



# Food trade disruption after global catastrophes

Florian Ulrich Jehn<sup>1</sup>, Łukasz G. Gajewski<sup>1</sup>, Johanna Hedlund<sup>2</sup>, Constantin W. Arnscheidt<sup>3</sup>, Lili Xia<sup>4</sup>, Nico Wunderling<sup>5,6,7</sup>, David Denkenberger<sup>1,8</sup>

<sup>1</sup>Alliance to Feed the Earth in Disasters (ALLFED), Wilmington, DE, USA

<sup>2</sup>Stockholm Environment Institute, Stockholm, Sweden

<sup>3</sup>Centre for the Study of Existential Risk, University of Cambridge, Cambridge, UK

<sup>4</sup>Department of Environmental Sciences, Rutgers University, NJ, USA

<sup>5</sup>Center for Critical Computational Studies (C<sup>3</sup>S), Goethe University Frankfurt, Frankfurt am Main, Germany

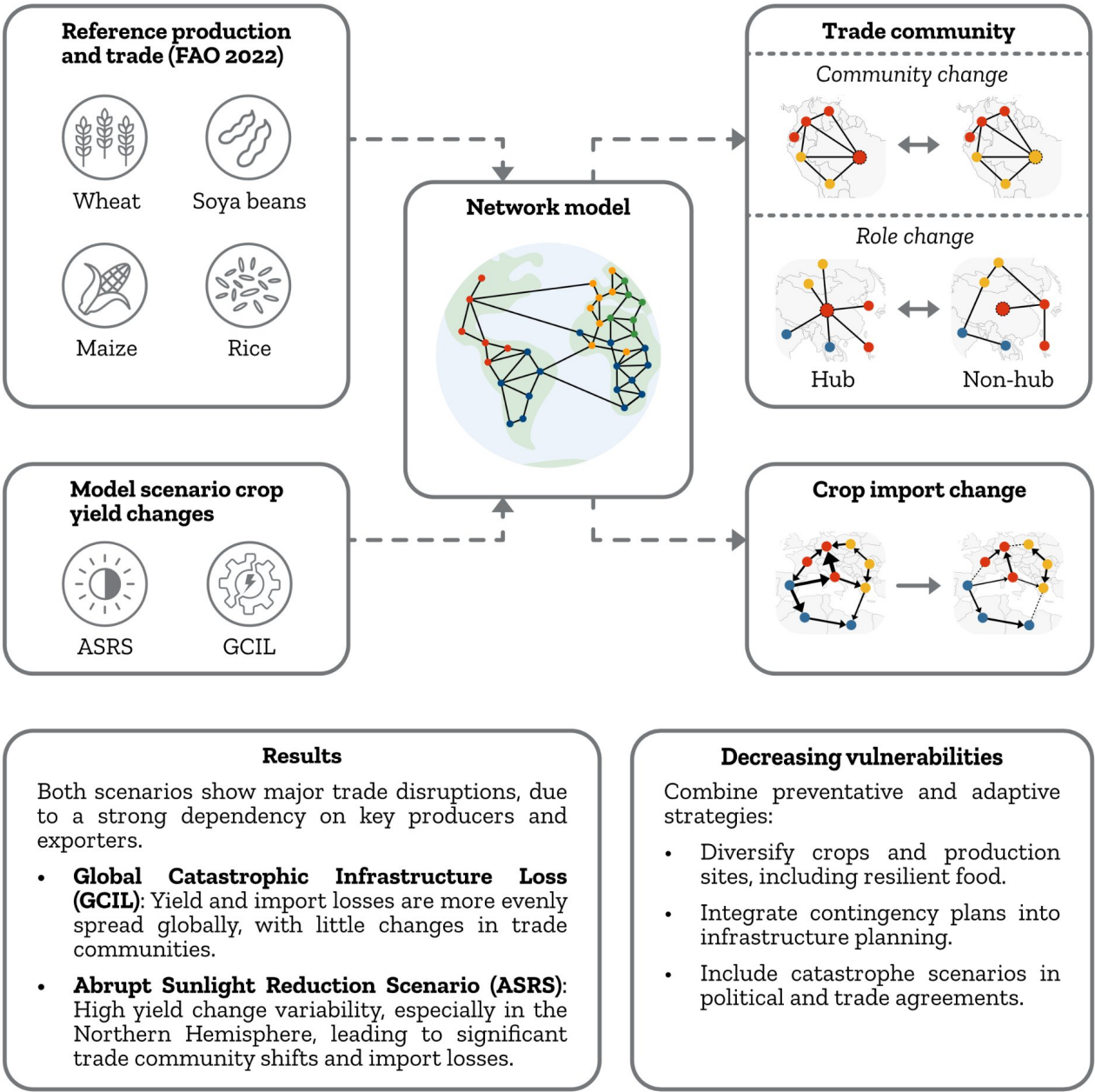
<sup>6</sup>Earth Resilience Science Unit, Potsdam Institute for Climate Impact Research (PIK), Member of the Leibniz Association, Potsdam, Germany

<sup>7</sup>High Meadows Environmental Institute, Princeton University, Princeton, NJ, USA

<sup>8</sup>Department of Mechanical Engineering, University of Canterbury, Christchurch, New Zealand

*Correspondence to:* Florian Ulrich Jehn (florian@allfed.info)

**Abstract.** The global food trade system is resilient to minor disruptions but vulnerable to major ones. Major shocks can arise from global catastrophic risks, such as abrupt sunlight reduction scenarios (e.g., nuclear war) or global catastrophic infrastructure loss (e.g., due to severe geomagnetic storms or a global pandemic). We use a network model to examine how these two scenarios could impact global food trade, focusing on wheat, maize, soybeans, and rice, accounting for about 60% of global calorie intake. Our findings indicate that an abrupt sunlight reduction scenario, with soot emissions equivalent to a major nuclear war between India and Pakistan (37 Tg), could severely disrupt trade, causing most countries to lose the vast majority of their food imports (50-100 % decrease), primarily due to the main exporting countries being heavily affected. Global catastrophic infrastructure loss with a comparable impact on yields as the abrupt sunlight reduction has a more homogeneous distribution of yield declines, resulting in most countries losing up to half of their food imports (25-50 % decrease). Thus, our analysis shows that both scenarios could significantly impact the food trade. However, the abrupt sunlight reduction scenario is likely more disruptive than global catastrophic infrastructure loss regarding the effects of yield reductions on food trade. This study underscores the vulnerabilities of the global food trade network to catastrophic risks and the need for enhanced preparedness.



## 31 1 Introduction

32 Humanity receives much of its food via the global trade network (D’Odorico et al., 2014; Janssens et al., 2020). However, with  
33 such interconnectedness comes the potential for large-scale systemic risk (Bernard de Raymond et al., 2021), where local failures  
34 can have cascading effects throughout the broader system. A significant component of the system’s vulnerability is its lack of  
35 diversity on all levels, ranging from seed varieties to the number of companies trading food and few but dominant exporters  
36 (Clapp, 2023; Hamilton et al., 2020; Nyström et al., 2019). Global trade has been described as "robust, yet fragile," capable of  
37 weathering more minor shocks but increasingly vulnerable to major ones (Foti et al., 2013; Ma et al., 2023; Wang et al., 2023).  
38 Such major shocks could come in the form of “tipping points”, and involve cascading interactions with other processes such as  
39 conflict and migration in a globally interconnected world (Centeno et al., 2023; Spaizer et al., 2023) or by multiple separate  
40 shocks happening at once (Baum et al., 2024). In this context, the World Economic Forum's Global Risk Report 2023 highlights  
41 food supply crises as one of the most severe risks in the coming years and decades (World Economic Forum, 2023).

42  
43 A key vulnerability in the food trade network lies in the potential disruption of the biggest food exporters (Clapp, 2023; Puma et  
44 al., 2015), and this vulnerability appears to be increasing over time (Ji et al., 2024; Ma et al., 2023). Currently, only five  
45 countries (China, United States, India, Russia and Brazil) are responsible for producing the majority of wheat, maize, rice and  
46 soya beans (Caparas et al., 2021), and these producers are especially vulnerable to disruptions of agricultural inputs (Ahvo et al.,  
47 2023). A stop of trade by, e.g., the United States could trigger cascading failures (Goldin and Vogel, 2010; Helbing, 2013; Ma et  
48 al., 2023), plausibly endangering the entire system. One possible reason for large yield shocks is synchronised multiple  
49 breadbasket failure, which means the simultaneous collapse of multiple major agricultural regions (Anderson et al., 2023; Gaupp  
50 et al., 2020; Kornhuber et al., 2023). Beyond this, there are various global catastrophic risk (GCR) scenarios which could involve  
51 large-scale food system disruption.

52  
53 GCR has been defined as the risk of “serious damage to human well-being on a global scale” (Bostrom and Cirkovic, 2008), and  
54 could occur due to a wide range of possible hazards. Here, we consider two specific scenarios particularly relevant to the food  
55 system. The first is global catastrophic infrastructure loss (GCIL), which could be triggered by High Altitude Electromagnetic  
56 Pulses (HEMPs) (Cooper and Sovacool, 2011; Wilson, 2008), geomagnetic storms (Baum, 2023; Cliver et al., 2022; Isobe et al.,  
57 2022), globally coordinated cyber attacks (Ogie, 2017), and extreme pandemics causing people to be unable or unwilling to work  
58 in critical industries (Denkenberger et al., 2021). These events, disrupting the electrical grid on a global scale and thus the  
59 production of inputs for the food system, like fertilisers, pesticides or fuel, could lead to a substantial reduction in global food  
60 yields (Moersdorf et al., 2024) and would thus further influence food trade.

61  
62 The second is that of abrupt sunlight reduction scenarios (ASRSs), which could result from nuclear war (Coupe et al., 2019;  
63 Toon et al., 2008), asteroid/comet/meteor (bolide) impacts (Chapman and Morrison, 1994; Tabor et al., 2020), or large volcanic  
64 eruptions (Rampino, 2002; Rougier et al., 2018). Such events could inject aerosol particles into the upper atmosphere, causing a  
65 significant drop in temperature and disrupting global agriculture (Coupe et al., 2019; White, 2013). A recent analysis of Xia et al.  
66 (2022) suggests that a nuclear war between Russia and the United States could lead to global yield reductions of up to 90% in the  
67 worst year following the war. Even a smaller nuclear war could disrupt global trade due to a massive spike in food prices  
68 (Hochman et al., 2022).

69  
70 The likelihood of large yield shocks may be substantial. For example, Rivington et al. (2015) estimate an 80% likelihood of a  
71 10% or greater global yield shock due to multiple breadbasket failure within this century. This probability combined with the  
72 probability of the abovementioned catastrophes, based on current estimates and preparations, moves to over 90% for this century

at least one of them happening (Barrett et al., 2013; Denkenberger et al., 2021, 2022; Karger et al., 2023), with the majority of the probability mass coming from multiple breadbasket failures. While these numbers are highly uncertain, they highlight that there is the need to understand better what might happen if yield shocks on such a scale occur.

While the impacts of climate change and extreme events on trade have been studied more in recent years (Hedlund et al., 2022; Thang, 2024), only limited research has been conducted regarding the effects of GCIL and ASRS on food production and trade. The research that does exist assumes that trade will continue as it is now or cease completely (Hochman et al., 2022; Rivers et al., 2024a; Xia et al., 2022). These simplifications reduce the enormous complexity of how our food system might react to global catastrophic risks. While some preliminary economic research on smaller nuclear conflicts has been conducted (Hochman et al., 2022), broader insight, especially into the consequences of a wider range of scenarios, is needed .

For an initial assessment of how global trade might evolve after such global catastrophes, we study the shifts of trade communities and trade flows caused by GCIL and ASRS in a global food trade network model (Hedlund et al., 2022). In this context, trade communities refer to groups of countries that trade extensively with one another. Understanding them and their changes allows a more targeted assessment of the disruptions caused by changes in yield. The model is intentionally simple, focusing on the direct effects of yield changes on trade without considering second-order economic aspects. Our initial analysis can serve as a foundation for future, more detailed economic assessments, while the model itself offers policymakers and scientists a practical tool to analyse the direct effects of food production shocks on global trade. Such assessments are important because they advance our understanding of how global catastrophes impact food trade, revealing the different implications of various shocks to the system. By modelling these shocks under different scenarios, we can better understand and predict changes in the global food trade system after major disruptions.

