





1 ABSESpy: An agent-based modeling framework for 2 social-ecological systems

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7 Summary

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

Statement of need

19 Social-ecological systems (SES) represent an integrated concept that recognizes the complex
20 and interdependent dynamics between human societies and ecological systems ([Folke et al.,
21 2010](#)). Consisting of decision-making agents (representing people, communities, organizations,
22 and environmental components) capable of following heterogeneous objectives ([Levin et al.,
23 2013](#)), SES has specific needs for research support from agent-based modeling.

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

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

35 Design structures

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

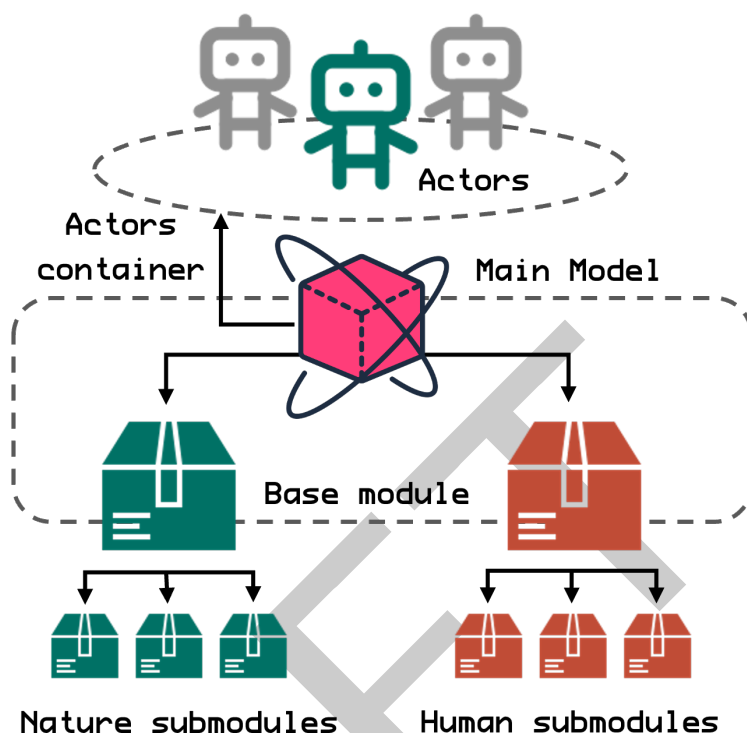


Figure 1: Structure of main components of ABSESpy and its Branch-Leaf architecture of modules.

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).

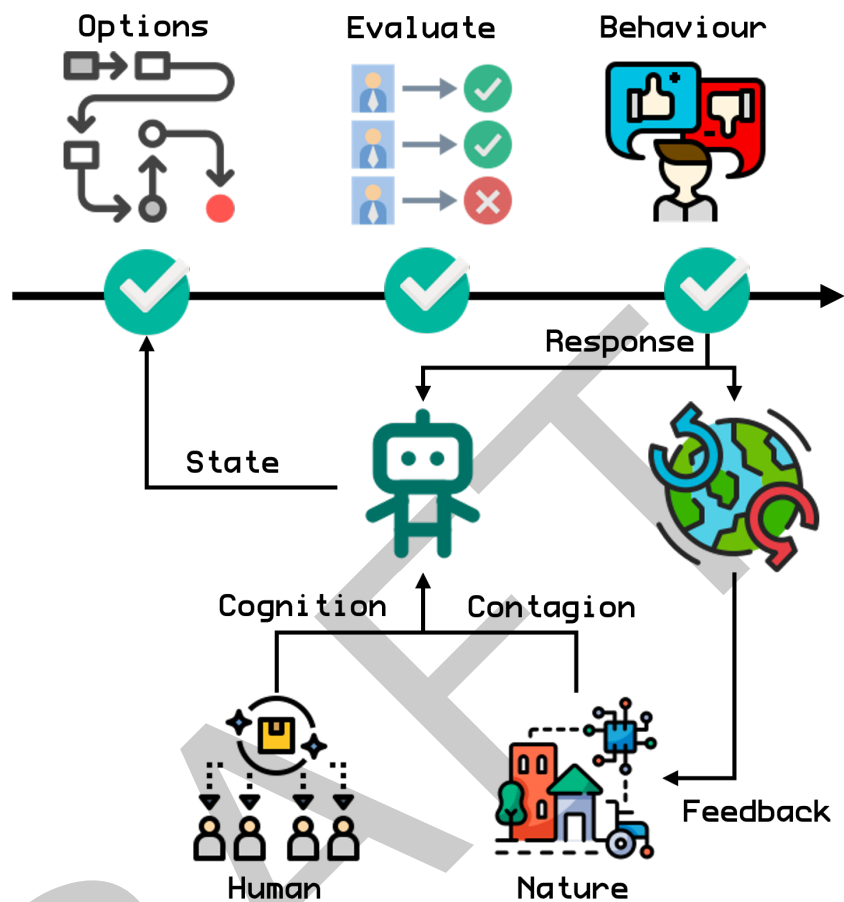


Figure 2: Decision-making workflow for simulating human behavior.

