

¹ ExerPy: An open-source framework for automated exergy analysis

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

Software

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Editor: 

Submitted: 24 July 2025

Published: unpublished

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⁸ Summary

⁹ ExerPy is an open-source Python framework that automates the exergy analysis of ¹⁰ thermodynamic systems. It integrates with the process-simulation software Aspen Plus®, ¹¹ Epsilon®Professional, and TESPy, through a unified JSON interface. ExerPy automatically ¹² identifies components and defines component- and system-level exergy balances. It also ¹³ provides key performance metrics, including exergy destruction and exergetic efficiency. By ¹⁴ accounting for both physical and chemical exergy and minimizing the need for manual ¹⁵ calculations, ExerPy enables consistent second-law assessments across a wide range of systems, ¹⁶ from simple thermodynamic cycles to complex plants. ExerPy facilitates the integration of ¹⁷ analysis across different tools and supports subsequent data processing through standardized ¹⁸ outputs.

¹⁹ Statement of need

²⁰ Exergy analysis is an effective tool for assessing the quality of energy and capability to generate ²¹ useful work. It facilitates the identification of thermodynamic irreversibilities within a system, ²² thereby offering a more comprehensive understanding of energy conversion processes. The ²³ quantification of exergy destruction enables researchers and engineers to develop strategies ²⁴ that enhance efficiency, reduce costs, and promote sustainable conversion technologies ([Meyer et al., 2009](#); [Petrakopoulou et al., 2017](#); [Tsatsaronis, 1993](#)).

²⁶ Despite its advantages, exergy analysis has not yet been widely integrated into most commercial ²⁷ software used for thermodynamic assessments, which primarily focus on energy and mass balance ²⁸ calculations. The calculation of both physical and chemical exergy of material streams, as ²⁹ well as an automated evaluation of the overall process, was seamlessly integrated into the ³⁰ open-source software TESPy ([Hofmann et al., 2022](#); [Witte et al., 2022](#); [Witte & Tuschi, 2020](#)). While this represented an important step toward facilitating the application of exergy ³¹ analysis, current exergy analysis efforts still rely heavily on user input, are prone to incorrect ³² interpretation of component balances, and lack interoperability with other open source or ³³ commercial tools. These shortcomings have driven the demand for specialized, user-friendly, ³⁴ automated open-source software that enables exergy-based analyses and interoperates with ³⁵ both commercial and open-source environments.

³⁷ To address these needs, ExerPy provides a Python-based solution that automates exergy ³⁸ analysis of energy-conversion systems via a JSON data interface. The tool includes an API that ³⁹ automatically connects to different process-simulation environments, autonomously identifies ⁴⁰ components and assigns exergy balances, enabling detailed and accurate exergy analysis across ⁴¹ the entire process. This level of automation streamlines the workflow, improving efficiency and

accuracy in applying exergy analysis, and thereby supports the optimization of energy-conversion systems from an exergy perspective.

Software design

ExerPy has been developed as an exergy-analysis tool that is independent of any specific process-simulation environment. A central architectural decision was to separate data acquisition and normalization from the exergy-calculation core. Simulation-specific adapters extract stream and component information from Aspen Plus®, Epsilon®Professional, or TESPy and convert it into a standardized JSON representation. The analysis module operates exclusively on this uniform component and connection based schema, which enables consistent component- and system-level balances, standardized outputs, and straightforward extension to additional analysis types (e.g., exergoeconomic analysis) without re-implementing simulation interfaces. Although TESPy had already incorporated an automated exergy-analysis tool (Hofmann et al., 2022; Witte et al., 2022; Witte & Tuschi, 2020), its integration within a simulator-centric codebase strongly couples analysis logic to a specific data model and limits interoperability. The extension of TESPy to effectively process simulation results from commercial tools would lead to an increase in complexity, dependencies, and maintenance requirements for users primarily interested in simulation. Therefore, a dedicated package was required to provide a single, maintainable, open-source framework that performs automated exergy analysis across both commercial and open-source tools, including user-supplied system data via JSON.