## **2 Methods**

### **2.1 Model setup**

The model we used was introduced by Hedlund et al., (2022); for the present analysis, we have re-implemented it in Python (Jehn and Gajewski, 2024) (<https://github.com/allfed/pytradeshifts>). The global trade network is described as a weighted directed graph with the countries as nodes and trade volumes between two countries as the weight of the edges connecting the nodes. In the model, we accounted for re-exports to represent point-of-origin-to-point-of-destination trade movements, meaning that the resulting data only contain the direct trade between countries without intermediaries (more information about this is in the Section 2.2 of the supplement and in Hedlund et al. (2022)). The model determines post-catastrophe trade by applying country-specific yield changes directly to export volumes. For example, if a country experiences a 30% yield reduction, all its exports decrease proportionally – by 30%. We do not introduce new connections, though trade connections can become 0 if the yield is reduced by 100%. Compared to the original model, we have added the option to remove countries from the analysis to simulate an overall inability to take part in trade (e.g. due to destruction after a nuclear war). Other additional functionality is described in the Supplement (Section 1).

To detect the communities in the trade network, we used the Louvain algorithm (Blondel et al., 2008), as implemented in NetworkX (Hagberg et al., 2008). It assigns every country a trade community, i.e., a group of other countries with which said country has the closest trade ties. As the Louvain algorithm is not deterministic, our model can be provided with a random seed parameter to ensure the reproducibility of the results.

112 The Louvain algorithm identifies communities by optimizing modularity, which measures the density of connections within  
113 communities versus connections between communities. The algorithm works iteratively:

- 114 1. It assigns each country to its own community
- 115 2. For each country, it evaluates whether moving it to a neighbor's community would increase modularity
- 116 3. After all possible improvements, it aggregates each community into a single node
- 117 4. It repeats the process until modularity cannot be further improved

118 This approach allows us to detect natural trading blocs based on connection patterns without imposing geographical constraints.

## 119 **2.2 Production and trade data**

120 The Food and Agriculture Organization of the United Nations (FAO) supplies annual data on crop production and bilateral trade  
121 for agricultural commodities. Our study utilised the most recent data available (2022), adjusting for re-exports and relies on crop  
122 production and trade matrix information in tonnes.

123 While research suggests a notable 'stickiness' in the trading system (Reis et al., 2020) and that countries tend to remain in the  
124 same trade communities for long periods (Ma et al., 2023), there can still be considerable changes over time, especially after  
125 major disrupting events like COVID-19 (Clapp and Moseley, 2020) or the Russian invasion of Ukraine (Jagtap et al., 2022;  
126 Zhang et al., 2024). We, therefore, used the most recent data (2022) to most accurately represent the current global food trade  
127 network. Our analysis focuses on wheat, rice, soya beans and maize. We used primary commodity data for wheat, maize and  
128 soya beans, and for rice, given that paddy rice is predominantly traded in processed forms, we used the milled equivalent in the  
129 FAO data. We focus on those crops because they are the most important staple crops, accounting for roughly two-thirds of  
130 calories and proteins consumed globally (D'Odorico et al., 2014).

131 We excluded bilateral trade flows falling below the 75th percentile in trade volume to concentrate on the main trade movements,  
132 following Hedlund et al. (2022). This maintained the majority of countries in the network. However, the results are robust across  
133 a wide range of percentile cut-offs, as trade is dominated by a small number of large exchanges (Figure S1).

## 134 **2.3 The impact of global catastrophic risk scenarios on yields**

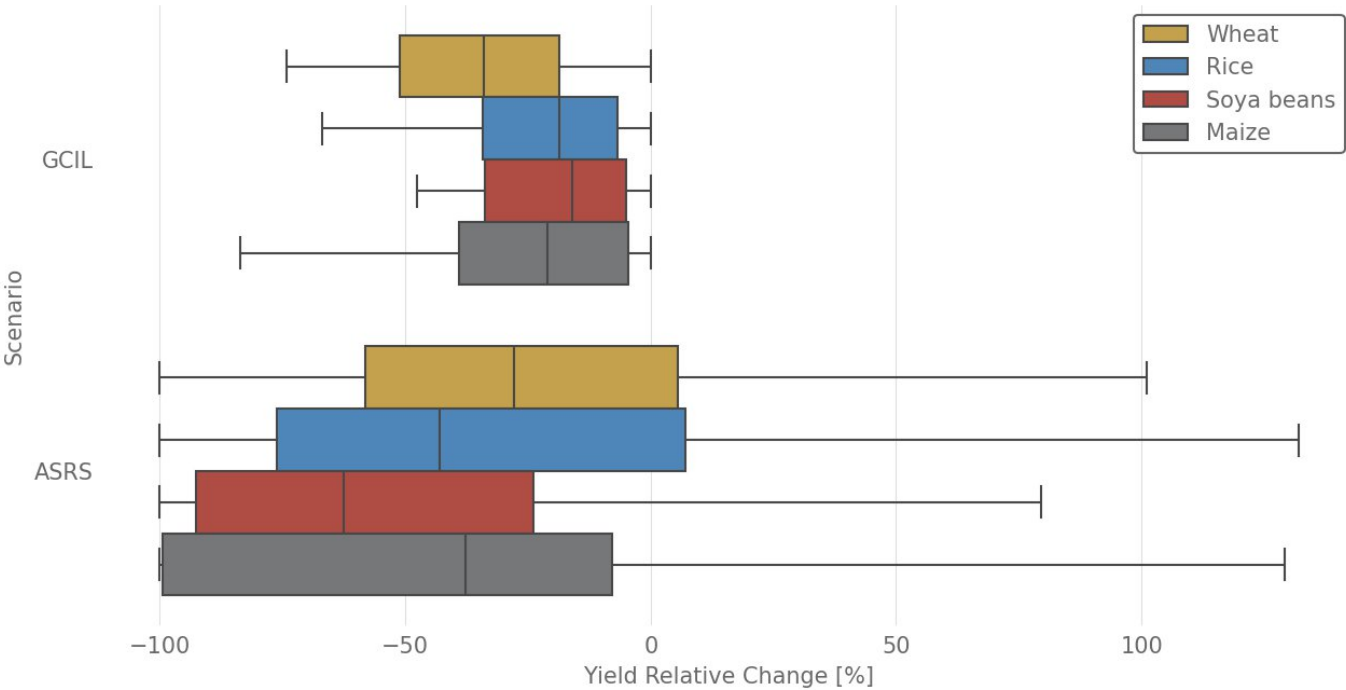
135 We focus on two main GCR scenarios: GCIL and ASRS (see introduction). We obtained yield losses for GCIL scenarios from  
136 Moersdorf et al. (2024). Moersdorf et. al (2024) assumed that if a GCIL happens, this will result in a global stop in the  
137 production of agricultural inputs like fuel, pesticides and fertilisers. Based on this they split their simulations into two phases.  
138 Phase 1 is the first year after GCIL with some stocks for fuel, pesticides and fertilisers remaining, while phase 2 simulates all  
139 following years, where all stocks are depleted. For our analysis, we used the phase 2 data to focus on the lowest yields. Since it is  
140 only available on a global (with a 5 arcmin resolution) and continental scale, we averaged the yield losses from global data for all  
141 points in each country. The resulting mean values of yield reduction differ slightly from the ones stated in Moersdorf et al.  
142 (2024) because: 1) Moersdorf et al., assigned weights using pre-catastrophe productivity, but as the nuclear war data is not  
143 productivity weighted, we used Moersdorf et al's unweighted data to ensure comparability between the two scenarios. The wider  
144 yield change distribution under ASRS compared to GCIL thus reflects genuine scenario differences rather than methodological  
145 artifacts. 2) In our model, the connections between countries are based on the actual amount traded (corrected for re-exports).  
146 Weighting the yield changes by their productivity would thus skew the results. Also, we aggregate on country level first instead  
147 of taking a global average. The scenario by Moersdorf et al. likely would have wide ranging consequences for society beyond  
148 yield impacts, as it assumes a disruption of the industrial base. These further disruptions are not modelled here.

149

150 For ASRSs we used the country-level nuclear war crop modelling data from Xia et al., (2022). We used nuclear war as a proxy  
151 for all ASRSs because nuclear war has the best climate model data available (Coupe et al., 2019), and the global impact on  
152 climate is possibly similar across different ASRS scenarios with similar magnitude. We used data for the third year after the  
153 nuclear war, as this represents the year with the lowest yields. To make the scenario more comparable with the GCIL scenario,  
154 we used the 37 teragram (Tg) scenario from Xia et al. (2022) as the main comparison. This is meant to simulate a nuclear war  
155 between India and Pakistan with 250 nuclear weapons of 100 kt explosive yield each and would thus equal a total of 25  
156 megatons of TNT.. In this scenario, some of the smaller and hotter countries experience increases in yield due to a better climate,  
157 and the climate model used with a horizontal resolution of 2 degrees cannot resolve such small countries correctly. Thus, we  
158 limit this effect to a maximum value compared to current yields to avoid unrealistically high values (Wheat: 100 %, Rice: 132 %,   
159 Soya Beans: 79 %, Maize: 129 %). Since more accurate crop growing models are not available for nuclear war, we determine  
160 this upper limit as the  $Q3+1.5(Q3-Q1)$ , where Q1 and Q3 are the 1st and 3rd quartile respectively (Tukey, 1977), of the data  
161 presented in Xia et al. (2022). Xia et al. (2022) did only model spring wheat. We are assuming here that spring wheat can be used  
162 as a proxy for wheat in general.

163

164 The ASRS with 37 Tg soot emissions has a median wheat yield decline similar to GCIL (Figure 1). Soya beans, maize and rice  
165 have more dissimilar ranges (Figure 1). This makes wheat the most comparable crop across the two scenarios, while also being  
166 the most traded and, therefore, our main focus; however, we also discuss the other crops and provide the figures for them in the  
167 supplement.



168

169 **Figure 1: Relative yield change (%) in all affected countries (combined) for both the global catastrophic infrastructure loss (GCIL) and**  
170 **the abrupt sunlight reduction scenario (ASRS), by crop (colour). The values for GCIL yield changes are taken from Moersdorf et al.**  
171 **(2024), and those for ASRS yield changes from Xia et al. (2022) (see Section 2.3 for details). The boxplot displays data distribution**  
172 **using five key summary points: the minimum, first quartile, median, third quartile, and maximum. The box spans from the first to the**  
173 **third quartile, with a line at the median. Whiskers extend to the smallest and largest values within 1.5 times the interquartile range**  
174 **from the quartiles. Outliers are circles beyond the whiskers. This is the same for all boxplots shown in this article.**

## 2.4 Trade communities before and after global catastrophes

The model allows a qualitative analysis of the changes by comparing the trade communities before and after the catastrophic event. To allow for a more quantitative comparison as well, we used a variety of measures (described below and in supplement section 1 and 2) for changes in trade communities alongside the overall complexity and robustness of the resulting trade networks.

### 2.4.1 Change

#### Jaccard distance

To assess how much the trade communities of all countries have changed before and after global catastrophes we used the Jaccard distance. This measure allows us to compare how similar/different two trade communities are. It finds the percentage of common countries between trade communities divided by the total number of elements between them. The Jaccard *similarity* (also called Jaccard index) is typically defined as the size of the intersection of two sets divided by the size of the union of these sets, and has a range from zero to one (Jaccard, 1901). The Jaccard distance ( $d_J$ ) is one minus the Jaccard similarity. Therefore, for any given country, we can look at the set of countries that are in the same community before and after the catastrophe and compute the Jaccard distance (dissimilarity score) for these sets.

Let  $A$  denote the set of community members of some country before a catastrophe and  $A'$  the set of community members of the same country after the catastrophe. We can then define the Jaccard distance  $d_J$  as:

$$d_J(A, A') = 1 - \frac{A \cap A'}{A \cup A'}. \quad (1)$$

In the context of this study, the Jaccard distance indicates how similar two trade communities are. A value of zero indicates that the trade community did not change, while a value of one indicates that the trade community has changed completely. The assumption here is that a larger change is bad, as countries build their infrastructure to accommodate their current trading partners and cannot be easily changed without preparation (Jagtap et al., 2022).

