

Re-framing the threat of global warming: an empirical causal loop diagram of climate change, food insecurity and societal collapse

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Abstract

There is increasing concern that climate change poses an existential risk to humanity. Understanding these worst-case scenarios is essential for good risk management. However, our knowledge of the causal pathways through which climate change could cause societal collapse is underdeveloped. This paper aims to identify and structure an empirical evidence base of the climate change, food insecurity and societal collapse pathway. We first review the societal collapse and existential risk literature and define a set of determinants of societal collapse. We develop an original methodology, using these determinants as societal collapse proxies, to identify an empirical evidence base of climate change, food insecurity and societal collapse in contemporary society and then structure it using a novel-format causal loop diagram (CLD) defined at global scale and national granularity. The resulting evidence base varies in temporal and spatial distribution of study and in the type of data-driven methods used. The resulting CLD documents the spread of the evidence base, using line thickness and colour to depict density and type of data-driven method respectively. It enables exploration of how the effects of climate change may undermine agricultural systems and disrupt food supply, which can lead to economic shocks, socio-political instability as well as starvation, migration and conflict. Suggestions are made for future work that could build on this paper to further develop our qualitative understanding of, and quantitative complex systems modelling capabilities for analysing, the causal pathways between climate change and societal collapse.

Keywords Climate change · Food security · Existential risk · Starvation · Conflict · Migration

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1 Introduction

Despite recent social protests and climate emergency declarations, efforts to mitigate climate change to-date are insufficient (Ripple et al. 2019). Greenhouse gas (GHG) emissions continue to rise and global warming above 3 °C is increasingly likely this century (Raftery et al. 2017). There is emerging evidence of amplifying feedbacks accelerating (Natali 2019) and dampening feedbacks decelerating (Walker et al. 2019). These feedbacks exacerbate the possibility of runaway global warming (Steffen et al. 2018), estimated at 8 °C or greater by 2100 (Schneider et al. 2019). Such temperature increases translate to a range of real dangers (The Center for Climate and Security 2020), shifting the narrow climate niche within which humans have resided for millennia (Xu et al. 2020).

Looking beyond the framing of 'global warming', there is concern that the effects of climate change may pose an existential risk to humanity, one that threatens 'societal collapse' or even extinction (Ord 2020). Understanding these worst-case scenarios is essential for good risk management (Kunreuther et al. 2013). Improving awareness of potential pathways through which climate change poses such a risk can help inform decision-making about interventions (Shepherd et al. 2018). Considering societal impacts that are more tangible for individuals, businesses and governments (Briggs et al. 2015), and better aligned with conventional risk priorities (Wagner and Weitzman 2015), may facilitate more effective action to mitigate climate change (Weber 2006).

A number of pathways through which climate change could cause societal collapse have been identified, one being via food insecurity (Gowdy 2020). Climate change is predicted to undermine agricultural systems and disrupt food supply (FAO, IFAD, UNICEF 2020), which may lead to economic shocks, socio-political instability as well as starvation, migration and conflict at local through to global scale (Rivington et al. 2015). While the climate science underpinning global warming estimates is well established (IPCC 2014), albeit subject to sensitivities, the uncertainties increase significantly when we start to consider these tangible societal impacts given the complex relationships involved (Butzer 2012). Our understanding of worst-case scenarios, and particularly of empirical evidence addressing the causal pathways through which climate change may cause societal collapse, is underdeveloped (Kemp 2020).

In this paper, we aim to identify and structure an empirical evidence base of the relationships between climate change, food insecurity and societal collapse. We do this using causal loop diagrams (CLD), a system dynamics approach that is useful for visualising the relationships between variables in a complex system (Randers 2000). This paper is organised as follows. In Section 2, we review the societal collapse and existential risk literature to refine the aim introduced above. In Section 3, we develop an original methodology to establish a new empirical evidence base and create a novel-format CLD of causal pathways between climate change and societal collapse. In Section 4, we present and discuss the results from the application of this methodology to the climate change, food insecurity and societal collapse causal pathway of interest. We conclude, in Section 5, by identifying avenues of future work that may build upon this paper.

2 Literature review

To refine the aim of this paper, introduced in Section 1, our review examines whether there is historical evidence of climate change as a mechanism of societal collapse and to what extent have causal pathways been documented to inform our understanding of climate change as an existential threat to contemporary society.

We first define the terms 'existential risk' and 'societal collapse' as used in this paper. Adopting Ord's (2020) definition, "an existential risk is a risk that threatens the destruction of humanity's long-term potential" be it incomplete destruction, such as societal collapse, or complete destruction, such as extinction. Adopting Kemp's (2019) definition, societal collapse is an "enduring loss of population, identity [and/or institutional] complexity"; it may be abrupt or gradual, but is typically rapid because it is notably transformative, and may be experienced by a local, national or the global community of people. Figure 1 presents a conceptual model of societal collapse, synthesised from the broader literature, to provide further contextual definition.

The rise and fall of civilizations has been documented since the earliest recordings of history and is increasingly studied to inform our understanding of societal collapse (Butzer and Endfield 2012). We consider two types of historical studies that provide insight into climate change as a mechanism of societal collapse in the past. We note that other mechanisms are also discussed in the literature, and there is debate about the role of different mechanisms in particular societal collapse events.

