

¹ District Energy Model (DEM): A Python framework for modelling renewable energy integration and flexibility at district scale.

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⁹ Summary

¹⁰ The transition to locally generated, decentralised, and renewable energy technologies is a promising pathway toward net-zero emissions and a decarbonised energy system. This transition requires systematic evaluation of potential future scenarios for technology adoption across multiple spatial scales. Computational energy system models are used for this purpose. ¹¹ The *District Energy Model (DEM)* is a Python-based multi-energy system model designed to simulate scenarios from the neighbourhood to the regional scale, with a focus on the ¹² integration of decentralised renewable energy technologies such as solar, wind, and biomass. ¹³ DEM can be used to execute simulation and optimisation studies at hourly resolution using ¹⁴ a snapshot-year methodology. It is released as an open-source Python library on [PyPi](#). The ¹⁵ model can be launched via a command-line interface, not requiring any Python programming ¹⁶ knowledge. Alternatively, it can be accessed programmatically through Python. DEM requires ¹⁷ two types of input: *configuration files* and *data files*. *Configuration files* specify the simulation ¹⁸ settings (e.g., included buildings, temporal scope, output variables) and define the energy ¹⁹ system configuration (e.g., scenarios, technologies, selected year). These configuration files ²⁰ are provided in YAML format ([Ben-Kiki et al., 2009](#)). Configuration parameters may also be ²¹ passed directly to DEM in Python. *Data files* contain model data such as energy demand ²² profiles, generation potentials, and ambient conditions. They are provided in Apache Feather ²³ format ([Apache Software Foundation, 2025](#)). For selected regions, these data files have been ²⁴ pre-compiled from public datasets and made available in a public repository, providing a fully ²⁵ parameterised model without requiring users to source or preprocess data. For example, data ²⁶ for a full parameterisation of DEM is available for all municipalities in Switzerland. ²⁷ ²⁸ ²⁹ ³⁰ ³¹ DEM comes with a detailed [documentation](#), which includes step-by-step instructions, descriptions of modelling approaches and methods, and links to related publications and research.³²

³³ Statement of need

³⁴ Several countries have defined national net-zero emission targets ([IPCC, 2023](#)). Switzerland, for ³⁵ example, aims to reach net-zero by 2050 ([The Federal Council, 2025](#)). Achieving such targets ³⁶ generally requires a structural shift from large, centralised generation to decentralised renewable ³⁷ resources, including solar, wind, and biomass ([Trutnevyyte et al., 2024; Van Liedekerke et al., 2025](#)). ³⁸ To support energy system planning and policy design, scenario evaluation must be ³⁹ carried out at local scales such as districts, municipalities, cities, or similarly sized regions. This ⁴⁰ includes analysing system-integrated deployment of renewable energy generation, conversion,

41 and storage technologies, assessing alternative demand trajectories, and identifying solutions
42 optimised for specific objectives such as cost, emissions, or security of supply.

43 DEM provides these functions with a specific focus on the use of local renewable energy
44 resources and the integration of decentralised technologies within local system boundaries.
45 Multiple energy system and demand scenarios can be defined, simulated, and compared.
46 Increased penetration of variable, distributed resources raises the relevance of supply- and
47 demand-side flexibility (Golmohamadi et al., 2024; Kachirayil et al., 2022). DEM models
48 several flexibility options, including flexible electric vehicle charging, thermal and electrical
49 storage, photovoltaic curtailment, and sector-coupling.

50 Existing multi-energy system models have been applied extensively in case studies of local
51 energy scenarios, but they typically target a single location. Each new application requires
52 reparameterisation and new data collection, including demand profiles, cost estimates, and
53 technology characteristics. Data acquisition and preparation dominate the modelling workload
54 in such studies. DEM removes this burden for selected regions by providing pre-compiled and
55 pre-processed datasets assembled from public sources. Simulation and optimisation studies can
56 therefore be executed with minimal configuration (e.g., selecting the buildings to include) while
57 maintaining full flexibility to replace any pre-configured dataset with user-defined data when
58 required. For regions not included in the provided dataset, users can construct the necessary
59 data using the specifications provided in the documentation. DEM's input data architecture
60 allows datasets to be provided at a large regional scale (e.g., an entire country) that can then
61 be used to run simulations on any spatial subset of that data, such as individual municipalities
62 or districts.

63 Optimisation is optional in DEM. Many scenario questions, such as assessing the impact
64 of a specific technology, do not require optimisation because the system configuration is
65 fixed, the technology capacities are predetermined, or the aim is to examine feasibility, energy
66 balances, or system interactions rather than to identify an optimal design. Simulation is also
67 appropriate when the objective is to reproduce prescribed operating behaviour, when data
68 are insufficient to support a reliable optimisation formulation, or when the computational
69 burden of optimisation is unjustified for large scenario ensembles. In these cases, DEM runs
70 simulations without invoking the optimisation module, yielding short computation times and
71 direct results. This distinguishes DEM from optimisation-centric energy system models that
72 rely exclusively on optimisation and therefore cannot execute non-optimised scenario analyses
73 with fixed configurations or purely exploratory simulations.