#### Within-community degree and participant coefficient

The functional cartography approach (Guimerà and Nunes Amaral, 2005) assumes that nodes within a network serve specific roles based on their connections within and across communities. A node's role is determined using two indices: one measuring its connectivity within its community ( $z$ ) and another assessing how its links are distributed among different communities ( $P$ ). The first index (the  $z$ -score) is defined as

$$z_i = \frac{K_i - \frac{K}{s_i}}{\delta_{K s_i}}, \quad (2)$$

where  $K_i$  is the number of links of country  $i$  within its trade community  $s_i$ ,  $\frac{K}{s_i}$  is the average number of links across all countries in  $s_i$ , and  $\delta_{K s_i}$  is the standard deviation of the number of links  $s_i$ . The trade communities are delineated with the Louvain algorithm (see section 2.1) The second index (the participation coefficient) is defined as

$$P_i = 1 - \sum_{s=1}^N \left( \frac{K_{is}}{k_i} \right)^2, \quad (3)$$

210  
 211 where  $K_{is}$  is the number of links of node  $i$  to nodes in community  $s$ ,  $k_i$  is the total number of links of node  $i$ , and  $N$  is the number  
 212 of communities.  
 213 These indices define a parameter space where different regions correspond to specific roles based on threshold values. Guimerà  
 214 and Nunes Amaral identified seven node roles:

- 215 1. **Hubs** (if  $z \geq 2.5$ ) and **non-hubs** (if  $z < 2.5$ ).
- 216 2. **Non-hubs** are further classified based on the  $P$ -dimension:
  - 217 ○ **Ultra-peripheral** (all or almost all links within their own community,  $P \leq 0.05$ ),
  - 218 ○ **Peripheral** (most links within their own community,  $0.05 < P \leq 0.62$ ),
  - 219 ○ **Connectors** (many links across different communities,  $0.62 < P \leq 0.80$ ),
  - 220 ○ **Kinless** (evenly distributed links across all trade communities,  $P > 0.80$ ).
- 221 3. **Hubs** are categorised as:
  - 222 ○ **Provincial hubs** (vast majority of links within their own community,  $P \leq 0.30$ ),
  - 223 ○ **Connector hubs** (many links to most other communities,  $0.30 < P \leq 0.75$ ),
  - 224 ○ **Kinless hubs** (evenly distributed links across all communities,  $P > 0.75$ ).

225 These roles represent different types of traders within the network, with provincial hubs being crucial for community cohesion,  
 226 kinless hubs for global network cohesion, and connector hubs playing important roles in both aspects (see Figure 4).

## 227 2.4.2 Centrality

228 Centrality is a measure of the importance of a node in the whole network. This metric allows us to identify the main importers  
 229 and exporters of food in our trade network. Here, we consider weighted degree centrality, which is calculated by dividing the  
 230 sum of all incoming/outgoing edge weights (the amount of food traded) for a given node by the sum of all incoming/outgoing  
 231 edge weights in the entire graph.

## 232 3 Results

### 233 3.1 Changes in wheat trade

#### 234 3.1.1 Shifts in trade communities

235 According to our modelling, the wheat trading communities (based on the Louvain algorithm, section 2.1) would evolve  
 236 differently during GCIL and ASRS. This can be seen in the distribution of the trade communities globally (Figure 2), but  
 237 especially in the amount of change that countries could undergo in their trade communities (Figure 3). For this part of the  
 238 analysis, we assume that all countries still participate in trade, even if they were involved in a nuclear exchange in the ASRS  
 239 scenario. We separately look at the impacts of a complete removal of countries from the trade network in section 3.3.1..

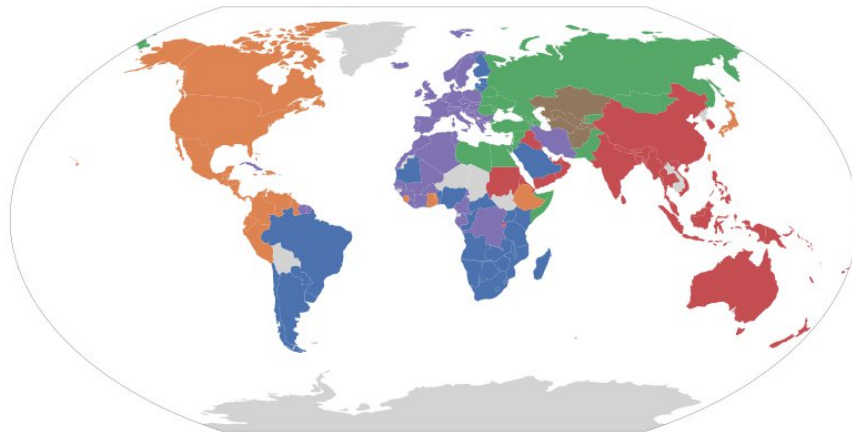


240 Under GCIL, most trade communities could remain relatively unchanged from the present configuration. Only a handful of  
241 countries, such as the United Kingdom, Ireland, Iran, Senegal, and the Democratic Republic of Congo, may experience a  
242 complete reconfiguration of their trade partnerships compared to the current state.

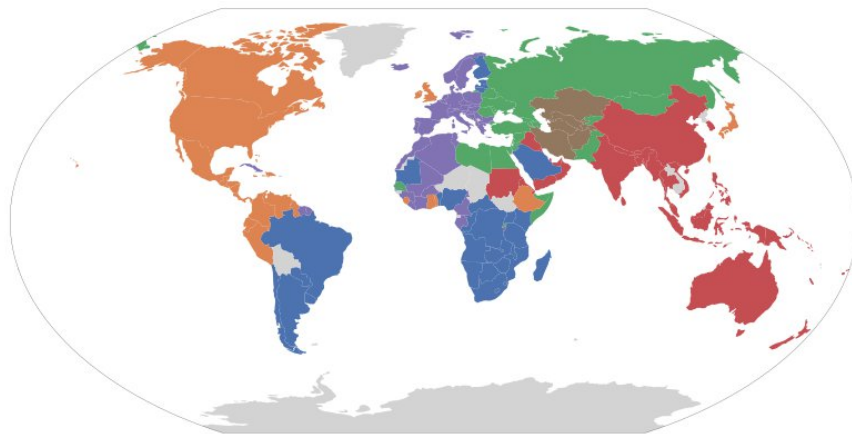
243 In contrast, the changes might be far more substantial in ASRSs. Nearly half of all countries could experience a shift in their  
244 trading partners, with eleven countries undergoing a complete or near-complete overhaul of their trade connections. Some  
245 countries affected are consistent with the GCIL scenario, like Iran and the Democratic Republic of Congo, while others, such as  
246 Peru or Finland, could be part of the transformed trade landscape.

247 The global distribution of trading communities (Figure 2) reveals that this significant shift is primarily due to the expansion of  
248 the trading community containing Russia. Today, this community comprises mainly Russia, Eastern Europe, and a portion of  
249 North Africa. In the ASRS, however, it extends across the Balkans and most of North Africa.

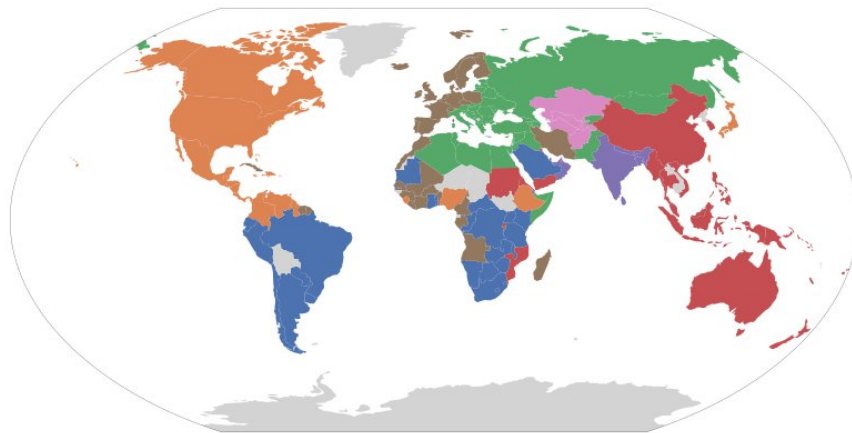
Trade communities for wheat with base year 2022



Trade communities for wheat with base year 2022  
in scenario: Global Catastrophic Infrastructure Loss



Trade communities for wheat with base year 2022  
in scenario: Abrupt Sunlight Reduction Scenario



250

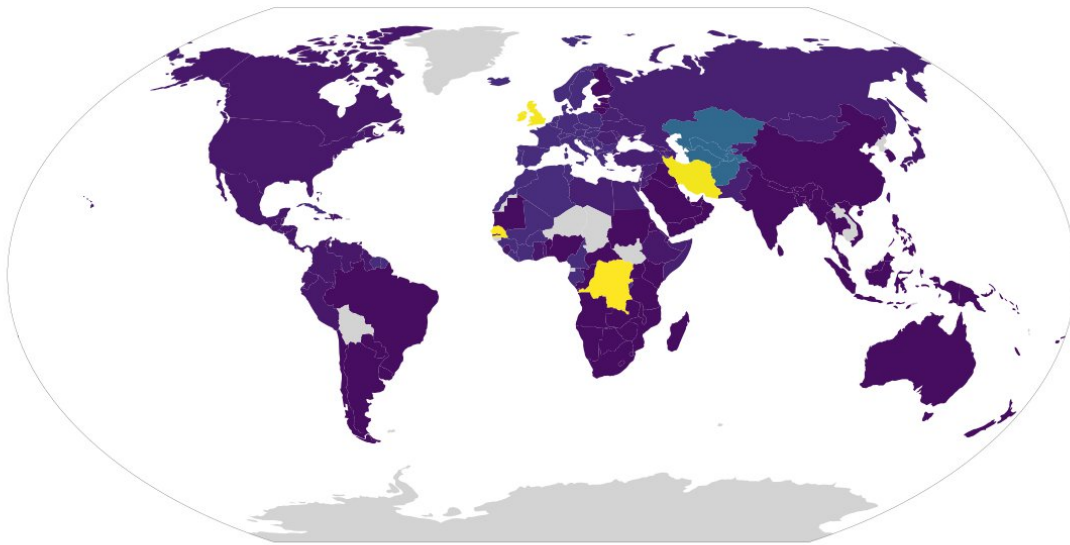
251

252

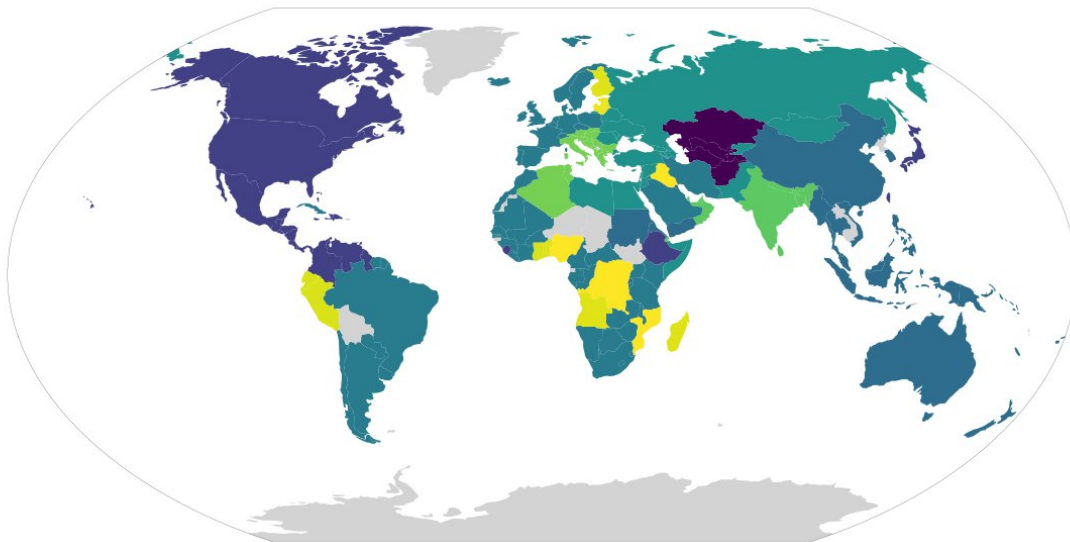
253

**Figure 2: Trade communities for wheat in 2022 after yield reduction due to global catastrophic infrastructure loss as well as abrupt sunlight reduction. The colours indicate trade communities. In the GCIL scenario, despite large drops in yields, global trade communities remain relatively unchanged. However, in the ASRS, the changes are more substantial.**

Jaccard Distance for wheat with base year 2022  
in scenario: Global Catastrophic Infrastructure Loss



Jaccard Distance for wheat with base year 2022  
in scenario: Abrupt Sunlight Reduction Scenario



