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A review of systems modelling for local sustainability

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Abstract

The 17 Sustainable Development Goals (SDGs) represent a holistic and ambitious agenda for transforming the world towards societal well-being, economic prosperity, and environmental protection. Achieving the SDGs is, however, challenged by the performance of interconnected sectors and the complexity of their interactions which drive non-linear system responses, tipping points, and spillover effects. Systems modelling, as an integrated way of thinking about and modelling multisectoral dynamics, can help explain how feedback interactions within and among different sectors can lead to broader system transformation and progress towards the SDGs. Here, we review how system dynamics, as a prominent systems modelling approach, can inform and contribute to sustainability research and implementation, framed by the SDGs. We systematically analyse 357 system dynamics studies undertaken at the local scale where the most important SDG impacts and their initiators are often located, published between 2015 (i.e., SDGs' inception) and 2020. We analyse the studies to illuminate strengths and limitations in four key areas: diversity of scope; interdisciplinarity of the approaches; the role of stakeholder participation; and the analysis of SDG interactions. Our review highlights opportunities for a better consideration of societal aspects of sustainable development (e.g., poverty, inequality) in modelling efforts; integrating with new interdisciplinary methods to leverage system dynamics modelling capabilities; improving genuine stakeholder engagement for credibility and impacts on the ground; and a more in-depth analysis of SDG interactions (i.e., synergies and trade-offs) with the feedback-rich structure of system dynamics models.

Keywords: Sustainability; systems thinking; systems modelling; SDGs; review.

1 Introduction

The United Nations 2030 agenda, commonly known as the SDGs, outlines 17 goals and 169 targets aiming for global *sustainability*, that is a balanced achievement of economic prosperity, social inclusion, and environmental protection (1). Since their adoption in 2015, the SDGs have been used across international, national, and local scales to set out development priorities and guide funding and policy interventions for their achievements (2). With the increasing adoption of the SDGs, the use of systems models has been also rapidly growing in support of planning for sustainability. Systems models are often used to better deal with the complexity and uncertainty inherent in understanding the long-term progress towards interacting goals and their synergies and trade-offs (3). Several specific modelling approaches, such as input-output analysis (4), computational general equilibrium (5), system dynamics (6), and agent-based modelling (7), as well as broader model-based paradigms such as integrated assessment modelling (8), transitions modelling (9), and exploratory modelling (10-12), have been used to systematically analyse sustainability in various ways.

Among past systems modelling efforts (13), system dynamics has been used historically for modelling complex feedback interactions, analysing interlinkages between sectors, and understanding non-linearities, radical change, tipping points, with many applications in the broader sustainability context (14, 15) as well as those directly related to the SDGs (16, 17). System dynamics has co-evolved with sustainability research over the past 50 years (18), starting in the early 1970s with Jay Forrester's World Model(s) for the Club of Rome (19). This model underpinned the seminal work of *The Limits to Growth* (20) which analysed global dynamics and projected the collapse of the socio-technological-natural system by the mid-21st century. Since then, the co-development of system dynamics and sustainability research has continued within different sectors including energy and environment (21), climate change (22, 23), and socio-cultural systems (e.g., poverty), within the broader context of sustainability (24, 25). Modelling with system dynamics has been also advanced by new supporting theories for sustainability from other areas. Among them is the use of system dynamics for *transitions modelling* (26-28) where the feedback-rich structure of these models is used to explain transformational change in societal system, i.e., the destabilisation of an existing regime (e.g., fossil fuels) and the emergence of several competing niches (e.g., renewable energies). Another example of advances from other areas is related to *robust decision-making under deep uncertainty* (29, 30) where system dynamics models with relatively simpler and faster simulation engines (compared to other models) allow the evaluation of many decision interventions under many plausible scenarios.

Following the adoption of the UN 2030 agenda in 2015, modelling sustainability with system dynamics gained momentum with a stronger tie to the SDGs. This included a range of sector-specific models related to individual SDGs (e.g., energy, water), nexus models that analyse interactions of multiple sectors and inform multiple SDGs (e.g., water-food-energy (31)), and integrated assessment models that capture socioeconomic processes with the Earth system dynamics and provide a systemic view of SDG interactions with their complex synergies and trade-offs (14-17, 25).

Current system dynamics studies of sustainability are diverse in scope, scale, model feedback structure, and their supporting analytical methods, leading to notable differences in their contribution to the field. A few seminal works have provided an overview of system dynamics as a general tool for learning about complex system behaviours (32, 33), and some studies have reviewed system dynamics in specific sectoral or disciplinary domains (e.g., transportation (34), supply chain (35),

1
2 and water and hydrology (36)). Despite this progress, no study has reviewed and mapped the
3 potential contributions of this diverse literature to understanding and informing sustainability as an
4 indivisible whole with its broad sectoral and disciplinary span. Moreover, although sustainability has
5 been always part of the system dynamics portfolio, the specific focus on the SDGs is still relatively a
6 niche area. Given that sustainability research is at a stage where it has need for expansion and
7 5 integration with other areas for better understanding SDG complexities (37), a review of related
8 systemic tools and applications across disciplines is timely and can be a point of reference. Such a
9 review can bridge different research communities in sustainability and related fields to further
10 develop and adopt systemic tools that are more effective for understanding the dynamics and
11 interactions underlying the SDGs.
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15 Here, we map and evaluate the key characteristics of system dynamics in support of planning for
16 sustainability, framed by the SDGs, through a systematic literature review (Section 2). We critically
17 analyse what system dynamics means for sustainability research and implementation and what the
18 emerging research around the SDGs could potentially learn and benefit from systems thinking and
19 modelling tools and concepts. We frame our analysis (Section 3) and its discussion (Section 4) in
20 15 four areas: the diversity of scope (i.e., the heterogeneity of context and analytical objective);
21 interdisciplinary approach (i.e., diversity of research areas that supports the analysis); the role of
22 stakeholder participation (i.e., collaborative approach, timing of collaboration, the type of
23 stakeholder groups involved); the analysis of interactions (i.e., SDGs studied together as a system).
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27 The review of the state-of-the-art and the analysis of related works in these four areas can be useful
28 20 for modellers and non-modellers alike. The review provides a knowledge base of available tools and
29 applications that can promote the systemic analysis of SDGs in a qualitative or quantitative manner.
30 This review leads us to give due attention to the state-of-the-art of system dynamics modelling for
31 sustainability with the explicit aim of evaluating their contributions for the SDGs (Section 3). It also
32 leads us to identify the current gaps and to look forward to approaches that are currently not part of
33 the standard toolkit, but whose use in the future can push research boundaries and bridge the two
34 25 scientific communities of systems modelling and SDG research (Section 4).
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38 In our review and analysis, we focus on sustainability at a local scale, namely in relation to
39 communities, small-scale businesses, cities, states or provinces, and sub-national regions. The
40 choice of local scale is prominent for two reasons. First, SDG progress is highly dependent on
41 30 knowing and managing the changes on the ground (38). Highly nuanced socioeconomic conditions
42 and capacities and the diversity of stakeholder interests necessitate bottom-up initiatives that can
43 be tailored to specific local characteristics (39). Bottom-up, grassroots efforts can be an opportunity
44 to address place-based needs and priorities while also recognising the interlinkages with
45 sustainability aspirations at higher scales. These give a crucial relevance to the local scale in the
46 study of SDGs as a context where both impacts and initiators are on the ground. Second, among
47 35 different systems modelling approaches, system dynamics is also a suitable approach for the
48 modelling of local dynamic mechanisms and small-scale case studies given its faster model building
49 process (e.g., compared to computational general equilibrium or optimisation models) and less data
50 requirement (e.g., compared to agent-based modelling), with many salient examples in the
51 literature.
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57 **2 Methods**

58 We undertook a systematic review consistent with the Preferred Reporting Items for Systematic
59 Reviews and Meta-Analyses (PRISMA) (40) via four main steps (Figure 1): (1) systematic literature
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search in Scopus; (2) screening of the search results for inclusion in review; (3) coding of included publications for data collection; (4) synthesis and analysis of collected data.

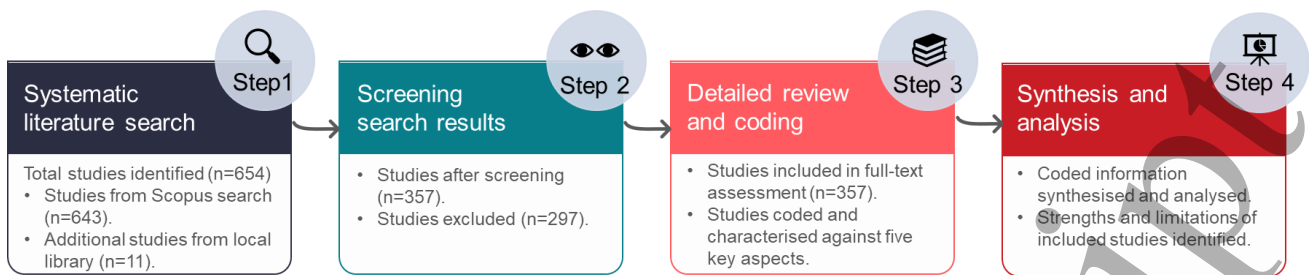


Figure 1. An overview of the steps in the systematic review.