The first type of historical study empirically investigates an individual societal collapse event using primary sources, including anthropological, archaeological and paleontological data. Based on such data analysis, natural climate change has been asserted as a mechanism of societal collapse in many of these case studies, as established by de Menocal (2001) and Weiss and Bradley (2001). For example, Hodell et al. (1995), Haug et al. (2003) and Medina-Elizalde and Rohling (2012) analyse paleoclimate data alongside the archaeological record to show that drought conditions, driven by climate change likely due to solar forcing, contributed to the collapse of the Classic Maya civilization of Mesoamerica in ~eight–tenth century CE. Weiss et al. (1993), Cullen et al. (2000) and Cookson et al. (2019) show that regional aridity, driven by climate change likely due to volcanic forcing, contributed to the collapse of multiple societies across Mesopotamia, including the Akkadian Empire in ~twenty-second century BCE. Similarly, natural climate change has been implicated in the collapse of multiple Late Bronze Age societies around the Mediterranean (Cline 2014), including Mycenaean Kingdoms in ~tweffth century BCE (Finné et al. 2017), the Harappan Civilization of South Asia in ~nineteenth century BCE (Giosan et al. 2012), the Angkor Empire of Southeast Asia in

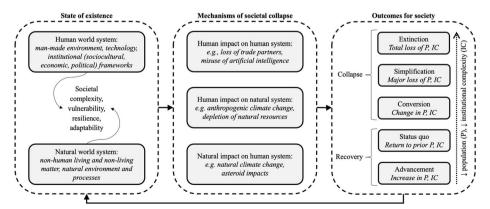


Fig. 1 Conceptual model of the overarching process of societal collapse

~fifteenth century CE (Buckley et al. 2010), multiple Chinese Dynasties (Zhang et al. 2006) and civilizations along the Silk Road (Li et al. 2016) during the previous millennium, the Norse Vikings of Greenland in ~ fifteenth century CE (Kintisch 2016) and the Tiwanaku Empire of Pre-Columbian South America in ~tenth century CE (Ortloff and Kolata 1993), amongst others.

These first type of studies establish precedence of natural climate change as a mechanism of societal collapse throughout history, demonstrating the risk that anthropogenic climate change similarly poses to contemporary society. However, the events examined occurred more than 100 years ago, with most dating back to ancient history, when societies were relatively isolated. Because these case studies pre-date contemporary society, they do not provide empirical evidence of anthropogenic climate change in context of today's highly interconnected society (Ehrlich and Ehrlich 2013).

Statistical evaluation of the frequency and significance of natural climate change relative to other mechanisms of societal collapse identified across these case studies has not yet been established within the literature. However, the second type of historical study does qualitatively examine collections of these case studies to develop theories of predominant modes of societal collapse. Three major modes are observed, as follows.

Fagan (2008) and McMichael (2017) focus on natural impact on the human system across multiple civilizations, concluding that natural climate change is predominant having significantly influenced human existence throughout history. Over the past 12,000 years, the natural and human systems developed within the stable climate niche of the Holocene Epoch (Runciman 2001). The associated geographic endowments governed human transition from band societies based on foraging to complex societies based on agriculture. Unfavourable subtle (e.g. weather variations) and drastic (e.g. natural disasters) shifts in climate influenced the collapse of complex societies either by direct loss of life or indirectly via resource insecurity. In particular, in this mode, typically, the loss of agriculture led to de-population via famine, migration or conflict due to food insecurity.

Ponting (1991), Wright (2004) and Diamond (2005) focus on human impact on the natural system across multiple civilizations, concluding that human overpopulation and overexploitation relative to the carrying capacity of the environment is predominant. Societal collapse via environmental degradation often involved unsustainable agriculture, exacerbated by natural climate change, leading to de-population as well as institutional breakdown via loss of economic stability and socio-political dysfunction due to magnified inequality. This mode aligns with early 'Malthusian catastrophe' (Malthus 1798), 'tragedy of the commons' (Hardin 1968) and 'overshoot-and-collapse' (Catton 1980) theories.

In their 12-volume magnum opus exploring the rise and fall of 28 civilizations, Toynbee (1961) concludes that "great civilizations are not murdered [but rather] they take their own lives" (Kemp 2019). Building on this, Tainter (1988), Acemoglu and Robinson (2012) and Johnson (2017) focus on human impact on the human system across multiple civilizations, concluding that societal complexity in relation to problem-solving inability (e.g. environmental degradation) and institutional dysfunction (e.g. inequality and oligarchy internally, trade ally and hostile neighbour relations externally) is predominant. As a society becomes more complex, it reaches a point beyond which "continued investment in complexity as a problem-solving strategy yields a declining marginal return" and it will be at risk of collapsing under its own weight via institutional breakdown and de-population (Tainter 1988). This mode

aligns with 'energy returned on energy invested' (Murphy and Hall 2010) theory, applied to explore societal collapse by Homer-Dixon (2006).

Diamond (1997), Turchin (2006) and Schwartz and Nichols (2010) examine why some civilizations have been able to thrive or recover, rather than collapse. They similarly conclude that societies have flourished due to combinations of favourable geographic endowment, managing their existence within the carrying capacity of the natural system, and co-operative action in problem-solving.

These second type of studies highlight that societal collapse involves a complex nexus of factors and dynamically interlinked events. For instance, Gibbon (1789) details how all three of the modes, described in the preceding five paragraphs, contributed to the collapse of the Roman Empire. These modes of societal collapse, although based on empirical evidence predating contemporary society, describe key aspects of the anthropogenic climate change problem faced today. While these studies describe causal pathways of relevance, to the best of our knowledge, no study has used CLDs to untangle the complexity and give structure to the dense information in this evidence base.

Across these historical studies, we observe no apparent temporal or spatial influence on the occurrence of societal collapse. Rather, societal collapse has been described as occurring in various forms, whether it be by known 'white-swan' or surprise 'black swan' events (Taleb 2007), in different geographic locations and times throughout history. Additionally, a quantitative statistical analysis by Arbesman (2011) shows that societal collapse has occurred randomly and independent of civilization life-spans. These qualitative and quantitative observations highlight that any society may be susceptible to collapse, much in-line with the *Red Queen hypothesis* of the *Law of Extinction* (Van Valen 1973).

