yet to be fully realized in SES researches. Current challenges, such as incorporating human decision-making, portraying socio-ecological networks, and modeling real-world systems, must be addressed (Schulze et al., 2017). Additionally, issues related to data availability, model validation, replicability, and transparency must be systematically tackled to enhance the reliability and applicability of ABM in this field (Gotts et al., 2019).

Developing and refining ABM approaches for social-ecological systems are crucial in light of these needs and challenges (Reyers et al., 2018). At the heart of this should be a modeling framework that is portable, scale-flexible, and capable of expressing the interaction of the decision-making agent with the natural environment or ecosystem. ABSESpy represents a significant advancement in this regard, offering several features that address the current gaps in SES modeling.

Design structures

ABSESpy introduces a Branch-Leaf architecture central to its functionality. It facilitates a clear separation of the human and natural subsystems within SES research, aligning with the requisite to enhance replicability and extensibility (Figure 1).

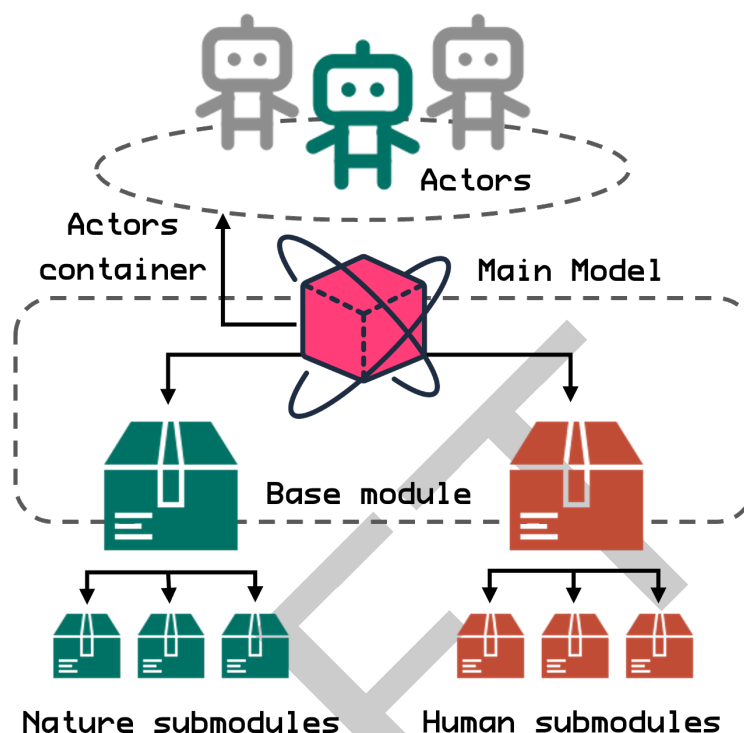


Figure 1: Structure of main components of ABSESpy and its Branch-Leaf architecture of modules.

39 Integrated by the MainModel, the two primary base modules are named as Base Human and
40 Base Nature, corresponding to components of a typical SES (Reyers et al., 2018). By this
41 architecture, ABSESpy enables the addition of specialized sub-modules, thus promoting a tailored
42 modeling approach. The extension mesa-geo is embedded as the base driver for the nature
43 subsystem so that most of the different geographic datasets are compatible (.tif, .nc, .json,
44 .shp, et al.).

45 In the SES context, ABSESpy conceptualizes agents as Actors managed within a unique
46 ActorsContainer and can be referred from a temporary ActorsList. In human sub-modules,
47 users can define a series of Actor's references by or link each other (between agent and
48 patch, or agent and agent) by inputting advanced query. It simplifies the agents' organization,
49 ensuring each actor can be searched, operated, and able to access global information.

50 Human-behavior modeling framework

51 ABSESpy recognizes the centrality of human behavior in SES and, as such, prioritizes the
52 workflow approaching its simulation. To this end, the framework provides an integrative
53 approach based on popular theories of conceptualizing human decision-making (Figure 2)
54 (Schlüter et al., 2017), (Beckage et al., 2022).