2.1 Systematic literature search

The first step aimed to systematically search for peer-reviewed articles in the Scopus database. Scopus was chosen because it covers a diverse range of sustainability and systems modelling journals related to the topic of this review. We started with a preliminary search to scope the diversity of keywords used in the related articles with a test search string: 'system dynamics' AND sustain* AND local*. This initial scoping in conjunction with our knowledge of the literature iteratively developed the final search string constructed by three keyword components:

- The first component captured studies which used *system dynamics*. We used 'system dynamics' as one keyword rather than 'system' or 'dynamics' independently which could refer to other meanings not necessarily related to the system dynamics approach.
- The second component was to include studies that are related to *sustainability*. Here, we used three keywords. The first was 'sustain*' to include articles with any combinations of sustainability, sustainable development, and sustainable development goals. The other two were 'environ*' and 'ecolog*' to include studies that used environment or ecology as the integral components of (and sometimes interchangeably with) sustainability. The current selected keywords were informed by the scope of this review (i.e., specific to sustainability) and aimed to return a manageable number of papers rather than being too restrictive or too open. We did not use 'SDG' as an independent keyword to minimise the exclusion of possible relevant system dynamics applications which contributed to various sustainability aspects (i.e., food, population, energy) so far but without mentioning the term *SDG* specifically. Even slightly expanding these keywords (e.g., 'health', 'energy') could return thousands of additional articles. However, we did not include these additional keywords as separate search terms as we only aimed for those system dynamics studies which were discussing these sectoral issues in a sustainable development or sustainability context rather than articles related to general issues in healthcare, fossil fuels, or the electricity market.
- The third component was to capture the diversity of *local* scale studies. We used five keywords of 'local*', 'rural*', 'cit*', 'urban*', 'region*' to reflect different sub-national scales.

A second reviewer independently evaluated the composition of keywords to maximise the coverage of relevant articles and limit the inclusion of irrelevant articles. Using the selected keywords, we constructed the following search string with a time span from the beginning of 2015 (i.e., SDGs inception) to (mid-) 2020 (i.e., when this review was conducted):

- TITLE-ABS-KEY("System Dynamics" AND (sustain* OR ecolog* OR environment*) AND (local* OR rural* OR cit* OR urban* OR region*)); LANGUAGES: (English) AND DOCUMENT TYPES: (article, review); SOURCE TYPE: (Journal); TIMESPAN: 2015–2020.

We excluded irrelevant subjects (e.g., neuroscience, physics and astronomy, biochemistry, genetics, and molecular biology) from our search results. This resulted in 643 total articles (without duplicates). In addition to the results of the search string, we also added 11 other relevant articles from our local library which did not appear in the search results with the selected keywords. Together, they formed 654 potential articles (Data S1) to be assessed for relevance in the next step.

We acknowledge that the scope of our search (SDG, local scale, post-2015) may have excluded some of highly cited system dynamics works for sustainability (e.g., those before 2015 or at national/global scales). However, this was a deliberate choice and an inevitable trade-off with the literature search terms. The significantly higher number of articles from further expanding the search scope could limit the ability to undertake a detailed review in the next steps while our selected search scope helped limit the articles to a manageable number with limited irrelevancies.

2.2 Screening search results

The second step involved screening collected studies for their relevance. We firstly read each study's title, abstract, and keywords in accordance with the PRISMA approach for systematic reviews. In the screening process, we excluded studies in our search results related to national or global scales which were incorrectly captured, those not using the system dynamics approach, those with the keyword 'sustain*' or 'ecology*' which had no topic relevance (e.g., 'sustainable' state of atoms in nuclear physics, 'sustainable' human body synthesis in biology), and those articles with no full text online access. As a result, the search results were narrowed down to those studies directly related to system dynamics for sustainability at the local scale. A second reviewer randomly checked around 10% of studies considered eligible and those ineligible in the first iteration. In total, 297 articles were deemed irrelevant, resulting in a final total of 357 articles (Data S2) for coding via a detailed assessment.

2.3 Detailed review and coding

In addition to bibliometric information (i.e., author, title, year, journal) of each study, we also collected complementary information by reviewing the abstract and full text of the final articles and coding them against four key areas (Data S2).

2.3.1 Diversity of scope

We represented the diversity of scope in collected articles in terms of: (a) the relevant SDGs that each article contributed to; (b) the geographical location of their case study (either as their main focus or as an illustrative example for method development); (c) the scale indicating what 'local' represented in each article (i.e., community/business, city/urban, state/province, sub-national regions); (d) the analytical objective (i.e., case-specific, methodological, review) from the use of system dynamics to better understand the nature and the type of questions answered and the way the results were interpreted within each context.

In terms of the analytical objective, we considered an article as case-specific when the primary focus was on addressing a sustainability problem/question, and system dynamics was used to inform the (real-world or hypothetical) application. These studies often included case-based models for developing contextualised insights for a specific problem. We considered an article as methodological when the primary focus was innovation in method and technique (e.g., to improve a

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2 model or develop a conceptual framework), and any case studies were used only as an illustrative
3 example or proof of concept for the new methods. In contrast to case-specific, this analytical
4 objective focused on generic insight development and adoption of a high-level of abstraction which
5 remained independent from case studies and only made loose links to empirical data. We
6
7 5 considered an article as a review if the primary focus was a comparative study of multiple methods,
8 models, applications in relation to sustainability. In most cases during the detailed review, it was
9 possible to associate a study to multiple objectives, but we only considered one of them as the
10 primary objective based on our judgement.
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12 2.3.2 Interdisciplinary approach

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14 10 As a research method, system dynamics has been used across disciplines, from social, to physical,
15 to natural sciences. To provide an overview of the intellectual contour of contributions from various
16 research areas, we specified the disciplinary lens associated with each study, i.e., the main
17 research area of the system dynamic application (e.g., water and hydrology, agricultural studies,
18 climate change) and methods from other research areas used in support of system dynamics. We
19 also specified the scientific approach (i.e., qualitative or quantitative) used to apply system
20 15 dynamics. An approach was deemed *qualitative* when the study's primary focus was on
21 conceptualisation based on systems thinking, and *quantitative* when the study primarily relied on
22 generation and analysis of quantitative data through simulation modelling. Anything in between was
23 considered *semi-qualitative* (or equally semi-quantitative). We acknowledge that in some cases,
24 there was no clear-cut boundary between qualitative and quantitative approach in the paper, and
25 the coding was subjective to the authors' judgement about which approach dominates each article.
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28 2.3.3 Stakeholder participation

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30 Given the increasing role of stakeholders in co-developing models for sustainability (41-43), we
31 specified the type and extent of stakeholder participation in the reviewed studies. We initially
32 specified collaboration with stakeholders to determine whether the study was *participatory* or *non-*
33 *participatory*. We considered a study participatory when stakeholder interactions were considered in
34 25 developing concepts/models, validating generated results, and/or communicating insights. We
35 considered a study as non-participatory when the analysis relied on models and data from
36 documents with limited or no direct stakeholder input. If a study was deemed participatory, we then
37 specified the collaborative approach to indicate whether participation was front-end (i.e., engaging
38 with stakeholders from early stages of problem definition and model development), back-end (i.e.,
39 engaging towards the end in validation of results and communication of outcomes), or both. We also
40 30 specified stakeholders involved to identify the type of stakeholder groups, including community
41 members (i.e., general public), sectoral practitioners (e.g., technical experts who are often involved
42 in the on-ground management), decision-makers (i.e., responsible for policy-making, budget
43 allocation, etc.), and researchers, who informed the analysis.
44
45 35

46 2.3.4 Interaction analysis

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48 No individual SDG on its own is sufficient, and the whole SDGs as a system of synergistic
49 reinforcements (44) is necessary for successfully implementing the 2030 Agenda. This signifies the
50 importance of interactions and dependencies among the SDGs. We specified whether each study
51 40 could be related to a single or multiple SDGs (and which SDGs were studied together) to highlight
52 to what extent past system dynamics research in local sustainability has covered, directly or
53 indirectly, the interactions between various societal, economic, and environmental sustainability
54 goals. While we identified interactions, we did not analyse whether the study was related to
55 synergies or trade-offs (i.e., positive or negative interactions, respectively) given that at least some
56 insights related to both can be often identified when a study discusses interactions.
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2.4 Synthesis and analysis

We synthesised the coded information across all included studies (Data S2) and provided an analytical summary from strengths and limitations of the current literature in relation to the four analysed areas listed in Section 2.3.

3 Results

This section provides an overview of the breadth of system dynamics modelling for sustainability, both in the general sense of the applications in the literature and in the sense of topical areas, objectives, case studies, techniques, and engagement approaches. The emphasis is on showing the diversity of the field and evaluating the contributions (i.e., areas of strengths and limitations), framed by the SDGs and at a local scale (Section 2).

3.1 Diversity of scope

In terms of the topical domain, system dynamics has been applied to local sustainability in the context of different SDGs, but the main focus has so far been limited to only a few goals (Figure 2a). From the 357 reviewed studies, the highest numbers were related to applications in clean water and sanitation (SDG6), sustainable cities and communities (SDG11), and industry, innovation, and infrastructure (SDG9) (24%, 24%, and 18% of studies, respectively). To illustrate, system dynamics applications related to SDG6 covered a diversity of topics such as urban water supply and waste water management (45, 46), drought and water security in dry regions (47), and irrigation and water management for food production and agriculture (48). Studies related to SDG11 spanned a wide range of topics too, including sustainable transportation and emissions reduction in cities (49), building and infrastructure improvement (50), and eco-cities and sustainable urban development (51), amongst others. Studies related to SDG9 discussed topics such as sustainable development of specific industries and industrial regions (52) and improving public infrastructure (e.g., natural disaster management (53), logistics and transportation (54)). The use of system dynamics, however, has been limited in the case of other SDGs. For example, no article among the reviewed studies focuses on local challenges in gender equality (SDG5), and there are only a few works on reduced inequalities (SDG10), poverty reduction (SDG1), and quality education (SDG4) at the local scale (see Section 4.1 for discussion).