From these historical studies, we observe sets of secondary determinants for each of the primary determinants introduced in Fig. 1, which are defined in Fig. 2. Considering a geographically bounded society, *emigration* refers to any permanent departure of population including both voluntary or forced migration, *conflict mortal-ity* accounts for deaths directly arising from any form of domestic or international conflict (e.g. due to war), and *natural mortality* accounts for deaths related to domestic environmental conditions (e.g. due to famine). The *loss of sociocultural norms, political structures* or *economic value* accounts for that which notably transforms the identity and institutions of the society.

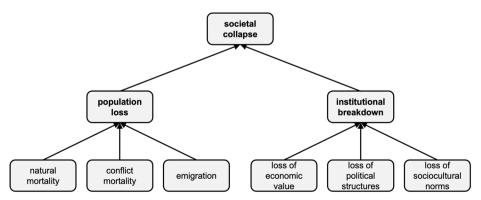


Fig. 2 Primary and secondary determinants of societal collapse for a bounded society

In addition to these historical studies, we consider the relatively nascent studies of existential risks (X-risks) that provide insight into how climate change may trigger societal collapse in the future.

Comprehensive surveys of X-risks reveal mechanisms that could cause the collapse of contemporary society. Bostrom and Ćirković (2008), Rees (2018) and Ord (2020) provide eminent scholarly treatment of the field, drawing from the academic literature. WEF (2020) and GCF (2020) produce global risk reports drawing from decision-makers and experts across intergovernmental and non-governmental organisations. These surveys establish that many historically observed mechanisms of societal collapse, including natural climate change, remain applicable as X-risks today. However, the state of existence of contemporary society has led to a different landscape in which these mechanisms apply, and to a number of unprecedented mechanisms, including anthropogenic climate change. Ehrlich and Ehrlich (2013) and Häggström (2016) note that although increased complexity, such as globalisation and technological advancement, can increase a society's resilience and adaptability, it can also increase vulnerability. For example, globalisation increases resilience to local agricultural production shocks through access to global markets; however, it also increases vulnerability through exposure to sudden reversal in connectivity, such as trade restrictions (Rivington et al. 2015). Some geoengineering technologies, for example, may enable society to mitigate and adapt to climate change; however, they may also increase vulnerability to termination shocks, where failure of the technology exposes society to sudden temperature increases (Morton 2016). In this highly interconnected landscape, 'synchronous' (Homer-Dixon et al. 2015) and 'cascading' (Buldyrev et al. 2010) failures create the potential for mechanisms and outcomes of societal collapse, once contained to a single localised civilization, to rapidly spread across multiple nations and impact humanity on a global scale.

Works by Lynas (2007), Wallace-Wells (2019) and Gowdy (2020) draw on the scientific climate change literature to explore hypothetical futures under best- to worst-case scenarios. The scenarios consider the feedbacks within the natural system that could worsen, as well as the potential for humans to mitigate, anthropogenic climate change. Shifts in average weather (e.g. temperature) and natural disasters (e.g. floods) affected by climate change could impact human mortality directly. These two effects, coupled with sea level rise due to melting of ice caps, could indirectly impact human mortality via degradation of the natural world system (e.g. land quality) and the human world system (e.g. infrastructure failures) resulting in resource and service insecurity. This insecurity could impact institutional stability, resulting in economic loss, political dysfunction and social unrest, as well as migration and conflict. The hypothetical outcomes for contemporary society against the threat of anthropogenic climate change range from dystopian (collapse) to utopian (recovery).

These futures studies identify endpoints of different causal pathways between anthropogenic climate change effects and potential impacts on the human world system, with the latter reflecting key determinants of societal collapse observed in the historical studies. Scholars have made limited in-roads to empirically investigating the top-level relationships between some of these endpoints using recent datasets. The direct links between climate change and the endpoint impacts of mortality, conflict and migration are, respectively, examined by Mora et al. (2017), Hsiang et al. (2013) and Hauer et al. (2020). The feedback between migration and conflict driven by climate change is examined by Abel et al. (2019). Burke et al. (2015), Sofuoğlu and Ay (2020) and Adger et al. (2013), respectively, examine the direct links between climate change and the endpoint impacts of economic loss, political instability and shifts in cultural norms. However, the complex bottom-level links between and surrounding these endpoints are generally ill understood (Theisen et al. 2013), and the strength of empirical evidence is poorly documented (Gemenne et al. 2014) from a systems science perspective. To the best of our knowledge, no study has empirically examined how the impacts of climate change could explicitly translate into societal collapse for contemporary society. We do not have a clear picture of climate change as a systemic risk to our globalised society; particularly at spatial scales accounting for the heterogeneity of individual identity, business governance and policymaking across nations, and international exchanges. This limits our ability to understand feedbacks, identify intervention points, develop quantitative models and inform strategies to minimise the risk of societal collapse occurring in the future (Bostrom 2013).

Given the insights from this review, we refine the aim of this paper as follows. Firstly, the empirical evidence base should specifically address contemporary society. Secondly, the CLD should be constructed at a scale and granularity that addresses the heterogenous characteristics of nations and international interactions. The refined aim of this paper is thus to identify an empirical evidence base of climate change, food insecurity and societal collapse in contemporary society and structure the evidence base with a CLD defined at global scale and national granularity.

3 Methodology

A two-stage framework, consisting of five steps, was developed to achieve the aim of this paper. For each step, below, we first introduce it generically and then describe its application to our specific analysis of the climate change, food insecurity and societal collapse causal pathway.