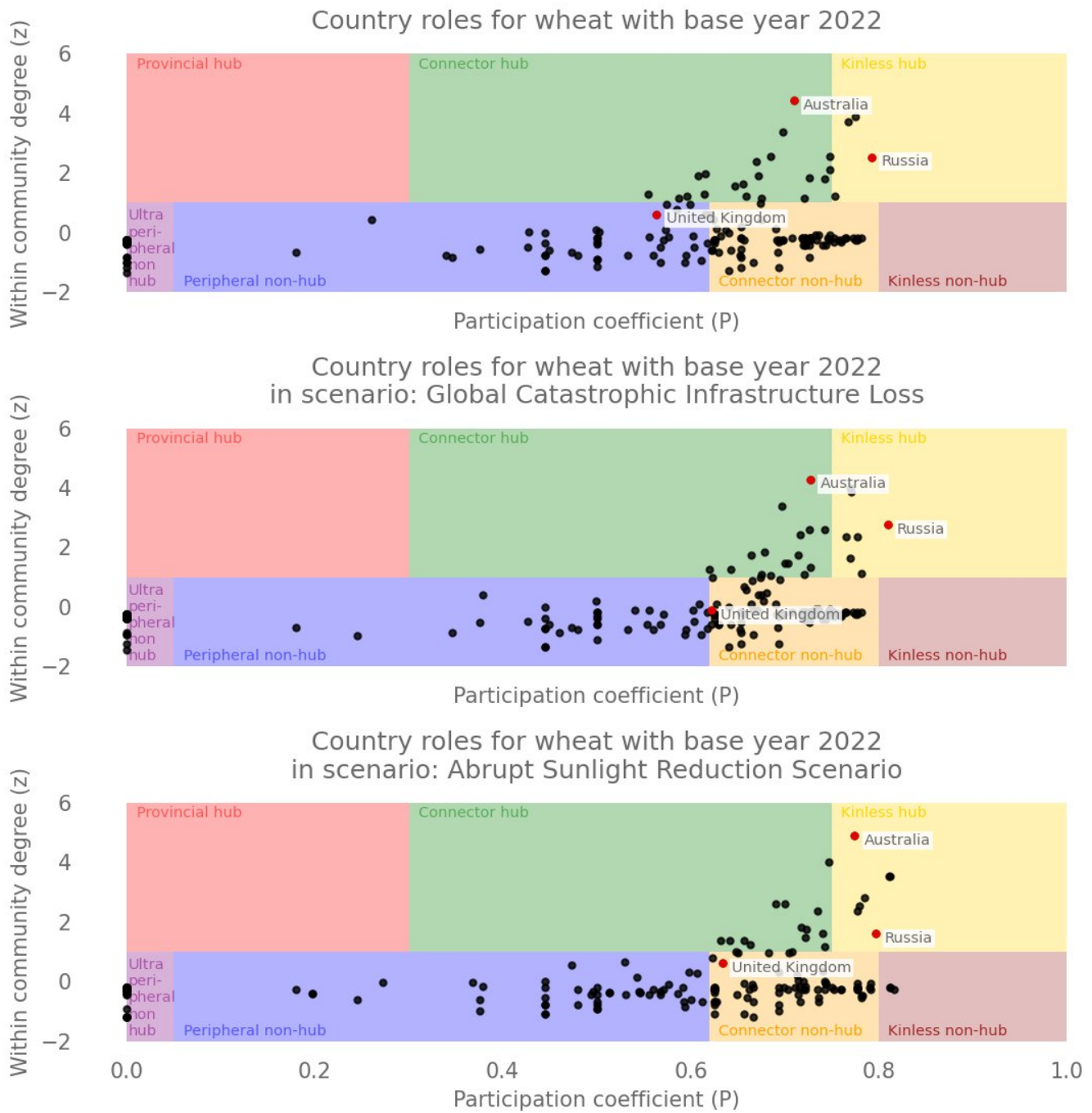
**Figure 3: Changes in wheat trade communities after yield reduction due to global catastrophic infrastructure loss as well as abrupt sunlight reduction, in comparison to the communities in 2022. Colours indicate the magnitude of change as the Jaccard distance. Yellow means the trade community of a country has changed completely, and dark blue that the country remains in the same trade community. Again, we see that changes in trade communities are much more pronounced in the ASRS.**

### 259 3.1.2 Community roles of countries

260 In contrast to the impact we can see in the trade communities, there are no significant shifts in community roles in the GCIL  
261 scenario and the default 37 Tg ASRS (Figure 4). However, we note that in larger ASRSs, country roles within trade networks do  
262 shift significantly. The 47 Tg scenario (Figure S2) reveals distinct transitions: some countries shift from non-hub connectors to  
263 peripheral non-hubs, while others become provincial hubs. This suggests countries lose connections both within and beyond their  
264 trade communities, with external connections most affected. Countries maintain fewer imports overall, but those remaining  
265 imports come primarily from within their trade community. These patterns indicate a potential tipping point between 37 Tg and  
266 47 Tg, where the system shifts from minimal change to substantial reorganization.

267  
268 Another way to assess country roles in the global trade network is through in- and out-degree centrality, which identifies key  
269 importing and exporting countries (Figure S3). In-centrality remains stable across all scenarios, reflecting the overall trade  
270 volume, although total imports decrease due to reduced yields. Out-centrality experiences more significant changes. Presently,  
271 Australia has the highest out-centrality, followed by the United States, France, Canada and Russia. This order remains largely  
272 unchanged after GCIL, though Russia's out-centrality slightly surpasses that of the United States and Canada. Likely because of  
273 its reduced use of agricultural inputs in comparison with the other countries. The most substantial shifts occur in the ASRS,  
274 however, where Russia, Canada, and the United States experience considerable yield losses, resulting in significantly reduced  
275 out-centrality. Meanwhile, Australia maintains its top position, with France and Argentina rising to second and third place,  
276 respectively.

277  
278  
279



**Figure 4: Country roles in the global wheat trade network in 2022 and after yield reduction due to global catastrophic infrastructure loss, as well as, abrupt sunlight reduction; based on within community degree and participant coefficient (see Section 2.4.1).**

### 3.1.3 Changes in trade flows

When examining the decline in imports by country, we observe greater impacts under ASRSs compared to GCIL (Figure 5). Ukraine and Argentina, which only export wheat, remain mostly unaffected in both scenarios. Under GCIL, most countries see a 20-30% reduction in imports, with some African and European nations experiencing up to a 40-60% decrease in imports.

In contrast to GCIL, ASRSs result in a broader range of import changes. Nations such as the United States, Norway, and Mongolia lose up to 100% of their wheat imports, primarily due to reduced yields in major wheat-exporting countries like Canada, the United States, Russia, and Ukraine. These changes are mirrored in both degree-centrality measures, indicating a

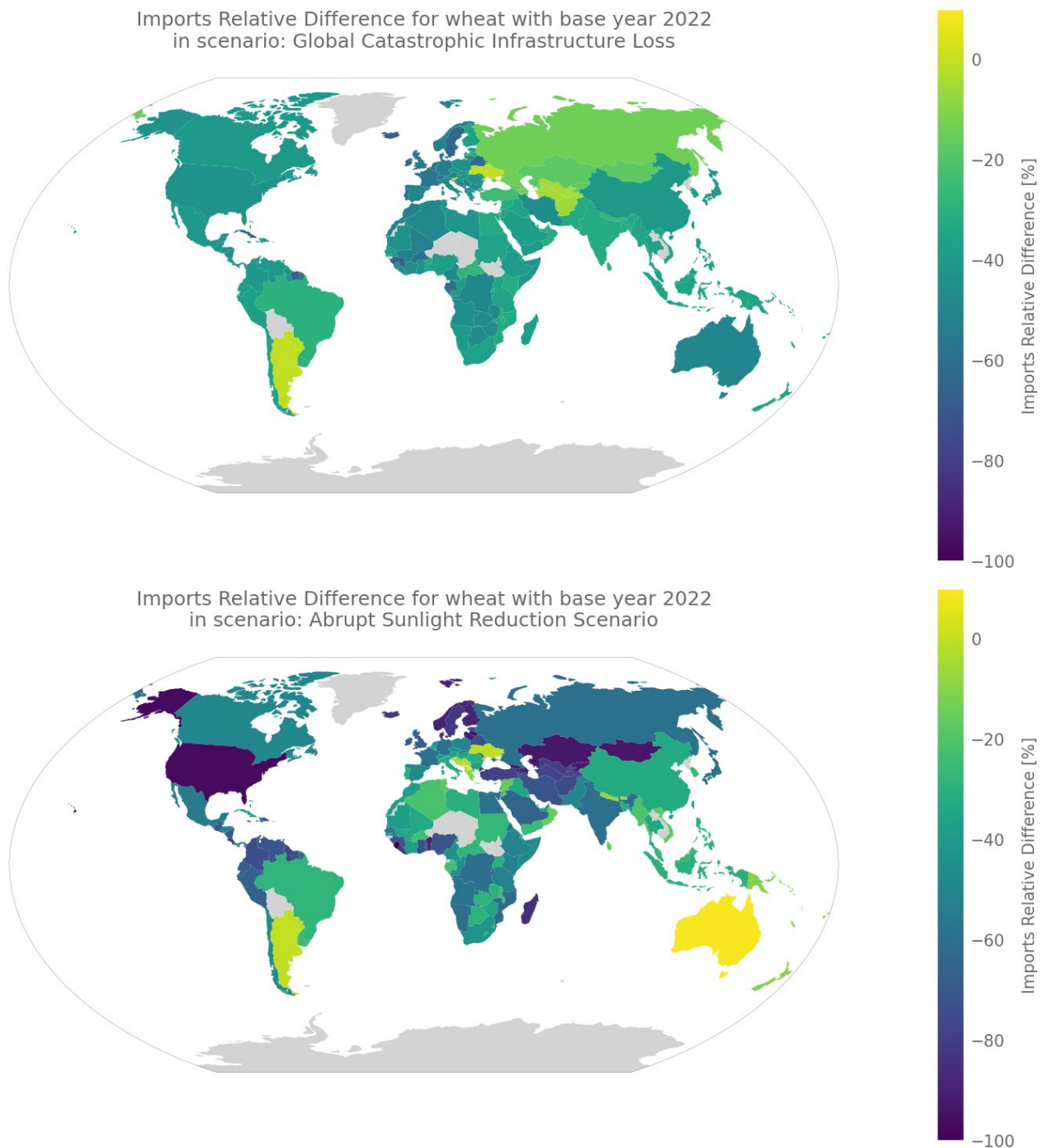
292 significant loss of centrality for these previously major exporting countries (Figure S3). This leaves Australia as the only  
293 remaining major exporter of wheat.

294  
295 We can also study the absolute changes in wheat imports (Figure S4). This highlights similar patterns across both GCIL and  
296 ASRS, albeit still with a higher impact in the ASRS. The strongest effects in both scenarios can be seen in China, Turkey,  
297 Indonesia and Egypt. All these countries import large amounts of wheat from Russia and Central Asian countries like  
298 Kazakhstan, which experience major yield losses in both scenarios. In particular, Turkey would experience a massive loss of  
299 wheat imports in absolute terms in an ASRS, with around 8 million tonnes of wheat imports lost.

300  
301 The impact to the trade network is also visible when examining remaining wheat production (Figure S5, S6). The scenarios differ  
302 markedly. The GCIL scenario (Figure S5) preserves more wheat production, particularly in Russia, where farming depends less  
303 on fertilizer than in Central Europe or the US. However, in ASRS, Russian production drops severely as lower temperatures  
304 make wheat growing nearly impossible. Australia, a major wheat exporter, continues production but at reduced levels. India  
305 produces the most wheat in ASRS, but primarily grows it for domestic use and may not become a major exporter.

306  
307 Additionally, we performed robustness checks of our results with different metrics. The shifts in trade patterns and the  
308 heightened impact of ASRSs are also evident in other metrics, like community satisfaction and node stability. Community  
309 satisfaction gauges the proportion of a country's trade within its trade community, while node stability indicates a country's  
310 ability to replace lost trade partners. Both metrics highlight the challenges faced by nations reliant on Russia and the United  
311 States. More information on those measures is provided in the supplement (section 3.6).