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

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

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

69 Real-world SES modeling enhancements

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

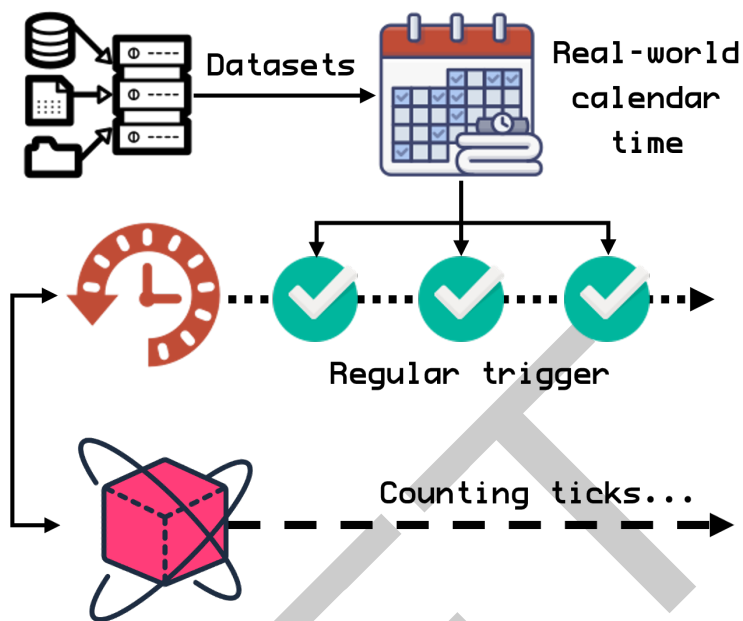


Figure 3: Calendar time module enhances real-world social-ecological system modeling approaches.

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

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

90 Positioning and comparison

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

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

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

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114 References

- 115 Beckage, B., Moore, F. C., & Lacasse, K. (2022). Incorporating human behaviour into Earth
116 system modelling. *Nature Human Behaviour*, 6(11), 1493–1502. <https://doi.org/10.1038/s41562-022-01478-5>
117
- 118 Folke, C., Carpenter, S. R., Walker, B., Scheffer, M., Chapin, T., & Rockstrom, J. (2010).
119 Resilience Thinking: Integrating Resilience, Adaptability and Transformability. *Ecology and*
120 *Society*, 15(4), 20. <https://doi.org/10.5751/es-03610-150420>
- 121 Gotts, N. M., Voorn, G. A. K. van, Polhill, J. G., Jong, E. de, Edmonds, B., Hofstede, G. J.,
122 & Meyer, R. (2019). Agent-based modelling of socio-ecological systems: Models, projects
123 and ontologies. *Ecological Complexity*, 40, 100728. <https://doi.org/10.1016/j.ecocom.2018.07.007>
124
- 125 Janssen, M., Alessa, L., Barton, C. M., Bergin, S., & Lee, A. (2008). Towards a Community
126 Framework for Agent-Based Modelling. *Journal of Artificial Societies and Social Simulation*,
127 *The*, 11.
- 128 Kazil, J., Masad, D., & Crooks, A. (2020). Utilizing Python for Agent-Based Modeling: The
129 Mesa Framework. In R. Thomson, H. Bisgin, C. Dancy, A. Hyder, & M. Hussain (Eds.),
130 *Social, Cultural, and Behavioral Modeling* (pp. 308–317). Springer International Publishing.
131 https://doi.org/10.1007/978-3-030-61255-9_30
- 132 Levin, S., Xepapadeas, T., Crépin, A.-S., Norberg, J., Zeeuw, A. de, Folke, C., Hughes, T.,
133 Arrow, K., Barrett, S., Daily, G., Ehrlich, P., Kautsky, N., Mäler, K.-G., Polasky, S., Troell,
134 M., Vincent, J. R., & Walker, B. (2013). Social-ecological systems as complex adaptive
135 systems: Modeling and policy implications. *Environment and Development Economics*,
136 *18*(02), 111–132. <https://doi.org/10.1017/S1355770X12000460>
- 137 Reyers, B., Folke, C., Moore, M.-L., Biggs, R., & Galaz, V. (2018). Social-Ecological Systems
138 Insights for Navigating the Dynamics of the Anthropocene. *Annual Review of Environment*
139 *and Resources*, 43(1). <https://doi.org/10.1146/annurev-environ-110615-085349>
- 140 Schlüter, M., Baeza, A., Dressler, G., Frank, K., Groeneveld, J., Jager, W., Janssen, M. A.,
141 McAllister, R. R. J., Müller, B., Orach, K., Schwarz, N., & Wijermans, N. (2017). A
142 framework for mapping and comparing behavioural theories in models of social-ecological
143 systems. *Ecological Economics*, 131. <https://doi.org/10.1016/j.ecolecon.2016.08.008>
- 144 Schulze, J., Müller, B., Groeneveld, J., & Grimm, V. (2017). Agent-based modelling of social-
145 ecological systems: Achievements, challenges, and a way forward. *Journal of Artificial*
146 *Societies and Social Simulation*, 20(2), 8.
- 147 Taghawi-Nejad, D., Tanin, R. H., Del Rio Chanona, R. M., Carro, A., Farmer, J. D., Heinrich,
148 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-*
154 *Based Modeling Framework in Python*.

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