3.1 Stage 1: establishing an empirical evidence base of societal collapse in contemporary society

Step I deploys societal collapse proxies via a key word search to identify 'evidence points', which in this instance may be considered data points, in the form of publications that empirically examine the causal pathway of interest in contemporary society.

The determinants defined in Fig. 2 provide these societal collapse proxies to establish the new empirical evidence base in lieu of historical societal collapse events pre-dating contemporary society. The *population loss* set are straightforward to isolate, consistent to measure across nations and describe tangible consequences. The *institutional breakdown* set are relatively less so. Thus, the societal collapse proxies adopted in this study were *natural mortality* (i.e. *starvation*, with respect to food insecurity), *conflict mortality* and *emigration*; subsequent studies could use the *institutional breakdown* set. Key words were selected based on terminology of climate change, food insecurity and the societal collapse proxies. Peerreviewed journal articles were chosen as the form of evidence point in this study; subsequent studies could use other publications, such as books and reports.

The keyword search was performed in Scopus. A record of the search is contained in the Supplementary Information (A.). Approximately 3000 publications were reviewed by reading the title, abstract and main body as needed. Evidence points were selected based on satisfaction of the following criteria: the publication (a) is a peer-reviewed, English-language, journal article; (b) uses empirical, data driven methods; (c) examines the period from 1990 to present (2019), representative of contemporary society; and (d) primarily examines the causal pathway

of interest. We made an exception to (a) to include the most recent 'Limits to Growth' book (Meadows et al. 2004), which was not itself a search result but documents the World3 model that was identified in the search results. We note that (b) precluded selection of review or essay-style publications; however, we found that these were often discussed in the literature review of selected evidence points, so were, nonetheless, accounted for indirectly.

This step resulted in a new empirical evidence base consisting of 41 evidence points, which are summarised in Fig. 4.

Step II defines a custom colour-coded typology for the new empirical evidence base. This typology is used in Stage 2, to construct a final CLD (f-CLD) in a novel format showing the spread of the evidence base across the system.

In this study, we were interested in the methodological spread as this provides information on data that may be useful for future studies. Four methodological categories were identified in the new empirical evidence base. Each evidence point was classified into one of these categories and assigned a colour coding, namely: quantitative complex systems model red; statistical analysis of quantitative dataset — blue; collection/analysis of qualitative interview/survey data — green; quantitative data-led case study/scenario — yellow.

The resulting typology of the new empirical evidence base is shown in Fig. 4.

3.2 Stage 2: constructing a novel-format causal loop diagram from the empirical evidence base

Step III involves creating an individual CLD (i-CLD) for each evidence point to clearly structure the complex causal relationships examined. These i-CLDs provide the building blocks from which to construct the f-CLD in Step IV.

The process to create an i-CLD is as follows. The corresponding evidence point was examined in its entirety to identify and record key information in the form of variables (nodes), links (arrowed lines) and relationship notation (positive or negative). Key information derived from the original data-driven content, i.e. the main analysis, of the evidence point was colour coded in the i-CLD according to the typology classification established in Step II. Any relationships hypothesised but not supported by the main analysis were coloured grey. Key information derived from other content, i.e. the literature review, of the evidence point was coloured black. The scale and granularity of the i-CLD was recorded as detailed in the evidence point. This process was repeated for each evidence point in isolation until a complete set of i-CLDs was produced for the new empirical evidence base.

All 41 i-CLDs created in this study are contained in the Supplementary Information (B.). One of the i-CLDs is shown in Fig. 3 as an example.

Step IV reconciles the set of i-CLDs into a standardised format in order to construct the f-CLD of the system of interest at the desired scale and granularity.

The standardisation process has two aspects. One aspect is related to component (variables and links) definition, necessary to maximise clarity of the f-CLD while covering all information contained in the evidence base. This addresses the typical challenge of CLDs becoming dense and overcomplicated, which decreases their utility. The other aspect is related to level of aggregation, necessary to ensure the f-CLD conveys information at the intended scale and granularity. The standardisation is an iterative process, as follows.

The \sim 950 variables from the set of 41 i-CLDs were recorded on a blank worksheet for the f-CLD, without links between them. A clustering approach was used to reconcile these variables into like groups. For each group, an overarching major node was isolated and the

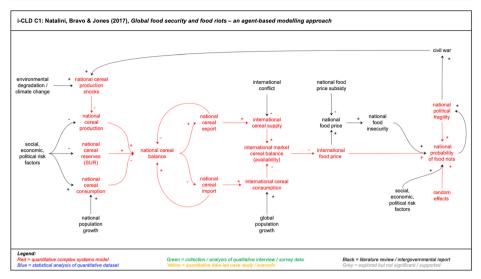


Fig. 3 One of the i-CLDs created for each of the 41 evidence points in the new empirical evidence base of climate change, food insecurity and societal collapse in contemporary society

i-CLD variables in the group were virtually deposited into a matrix for that major node. For example, *drought, sea level rise* and *crop disease* were some of the i-CLD variables clustered into an *environmental risk factors* f-CLD major node matrix. The f-CLD major nodes were defined at a level of aggregation representative of a nation. Doing so effectively scaled down any global or regional aggregation, and scaled up any sub-national or local aggregation, in the i-CLD variables. For example, *household food imports* was an i-CLD variable of local aggregation that was scaled up to national *food imports (trade)* in the f-CLD.