312



**Figure 5: Relative changes in wheat imports after global catastrophic infrastructure loss and abrupt sunlight reduction scenarios in comparison to today.**

### 3.2 Trends in rice, maize and soya beans

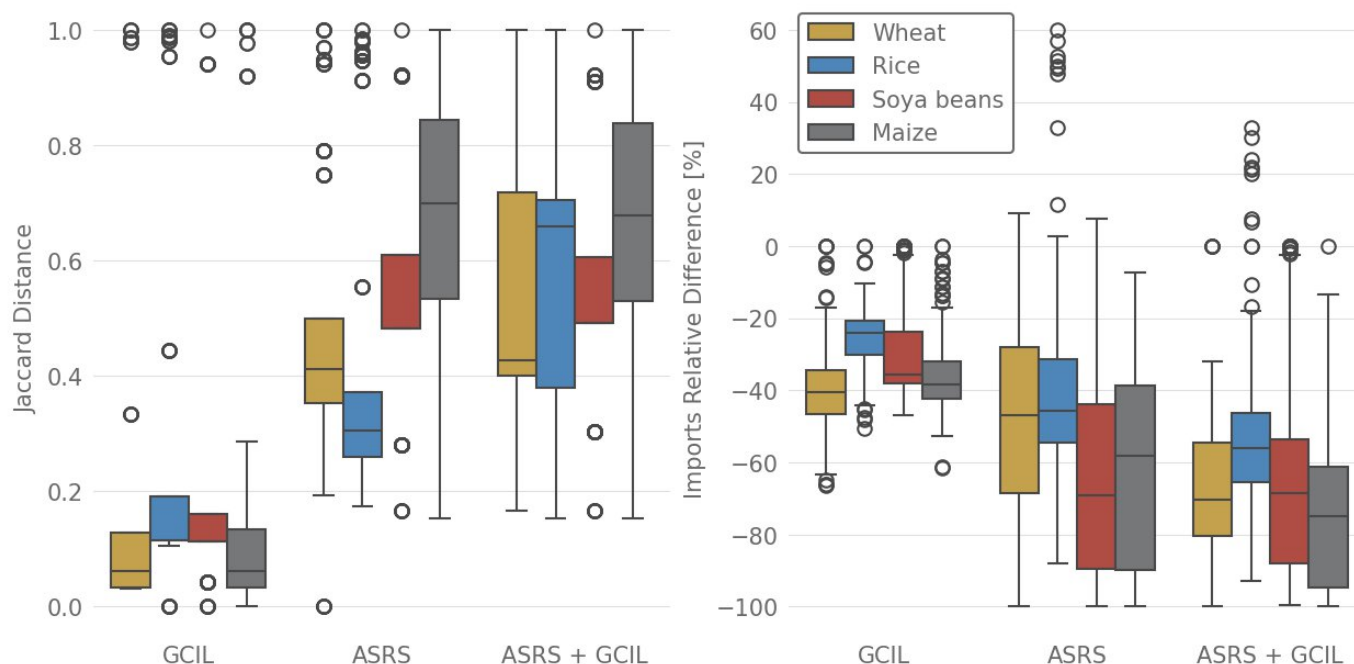
#### 3.2.1 Overall pattern and comparison across scenarios

The patterns observed in the wheat data are also evident in rice, maize, and soya beans (Figure 6). Across all crops, the impact of ASRSs is larger than GCIL. This is especially true for the outliers in the distribution. In the case of wheat, for instance, while the median remains similar across scenarios, certain countries experience a complete loss of imports under abrupt sunlight reduction, which does not happen in GCIL. Considering the variations in yield reduction (Figure 1), it becomes clear that at 37 Tg of soot emissions, the effects are generally comparable for both scenarios when it comes to yield reductions. However, the range of



323 impacts and change in trade communities would be much more extensive in ASRSs. Additionally, the most affected countries  
 324 vary between crops due to differing trade volumes across world regions.

325 Combining the effects of ASRS and GCIL, which could occur during a nuclear war that influences climate and disables industry  
 326 due to direct destruction and HEMP, has a very severe impact on food trade. However, the overall impact is less than the sum of  
 327 their individual effects. Many countries severely affected by ASRS have already experienced significant yield losses and the  
 328 additional disruption due to GCIL has thus little effect. Nonetheless, this combined catastrophe would severely impact yields and  
 329 food trade.



330  
 331 **Figure 6: Jaccard Distance and reduction in imports, for each country and crop, for Global Catastrophic Infrastructure Loss (GCIL),**  
 332 **and the Abrupt Sunlight Reduction Scenario (ASRS).**

### 334 3.2.2 Rice

335 For rice, the import reduction and trade community disruptions are similar between abrupt sunlight reduction and GCIL,  
 336 differing mainly in magnitude. Under GCIL, most countries typically lose around 20-30% of their rice imports, whereas it ranges  
 337 from 30-50% under ASRSs. Most countries also maintain much of their pre-catastrophe trading community, with exceptions  
 338 including Russia, Ukraine, and Indonesia. However, a majority of the countries experience at least some shift. This more limited  
 339 degree of change in comparison to wheat is also evident in community roles, which remain largely consistent across all  
 340 scenarios. This stability can be attributed to India's prominent role as the leading rice exporter, its relatively low reliance on  
 341 agricultural inputs compared to other countries, and it is still relatively high temperatures during ASRSs, thereby stabilising the  
 342 rice trade network even during catastrophes. See supplement section 4.1 for the figures showing the trends described here.

### 343 3.2.3 Maize

344 In GCIL, the impact on maize is evenly spread worldwide. However, under ASRSs, there's a stark contrast between the Northern  
 345 and Southern Hemispheres. Nearly all Northern Hemisphere countries lose most or all of their maize imports, while in the  
 346 Southern Hemisphere, South America, much of Africa, and Southeast Asia maintain some imports, mainly from less affected  
 347 regions like South America. Many countries switch to the connector non-hub role, likely as most countries experience low trade



348 volumes overall. The maize trade network in ASRSs exhibits low stability and is heavily affected by the removal of the other  
349 major exporters, after the United States' decline in importance due to yield reductions. See supplement section 4.2 for the figures  
350 showing the trends.

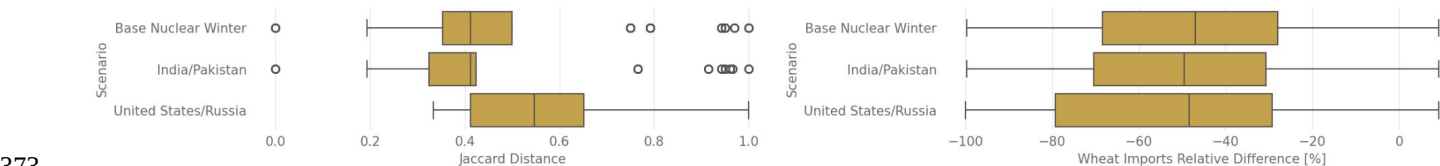
351 **3.2.4 Soya beans**

352 Regarding soya beans, there is a shift in the distribution of affected countries compared to wheat. Many African countries remain  
353 relatively unaffected, primarily due to their low trade volumes. Under GCIL, most countries face a similar reduction, roughly 20-  
354 40%, in imports. In ASRSs, the patterns resemble those of wheat, except for South-East Asia and Oceania. These regions still  
355 receive wheat imports from Australia and each other, but their soya bean imports mainly come from the United States, resulting  
356 in a decline. This trend is reflected in trade communities, which remain mostly stable for GCIL but converge into two primary  
357 and one minor communities for ASRSs. Soya bean export is heavily concentrated in the United States, so a sharp yield decline  
358 there disrupts trade communities significantly. Only countries importing soya beans from Brazil maintain higher import levels,  
359 and the trade community with Brazil stays very stable. Similarly to wheat, the role of countries in their communities shifts, with  
360 most staying the same for GCIL but losing much connectivity in ASRSs. Another notable deviation from wheat patterns lies in  
361 network vulnerability to node removal. With only two major exporters, the United States and Brazil, if the United States is  
362 already affected by yield reduction, the network becomes less stable, experiencing another shock when Brazil is removed. See  
363 supplement section 4.3 for the figures showing these trends.

364 **3.3 Comparison of nuclear war scenarios**

365 **3.3.1 Impact of removing countries**

366 The ASRS data is based on nuclear war simulations. To explore these further, we simulated the removal of Russia/United States  
367 and Pakistan/India from the 37 Tg scenario (Figure 7). We compared these scenarios with the ASRS that includes all countries  
368 and the wheat trade of today. The findings reveal that while removing these countries affects both trade communities and overall  
369 imports (Figure 7), the effect of the yield reduction due to abrupt sunlight reduction is already so big that the removal of those  
370 countries is negligible. Thus, if countries involved in a nuclear conflict were to cease as trading partners due to the destruction of  
371 their territories, it would cause additional disruptions to the global trade network beyond those due to the yield reductions, but  
372 only marginally and for a subset of countries.



373 **Figure 7: Country removal impact on wheat imports in nuclear war scenarios.**

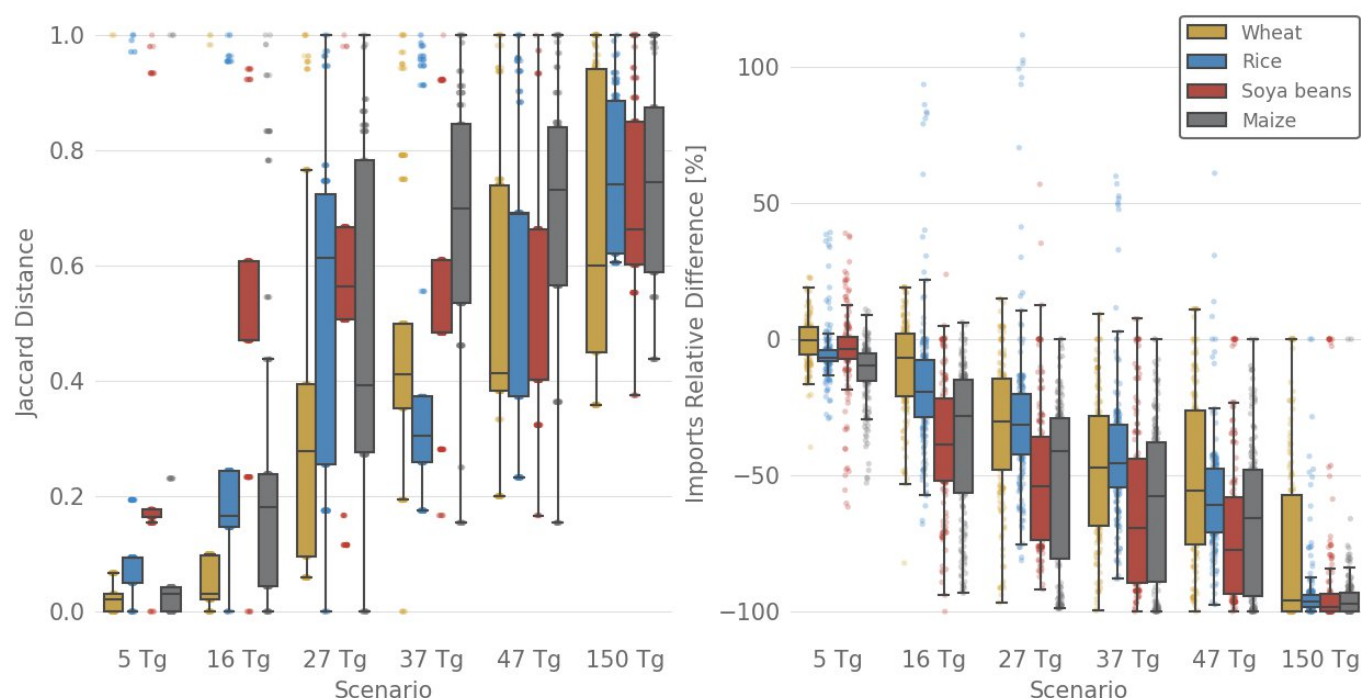
375 When simulating the gradual removal of nodes, the results indicate that removing random nodes causes a slow but steady decline  
376 in network stability. In contrast, specifically targeting the most active exporting nodes results in a rapid decline in stability,  
377 leading to network collapse after removing 10-20% of these crucial nodes. Further details can be found in Supplement Section  
378 3.7.