In terms of the geographical location, case studies presented in the system dynamics papers reviewed were concentrated in a few developing countries, whereas those in developed countries were less frequent and were also slightly more evenly distributed (Figure 2b). In both cases, only a few countries dominated the case studies related to each SDG (which could be due to the popularity of system dynamics as a field in their tertiary education system, and therefore a larger number of scientific papers). While there were common issues (e.g., sustainable food and agriculture in SDG2, sustainable cities and communities in SDG11) that were discussed frequently in both developed and developing countries, there were also some sustainability problems that were more popular in one compared to another. For example, issues related to health and well-being in SDG3 and partnership for the goals in SDG17 were less discussed overall and in developing countries in particular, compared to other goals (Data S2).

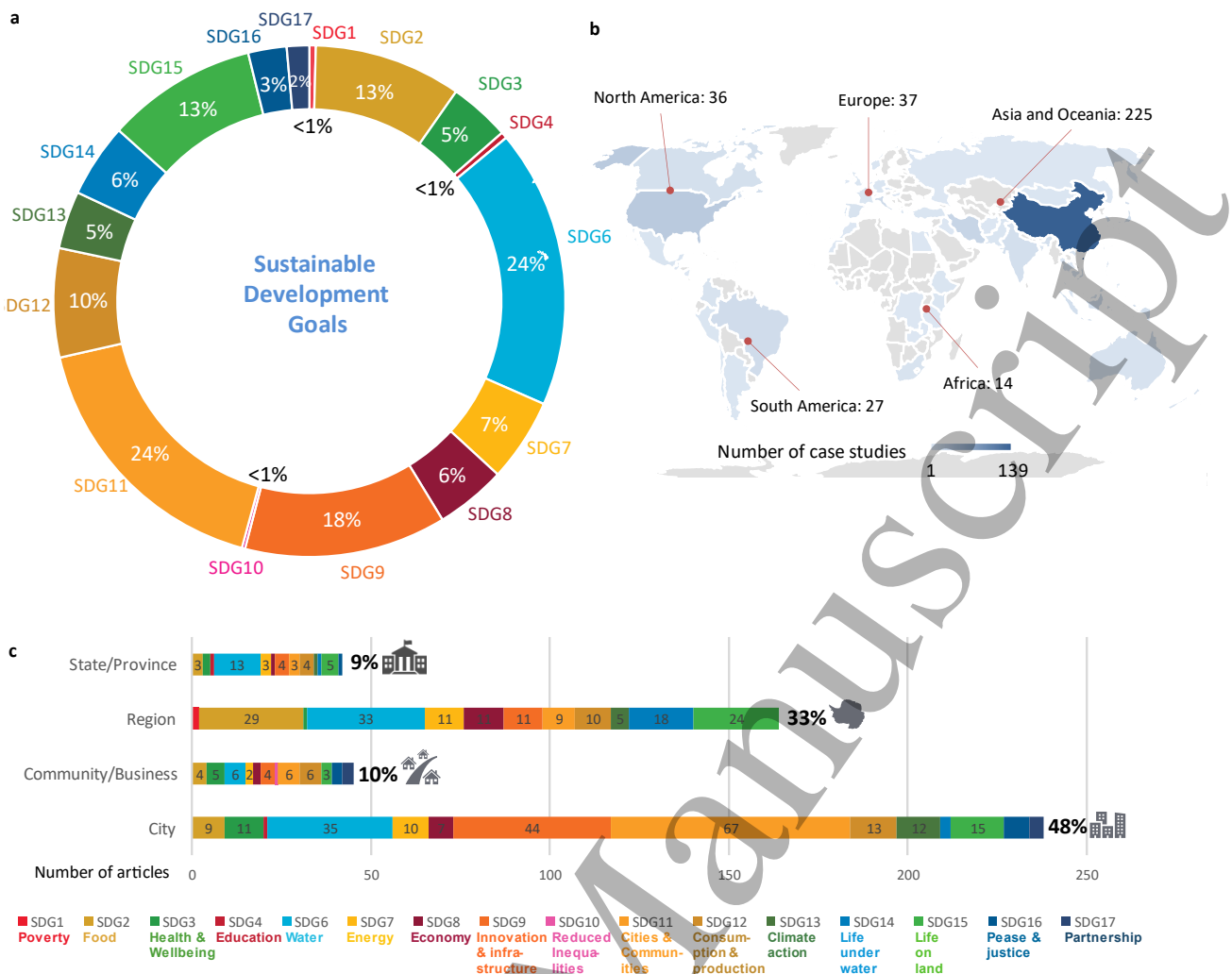


Figure 2. The diversity of scope in terms of (a) topical domain, (b) geographical location, (c) spatial scale. In (a), a study can cover multiple SDGs, hence the fractions in the pie chart add up to more than 100%, and the studies covering multiple SDGs are discussed in Section 3.4. Numbers inside stacked bar chart (c) show the number of studies per SDG. Numbers not visible in (b) and (c) are for those categories with limited space for text annotation. SDG5 (gender equality) is excluded from (a) to (c) as there is no study in the review related to this goal.

In terms of the spatial scale (Figure 2c), most studies focused on cities (48%) where the concentration of political power and capital tends to be, and where some of the most pressing local sustainability challenges, such as pollution (55), logistics, traffic, and mobility issues (56), and management of infrastructure projects (57), have a critical condition. Cities are also important in the study of local sustainability given their dense population and therefore their high vulnerability to natural and anthropogenic threats. Regions formed the second most common study scale (33%), mostly dealing with the management of natural resources (53), amongst other topics (e.g., the development of industrial regions (58), cross-comparisons (or meta-analysis) between performance of multiple regional areas (59)). Rural or sub-urban communities, local businesses, and states (or provinces) were presented in case studies less frequently (each ≤10%).

The results showed potential links between the scale and the SDG(s) of focus across the reviewed studies. To illustrate, studies related to SDGs 14 and 15 (i.e., life below water and life on land) were mainly discussed at the regional scale where the boundaries, drivers, and impacts of natural habitat change (e.g., forests, lakes, and wetlands) extend beyond a single city or a local community and can include large regions, sometimes even span multiple countries. On the other hand, SDGs 6 and 9 were discussed more frequently in applications at the city scale where issues such as storm and

waste water management (60), and logistics and transportation (61) are often more relevant dimensions of sustainability.

Similar relationships were also observed between geographical location and study scale.

Applications in developing countries focused on the city scale (50%) more frequently compared to other scales (e.g., 30% at the regional scale). One reason for this relationship is the growing population of cities in developing countries (e.g., 90% of global urban population increase has occurred in Asia and Africa (62)) and their rising sustainability challenges such as pollution, traffic, and logistics. Conversely, sustainability issues among studies in developed countries were more evenly distributed between cities (45%) and regions (42%) where the management of natural resources and ecosystems is often a key focus. Another reason is that developed countries have historically provided better socioeconomic conditions, but had created large environmental and material footprints (e.g., increasing emissions and food waste) in their surrounding environment which need to be addressed for the successful implementation of the SDGs (44). While studies at the scale of local communities were limited in general across all geographies, those in developed countries still had a larger share (11%) compared to those in developing countries (8%). This can reflect the emergence of local communities as a niche area of application and that more bottom-up initiatives in achieving sustainability exist among developed countries (38).

With respect to the analytical objective, most of the reviewed studies (63%) were classified as methodological improvement where the main focus was on model (or a conceptual framework) development, and the case study application was primarily a demonstration of the methods (Figure 3). These results reflected the diversity of models that existed in the system dynamics area and the fact that new models (both in terms of structure and parametrisation) were often being developed to suit a given problem (as opposed to other modelling domains where a handful of reference models are often applied across cases). One example of these studies was Liu *et al.* (63) where they focused on developing a specific decision support system to inform the impact of policy interventions on land-use change. Their primary contribution was methodological innovation via integrating multiple modelling paradigms (i.e., system dynamics, agent-based, cellular automata models). The popularity of method development was a common feature across studies related to most of the SDGs (e.g., 75%, 68%, and 63% of studies related to SDG9 on industry, innovation, and infrastructure; SDG11 on sustainable cities and communities; and SDG6 on clean water and sanitation, respectively).