The ~1150 links from the set of 41 i-CLDs were reconciled into arrowed lines between the major nodes in the f-CLD. This sometimes-required interpretation of implied causality in the i-CLD relationships in order to route them across the major nodes in the f-CLD. For example, where an i-CLD showed a direct link from *international food price* to *conflict* variables, this was routed using arrowed lines from *international food price* to national *food price* to *food accessibility* to *food insecurity* and finally to *conflict* major nodes defined in the f-CLD. Where there was a discrepancy between relationship descriptions, the relationship with the most supporting i-CLDs was adopted.

The interim f-CLD produced at the end of each standardisation iteration was examined to determine whether the major node definition could be refined to maximise clarity. For example, in one iteration *water* and *land* were defined as separate major nodes, but examination determined that each had the same arrowed lines to other major nodes; therefore, another iteration was undertaken with *water* and *land* now clustered under a single *natural resources* major node in order to minimise redundant arrowed lines. This process was iterated several times until an f-CLD had been constructed at an appropriate level of detail for this study. Additionally, relevant literature reviewed in Section 2 (e.g. IPCC 2014) was cross-referenced, but not included as evidence points, to ensure comprehensive coverage of key relationships in the f-CLD.

The standard-format f-CLD, consisting of un-coloured and un-weighted components, resulting at the end of this step is contained in the Supplementary Information (C.).

Step V maps each i-CLD to the f-CLD using a weighted (line thickness) typology (colour coded) approach. This visually documents the spread of the evidence base across the system described by the f-CLD.

The process to map an i-CLD to the f-CLD is as follows. Each variable (node) of the i-CLD was assigned to its corresponding major node(s) in the f-CLD. Each link (arrowed line) of the i-CLD was assigned to a corresponding route along the arrowed lines in the f-CLD. Each time an arrowed line in the f-CLD had an i-CLD link assigned to it, an incremental weighting of one-unit line thickness in the corresponding typology colour-coding of the i-CLD link, was added to the f-CLD arrowed line. This process was repeated for each of the 41 i-CLDs until all had been mapped to the f-CLD. A record of this process for each of the 41 i-CLDs is contained in the Supplementary Information (D.).

The novel-format f-CLD, consisting of colour-coded and weighted components, resulting at the end of this final step is presented in Fig. 5.

4 Results and discussion

The new empirical evidence base and novel-format CLD of climate change, food insecurity and societal collapse in contemporary society resulting from the application of our original methodology (Section 3) are discussed in turn below.

4.1 Empirical evidence base of climate change, food insecurity and societal collapse in contemporary society

The new empirical evidence base (Section 3, Step I), along with its colour-coded typology (Section 3, Step II), is presented in Fig. 4. It consists of 41 evidence points, of which 9 examine the *natural mortality* (i.e. *starvation*, with respect to food insecurity), 20 the *conflict mortality* and 12 the *emigration* societal collapse proxy, alongside other human and natural world system factors. We discuss three key aspects of the evidence base, namely temporal and spatial distribution, data-driven method distribution and advantages of each data-driven methods, below.

The temporal scale and granularity of study varies across the evidence base; however, our methodology limited the possible scale of study to the period from 1990 to present, representative of contemporary society. Within this period, approximately half of the evidence points cover a scale of less than one decade and the other half a scale of greater than one decade. Approximately half of the evidence points conduct analyses at yearly granularity and the other half conduct analyses at granularity greater than one year, with only a few studies conducting analyses at monthly granularity. The spatial scale and granularity of study varies across the evidence base. Approximately one third of the evidence points investigate the system at a global scale, with the remaining two thirds focusing on regional or national scales, primarily in Africa as well as the Middle East and Asia. Approximately half of the evidence points analyse the causal pathway at sub-national granularity, with the other half primarily focusing on national-level granularity. This variation provided different coverage of the complex relationships within the system, which was informative for constructing our CLD.

The distribution of data-driven methods used across the evidence base is notably different for each societal collapse proxy. Evidence points for *natural mortality* mostly use collection/ analysis of interview/survey data. This is likely because the minimum daily food intake for