379 **3.3.2 Impact of soot emission magnitude after nuclear exchange**

380 Assessments of the impacts for nuclear war scenarios of different magnitudes (Figure 8) show a consistent pattern across all  
381 crops analysed. While the most significant impacts can be seen in the worst nuclear war with 150 Tg of soot emitted (nuclear

382 winter), the effects would already be quite severe at 37 Tg (nuclear fall). The 37 Tg scenario engenders a substantial of about  
 383 60 % import loss, suggesting that trade would be massively impacted in the 37 Tg case. However, for most countries, food  
 384 imports would have ceased almost entirely in a 150 Tg scenario. In addition, even at merely 5 Tg, some countries could  
 385 experience a 50 % loss of maize, and at 16 Tg, a considerable number of countries have import reductions from 40 % to almost  
 386 100 % across all crops.

387 While trends remain comparable across all crops, including major import changes, wheat seems to be the least affected. Soya  
 388 beans experience a stark shift in trade communities at as low a magnitude as 5 Tg. However, the change then stays relatively  
 389 constant for all other magnitudes.



390  
 391 **Figure 8: Relative change in imports and Jaccard distance for the four primary crops and nuclear war climate changes resulting from**  
 392 **the emission of 5, 16, 27, 37, 47, and 150 teragrams (Tg) of soot across all countries. Coloured points represent individual countries.**

## 393 4 Discussion

394 Overall, the main finding of this study is that the two food-system relevant GCR scenarios we have considered may affect  
 395 agriculture quite differently in both the magnitude of their effects as well as the spatial distribution, suggesting that they will  
 396 need different mitigation strategies to increase societal resilience against them. ASRS will be challenging as it will hit a fraction  
 397 of countries very hard, while leaving others mostly unaffected. GCIL on the other hand would affect all countries, but on a  
 398 similar magnitude.

399  
 400 Our results show clear differences between the effects of ASRSs and GCIL on food trade. The scenarios have different effects on  
 401 how much trade communities are disrupted, the decrease in overall imports, and the roles of countries within their trade  
 402 communities. Across all these measures, ASRSs lead to much larger disruption than GCIL, even for a similar net global yield  
 403 loss. This is due to the way these global catastrophes play out and the spatial distribution of their effects.

404

When both scenarios are combined to simulate the co-occurrence of both kinds of catastrophes, the impacts increase and result in food import losses in the range of 70-100 % for many countries. The impact on yields (Figure 1) and trade (Figure 6) change for both catastrophes in a similar way. We can see that the median losses are similar for both trade and production. However, while there are still countries that will likely see little impact on their food production by direct effects of the catastrophes, almost all countries experience a major loss in their food imports. This is due to the countries that are least affected are usually not major exporters. For GCIL the least affected countries are those that have very low input agriculture, which is usually also not very productive, while for ASRS the positive effects are mostly in countries which are too warm now for most agriculture.

As Moersdorf et al. (2024) have shown, in a GCIL scenario, the countries hit the hardest are those doing the most intense agriculture when it comes to industrial input like fertilisers. This is also in line with other research studying the impact of losing these inputs (Ahvo et al., 2023). These highly productive countries are also typically the countries that export the most food. Also, the effects are felt in all countries globally with no exceptions, as industrial inputs are in use worldwide. This results in a very homogenous impact on food trade, where most countries experience a relatively similar level of trade disruption.

On the other hand, for ASRSs, we see a much larger split between which countries are more or less affected. Generally, the higher the country's latitude, the more it is affected (Coupe et al., 2019). In addition, countries in the Northern Hemisphere are affected more overall. This is partly because nuclear war would most likely occur in the Northern Hemisphere (Coupe et al., 2019), resulting in somewhat lower soot concentrations in the Southern Hemisphere. In addition, the Southern Hemisphere has more land closer to the equator and more ocean (which acts as a temperature buffer), suggesting that the Northern Hemisphere may still be more affected even for ASRSs that do not involve nuclear war. These factors may lead to an especially large yield decline in the United States, Canada, Central and Northern Europe and Russia. These are all major food exporters, particularly for wheat or, in the case of the United States, for all major crops. This loss of exports from the major exporting countries cascades across the whole system. For all crops we can see significant changes in trade communities and large declines in the amount of imported food. This is especially true for maize, as maize is not very cold tolerant and, therefore, especially vulnerable to drops in temperature.

Focusing more specifically on the nuclear war scenario, we can see that the main effects of these disruptions are due to the yield decline. The complete removal of the countries involved in a war only introduces little additional shifts in the overall imports. However, removing Russia and the United States brings additional disruptions to the trade communities, as both countries are the anchors in their respective trade communities. We can also observe that the effects of the nuclear war increase considerably with rising amounts of soot ejected into the atmosphere. While a 5 Tg emission has relatively minor effects, except in soya bean trade communities, the effects quickly grow with higher soot emissions.. This emphasises that even if nuclear weapons were used, it is extremely important to limit further escalation in order to prevent additional disruption to the global food system.

#### **4.1 Implications for the food system**

Clapp (2023) identifies three primary vulnerabilities in the food system: 1) dependence on a limited number of staple crops, 2) domination by a small group of major exporters, and 3) concentration of food trade among a few companies. While we did not explore the role of companies, the issues of reliance on a few staple crops and dominance by major exporters are also evident here. The main vulnerability is the extremely central role of the United States in the food trade network. Every scenario resulting in yield reduction in the United States or even its complete removal from the system will result in massive cascading disruptions, both in the overall communities and the amount of food traded. The vulnerability would decrease considerably if the food system were less concentrated in the United States. Other studies of the current trade network show a high dependency on other major

exporters, such as Brazil and Russia (Ji et al., 2024). Our results also indicate that a disturbance of these nodes is very significant. For instance, Australia would be the last remaining major exporter of wheat during an ASRS (Figure S3). Therefore, if this country were to stop exporting, the wheat trade would effectively end globally.

We know that complex networks become more susceptible to perturbations as they get more centralised (Wiliński et al., 2013), and the food system is getting increasingly centralised and concentrated (Clapp, 2023). This means that if we do not alter our approaches to food trading, we will get more and more vulnerable to major shocks and the kinds of scenarios we have described. There are some indications that this global concentration of trade might be beginning to change (Kang et al., 2024; Mamonova et al., 2023), as more countries rethink how they handle food and trade more generally. Whether these trends continue depends on how the geopolitical situation develops in the coming years and decades.

Our research also shows that the ASRS has a much wider range of effects. Some countries could even increase imports, as their neighbouring countries profit from the changed climate (e.g. more precipitation and cooler temperatures in some semi-arid regions), while others could lose all of their incoming food products. This means that recommendations to tackle these scenarios have to be tailored to specific countries, as there can be no approach that applies to all countries. For GCIL, more general recommendations might be possible, as all countries are affected similarly.

Recent studies have compiled lists of countries that have experienced substantial food import shocks in the past (Zhang et al., 2023b). While there is some overlap between these countries and the ones affected the most here, it also becomes clear that especially the Central European countries, as well as the major exporting countries, have not experienced large food import shocks on that scale in modern history. This indicates that these countries have no experience with import shocks and are possibly less prepared to handle the scenarios described in this study.

Shifting dietary patterns could also help decrease vulnerability to trade disruptions. A move toward plant-based diets with more locally-produced fruits, vegetables, and legumes could reduce dependence on international grain trade, as much of the currently traded grain (especially soya beans and maize) is used for animal feed rather than direct human consumption. This conversion of grain to animal products is inefficient from an energy perspective. However, this strategy presents a trade-off: while reducing animal feed imports would decrease trade dependencies, it might also reduce the system's overall flexibility. Current livestock systems, despite their inefficiencies, create a buffer by maintaining large stores of grain that could be redirected to human consumption during crises. Additionally, ruminants can digest cellulose that humans cannot, potentially providing an additional food source during catastrophes. Therefore, while dietary shifts toward plant-based foods could improve local food security under normal conditions, maintaining some animal agriculture may provide valuable system redundancy for extreme scenarios. The optimal balance likely varies by region based on local agricultural conditions and trade relationships.

Similarly, more strategic use of agricultural land could enhance resilience. Currently, significant agricultural capacity is devoted to non-food purposes – particularly biofuel production and crops like tobacco. While biofuel crops are often heavily subsidized, transitioning to electric vehicles would be more energy efficient and free up land for food production. The land used for tobacco cultivation could be repurposed for food crops, providing dual benefits of improved food security and public health. Additionally, reducing food waste, which accounts for approximately one-third of food production in many developed countries, represents a readily available opportunity to build resilience (Alexander et al., 2017). However, as with dietary shifts, these changes present trade-offs. Some biofuel infrastructure could potentially be repurposed to produce food during crises, similar to how breweries can be converted to produce sugar from cellulose (Throup et al., 2022). Moreover, maintaining diverse

agricultural systems and processing capabilities, even for non-food crops, helps preserve farming knowledge and infrastructure that could be valuable during catastrophes. The key is finding a balance between efficient land use under normal conditions and maintaining adaptable agricultural systems that can respond to major disruptions.

## 4.2 Study Limitations

The research presented is, to our knowledge, the first to take a more nuanced look into what might happen to the food trade system after global catastrophe, meaning that there is much room for improvement in future work. We consider the main limitations of our study to be:

- We only looked at the direct effects of yield reduction on trade flows and did not consider any additional adaptations. For example, it seems likely that many countries would introduce export bans if their own yields dropped significantly, worsening the overall situation. This means that our study can be seen as the minimal amount of change that one can expect to happen after global catastrophes by the yield changes alone, barring the introduction of resilient food adaptations to counter the loss of yields (Pham et al., 2022). Further research is needed to understand how societies might react to the effects explained here.
- GCIL also includes the assumption that we lose the majority of our mechanisation and transportation. This is not modelled in this study, but plausibly could have major implications beyond the impact on yields and make it more catastrophic than ASRSs. A GCIL would disrupt fossil fuel production, hampering international trade. However, possible interventions include retrofitting ships to be wind powered (Abdelkhalik et al., 2016) or wood gasification to replace fossil fuels (Nelson et al., 2024).
- We studied the four major food crops in isolation to understand what effects might play out on that level. However, the food system also consists of other parts like fisheries or livestock. While those are also predicted to decline after a global catastrophe (Scherrer et al., 2020; Xia et al., 2022), it remains unclear how the totality of all food trade might be affected by global catastrophes. Livestock would be more strongly affected than major crops because it mostly depends on them; whereas fisheries, while less affected than crops, make up a small percentage of global caloric requirements (<2%).
- Additional layers of interaction from non-food products through social dynamics to economic policies could be considered in a multi-layer network model, which has been shown to be impactful and effective in other scientific disciplines (De Domenico, 2023; Kivelä et al., 2014; Paluch et al., 2021).
- We treated nuclear war simulations as a proxy for large size impact over the land and super volcanic eruptions. While this is a reasonable assumption, the results might end up very different, especially if the impact/eruption happens in the Southern Hemisphere, as nuclear war scenarios usually only involve the Northern Hemisphere, as there are no nuclear weapon states in the Southern Hemisphere. Although these extreme events all produce large amounts of aerosols in the upper atmosphere, which block sunlight and cause significant cooling, the compositions of the aerosols differ. This results in variations in the duration of the cooling and some climate impacts. Recent research indicates that a simulated volcanic winter shows similar trends (Enger et al., 2024) to previous studies on nuclear winter (Coupe et al., 2019), although volcanic winters are likely to be shorter in duration. Additionally, there is a possibility that multiple mid-sized volcanic eruptions could occur simultaneously, releasing enough sulphate aerosol to cool the Earth significantly.
- Even for such a relatively well-studied global catastrophe as nuclear winter, there is still much we do not understand. For example, the work of Coupe et al., (2023) suggests that nuclear winter can paradoxically lead to a decrease in Antarctic sea ice despite global cooling. As our understanding of global catastrophic risks increases, we may see shifts in our expected effects on the food system.

- Trade is only a part of the global value chain, and if we look at the whole value chain, we can expect many more disruptions (Ibrahim et al., 2021).
- The modeling of Xia et al. (2022), which we use to calculate yield reductions during a nuclear war, assumes the usage of spring wheat. However, during an ASRS, wheat producers could switch to winter wheat, which is more resistant to cooler temperatures and frost and generally has slightly higher yields than spring wheat. Therefore, the wheat yields in this study are potentially underestimated. We only consider the global aspects of the catastrophes. However, there are a variety of plausible scenarios where regional effects could have global repercussions. For example, the food system has several so-called choke points (Bailey and Wellesley, 2017; Key et al., 2024; Wellesley et al., 2017), where much food trade is funnelled through a small geographic area. Some of these choke points are near volcanoes and could be severely affected by eruptions (Mani et al., 2021). Should these choke points close in the aftermath of a global catastrophe, the disruption of the food system would further increase.