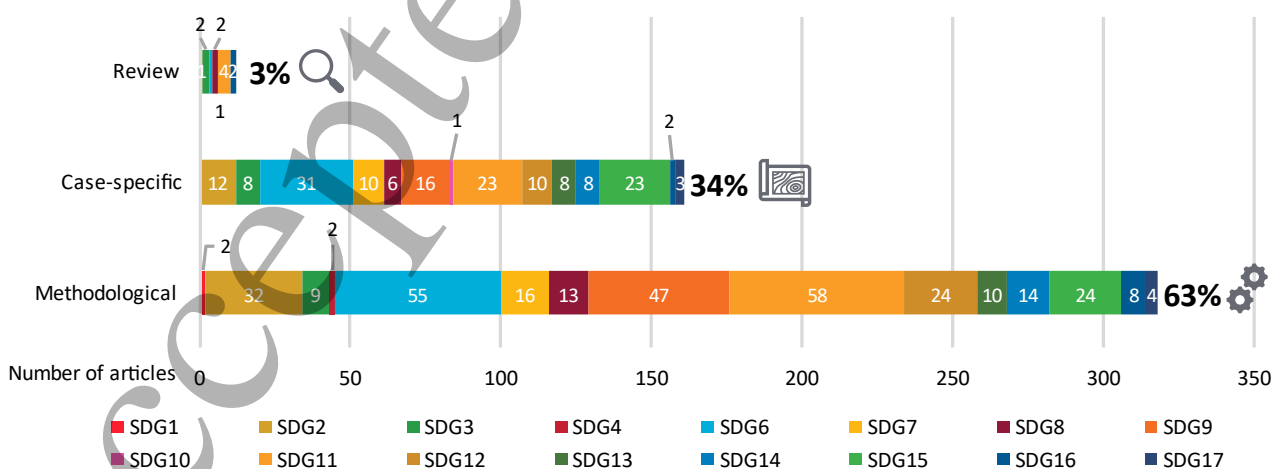


Figure 3. Analytical objective overview. SDG5 (gender equality) is excluded as there is no study in the review with a direct link to this goal.

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2 34% of the reviewed articles were classified as case-specific where an application in addressing a
3 sustainability challenge was the primary aim of the study. Any model development was assumed to
4 be used for answering related empirical questions. For example, (64) aimed to understand how
5 large-scale changes affect dryland sustainability as an empirical research question. They used
6 system dynamics as a tool to assess the interactions of various large-scale changes with dryland's
7 5 endogenous sustainability drivers and to quantify their impacts. Among the SDGs with the highest
8 share of case-specific articles were SDG15 on life on land (49% of related articles), SDG13 on
9 climate action (44% of related articles), and SDG17 on partnerships for the goals (43% of related
10 articles). Finally, 3% of the articles were classified as review where the aims were the evaluation of
11 related studies about a specific sustainability issue (65), a comparison of methods for a certain class
12 of local challenges (66), amongst others.
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16 17 **3.2 Interdisciplinary approach**

18 An important feature of system dynamics is its extensive connections with other established fields.
19 The results showed the study of local sustainability with system dynamics through the lens of a
20 broad range of (sometimes overlapping) disciplinary areas associated with different SDGs (Figure
21 5 4b). By frequency, these fields included sustainable development (32%), water and hydrology
22 (18%), and agricultural studies (11%), to less common areas such as climate change (3%),
23 transportation (4%), and ecology (4%). Despite association to a research area, most of the reviewed
24 studies were identified as interdisciplinary as they integrated a mix of methods across disciplines in
25 support of system dynamics, or their applications lay at the intersection of multiple sectors (e.g.,
26 water resource management and aquatic ecology (67), energy and climate (68)). The integration of
27 20 methods often aimed to support early modelling steps such as problem formulation and dynamic
28 hypothesis development. For example, Walters and Javernick-Will (69) integrated system dynamics
29 with social science techniques (e.g., survey, Delphi method) in a participatory process to specify
30 factors affecting rural water services and their dynamics, to inform the modelling of feedback
31 interactions. Integration sometimes aimed to enhance the model building process. An example was
32 the integration of system dynamics with other modelling approaches such as agent-based modelling
33 (70) or computational general equilibrium modelling (71) to better account for the heterogeneities in
34 25 actors' behaviour and decision-making. Another example was the integration of system dynamics
35 with Bayesian Networks (72) to effectively deal with missing data and uncertainty. Integration with
36 other methods was also used for the analysis of the modelling results. An example for this type of
37 integration was the use of scenario analysis for policy evaluation under alternative future conditions
38 and for coping with uncertainties and risk assessment in model projections (73) (see Section 4.2 for
39 discussion). Further details about how system dynamics links with other methods and for what
40 30 SDGs are available in Data S2.
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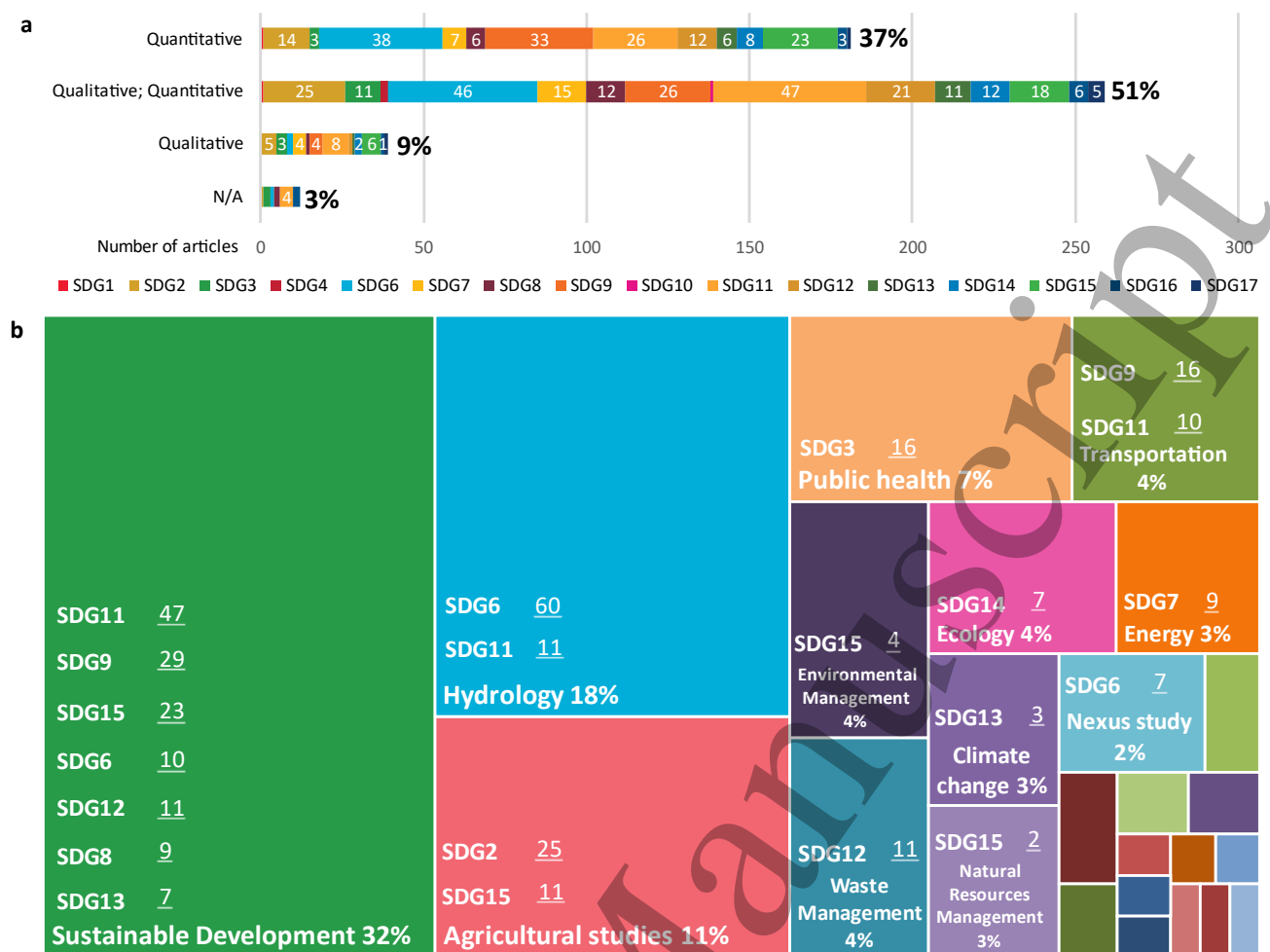
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Figure 4. Overview of interdisciplinary approaches used in support of system dynamics. (a) is share of qualitative vs. quantitative approach and (b) is the disciplinary lens from which each study adopts system dynamics. N/A in (a) represents studies (i.e., review articles) where our definitions of qualitative and quantitative approach were not applicable. SDG5 (gender equality) is excluded from (a) as there is no study in the review with a direct link to this goal. Each tile in (b) represents a disciplinary area, and the listed SDGs inside each tile represent those discussed the most from that respective disciplinary lens. Numbers inside (a) and (b) show the number of studies per SDG. Numbers and labels not visible in (a) and (b) are for those categories with limited space for text annotation.

Both qualitative and quantitative methods were adopted (Figure 4a). However, the majority of the reviewed studies (51%) adopted a mixed qualitative and quantitative approach. A common example of a mixed approach was when insights from a conceptual framework (e.g., cognitive causal loop diagram, scenario narratives) developed with stakeholders complemented and supported computer simulations and the quantification of system behaviour. 37% of the reviewed studies primarily adopted a quantitative approach. Exploring non-linear complex interactions (74) and analysing the effects of long-term uncertainties on system behaviour (75) were two examples of studies with a primarily quantitative approach. 9% of the studies also adopted a primary qualitative approach aiming, for example, at conceptualising system processes (76) or providing a detailed account of heterogeneities in system dynamics (77).