 Model (SDM) Statistical analysis Interview/survey Model (ABM) Statistical analysis 	Meadows et al, 2004 Agboola, 2017 Lee et al, 2016 Campbell et al, 2009 Fieldderjohann et al, 2016 Gundersen et al, 2016 Gundersen et al, 2016 Walker et al, 2019 Natalini et al, 2017 Arezki & Brueckner, 2014 Bellemare, 2014 Berzareva & Lee, 2013 Caruso et al, 2017 Koren & Bagozzi, 2016 Hendrix & Haggard, 2015 Jones et al, 2017 Koren & Bagozzi, 2016 Naystadt et al, 2014 Natalini et al, 2015 Newman, 2018 Pinstrup-Andersen & Shimokawa, 2008 Raleigh et al, 2014	Global Global (114 countries) Global (95 countries) Indonesia India Canada Nigeria, Ethiopia India USA Global (213 countries) Global (213 countries) Global (213 countries) Global (213 countries) Indonesia Africa, Asia (49 countries, 55 cities) Africa (40 countries) Global (416 countries) Global (416 countries) Global (79 countries) Global (79 countries) Global (79 countries)	Global National National Sub-national Sub-national Sub-national Sub-national National National Sub-national Sub-national Sub-national Sub-national National National National National National National National National National National National	1900-2004, fwd sim, 1995-2009 2001-2011 2000-2003 2005-2003 2005-2013 2005-2013, kwd sim, 1970-2007 1990-2011 1990-2011 1991-2011 1991-2010 1991-2011 1991-2010	Yearly Yearly Period Period Yearly Period Period
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	Raleigh et al, 2015		National	1980-2005	Yearly
		Africa (24 countries, 113 markets)	Sub-national	1997-2010	Monthly
		Africa (40 countries)	National	1990-2012	Monthly
			National		
	van Weezel, 2016	Africa (45 countries)		1990-2011	Monthly
	Weinberg & Bakker, 2015	Global (71 countries)	National	1972-2007	Yearly
17 18 19 19 20 11 Emigration Model (ABM)	Wischnath & Buhaug, 2014	India	Sub-national	1980-2011	Yearly
	Legwegoh et al, 2015	Cameroon	Sub-national	2008	Period
	Lunt et al, 2016	Global	National	2016, projection	Period
	Sternberg, 2012	Global (Egypt, China)	National	2010-2011	Period
. ,					Period, yearly
22 Interview/survey 33 44 55 56 57 59 50 51 11 Data-led/scenario 12					Period
	Milan & Ruano, 2014		Sub-national		Period
	Murali & Afifi, 2014		Sub-national		Period
	Panter-Brick & Eggerman, 1997	Nepal	Sub-national	1996 (assumed)	Period
	Rademacher-Schulz et al, 2014	Northern Ghana	Sub-national	2011	Period
	Warner & Afifi, 2014	Global (S/SEAsia, SSAfrica, LatAm)	Sub-national	2011-2012	Period
	Doos, 1994	Global	Global, national	1990-94, projection	Yearly
	Puma et al, 2018	Global	National	2011-2018	Yearly
	Model (ABM) Interview/survey Data-led/scenario	Interview/survey Afifi et al, 2014 Etzold et al, 2014 Jacobson et al, 2019 Milan & Ho 2014 Murali & Afifi, 2014 Panter-Brick & Eggerman, 1997 Rademacher-Schutz et al, 2014 Warner & Afifi, 2014 Data-led/scenario Doos, 1994 Puma et al, 2018	Interview/survey Affi et al, 2014 Tanzania Etzold et al, 2014 Bangladesh Jacobson et al, 2019 Cambodia Mian & Ho 2014 Peru Mian & Ruano, 2014 Guaternala Murali & Affi, 2014 India Panter-Brick & Eggerman, 1997 Nepal Rademacher-Schulz et al, 2014 Northern Ghana Warrer & Affit, 2014 Giobal Otata-led/scenario Doos, 1994 Giobal	Interview/survey Affi et al, 2014 Tanzania Sub-national Etcold et al, 2014 Bangladesh Sub-national Jacobson et al, 2019 Cambodia Sub-national Mian & Ho 2014 Peru Sub-national Mian & Rouno, 2014 Guatemala Sub-national Murali & Affi, 2014 India Sub-national Panter-Brick & Eggerman, 1997 Nepal Sub-national Panter-Brick & Eggerman, 1997 Northern Ghana Sub-national Warmer & Affi, 2014 Northern Ghana Sub-national Data-led/scenario Doos, 1994 Global Global Puma et al, 2018 Global National	Interview/survey Affif et al, 2014 Tanzania Sub-national 2012 Etcold et al, 2014 Bangladesh Sub-national 2011 2011 2016 Jacobson et al, 2019 Cambodia Sub-national 2016 2011 2011 Mian & Ho 2014 Peru Sub-national 2011 2011 2011 Mian & Ruano, 2014 Guatemala Sub-national 2011 2011 2011 Murait & Affit, 2014 India Sub-national 2011 2011 2011 Panter-Brick & Eggerman, 1997 Nepal Sub-national 2011 2011 Rademacher-Schulz et al, 2014 Northern Ghana Sub-national 2011-2012 Data-led/scenario Doos, 1994 Global (S/SEAsia, SSAfrica, LatAm) Sub-national 2011-2012

Fig. 4 Summary and custom colour-coded typology of the new empirical evidence base of climate change, food insecurity and societal collapse in contemporary society. A full reference list is contained in Supplementary Information (E)

human survival is well established (FAO 2004); as such, statistical analysis of food and mortality data sets would not yield significantly new insights into thresholds whereas interviews/surveys can provide insight into an individual's circumstances influencing this relationship. Evidence points for *conflict mortality* mostly use statistical analysis of existing datasets. This likely reflects the interest in rigorously curated conflict datasets, such as UCDP/ PRIO (2019), across the conflict and peace fields. Evidence points for *emigration* mostly use collection/analysis of interview/survey data, likely because this provides nuanced insight into an individual's decision to migrate. It may also be due to data availability and quality challenges that limit quantitative statistical analyses, which are being addressed by groups such as IOM GMDAC (2019). Amongst these data challenges, it is important to recognise the issue of reconciling different types of voluntary and forced migration with causal drivers, given the complex social, economic and political factors at play; this challenge similarly

applies to the other societal collapse proxies but is particularly noted in the migration studies. We observe from these studies that a food insecurity threshold for *natural mortality* is well established but thresholds for *conflict mortality* and *emigration* are not. Indeed, distinguishing causal drivers within datasets and defining quantitative thresholds for these determinants remains a 'grand challenge' (Kintigh et al. 2014).

Each data-driven method offers different advantages. The complex systems models each describe 'chunks' of the system at different scale and granularity. The models provide mathematical definition, are calibrated to real-world data and enable quantitative simulation of key relationships in the system. The statistical analyses quantitatively examine relationships between a dependent variable and one or more independent variables within the system, which can be used as a mathematical basis for extending modelling capabilities. The collection/ analysis of interview/survey data provides insight into qualitative aspects of human perspective and decision-making that quantitative data sets cannot provide directly. The data-led case study/scenarios combine quantitative data with qualitative expert interpretation to better understand global trends and forecasts. These latter two methods can also be used to inform the development of modelling capabilities, the scenarios analysed by such models and their application in decision-making processes. Collectively, these different data-driven methods can yield useful insights into the nuances of relationships in the system of interest.