### **4.3 Comparison to climate change**

The model employed in this study was originally developed to study the effects of climate change on food trade (Hedlund et al., 2022). We can see that the impact of a rather severe climate change scenario based on RCP 8.5 has considerably lower effects than the catastrophes explored here and even results in an increase in imports for almost all countries (Figure S25). For all crops the trade communities stay mostly the same, while they would be much more disrupted in our scenarios. A similar pattern holds up for all crops considered. These differences are likely due to the different magnitudes of the catastrophes considered. For RCP 8.5, a land surface air mean temperature increase of around 5°C is expected by 2100 (Zhang et al., 2023a), while for a 37 Tg nuclear fall scenario, a land surface air mean temperature drop of up to 8 °C is predicted in the 3rd and 4th year after the nuclear war. (Xia et al., 2022). Therefore, the ASRS considered here not only has the larger temperature change, but also in a much shorter time period. Also, in the case of climate change, the countries that will be more affected are those closer to the equator (Frame et al., 2017). Since the main exporting countries are mostly at higher latitudes, they will be less affected by climate change, contributing to a more stable food trade in comparison to the scenarios we explored.

### **4.4 Gaining a deeper understanding of how global catastrophes impact the food system.**

#### **4.4.1 Research gaps**

The research presented here is a first step in understanding what might happen to food trade after global catastrophes. However, there are still a wide range of factors we do not understand. With the introduction of terms like multiple-breadbasket failures, food system research has increased in scope (Clapp, 2023; Jahn, 2021; Nyström et al., 2019; Savary et al., 2020). Still, this kind of research does not consider events where all countries are affected simultaneously and on a scale not seen in modern history, leaving the effects of global catastrophic risks unexplored. This means that global food system research should also include global catastrophic risk in order to have all angles covered. Due to this general lack of focus on global catastrophes, we outline specific topics that warrant further attention:

- Understanding how global catastrophic risk might affect different parts of the global population by socio-demographic metrics. We know that climate impacts are felt differently depending on how rich the country is (Quante et al., 2024) and also increase wealth inequality (Méjean et al., 2024). Therefore, it is likely that these differences also exist as a consequence of global catastrophes.
- While there is little research on the effects of the dependency on very few food trading companies (Clapp, 2023), there is none when it comes to the question of how this might affect the outcomes of global catastrophic risk scenarios.

- There exists some research that acknowledges the potential cascading effects and systemic risk of an ASRS, like nuclear war, for instance, recent summaries by Green (2024) or Glomseth (2024), but for many of the events that could cause GCIL, we know only very little about the potential cascading effects. Beyond this, even sophisticated modeling efforts like Xia et al. (2022) have limitations - they did not account for several factors that could further impact agriculture after nuclear war, such as changes in irrigation water availability, increased surface ozone levels, ultraviolet light damage, effects on pollinators, and killing frost risk. For many of the events that could cause GCIL, we know even less about the potential cascading effects and systemic risks.
- We need more understanding of the effects of catastrophes like geomagnetic storms and how the loss of industrial inputs might affect agriculture. There is some global research on the direct effects (Cliver et al., 2022; Isobe et al., 2022; Rivers et al., 2024b) but less on the indirect effects, especially on agriculture (Moersdorf et al., 2024). There are some recent research studies which explore similar effects yet do not frame it in regards to global catastrophic risk but instead as a general disruption in the trade of industrial inputs for agriculture (Ahvo et al., 2023; Sandström et al., 2024).
- There is a good chance that catastrophes will not happen in isolation but interact with each other and existing vulnerabilities. An example is the possible interaction between nuclear winter and planetary boundaries (Jehn, 2023) or termination shock caused by civilization collapse (Baum et al., 2013). These are only two of the possible interactions, and many others are entirely unexplored (for example, having a major geomagnetic storm during a pandemic).
- Our food system is not reliant on the food trade network alone but on a highly complex supply chain with many interacting goods and services (Ibrahim et al., 2021), also consisting of many non-food items. It would be valuable to understand how these might react to the scenarios described in this manuscript. There has been some work to study this for current conditions (Deteix et al., 2024), but not with a focus on global catastrophes.
- We do not know what might happen after the initial effects play out, as this paper only describes the minimal amount of change that is expected to happen due to the yield changes alone. However, if we look into history, we can see that such disruptions of trade networks can have massive consequences. If they unravel the whole network, countries lose access to many goods they need, leading to internal problems and possibly collapse, as happened in the Late Bronze Age (Linkov et al., 2024). Important insights could be gained here by applying insights from quantitative history to the last 100 years, as proposed by Hoyer et al. (2024). This could be built upon by using historical worst cases and using them as downward counterfactuals to create more realistic and comprehensive scenarios (Woo, 2019).

Furthermore, all those research topics that need further exploration and studies like ours should be regularly re-assessed. As the Russian invasion of Ukraine has shown, major disruptions in the food network can and are likely to happen again (Miller et al., 2024). They reshuffle existing trade connections, making research like this less accurate as time passes.

#### **4.4.2 Decreasing vulnerability to global hazards**

Since the global food system is vulnerable to major disruptions, it is of high priority to decrease these vulnerabilities. Myers et al. (2022) suggest a list of interventions that could decrease the vulnerability of agriculture to climate change. Some of these suggestions would also help here, like having more diverse crops to ensure flexibility with respect to climate conditions or strengthening international trade agreements to ensure that the flow of food is stable. This also ties in with the criticism of concentration in the food system by Clapp (2023). These concentrations on all levels of the food system increase the risks of collapse and need to be decreased, especially for the safety of people in net food importing countries (Yıldırım and Önen, 2024).

#### 604 4.4.3 Increasing resilience after a global catastrophe

605 It is not only important to decrease the risk of a hazard spiralling into a catastrophe, but also to prepare if it happens despite  
606 precautions (Cotton-Barratt et al., 2020). The complex events following the described catastrophes would constitute major crises,  
607 but historical evidence suggests societies can withstand such a polycrisis by building resilient infrastructure, maintaining the  
608 ability to respond effectively at scale, and having high social cohesion (Hoyer et al., 2023). We should increase the overall  
609 resilience of the food system and see the resilience of our food supply chains not as something that aims to bring back a system  
610 to the status before the catastrophe but as a system that is able to persist, adapt and transform even under intense pressure  
611 (Wieland and Durach, 2021). This can be accomplished by a variety of strategies concerning infrastructure, politics and  
612 technology (Jagtap et al., 2024). One way is to incorporate contingency plans into our infrastructure. The Russian invasion of  
613 Ukraine has shown that it is very difficult to change your trading partners on short notice without a plan or infrastructure (Jagtap  
614 et al., 2022) in place and that countries usually rather try to strengthen existing trade connections instead of establishing new  
615 ones (Baum et al., 2024). If plans are drawn up that highlight what is needed for different scenarios, this could be taken into  
616 account when new infrastructure is built. Also, our food system is very dependent on large amounts of industrial inputs like  
617 fertilisers or water use. This has been identified as one the main problems in agriculture right now (Foley et al., 2011). If we  
618 could reduce the need for inputs now, this would both increase sustainability, but also make it easier to cope after catastrophe  
619 when fewer inputs are available. Another important avenue is to ensure there is a variety of resilient foods that could be scaled  
620 up massively if other parts of the food system fail. Examples for ASRSs include seaweed (Jehn et al., 2024), protein from natural  
621 gas (García Martínez et al., 2022) hydrogen (García Martínez et al., 2021), sugar from fibre (Throup et al., 2022), and  
622 greenhouses (Alvarado et al., 2020). The crops we use are also adapted to current climate conditions and show very little  
623 diversity (Clapp, 2023). This low diversity in crops has recently also been highlighted as an inhibiting factor in maintaining crop  
624 production during ASRS (McLaughlin et al., 2024). Finally, establishing political agreements (for example trade agreements that  
625 also consider global catastrophes) before catastrophes could reduce the need to negotiate in the aftermath of a global catastrophe.  
626 For example, Wellesley et al. (2017) discuss this in the context of choke points that critical food corridors could be agreed upon  
627 in collaboration with the United Nations and the World Food Programme to offer alternative routes should the choke points  
628 become blocked.

#### 629 5 Conclusion

630 Our research highlights the substantial impact of global catastrophic risks on the food system, both directly through yield  
631 reductions and indirectly via trade disruptions. Among the scenarios we studied, abrupt sunlight reduction scenarios disrupt trade  
632 communities more than global catastrophic infrastructure loss due to their uneven spatial distribution, particularly affecting  
633 higher-latitude countries that are key food exporters. Our analysis focuses solely on yield reduction effects and does not consider  
634 second-order economic effects and political events. Even so, the impacts are already substantial. If second-order effects would be  
635 taken into account, it is plausible that GCIL could lead to a larger disruption, as it directly impacts the industrial base that is  
636 needed to cope with catastrophes.

637 The results show that in both kinds of scenarios, the food system would be massively disrupted, underscoring the urgent need for  
638 better preparation. The food system's reliance on a few major exporters, especially the United States, amplifies its vulnerability.  
639 This concentration means that any yield reduction or removal of these countries from the trade network results in major  
640 disruptions. We suggest diversifying crop production, securing trade agreements, and developing resilient food sources that can  
641 be rapidly scaled in crisis scenarios.



642 We need both preventive and adaptive strategies to safeguard the global food system. Future research should continue to explore  
643 these dynamics, incorporating broader aspects of the food supply chain and potential cascading effects. Such efforts are crucial,  
644 especially in light of recent global disruptions like COVID-19 and the Russian invasion of Ukraine, which have highlighted the  
645 food system's vulnerabilities. Successfully navigating global catastrophes requires understanding and preparation, necessitating  
646 both research efforts and policy interventions.

#### 647 **Author contributions**

648 Conceptualization: FUJ, JH

649 Data curation: FUJ, LGG

650 Formal analysis: FUJ, LGG

651 Funding acquisition: DD

652 Investigation: FUJ, LGG

653 Methodology: FUJ, LGG

654 Project administration: FUJ

655 Software: FUJ, LGG

656 Supervision: FUJ

657 Validation: FUJ, LGG, CWA

658 Visualization: FUJ, LGG

659 Writing-original draft: FUJ

660 Writing-review & editing: FUJ, LGG, JH, CWA, LX, NW, DD

#### 661 **Data and code availability**

662 The most recent data can be directly downloaded from the Food and Agriculture Organization:

663 1) Trade: <http://www.fao.org/faostat/en/#data/TM>

664 2) Production: <http://www.fao.org/faostat/en/#data/QC>

665 The model code (with additional documentation) can be found at: <https://github.com/allfed/pytradeshifts> (Jehn and Gajewski,  
666 2024).

#### 667 **Acknowledgements**

668 We would like to thank Juan B. García Martínez and Matt Boyd for providing valuable comments to improve this manuscript.