3.3 Stakeholder participation

Despite the potential benefits of engaging with stakeholders in support of modelling, only 28% of the reviewed articles were identified as participatory system dynamics studies (Figure 5a). The degree of participatory engagement also varied across the SDGs. For example, despite a limited number of articles found, applications related to SDG4 on quality education, SDG10 on reduced inequalities, and SDG17 on partnerships for the goals had the highest share of participatory articles (100%,

100%, and 71%, respectively). However, the majority of articles related to other SDGs, such as SDG13 on climate action, SDG9 on innovation and infrastructure, and SDG6 on clean water and sanitation had a limited share of participatory articles (11%, 16%, and 16%, respectively) (see Section 4.3 for discussion).

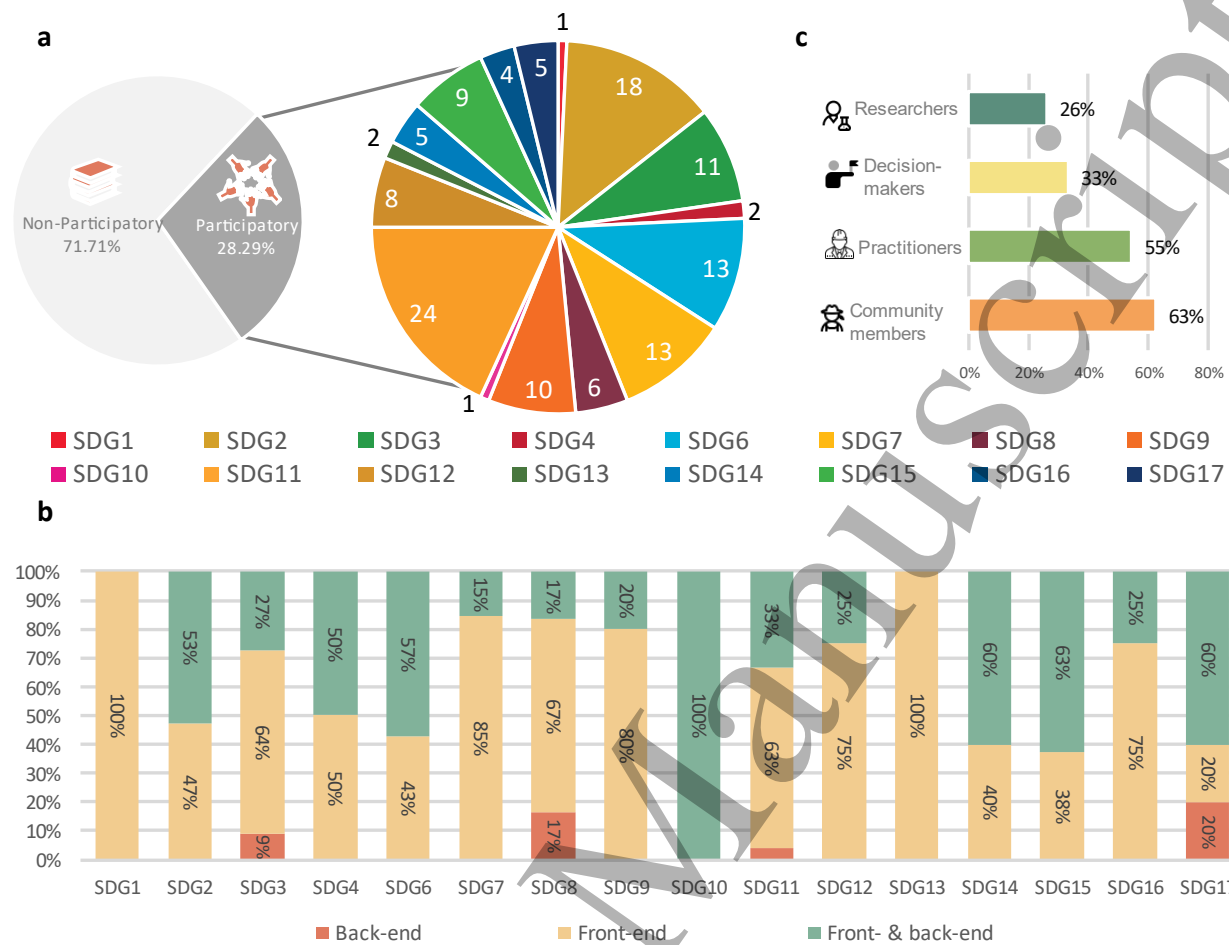


Figure 5. Overview of stakeholder participation. (a) shows share of participatory to non-participatory research across all included studies and in each SDG where the numbers represent the number of studies. (b) shows the timing of engagement across studies per each SDG. (c) shows to what extent different types of stakeholders were involved across the reviewed studies. See Section 2.3 for the definition of these stakeholder groups. A study can have multiple stakeholder groups involved, hence bars in (c) add up to more than 100%.

Among participatory system dynamics modelling studies (Figure 5b), stakeholders mostly had front-end engagement (59%) to inform problem definition and model development, for example, in demarcating the system boundary with inputs from the community (78) and co-developing a dynamic hypothesis with stakeholders to test and evaluate with models (79). 3% of the included studies were classified as back-end engagement in support of the validation of results and communication of outcomes with stakeholders. Stakeholder engagement in 38% of the included studies was identified as both front-end and back-end, for example to inform problem scoping and boundary setting during model development as well as to validate model results and ensure robust outcomes negotiated among participants towards the end of the project (80). There were also often multiple stakeholder groups involved in participatory system dynamics studies (Figure 5c). Among stakeholders, community members formed the most common group (63%), followed by practitioners (55%), decision-makers (33%), and researchers (26%) (note that more than one group can be associated to each study). The involvement of different stakeholder groups, however, varied across the SDGs. For example, practitioners were the most common stakeholder group involved in SDGs 6, 9, and 13 (i.e., clean water and sanitation, innovation and infrastructure, and climate action)

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2 whereas community members were more engaged in SDGs 2, 3, and 4 (i.e., food and agriculture,
3 well-being, and quality education).
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5 **3.4 Interaction analysis**

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7 System dynamics is a suitable tool for understanding and explaining the co-development of systems
8 underlying different goals as an indivisible whole. From the included studies, 133 articles were
9 associated with more than one SDG and captured complex (positive or negative) interactions among
10 social, economic, and environmental aspects of local sustainability (Figure 6). Among them,
11 sustainable cities and communities (SDG11) had the largest share of studies considering interactions
12 (59 studies). Cities and communities were a major part of the study of sustainability at the local scale,
13 and therefore many of the collected studies were associated with SDG11, in addition to links to other
14 goals. Notable interactions from sustainable cities and communities were with industry, innovation,
15 and infrastructure (SDG9) (22 studies). The interactions between these two goals were through
16 various targets. One example was via the city and community links to sustainable and resilient
17 transportation and logistics and supply chain infrastructure with affordable and equitable access that
18 supports economic development as well as human well-being (Target 9.1) (81). System dynamics
19 research on cities and communities also substantially interacted with clean water and sanitation
20 (SDG6) (14 studies) via targets under SDG6 such as access to safe and affordable drinking water
21 (Target 6.1) (82), improving water quality and wastewater treatment (Target 6.3) (83), addressing
22 water security and water scarcity (Target 6.4) (84), and integrated water resource management
23 (Target 6.5) (68). Sustainable cities and communities had the third highest share of interactions with
24 life on land (SDG15) (7 studies). Notable examples of these interactions were through sustainable
25 management of land and urban afforestation (85, 86), conservation and restoration of freshwater
26 ecosystems and urban wetlands (Target 15.1) (87), and the design of eco-friendly solutions to
27 enhance urban living (51) (Targets 15.1 to 15.3).
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31 Other prominent interactions were also observed in the results. The first interaction was between
32 responsible consumption and production (SDG12) and decent work and economic growth (SDG8)
33 (7 studies), for example, how sustainable tourism activities can create jobs while using natural
34 resources efficiently (88, 89). The second interaction was between zero hunger (SDG2) and life
35 below water (SDG14) (7 studies), for example, focusing on the management of commercial fishing
36 with a balance between stock replacement and prices of fish to supply food (90). The third
37 interaction was between clean water and sanitation (SDG6) and responsible consumption and
38 production (SDG12) (6 studies), for example, looking at the management of reactive nitrogen flows
39 from food production in wastewater systems (91). Having a high number of studies focusing on
40 each of these interactions was the result of similar (or closely related) targets under their respective
41 SDGs. For instance, observed interactions between decent work and economic growth (SDG8) and
42 responsible consumption and production (SDG12) were the result of two closely related targets
43 under these goals, i.e., decouple economic growth from environmental degradation via 'resource
44 efficiency in consumption and production' (Target 8.4) and 'achieve the sustainable management
45 and efficient use of natural resources' (Target 12.2), respectively (see Section 4.4 for discussion).
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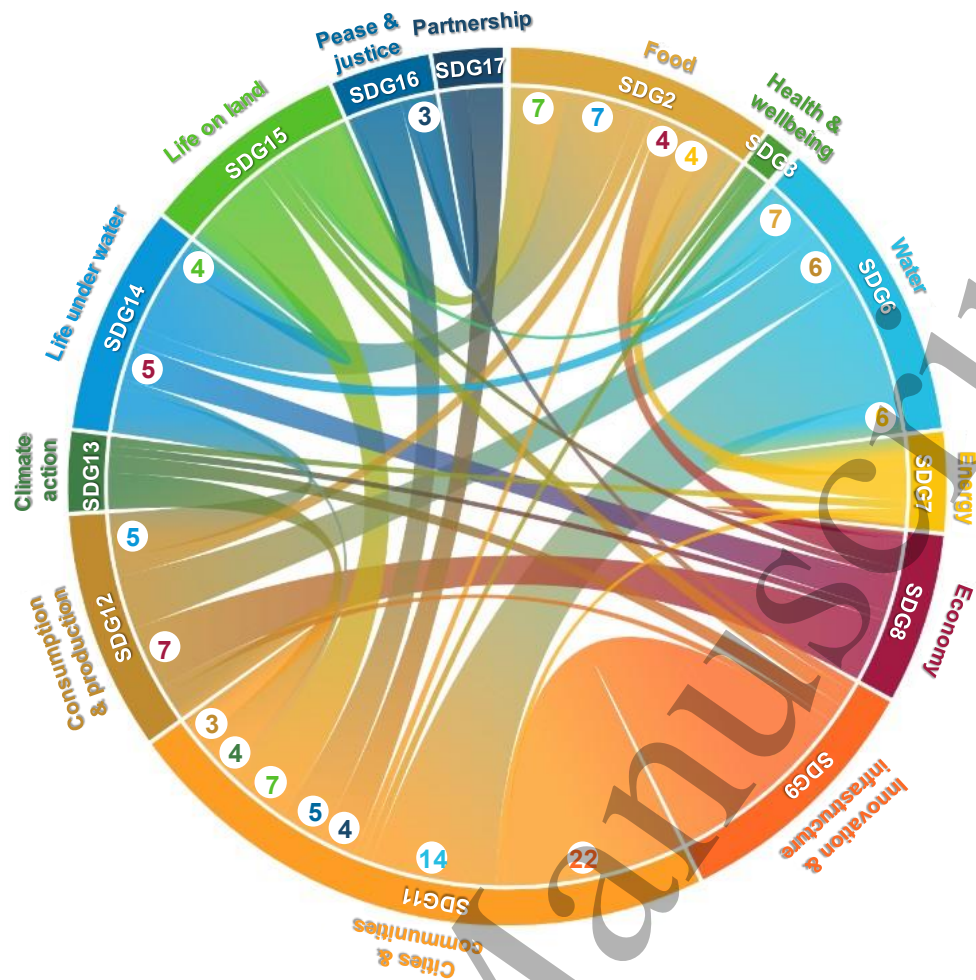