4.2 Causal loop diagram of the climate change, food insecurity and societal collapse in contemporary society at global scale and national granularity

The main result of this paper is the CLD (the f-CLD from Section 3, Step V), presented in Fig. 5. It structures the relationships between climate change, food insecurity and societal collapse as described in our new empirical evidence base (presented in Fig. 4 and discussed in Section 4.1.). We discuss three key aspects of the CLD, namely insights related to the spread of empirical evidence, the qualitative complex system depicted, and quantitative complex system modelling, below, alongside consideration of well-established benefits and limitations of CLDs.

Our CLD is presented in a novel format that documents the spread of our empirical evidence base. We use line thickness and colour, respectively, to depict the density and type of the data-driven methods used by the empirical evidence points to analyse a given link between two variables.

Doing this aids comprehension of where existing work has been focused with respect to the climate change, food insecurity and societal collapse causal pathway. It may also help with the identification of gaps in existing analyses. For example, we can see that the link between *food insecurity* and *conflict* has been investigated mostly by evidence points using statistical analyses (blue), whereas the links between *food insecurity* and *migration*, and *food insecurity* and *natural mortality*, have been investigated mostly by evidence points using interviews/ surveys (green). This hints that it may be useful to investigate the former using quantitative statistics, and the latter using qualitative interviews/surveys, to gain further insights offered by the different data-driven methods as described in Section 4.1.

It is important to recognise that our CLD may show negligible density for important links or even be missing important variables and/or links, either because they have not yet been studied or because our key word search failed to identify evidence points that have studied them. For example, our study focused on the climate change, food insecurity and societal collapse causal pathway, so the density of our empirical evidence is concentrated along links central to this pathway; whereas, the links between peripheral variables in the system, such as between *fertility* and *births*, show a lower

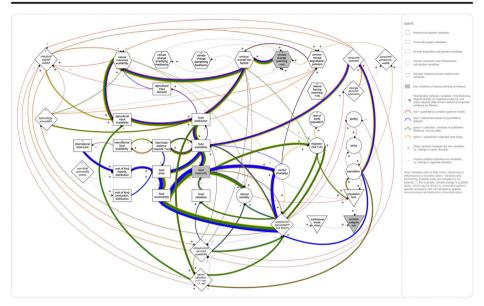


Fig. 5 Causal loop diagram of climate change, food insecurity and societal collapse in contemporary society at global scale and national granularity. Variables are depicted as nodes in five different shapes, indicating different sub-systems. Links between variables are depicted as arrowed lines, indicating the direction of the relationship. Each link has a positive (+) or negative (-) notation, indicating that the two variables change in the same direction or opposite direction, respectively. The density and type of data-driven method of the empirical evidence base, from which the causal loop diagram was constructed, are depicted by line thickness and colour, respectively

density of empirical evidence. Similarly, our use of the *population loss* set of societal collapse proxies means that the evidence base details *natural mortality, conflict mortality* and *emigration*; whereas, the *institutional breakdown* set are not detailed. In considering this issue, our methodology attempted to maximise the rigour and transparency of our study by documenting the spread of our empirical evidence base to help make the reader aware of exactly how much and what type of evidence was supporting the CLD presented here.

Further, we can see that while empirical studies have linked *climate change* via *food insecurity* to our societal collapse proxies of *natural mortality*, *conflict mortality* and *emigration*, we found no empirical studies linking these proxies to the explicit term of *societal collapse*. This was expected given the motivation of this study (Section 1) and is due to the fact that there are no contemporary events of societal collapse, under the same definition as those in the historical studies pre-dating contemporary society, that enable these links to be empirically studied (Beard et al. 2020).

Having considered the spread of empirical evidence, we now consider the complex system documented. A key benefit of CLDs is that they simply present a myriad of information in a single diagram; in doing so, CLDs enable comprehension of the structure and behaviour of complex systems, including feedbacks, intervention points and far-reaching interdependencies (Sterman 2011). Our CLD visually depicts a system of 39 variables, 105 links and 32,000 feedback loops,¹ integrating information from different fields including climate science, food security, conflict, migration and health research.

¹ Calculated using Vensim ® PLE software, Ventana Systems Inc.

Walking through the CLD at a high-level, we can see how *population* growth and *lifestyle* emissions, influenced by institutional/demographic factors (e.g. emission reduction incentives), combine to directly drive *climate change*. Similarly, they indirectly drive *climate* change via consumer demand on food production, which produces emissions directly (e.g. ruminant livestock) and indirectly via industrial capital/output (e.g. processing factories). The environmental risk factors (e.g. extreme weather events) of climate change may cause losses of food production either directly (e.g. plant disease) or indirectly via agricultural input availability (e.g. loss of water source for irrigation). A country's food availability is influenced by domestic food production and international food trade. Food accessibility is influenced by its food price, which responds to domestic (e.g. cost of food production and distribution) and international (e.g. international food price) markets, and institutional/demographic factors (e.g. food subsidies). Food utilization is influenced by infrastructure/services (e.g. education) and *institutional/demographic factors* (e.g. cultural traditions). *Food insecurity* is underpinned by these three pillars of food availability, food accessibility and food utilization. For a given country, food insecurity can drive natural mortality (i.e. starvation), conflict and migration, contributing to *population loss*, as well as economic shocks and socio-political instability, contributing to *institutional breakdown*, which exacerbates the risk of *societal collapse*.