Features

ExerPy is divided into two main modules: the data-processing module, which manages the extraction and preparation of simulation data, and the exergy-analysis module, which conducts the detailed exergy calculations. The initial implementation supports Epsilon®Professional, Aspen Plus®, and TESPy. The architecture is outlined in the following sections and is shown in Figure 1.

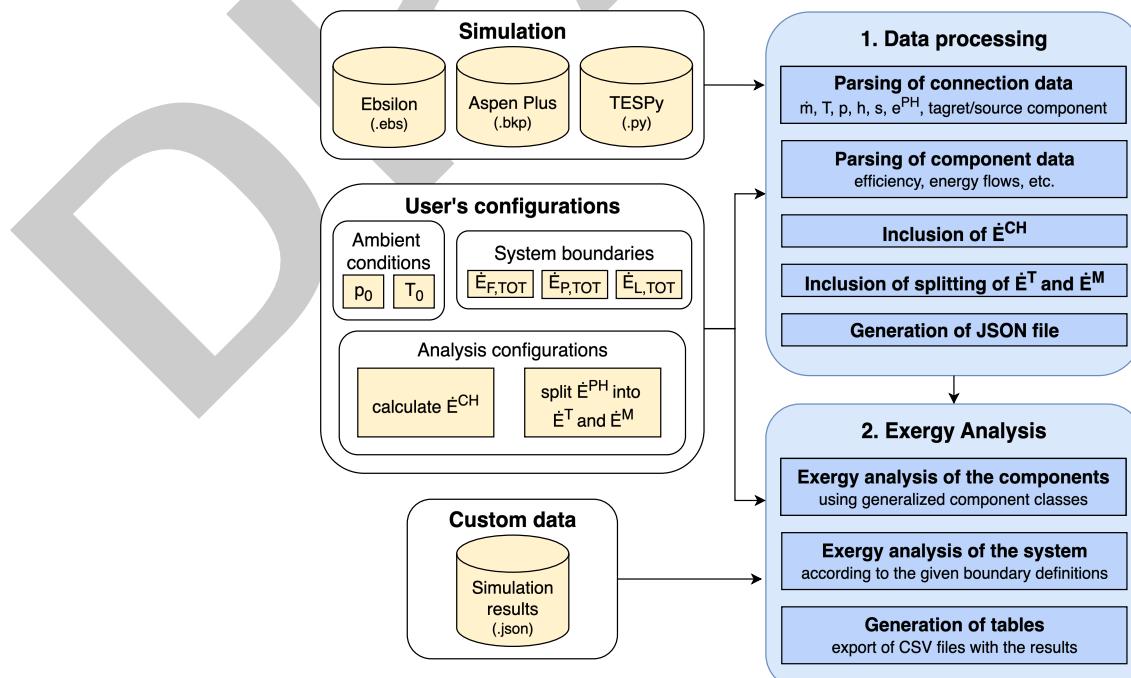


Figure 1: Structure of the ExerPy framework.

67 Data processing

68 A workflow begins with the parsing of simulation data from models created in
69 Epsilon®Professional, Aspen Plus®, or TESPy, using the respective functions: `from_epsilon`,
70 `from_aspen`, and `from_tespy`. It is important to note that the physical exergy, calculated
71 from the entropy and enthalpy of the streams, is obtained directly from the simulation tools.
72 Independent of the simulation software, users also have the option to supply their own JSON
73 file using `from_json`, which must conform to the required format for exergy analysis.

74 During the parsing process, connection data such as mass flow rate (m), temperature (T),
75 pressure (p), enthalpy (h), entropy (s), and physical exergy (e^{PH}) are parsed, along with
76 the identification of target and source components. In addition, component data including
77 efficiency, energy flows, and other relevant thermodynamic properties are also extracted.
78 Ambient conditions can be taken directly from the simulation, or specified manually by the
79 user.

80 ExerPy also allows the splitting of physical exergy into thermal (e^T) and mechanical (e^M) parts.
81 This separation enables a more comprehensive analysis of thermodynamic processes, especially
82 for components operating below ambient temperature (Morosuk & Tsatsaronis, 2019). These
83 values are calculated using the native property functions of the simulation tools. In the initial
84 release of ExerPy, this separation is not yet supported in Aspen Plus® due to limited access to
85 thermodynamic functions, but it is planned for a future update.