Figure 6. Overview of interactions between pairs of SDGs studied across the reviewed articles. Each SDG is represented by a fragment on the outside of the circle. Arcs drawn between each SDG represent interactions. The size of the arc is proportional to the number of studies, also annotated in small circles for major interactions. Annotated numbers are not repeated on both sides of each arc and are only on one side connected to the SDG with higher total interactions. SDG5 (gender equality) and SDG10 (reduce inequality) are excluded as no study in the included studies focused on interactions with these goals. The Chord diagram was developed using the Python-based Chord package (92).

4 Discussion

This section expands on the major gaps that were identified in Section 3 to show their implications for SDG research and then discusses potential ways to address these gaps in the future, in each of the four areas of our review (Section 2.3). In discussing the gaps and future priorities, the emphasis is on highlighting the ways that system dynamics and SDG research communities can co-develop and learn from each other's key issues and capabilities and also from those capabilities of other related fields.

4.1 Diversity of scope

Despite the diversity of the reviewed articles in terms of their scope, system dynamics studies in relation to the societal SDGs such as poverty (SDG1), quality education (SDG4), and inequality in its all forms (SDGs 5 and 10), have remained underdeveloped (Section 3.1). Societal SDGs in the reviewed studies were often modelled based on the 'average' properties of the (e.g., population, spatial, economic) systems with 'representative' indicators (e.g., standard gross domestic product per capita for modelling economy) and at an aggregated scale (e.g., general urban problems rather than issues among communities). One of the reasons why system dynamics models in relation to societal SDGs were limited is that the complex feedback relationships involved, such as the strong causalities among economic depression, natural disasters, income inequality, epidemic diseases,

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2 poor education, and poverty, are so far much less understood or explored across scales. Despite
3 this limited understanding, the influence of these societal SDGs and their policy interventions on
4 sustainable development are too significant to be ignored in system dynamics modelling. The
5 influence of societal SDGs becomes even more significant at the local level where the distributional
6 effects of societal SDGs appear (e.g., differences across various age cohort, socioeconomic status,
7 gender, identity, ethnicity) and the disaggregation of progress becomes important (e.g., differences
8 in progress between communities, cities, regions). This signifies a need to improve the links
9 between environmental SDGs (e.g., climate change, environmental footprints) and their related
10 societal goals (e.g., poverty, inequality) as causes and consequences (93) in system dynamics
11 models for the SDGs at any scale, especially at the local level.
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15 There are different ways for improving the links to societal SDGs in system dynamics models. First,
16 factors such as values, preferences, and distributional impacts can be endogenised in model
17 structural components for the modelling of societal SDGs such as poverty and inequality. An
18 example is in differentiating between gender and age cohorts in school enrolment and graduation in
19 a model structure to compute inequality in form of access to education across a population. Future
20 studies can endogenise similar human factors through improving current assumptions in models via
21 interdisciplinary learning and interaction with social sciences theories (94). A successful example of
22 learning and interactions with other theories can be found in the transitions modelling field as a
23 growing niche in computational social science for developing innovative models of co-evolution of
24 behaviours and technologies (9). Modelling societal SDGs endogenously would also require further
25 empirical research to understand and generalise common societal patterns important for modelling
26 and integration with other modelling techniques (e.g., agent-based modelling) that can help in the
27 (bottom-up) implementation human behaviour (94).
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33 Second, societal SDGs can also be incorporated beyond individual components, and through the
34 modelling of their feedback mechanisms with each other and with other biophysical systems.
35 Scientific communities tend to be closed in their fields (e.g., hydrology, agriculture, energy, climate)
36 often focusing on biogeophysical variables with limited linkages with their social drivers. However,
37 better linkages and an integrated interface between social and economic with biogeophysical
38 variables can improve the models. Such improved models facilitate a better understanding of the
39 emergent behaviour of complex socio-ecological systems arising from the interactions of societal
40 and environmental SDGs. For example, how deforestation (SDG15) can result from the interaction
41 of limited education (SDG4) as a key driver for diet change with high meat consumption (SDG2) and
42 therefore faster destructive land-use change. Few previous system dynamics studies have shown
43 examples of the modelling of human behaviour and its interactions with other environmental SDGs
44 such as diet shift (14) and climate change (95). Future studies can further integrate the dynamics of
45 societal and environmental SDG feedback interactions through learning from other areas that have
46 tackled multisectoral dynamics, such as network analysis, complexity science (96), and integrated
47 assessment (8).
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53 Third, the societal SDGs can be incorporated through exogenous scenarios (97, 98). The inclusion
54 of scenario drivers (e.g., in relation to gender, race, age, health) in models can help better capture
55 intersectional inequalities (e.g., gender wage gap, mental health and well-being of people of
56 different racial backgrounds) driven by external forces and assumptions. Some of these scenario
57 drivers are known and are part of the climate and sustainability scenario frameworks (99). System
58 dynamics applications need to parameterise and adopt these scenarios in their models.
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2 There are previous examples of the incorporation of societal SDGs through the ways mentioned
3 above (i.e., models, interactions, scenarios). Szetey *et al.* (39) recently used the SDGs to map a
4 balanced representation of (both societal and environmental) scenario drivers in a local community
5 for a future system dynamics modelling work. Pedercini *et al.* (25) endogenised societal SDGs such
6 as inequality and well-being in a system dynamics model. Randers *et al.* (16) also used the SDGs to
7 understand the interactions and the entanglement of humanity and the biophysical environment to
8 analyse environmental damage following socio-economic developments.
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11 **4.2 Interdisciplinary approach**

12 A diversity of interdisciplinary methods was used to leverage the capabilities of system dynamics in
13 the modelling of local sustainability. Past studies discussed some of the opportunities for integration
14 (Section 3.2). Examples of integration were with discrete event simulation to incorporate process
15 flow mapping features in system dynamics (100); with agent-based modelling to better capture actor
16 behaviour and system heterogeneities (101); with fuzzy cognitive mapping to engage with
17 stakeholders and use human opinion in the evaluation of causal relationships (102); with scenario
18 analysis techniques to better account for alternative futures in modelling (103); with Bayesian
19 networks to better manage missing data and uncertainty (104). Despite past efforts, integrations
20 with new methods have not necessarily been guided to address the specific SDG characteristics,
21 and therefore the opportunities and challenges of these SDG-motivated integrations have not been
22 yet investigated or fully operationalised.
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28 One of the specific characteristics in SDG analysis is the presence of deep uncertainties about the
29 future (105) (e.g., technological breakthrough, political instability, ecological collapse) which can
30 impact the validity of modelling assumptions and challenge the robustness of model results. To
31 better address uncertainty, integration with new methods from other areas, such as robust decision-
32 making (106) and exploratory modelling (107-109), has been suggested. Key to these new methods
33 is the systematic generation and impact assessment of alternative plausible assumptions (i.e., many
34 future projections which span the range of uncertainty across scenario, policy, model structure, and
35 model parameter settings). Despite a few past studies integrating system dynamics with these
36 methods at the national scale and in areas such as energy policy (30) and resource management
37 (29), no such integration exists at the local scale and directly in relation to the SDGs. A recent
38 synthesis of exploratory modelling (10) developed a typology of methods that can allow system
39 dynamics models to explore future possibilities and their implications in the multidimensional output
40 space of the SDGs. To illustrate, the design of experiments (29) can be used to randomly sample
41 from the uncertainty assumptions in system dynamics models, generate an ensemble of model
42 runs, and create an output space of projected futures with dimensions defined by the suite of
43 sustainability indicators from the SDGs. Stress-testing and scenario discovery (107, 110) can be
44 used to post-process the model runs and find crucial tipping points, scenario driver settings,
45 and policy choices essential to meet SDGs, or conversely settings that result in failure. Multi-
46 objective robust optimisation (108, 111) can be used to conduct new computational searches of the
47 uncertainty and output spaces to identify robust, adaptive pathways that can manage synergies
48 and trade-offs among SDGs in the face of future uncertainty. Available computational tools also
49 exist (109, 112) which can support the implementation of these methods with system dynamics
50 models as their simulation engine.
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58 Another important SDG characteristic is the many interactions, such as trade-offs and synergies,
59 which prevail between targets and between policy interventions affecting progress towards SDGs
60 (113). Ignoring these interactions may risk reversing progress through the spillover effects of one