74 Initially developed within the framework of a research project, DEM is designed for a di-
75 verse target audience extending from academia and research projects to decision-makers in
76 municipalities, energy utilities, and the industrial sector.

77 Modelling approach

78 DEM simulates energy flows within a defined district using a hybrid bottom-up and top-down
79 modelling approach. A “district” can represent any spatial scale, from a small group of buildings
80 to an entire municipality or city. Building-level attributes are modelled individually (e.g., type,
81 location, size, age, heat and electricity demand, heating system, and on-site solar potential).
82 Other parameters are defined at district scale, including wind and biomass resources, ambient
83 conditions, and mobility demand. Each simulation is constructed from three elements: a set of
84 available resources (e.g., wind, solar, biomass, hydro), a set of technologies for generation,
85 conversion, and storage, and a set of demand profiles for heat, electricity, and mobility. These
86 elements interact through defined flows of resources and energy carriers such as electricity
87 and heat. An example system layout is shown in Figure 1. DEM imposes no fixed limit on
88 the number of buildings included, allowing customised definitions of district boundaries and
89 building selections.

90 The workflow consists of: (1) input-data collection; (2) model parameterisation and config-
 91 uration; (3) scenario generation; (4a) simulation; (4b) optimisation (optional); (5) output
 92 generation.

93 The optimisation module in DEM is implemented using the Calliope framework ([Pfenninger &](#)
 94 [Pickering, 2018](#)), which is based on the Pyomo optimisation programming environment ([Hart](#)
 95 [et al., 2011](#)).

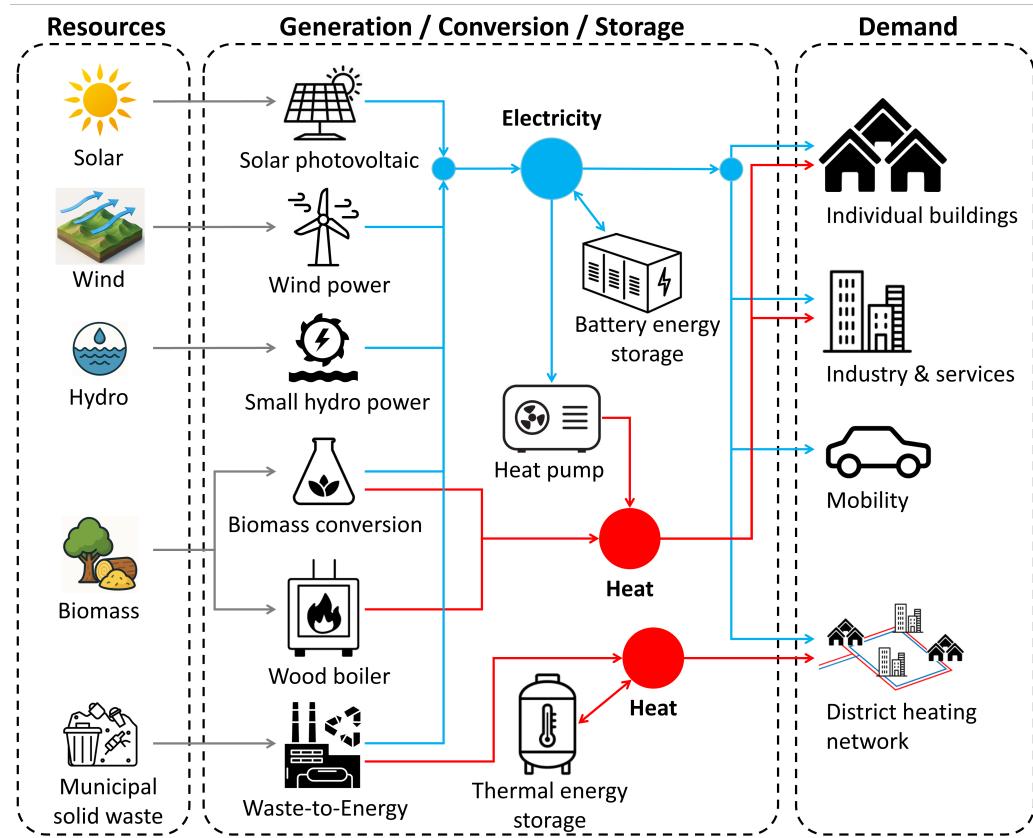


Figure 1: Schematic of an exemplary district energy system showing resources, generation, conversion, and storage technologies, and associated heat, electricity, and mobility demands. DEM supports many more technologies and scenarios than those illustrated here, as detailed in the [documentation](#).

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