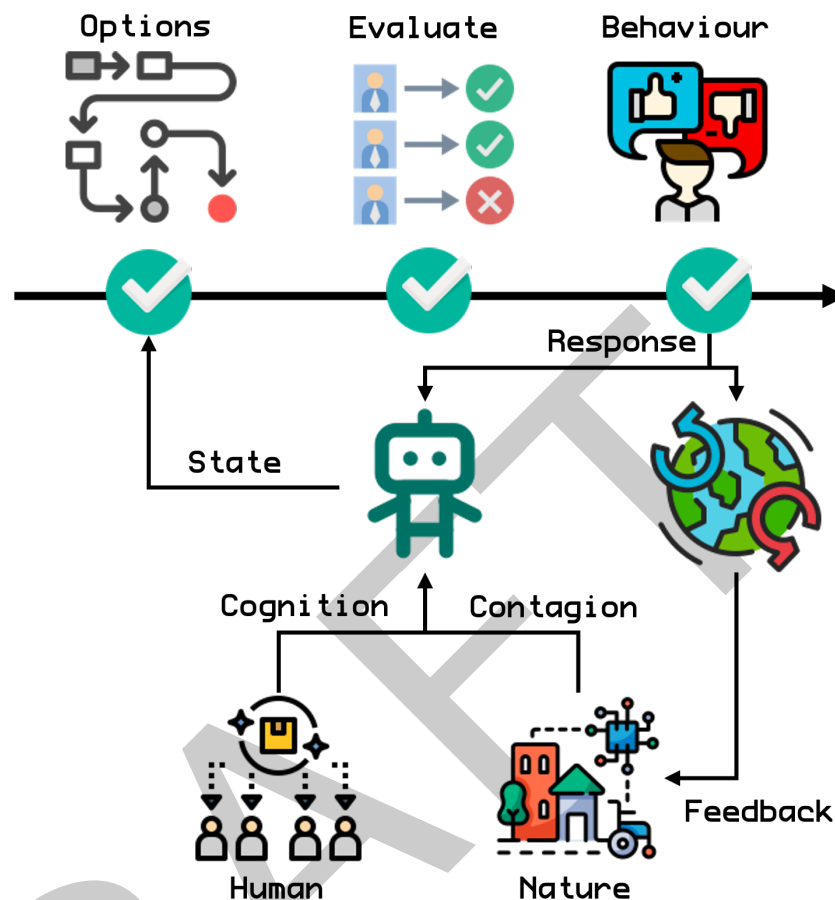


Figure 2: Decision-making workflow for simulating human behavior.

When practicing, ABSESpy provides an advanced behavior simulation framework, including the following main steps:

1. **Perceptions:** From direct environmental observations to social communications, users can define a perception variable to represent how agents gather information and form their understanding of the environment.
2. **Decision-making:** By evaluating the potential choices of a decision, decision-making logic can be implemented to capture how human agents might process information and select courses of action.
3. **Response:** Consequent to decision-making, agents exhibit responses for actualizing their strategies —e.g., spatial relocation, attribute changes, altering environment, or other forms of interaction.

By translating theoretical constructs into user-friendly, operational components, ABSESpy empowers researchers to bridge the gap between conceptual models and their tangible application in SES.

Real-world SES modeling enhancements

ABSESpy integrates an innovative time control mechanism to bridge the gap between ABMs and real-world SESs. These are attributions from a TimeDriver module that manages the association of ABM with real-world time (Figure 3).