1 goal's achievement at the cost of worsening several other goals. To make coherent policies and
2 help policy-makers understand which goals and targets can be achieved together and which trade-
3 offs need be made, the SDG interactions need to be at the centre of any system dynamics model
4 for sustainability. To identify synergies and trade-offs within and across SDGs for modelling,
5 correlation analysis from statistics is a method that can provide a measure of how variables are
6 related (44, 114). However, this data-driven analysis fails to address the key question of causal
7 inference in SDG interactions, and needs to be supported with other complementary methods.
8 Scoring-based approaches can be used to measure interaction intensity and direction based on the
9 judgements of experts and stakeholders (115) to inform causal relationships for system dynamics
10 modelling. Network analysis techniques can be used to explore more characteristics of interactions
11 among SDGs: interaction strength, nature of interactions (such as interdependence, constraints,
12 and reinforcement), and the 'nexus' relationship (116-118). In addition, interactive visual analytics
13 (119-121) can be used to facilitate stakeholder involvement and decision-making with the outcomes
14 of system dynamics models by visualising co-achievement of multiple SDGs and associated policy
15 pathways (113) as a result of trade-offs and synergies.

21 4.3 Stakeholder participation

22 System dynamics modelling was designed to be participatory from its inception, to involve
23 stakeholders and decision-makers throughout various phases of modelling, and to benefit from this
24 process in addition to the model as the final product (122). Stakeholder participation in system
25 dynamics modelling, also known as group model building, ensures that the model combines
26 scientific and local expert knowledge about a system (123). It aligns the mental models of
27 stakeholders and creates a shared knowledge, understanding, and meaning (124). It is also shown
28 to create consensus about both the causes and solutions of a problem, and to reduce conflict and
29 build trust among the participating stakeholders (125, 126).

30 Despite the group model building tradition in system dynamics and increasing integration of
31 modelling with genuine stakeholder engagement in sustainability research (41), more than 70% of
32 the reviewed studies still did not have (or did not explicitly mention) engagement with stakeholders
33 (Section 3.3). This transdisciplinary collaboration with stakeholders was limited across many of the
34 SDGs (e.g., SDG6 on clean water and sanitation, SDG8 on decent work and economic growth).
35 This gap can be justified in some cases in relation to the nature of the SDGs under study, for
36 example when they were more related to the Earth and biophysical processes with limited human
37 interaction and a stronger role for conventional disciplinary approaches. However, limited
38 participation in respect of cultural, political, and societal SDGs, which have a strong role for human
39 interaction (e.g., transforming human lifestyle and consumption patterns in SDG12, improving
40 governance arrangements in SDG16) can only indicate a gap and the need for further engagement
41 with stakeholders in their modelling with system dynamics.

42 Future system dynamics studies in relation to the SDGs can address the identified gap for more
43 participatory research to engage with stakeholders who will be most impacted by solutions
44 implemented in line with the SDGs. Participatory activities can be incorporated in various stages
45 from system conceptualisation to model use (127-129). In the conceptualisation stage, qualitative
46 participatory approaches, e.g., systems mapping with causal loop diagrams, can help delineate
47 problem boundaries and project scope through inclusive dialogues for translating the SDGs at the
48 local level. Stakeholder interests are also diverse with competing social, economic, and
49 environmental priorities among different groups, necessitating deliberation for compromising
50 between competing priorities that matter most to different stakeholder groups. Participatory system

1
2 dynamics can enable stakeholders to negotiate trade-offs and synergies among their divergent
3 priorities with a view of navigating diverse agendas, ensuring all voices are heard equally, and
4 moving towards a consensus and shared understanding about the problem and solutions.
5 Engagement in conceptualisation will later help enhance perceived credibility and usefulness of the
6 model among stakeholders (130).
7 5

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9 In the model use stage, participatory development and assessment of model-based scenarios
10 strengthen the understanding of system structures and resulting dynamic behaviour, why some
11 policy options work or not, and how they can be improved, contributing to trust building as well as
12 the policy-relevance of the eventual research findings. Engagement for model use can also help in
13 identifying meaningful actions for the attainment of sustainability priorities based on the local
14 knowledge of the resource gaps and the efficient allocation of resources for effective actions.
15 Interactive simulation environments (131, 132), which are enabled by the computational advantages
16 of system dynamics modelling, are particularly useful in this stage since they provide the
17 stakeholders with a first-hand experience of testing their own assumptions and policy options.
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21 Stakeholder engagement through participatory processes can have an important role in SDG
22 modelling at any scale (42). However, this role can be more important at the local scale where
23 stakeholder engagement can help in framing what sustainability could mean to local people and in
24 downscaling the high-level SDG agenda, aligned with internal resources and capacities and guided
25 by local needs and priorities (39, 133). In a smaller social entity (e.g., community, city), the
26 stakeholder knowledge is often richer and commitment is stronger, providing a homogenous source
27 of information for better modelling. Stakeholder actions also have higher impacts at the local scale
28 (i.e., the effects of every action can be significant), and therefore it becomes important that all
29 stakeholders have a sense of ownership about models and what they recommend for effective
30 implementations.
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35 4.4 Interaction analysis