86 After data parsing is complete, chemical exergy is calculated from stream composition following
87 TESPy's approach (Hofmann et al., 2022) and the reference environment developed by Ahrendts
88 (Ahrendts, 1977). For pure substances, values are obtained from tabulated data in the selected
89 thermodynamic model. For mixtures (e.g., air, flue gas), ExerPy computes specific chemical
90 exergy from constituent molar fractions using standard assumptions (ideal behavior for gas
91 mixtures, with adjustments for condensables such as water). Finally, all parsed and calculated
92 data are consolidated into a standardized JSON file, independent of the simulation tool used,
93 including all the necessary information for a comprehensive exergy analysis.

94 Exergy Analysis

95 The framework performs exergy analysis at both the component and system levels. Each
96 component of the system—such as turbines, compressors, and heat exchangers—is represented
97 by a Python class that automatically assigns the exergy of the fuel and the exergy of the
98 product of the component. Using these definitions, ExerPy calculates relevant metrics for each
99 component (i.e., the exergy destruction and exergetic efficiency). Thermal energy losses of
100 components are included in their exergy destruction, and streams discharged to the environment
101 are treated as exergy losses of the overall system. This approach provides coherent calculations
102 of the inefficiencies of individual component and supports targeted optimization.

103 At the system level, the total exergy balance is established by evaluating the exergy of streams
104 crossing the system boundaries. To perform this analysis, and in the current release of the
105 tool, it is necessary for the user to specify the product, the fuel, and the exergy loss of the
106 overall process. The system-level exergy analysis yields the overall exergetic efficiency and the
107 total exergy destruction of the overall system. Finally, the framework allows the results to be
108 exported as CSV files for further examination and integration into additional workflows.

109 Validation

110 Validation has been conducted based on three different case studies documented in the online
111 documentation of the framework (Tomasinelli et al., 2025). The results of the exergy analysis of
112 a combined cycle power plant simulated with Aspen Plus® and with TESPy show a maximum
113 difference of 1% compared to the simulation results from Epsilon®Professional, validating the
114 accuracy and confirming the flexibility of the tool. Additional applications and validation of

¹¹⁵ ExerPy, e.g., the CGAM process (Valero et al., 1994) and a heat pump, are also available in
¹¹⁶ the documentation and on the GitHub repository.

¹¹⁷ Research impact statement

¹¹⁸ The demand for automated, publicly accessible exergy-analysis workflows is evidenced by the
¹¹⁹ prior adoption of TESPy's exergy-analysis functionality in peer-reviewed educational (Hofmann
¹²⁰ et al., 2023) and applied studies (Barandier et al., 2023; Fritz et al., 2024; Hofmann et al.,
¹²¹ 2024). In addition to this, the near-term significance of automation has been emphasized in
¹²² perspective work on the future of exergy-based methods, explicitly discussing the relevance of
¹²³ streamlined and automated implementations (Tsatsaronis, 2024).

¹²⁴ ExerPy translates this demonstrated demand into a community-ready software by offering a
¹²⁵ dedicated, versioned, and openly developed package within the oemof ecosystem. Community-
¹²⁶ readiness signals include an OSI-approved license, public source repository, automated test
¹²⁷ infrastructure, and user documentation with working examples and contribution guidance,
¹²⁸ supporting reproducible analyses and facilitating external adoption.

¹²⁹ AI usage disclosure

¹³⁰ Generative AI tools (multiple ChatGPT models, Claude Code, and DeepL Write) were used
¹³¹ to generate and modify portions of the codebase and to support language editing of this
¹³² manuscript. All AI-assisted changes were reviewed and validated by the authors.

¹³³ Acknowledgements

¹³⁴ Parts of this work were funded by the German Federal Ministry for Economic Affairs and
¹³⁵ Climate Action through the research project SecöndLife, grant number 03EI1076A.

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