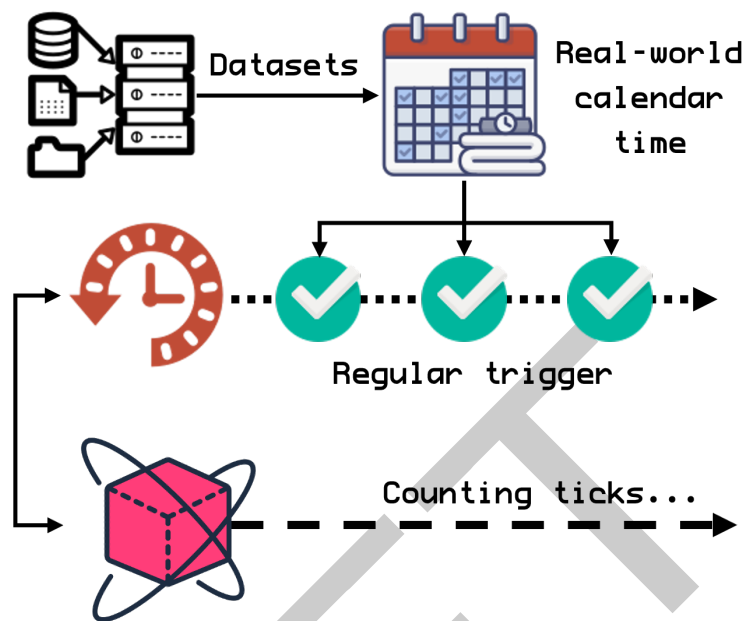


Figure 3: Calendar time module enhances real-world social-ecological system modeling approaches.

In addition to the standard tick-based time advancement, users can implement two temporal modes for matching the diverse scales of SES phenomena. (1) In a “Duration Mode,” users can define the length of time that each simulation step represents, thus allowing for variable temporal resolutions. This capability enables the model to represent time intervals vary from minutes to years, depending on the specific requirements of the SES being modeled. (2) The “Irregular Mode” addresses the non-uniformity of specific SES processes; this mode allows for irregular time steps, whereby different simulation intervals can represent varying lengths of time. This flexibility is crucial when modeling events that do not follow a linear timeline, such as erratic ecological phenomena or sporadic human activities.

A calendar schedule enables ABSESpy to import and utilize dynamic, temporal datasets. ABSESpy automates the updating of variables with time-series data, negating the need for manual data refreshes and recalculations. It supports real-time data feeds, ensuring that the model reflects current conditions. The ABSESpy introduces a time-based event handler (`time_condition` decorator) based on the same idea. By leveraging this decorator, temporal conditions for executing events can be set, enabling simulations to react to time-specific triggers. This aspect is especially pertinent for phenomena with distinct temporal patterns, like migratory behaviors or seasonal cycles.

Positioning and comparison

ABSESpy facilitates independent module creation, enabling an advanced human behavior framework and providing sophisticated time control and data integration tools. ABSESpy allows a more accurate and nuanced representation of SES dynamics, meeting the intricate requirements of real-world problem-solving and decision-making support. Its goal is to become a specialized package for the emerging SES field based on the mesa project, similar to the existing abce (a package aimed at providing an economic problem modeling framework, also a mesa package) (Taghawi-Nejad et al., 2017). Therefore, ABSESpy can take advantage of most of the benefits from the related projects (e.g., mesa (Kazil et al., 2020) and mesa-geo (Wang et al., 2022)), such as visualization and geographic data processing.

A possible competitor is AgentPy, but its goal remains to be a general ABM framework. Due to the need for more mature geographic data processing extensions like mesa-geo, building

ABSESpy on top of the mesa project allows users to deal with real-world SES problems while putting less coding effort into setting up their model projects. Currently, many open-source SES models are published on the platform CoMSES (Janssen et al., 2008); they primarily serve as heuristic models using netlogo (Tisue & Wilensky, 2004) software as their modeling foundation. The visible advantage of ABSESpy lies in its well-structured design, which is suitable for large-scale SES modeling projects. It calls upon vast amounts of actual data for real-world problem modeling rather than merely heuristic modeling. Its tree-like structure allows ABSESpy users to couple models together, maximizing Python's advantages as a "glue language".

Acknowledgment

This research has been supported by the National Natural Science Foundation of China (grant no. 42041007) and the National Natural Science Foundation of China Joint Fund for Scientific Research on Yellow River (grant no. U2243601).