Beyond a given country suffering increased *natural mortality*, famines (i.e. *food insecurity*) can place pressure on international humanitarian efforts (i.e. *institutional risk factors*). Conflict may occur domestically or internationally and can feedback to exacerbate *food insecurity* and institutional fragility (i.e. *institutional risk factors*). Potential mass *emigration* can increase pressure on *food availability, natural resources* and *infrastructure/services* in the destination nation, which can lead to socio-cultural tensions (i.e. *institutional risk factors*) that fuel *conflict. Food insecurity* can also directly contribute to *institutional risk factors* such as social unrest, political instability and economic inequality, which increase the risk of *societal collapse* due to *institutional breakdown*, that may also cascade internationally. While already fragile states are expected to be hit the worst directly, these insights reveal the indirect ramifications of climate change on our globalised society (Kemp 2020), with serious consequences for humanity's 'existential security' (Sears 2020).

While some of these relationships may appear obvious, it is the act of bringing this information, which may otherwise be siloed and thus preventing consideration of the full story, together in one place that is of value (Sterman 2011). In doing so, our CLD attempts to provide readers with the opportunity to explore the climate change, food insecurity and societal collapse causal pathway, consider worst-case scenarios that we want to avoid, develop transformative narratives of "where we want to go" and think about interventions that may help us attain this desired future (Hinkel et al. 2020).

It is important to appreciate that CLDs are only as good as their information inputs; our CLD documents relationships based on information portrayed in our empirical evidence base as well as our interpretation of that information. As such, there exist challenges and limitations (Cernev and Fenner 2020). For instance, CLDs may mask variability of relationships in different contexts and locations, because they can only depict a single scale and granularity. The portrayal of explicit causality between variables in a CLD is a challenge as this can often work in both directions rather than one. CLDs can often become either too complicated or too simplified, which undermines their usefulness. In considering each of these issues, our original methodology attempted to maximise the rigour and transparency of our study by first documenting the information in each evidence point with an i-CLD and then consistently applying, and recording, the iterative process of reconciling the variables and links from each i-

CLD to construct the f-CLD at the selected global scale and national granularity. In doing so, we sought to enable the reader to be aware of the nuances of the different scales and granularity of information underpinning our CLD, as well as our process of carefully reconciling causality, over 950 variables to 39 variables and 1150 links to 105 links to maximise the information conveyed while balancing readability.

It is also important to note that, due to their qualitative and static nature, CLDs do not enable us to comprehend the dynamics of the system, including nonlinear and emergent behaviour, non-intuitive quantitative results and time delays (Sterman 2011). Complex systems models, although with their own challenges and limitations (Sterman 2002), provide the opportunity to quantitatively analyse the dynamics of a system and gain insights into the potentially far-reaching impacts of our decisions (Sterman 2000). However, complex systems models that explicitly examine societal collapse in contemporary society are underdeveloped. The World3 system dynamics model (Meadows et al. 2004) — an evidence point in this study (refer Supplementary Information D. for respective i-CLD) — is the eminent model of relevance, with only a limited number of studies building on it (e.g. Ansell and Cayzer 2018). World3 examines the potential for 'overshoot-and-collapse' given population and industrial growth within the finite carrying capacity of the natural world system, implicitly accounting for climate change and explicitly accounting for food availability.

The information contained in our CLD and empirical evidence base may be useful in identifying and informing opportunities to improve these existing complex systems modelling capabilities for climate change, food insecurity and societal collapse scenarios. For example, our CLD highlights important factors at global scale and national granularity that World3 does not incorporate because it is defined at global scale and granularity (Meadows et al. 2004). World3 does not distinguish heterogenous characteristics of nations, such as distribution of population or geographic endowment of natural resources. It also does not account for international interactions, such as food trade, conflict and migration. Relatedly, World3 evaluates societal collapse only by natural mortality (defined by *food availability*, age and pollution) and does not include the other two *population loss* secondary determinants, as noted in the previous sentence, nor the three institutional breakdown secondary determinants. While our empirical evidence base may provide useful direction to datasets, it is important to note that quantitatively defining these relationships, particularly thresholds as discussed in Section 4.1., remains a key challenge of developing complex systems models. Nonetheless given that individuals associate with national identity, business governance and policymaking are concentrated at national level, and international interactions underpin the functioning of contemporary society it could be valuable to model societal collapse risk profiles of different nations to inform the prioritisation and development of intervention strategies.

5 Conclusions and future work

This paper identified an empirical evidence base of climate change, food insecurity and societal collapse in contemporary society and structured the evidence base using a novel-format CLD defined at global scale and national granularity.

Two types of future work could extend from the results of this paper. Identification of gaps in the spread of evidence across the CLD may guide future data-driven efforts to examine these causal relationships and define thresholds. The CLD and evidence base may be used to develop quantitative modelling capabilities, particularly by transforming the structure of World3 to account for heterogenous national characteristics and international interactions. Three types of future work could

extend from the methodology and literature synthesis. The causal pathway examined in this paper could be further detailed by re-applying the methodology using the *institutional breakdown* set of societal collapse proxies instead of the *population loss* set. The methodology, using either set of societal collapse proxies, could be applied to detail other causal pathways between climate change and societal collapse. The methodology, excluding the contemporary time period limitation, could be applied to document the information in the historical studies identified in the literature review. Similarly, the methodology could be applied to construct CLDs at different scales and granularities.

It is hoped that this paper has contributed to developing our understanding of the causal pathways through which climate change poses an existential risk to humanity and facilitates opportunities for future work.

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Data availability Not applicable. All relevant information is contained in the Manuscript or Electronic Supplementary Material.

Compliance with ethical standards

Conflict of interest The authors declare that they have no known conflict of interest, including competing affiliations, funding sources, financial or personal relationships that could be perceived, either directly or indirectly, as potential sources of bias with respect to this paper.

Code availability Not applicable.

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