36 While previous system dynamics applications analysed issues related to multiple development goals
37 together (Section 3.4), a specific focus on synergies and trade-offs as a key SDG feature and on
38 common dynamic mechanisms which generate these interactions remained elusive. Structuring and
39 evaluating these interactions and their underlying mechanisms upon which the SDGs operate are,
40 however, crucial for understanding alternative ways that cooperation (i.e., synergies) and conflicts
41 (i.e., trade-offs) can emerge and designing coherent solutions which are structurally nonobstructive
42 across the SDGs (134). For instance, strong interactions were identified between SDG11
43 (sustainable cities and communities) and several others (e.g., SDGs 6, 9, 13, 15) (Section 3.4;
44 Figure 6), which were related to the historical type of development, focusing on socio-economic
45 growth regardless of the environmental and material footprint. Such interactions exemplify a generic
46 trade-off pattern that is not specific to these two goals and can equally occur across a multitude of
47 systems varying in scope and goals. These recurring patterns, if understood correctly in relation to
48 their underlying structures, can be reverted in the future with more effective policy agenda aimed at
49 maximising the overall benefits across all SDGs.
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55 Future research and practical decision-making can use available system dynamics tools, such as
56 *system archetypes* (135), to specifically focus on SDG interactions and identify how synergies and
57 trade-offs can emerge from underlying systemic structures and their dynamic mechanisms. System
58 archetypes as a category of systemic tools enable the shift from simple behavioural correlations to
59 structural causality in the analysis of interactions where the aim is to identify generic explanations
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2 for classes of phenomena. System archetypes are applicable at any scales (e.g., regional, national).
3 The study of SDG interactions and the use of system archetypes in particular are important at the
4 local level, as the focus of this review. Heterogeneities in geography, governance, and resources
5 across different local contexts signify the need to understand and interpret SDG interactions in
6 relation to their local specificities (115, 136). Using system archetypes to obtain patterns and
7 5 classes of these local SDG interactions would allow comparability across multiple areas (e.g.,
8 communities, cities) and the transferability of lessons learned from one area to another.
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11 To demonstrate how system archetypes can help with understanding causes and effects of a
12 number of policy decisions related to the SDGs, we discuss three examples as an illustration
13 (Figure 7). *Drifting Goals* is one system archetype where two actions are possible to reduce the gap
14 10 (Figure 7). *Drifting Goals* is one system archetype where two actions are possible to reduce the gap
15 between the current state and the goal: either taking a corrective action which usually takes time, or
16 simply lowering the goal ambition (Figure 7a). An example is provided from governments' goal of
17 achieving certain carbon emissions reduction (SDG13) targets by 2030. In order to achieve this,
18 several crucial policy decisions must be made, such as the transition to cleaner (e.g., renewable)
19 forms of energy in contrast with fossil fuels (SDG7), both eventually creating balancing (or goal-
20 15 seeking) feedback mechanisms towards reducing the emission target gap to zero. Such transition
21 would likely take a long time to be completed, and in turn a significant delay can be expected before
22 recording remarkable drops in emissions. At the same time, in the short-term, considerable
23 disruption to the economy and the job market (SDG8) might arise, especially for states and regions
24 heavily reliant on fossil fuels. Combining such impact on the economy and the lack of evident
25 emissions reductions in the short-term, there might be pressure towards the government to reduce
26 the emissions targets instead (e.g., lower emissions cut, longer timeframe). Here, a trade-off can
27 20 clearly arise, between the need for climate action and transition to renewables (SDGs 7 and 13),
28 and the goal of short-term economic growth (SDG8).
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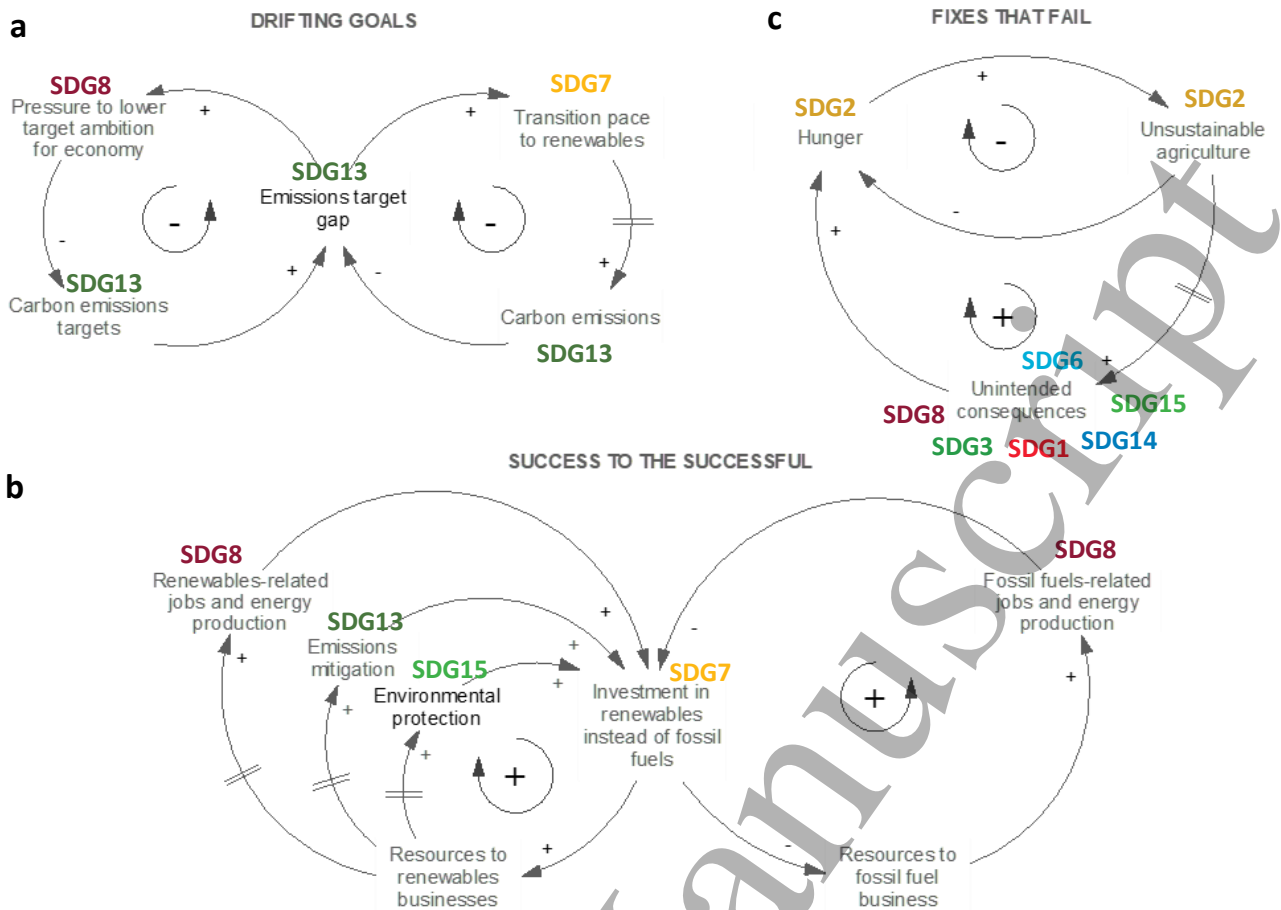


Figure 7. Three system archetypes to demonstrate examples of synergies and trade-offs between SDGs. (a) Drifting Goals, (b) Success to the Successful, and (c) Fixes that Fail. Arrows show interactions and the signs on arrows show the polarity of interaction (negative: trade-off; positive: synergy). The signed arrow loops indicate the feedback mechanisms where the positive sign represents a reinforcing loop (i.e., increase exponentially over time) and the negative sign represents a balancing loop (i.e., seek to achieve a goal). Relevant SDGs are identified next to each variable.

At the same time, the pace of transition to renewables, which relies on the investment in renewables instead of fossil fuels, can be also explained through another system archetype, called *Success to the Successful* (Figure 7b). In this archetype, the more success a certain sector/approach has had, the more resources are likely to be allocated to it in the future. This creates reinforcing feedback mechanisms, exponentially increasing the dominance of a sector or approach over time. In our example, the fossil fuel industry has received more resources and in turn delivered jobs, economic growth, and energy security in history. In this context, it is difficult for a niche and emerging system such as the renewable energy industry, as the most sustainable solution in the long run, to get attraction despite its apparent benefits. This is where the importance of a full system conceptualisation, beyond individual aspects, becomes evident: there are other variables to be included in the system, which might affect what we nowadays define as ‘success’, which is related to other SDGs (e.g., climate change mitigation, protection of environment and air quality), and will affect the final decision on resource allocation.

The complexity of the interconnection of factors within and across SDGs is more evident in the final example, which relies on the *Fixes that Fail* archetype where a short-term fix to an issue causes longer-term unintended consequences which may ultimately even aggravate the original problem (Figure 7c). Our example focuses on SDG2. While within SDG2, there is a focus both on zero hunger and sustainable agriculture, one region/community could address hunger and malnourishment by boosting agriculture with unsustainable practices, which might look like the best

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2 solution for the short term and increase food production in a goal-seeking behaviour. However,
3 several side effects can become apparent with reinforcing feedback mechanisms that can
4 substantially deteriorate the situation over a longer time horizon. Among these side effects are: soil
5 tilling, which can lead to unhealthy soil, malnourished plants, and in turn poor-quality food; soil
6 erosion, which can cause drinking water contamination (SDG6) (137); unsustainable agriculture
7 5 planning practices including deforestation, which causes loss of biodiversity (SDG15) and
8 increasing risk of disease pandemics (SDG3); mono-cropping, which increases agricultural water
9 demand (SDG6) (138) well as the need for chemicals and fertilisers, which can cause health issues
10 (SDG3) and whose runoff can cause eutrophication (SDG6) and marine and aquatic ecosystem
11 damage (SDG14). Over the long-term, such side effects will impact the country's economic growth
12 (SDG8), potentially increasing poverty (SDG1), and ultimately leading to a long-term failure in
13 10 addressing food security and food quality (SDG2).
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18 More archetypes can be further identified and extracted from the proposed examples and from other
19 SDG interactions. Given that previous studies (44) identified similar correlations (e.g., between
20 15 SDGs 3 and 6) for several regions of the world, it is possible to backup such archetypes with
21 historical data confirming or refining such conceptualisations. Overall, the goal for future research
22 and practice should be to identify and better represent such common behaviours in relation to their
23 underlying structural interactions, to move from simple association and correlations among SDGs to
24 a full understanding of causes and effects in such complex systems. This will ultimately enable a
25 better identification of effective strategies to maximise synergies and minimise trade-offs among
26 20 SDGs.
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30 **5 Conclusions**

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32 The integrated analysis of sustainable development with its diverse goals and sectoral areas
33 requires systemic approaches that can be extendable to various contexts, from well-being and
34 social inclusion, to economic development, to environmental and ecological protection, and from
35 25 local to national to global scales. Given the past methodological developments and the diversity of
36 applications in relation to almost all sustainability areas, system dynamics is well placed to address
37 some of the key challenges in the study of the SDGs. It can help in understanding the multisectoral
38 dynamics behind SDG interactions and representing a diverse set of (i.e., societal, economic,
39 environmental) sustainability indicators needed to measure the progress towards sustainable
40 30 development. The diversity of analytical objectives and the extent of participatory research observed
41 through the current review demonstrated this approach's flexibility to work under varying contextual
42 conditions across local-scale applications. The wide range of interdisciplinary methods used in
43 combination with system dynamics also showed opportunities for integration to overcome the
44 35 limitations of this approach through the strengths of other supporting methods. This richness of
45 models and their diverse applications reviewed in this article can provide an important knowledge
46 base for further adoption and improvement of system dynamics studies with applications related to
47 sustainability. Having said that, past studies were not complete in every aspect and had limitations
48 40 to be addressed in the future. Researchers can cover some of these gaps by improving the links
49 between societal factors and biophysical processes in sustainable development in models,
50 engaging with a wider variety of stakeholders during the modelling process, and understanding the
51 underlying systematic structures that drive interactions among competing agendas based on
52 archetypes.
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Supplementary Materials

Data S1. Collected articles from the literature search.

Data S2. Included articles in the review and their characterisation.

Code and Data Availability

The datasets (Data S1 and S2) generated during this study are available from <https://doi.org/10.5281/zenodo.5094092>. Further information and requests for resources and reagents should be directed to and will be fulfilled by the corresponding author.

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