References

- Beckage, B., Moore, F. C., & Lacasse, K. (2022). Incorporating human behaviour into earth system modelling. *Nature Human Behaviour*, 6(11), 1493–1502. <https://doi.org/10.1038/s41562-022-01478-5>
- Folke, C., Carpenter, S. R., Walker, B., Scheffer, M., Chapin, T., & Rockstrom, J. (2010). Resilience thinking: Integrating resilience, adaptability and transformability. *Ecology and Society*, 15(4), 20. <https://doi.org/10.5751/es-03610-150420>
- Gotts, N. M., van Voorn, G. A. K., Polhill, J. G., Jong, E. de, Edmonds, B., Hofstede, G. J., & Meyer, R. (2019). Agent-based modelling of socio-ecological systems: Models, projects and ontologies. *Ecological Complexity*, 40, 100728. <https://doi.org/10.1016/j.ecocom.2018.07.007>
- Janssen, M., Alessa, L., Barton, C. M., Bergin, S., & Lee, A. (2008). Towards a community framework for agent-based modelling. *Journal of Artificial Societies and Social Simulation*, 11, 11.
- Kazil, J., Masad, D., & Crooks, A. (2020). Utilizing python for agent-based modeling: The mesa framework. In R. Thomson, H. Bisgin, C. Dancy, A. Hyder, & M. Hussain (Eds.), *Social, cultural, and behavioral modeling* (pp. 308–317). Springer International Publishing. https://doi.org/10.1007/978-3-030-61255-9_30
- Levin, S., Xepapadeas, T., Crépin, A.-S., Norberg, J., de Zeeuw, A., Folke, C., Hughes, T., Arrow, K., Barrett, S., Daily, G., Ehrlich, P., Kautsky, N., Mäler, K.-G., Polasky, S., Troell, M., Vincent, J. R., & Walker, B. (2013). Social-ecological systems as complex adaptive systems: Modeling and policy implications. *Environment and Development Economics*, 18(02), 111–132. <https://doi.org/10.1017/S1355770X12000460>
- Reyers, B., Folke, C., Moore, M.-L., Biggs, R., & Galaz, V. (2018). Social-ecological systems insights for navigating the dynamics of the anthropocene. *Annual Review of Environment and Resources*, 43(1). <https://doi.org/10.1146/annurev-environ-110615-085349>
- Schlüter, M., Baeza, A., Dressler, G., Frank, K., Groeneveld, J., Jager, W., Janssen, M. A., McAllister, R. R. J., Müller, B., Orach, K., Schwarz, N., & Wijermans, N. (2017). A framework for mapping and comparing behavioural theories in models of social-ecological systems. *Ecological Economics*, 131. <https://doi.org/10.1016/j.ecolecon.2016.08.008>
- Schulze, J., Müller, B., Groeneveld, J., & Grimm, V. (2017). Agent-based modelling of social-ecological systems: Achievements, challenges, and a way forward. *Journal of Artificial Societies and Social Simulation*, 20(2), 8. <https://doi.org/10.18564/jasss.3423>
- Taghawi-Nejad, D., Tanin, R. H., Del Rio Chanona, R. M., Carro, A., Farmer, J. D., Heinrich, T., Sabuco, J., & Straka, M. J. (2017). ABCE: A python library for economic agent-based

- 149 modeling. In G. L. Ciampaglia, A. Mashhadi, & T. Yasseri (Eds.), *Social informatics* (pp.
150 17–30). Springer International Publishing. https://doi.org/10.1007/978-3-319-67217-5_2
- 151 Tisue, S., & Wilensky, U. (2004). Netlogo: A simple environment for modeling complexity.
152 *International Conference on Complex Systems*, 21, 16–21.
- 153 Wang, B., Hess, V., & Crooks, A. (2022). Mesa-geo: A GIS extension for the mesa agent-based
154 modeling framework in python. *Proceedings of the 5th ACM SIGSPATIAL International*
155 *Workshop on GeoSpatial Simulation*, 1–10. <https://doi.org/10.1145/3557989.3566157>

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