669

670 Special thanks to Alex van Domburg who designed and created the visual abstract.

671

672 Lili Xia is supported by U.S. National Science Foundation grants AGS-2017113 and the Future of Life Institute. Nico

673 Wunderling is grateful for funding provided by the Center for Critical Computational Studies and the Pb-TIP project at the

674 Potsdam-Institute for Climate Impact Research.

675

#### 676 **Competing interest**

677 The authors declare no competing interest.

## 678     **References**

- 679     Abdelkhalik, M., Denkenberger, D., Griswold, M., Cole, D. D., and Pearce, J.: Providing Non-food Needs if Industry is  
680     Disabled, IDRC DAVOS 2016. Integrative Risk Management - Towards Resilient Cities, 2016.
- 681     Ahvo, A., Heino, M., Sandström, V., Chrisendo, D., Jalava, M., and Kummu, M.: Agricultural input shocks affect crop yields  
682     more in the high-yielding areas of the world, *Nat Food*, 4, 1037–1046, <https://doi.org/10.1038/s43016-023-00873-z>, 2023.
- 683     Alexander, P., Brown, C., Arneth, A., Finnigan, J., Moran, D., and Rounsevell, M. D. A.: Losses, inefficiencies and waste in the  
684     global food system, *Agricultural Systems*, 153, 190–200, <https://doi.org/10.1016/j.agsy.2017.01.014>, 2017.
- 685     Alvarado, K. A., Mill, A., Pearce, J. M., Vocaet, A., and Denkenberger, D.: Scaling of greenhouse crop production in low  
686     sunlight scenarios, *Science of The Total Environment*, 707, 136012, <https://doi.org/10.1016/j.scitotenv.2019.136012>, 2020.
- 687     Anderson, W., Baethgen, W., Capitanio, F., Ciais, P., Cook, B. I., Cunha, C. G. R. da, Goddard, L., Schauburger, B., Sonder, K.,  
688     Podestá, G., van der Velde, M., and You, L.: Climate variability and simultaneous breadbasket yield shocks as observed in long-  
689     term yield records, *Agricultural and Forest Meteorology*, 331, 109321, <https://doi.org/10.1016/j.agrformet.2023.109321>, 2023.
- 690     Arnscheidt, C. W., Kemp, L., and Lenton, T.: Climate economic methods do not capture worst-case risk, 2024.
- 691     Bailey, R. and Wellesley, L.: Chokepoints and Vulnerabilities in Global Food Trade, 2017.
- 692     Barrett, A. M., Baum, S. D., and Hostetler, K.: Analyzing and Reducing the Risks of Inadvertent Nuclear War Between the  
693     United States and Russia, *Science & Global Security*, 2013.
- 694     Baum, S., Laber, M., Bruckner, M., Yang, L., Thurner, S., and Klimek, P.: Adaptive Shock Compensation in the Multi-layer  
695     Network of Global Food Production and Trade, <https://doi.org/10.48550/arXiv.2411.03502>, 21 November 2024.
- 696     Baum, S. D.: Assessing natural global catastrophic risks, *Nat Hazards*, 115, 2699–2719, [https://doi.org/10.1007/s11069-022-](https://doi.org/10.1007/s11069-022-05660-w)  
697     05660-w, 2023.
- 698     Baum, S. D., Maher, T. M., and Haqq-Misra, J.: Double catastrophe: intermittent stratospheric geoengineering induced by  
699     societal collapse, *Environ Syst Decis*, 33, 168–180, <https://doi.org/10.1007/s10669-012-9429-y>, 2013.
- 700     Bernard de Raymond, A., Alpha, A., Ben-Ari, T., Daviron, B., Nesme, T., and Tétart, G.: Systemic risk and food security.  
701     Emerging trends and future avenues for research, *Global Food Security*, 29, 100547, <https://doi.org/10.1016/j.gfs.2021.100547>,  
702     2021.
- 703     Blondel, V. D., Guillaume, J.-L., Lambiotte, R., and Lefebvre, E.: Fast unfolding of communities in large networks, *J. Stat.*  
704     *Mech.*, 2008, P10008, <https://doi.org/10.1088/1742-5468/2008/10/P10008>, 2008.
- 705     Bostrom, N. and Cirkovic, M. M.: *Global Catastrophic Risks*, Oxford University Press, Oxford, Oxford, 577 pp., 2008.
- 706     Caparas, M., Zobel, Z., Castanho, A. D. A., and Schwalm, C. R.: Increasing risks of crop failure and water scarcity in global  
707     breadbaskets by 2030, *Environ. Res. Lett.*, 16, 104013, <https://doi.org/10.1088/1748-9326/ac22c1>, 2021.
- 708     Centeno, M. A., Nag, M., Patterson, T. S., Shaver, A., and Windawi, A. J.: The Emergence of Global Systemic Risk, *Annual*

709 Review of Sociology, 41, 65–85, <https://doi.org/10.1146/annurev-soc-073014-112317>, 2015.

710 Centeno, M. A., Callahan, P. W., Larcey, P. A., and Patterson, T.: Globalization and Fragility: A Systems Approach to Collapse,  
711 in: *How Worlds Collapse*, Routledge, 2023.

712 Chapman, C. R. and Morrison, D.: Impacts on the Earth by asteroids and comets: Assessing the hazard, *Nature*, 367, 33–40,  
713 <https://doi.org/10.1038/367033a0>, 1994.

714 Clapp, J.: Concentration and crises: exploring the deep roots of vulnerability in the global industrial food system, *The Journal of*  
715 *Peasant Studies*, 50, 1–25, <https://doi.org/10.1080/03066150.2022.2129013>, 2023.

716 Clapp, J. and Moseley, W. G.: This food crisis is different: COVID-19 and the fragility of the neoliberal food security order, *The*  
717 *Journal of Peasant Studies*, 47, 1393–1417, <https://doi.org/10.1080/03066150.2020.1823838>, 2020.

718 Cliver, E. W., Schrijver, C. J., Shibata, K., and Usoskin, I. G.: Extreme solar events, *Living Rev Sol Phys*, 19, 2,  
719 <https://doi.org/10.1007/s41116-022-00033-8>, 2022.

720 Cooper, C. and Sovacool, B. K.: Not Your Father’s Y2K: Preparing the North American Power Grid for the Perfect Solar Storm,  
721 *The Electricity Journal*, 24, 47–61, <https://doi.org/10.1016/j.tej.2011.04.005>, 2011.

722 Cotton-Barratt, O., Daniel, M., and Sandberg, A.: Defence in Depth Against Human Extinction: Prevention, Response,  
723 Resilience, and Why They All Matter, *Global Policy*, 11, 271–282, <https://doi.org/10.1111/1758-5899.12786>, 2020.

724 Coupe, J., Bardeen, C. G., Robock, A., and Toon, O. B.: Nuclear Winter Responses to Nuclear War Between the United States  
725 and Russia in the Whole Atmosphere Community Climate Model Version 4 and the Goddard Institute for Space Studies ModelE,  
726 *Journal of Geophysical Research: Atmospheres*, 124, 8522–8543, <https://doi.org/10.1029/2019JD030509>, 2019.

727 Coupe, J., Harrison, C., Robock, A., DuVivier, A., Maroon, E., Lovenduski, N. S., Bachman, S., Landrum, L., and Bardeen, C.:  
728 Sudden Reduction of Antarctic Sea Ice Despite Cooling After Nuclear War, *Journal of Geophysical Research: Oceans*, 128,  
729 e2022JC018774, <https://doi.org/10.1029/2022JC018774>, 2023.

730 De Domenico, M.: More is different in real-world multilayer networks, *Nature Physics*, 19, 1247–1262,  
731 <https://doi.org/10.1038/s41567-023-02132-1>, 2023.

732 Denkenberger, D., Sandberg, A., Tieman, R. J., and Pearce, J. M.: Long-term cost-effectiveness of interventions for loss of  
733 electricity/industry compared to artificial general intelligence safety, *European Journal of Futures Research*, 9, 11,  
734 <https://doi.org/10.1186/s40309-021-00178-z>, 2021.

735 Denkenberger, D., Sandberg, A., Tieman, R. J., and Pearce, J. M.: Long term cost-effectiveness of resilient foods for global  
736 catastrophes compared to artificial general intelligence safety, *International Journal of Disaster Risk Reduction*, 73, 102798,  
737 <https://doi.org/10.1016/j.ijdr.2022.102798>, 2022.

738 Deteix, L., Salou, T., and Loiseau, E.: Quantifying food consumption supply risk: An analysis across countries and agricultural  
739 products, *Global Food Security*, 41, 100764, <https://doi.org/10.1016/j.gfs.2024.100764>, 2024.

740 D’Odorico, P., Carr, J. A., Laio, F., Ridolfi, L., and Vandoni, S.: Feeding humanity through global food trade, *Earth’s Future*, 2,  
741 458–469, <https://doi.org/10.1002/2014EF000250>, 2014.

Enger, E. R., Graversen, R. G., and Theodorsen, A.: Radiative forcing by super-volcano eruptions, <https://doi.org/10.22541/au.170967725.56351509/v1>, 5 March 2024.

Foley, J. A., Ramankutty, N., Brauman, K. A., Cassidy, E. S., Gerber, J. S., Johnston, M., Mueller, N. D., O'Connell, C., Ray, D. K., West, P. C., Balzer, C., Bennett, E. M., Carpenter, S. R., Hill, J., Monfreda, C., Polasky, S., Rockström, J., Sheehan, J., Siebert, S., Tilman, D., and Zaks, D. P. M.: Solutions for a cultivated planet, *Nature*, 478, 337–342, <https://doi.org/10.1038/nature10452>, 2011.

Foti, N. J., Pauls, S., and Rockmore, D. N.: Stability of the World Trade Web over time – An extinction analysis, *Journal of Economic Dynamics and Control*, 37, 1889–1910, 2013.

Frame, D., Joshi, M., Hawkins, E., Harrington, L. J., and de Roiste, M.: Population-based emergence of unfamiliar climates, *Nature Clim Change*, 7, 407–411, <https://doi.org/10.1038/nclimate3297>, 2017.

García Martínez, J. B., Egbejimba, J., Throup, J., Matassa, S., Pearce, J. M., and Denkenberger, D. C.: Potential of microbial protein from hydrogen for preventing mass starvation in catastrophic scenarios, *Sustainable Production and Consumption*, 25, 234–247, <https://doi.org/10.1016/j.spc.2020.08.011>, 2021.

García Martínez, J. B., Pearce, J. M., Throup, J., Cates, J., Lackner, M., and Denkenberger, D. C.: Methane Single Cell Protein: Potential to Secure a Global Protein Supply Against Catastrophic Food Shocks, *Frontiers in Bioengineering and Biotechnology*, 10, 2022.

Gaupp, F., Hall, J., Hochrainer-Stigler, S., and Dadson, S.: Changing risks of simultaneous global breadbasket failure, *Nat. Clim. Chang.*, 10, 54–57, <https://doi.org/10.1038/s41558-019-0600-z>, 2020.

Glomseth, R. E.: Resilience Beyond Food Production - Trade and Supply Chains in the Nuclear Winter Context, Zenodo, <https://doi.org/10.5281/zenodo.10720784>, 2024.

Goldin, I. and Vogel, T.: Global Governance and Systemic Risk in the 21st Century: Lessons from the Financial Crisis, *Global Policy*, 1, 4–15, <https://doi.org/10.1111/j.1758-5899.2009.00011.x>, 2010.

Green, W.: Nuclear War Impacts on Distant, Non-Combatant Countries, Toda Peace Institute, 2024.

Guimerà, R. and Nunes Amaral, L. A.: Functional cartography of complex metabolic networks, *Nature*, 433, 895–900, <https://doi.org/10.1038/nature03288>, 2005.

Hagberg, A., Schult, D., Swart, P., and Hagberg, J. M.: Exploring Network Structure, Dynamics, and Function using NetworkX, 2008.

Hamilton, H., Henry, R., Rounsevell, M., Moran, D., Cossar, F., Allen, K., Boden, L., and Alexander, P.: Exploring global food system shocks, scenarios and outcomes, *Futures*, 123, 102601, <https://doi.org/10.1016/j.futures.2020.102601>, 2020.

Hedlund, J., Carlsen, H., Croft, S., West, C., Bodin, Ö., Stokeld, E., Jägermeyr, J., and Müller, C.: Impacts of climate change on global food trade networks, *Environmental Research Letters*, 17, <https://doi.org/10.1088/1748-9326/aca68b>, 2022.

Helbing, D.: Globally networked risks and how to respond, *Nature*, 497, 51–59, <https://doi.org/10.1038/nature12047>, 2013.

774 Hochman, G., Zhang, H., Xia, L., Robock, A., Saketh, A., Mensbrughe, D. Y. van der, and Jägermeyr, J.: Economic incentives  
775 modify agricultural impacts of nuclear war, *Environ. Res. Lett.*, 17, 054003, <https://doi.org/10.1088/1748-9326/ac61c7>, 2022.

776 Hoyer, D., Bennett, J. S., Reddish, J., Holder, S., Howard, R., Benam, M., Levine, J., Ludlow, F., Feinman, G., and Turchin, P.:  
777 Navigating polycrisis: long-run socio-cultural factors shape response to changing climate, *Philosophical Transactions of the*  
778 *Royal Society B: Biological Sciences*, 378, 20220402, <https://doi.org/10.1098/rstb.2022.0402>, 2023.

779 Hoyer, D., Holder, S., Bennett, J. S., François, P., Whitehouse, H., Covey, A., Feinman, G., Korotayev, A., Vustiuzhanin, V.,  
780 Preiser-Kapeller, J., Bard, K., Levine, J., Reddish, J., Orlandi, G., Ainsworth, R., and Turchin, P.: All Crises are Unhappy in their  
781 Own Way: The role of societal instability in shaping the past, <https://doi.org/10.31235/osf.io/rk4gd>, 17 February 2024.

782 Ibrahim, S. E., Centeno, M. A., Patterson, T. S., and Callahan, P. W.: Resilience in Global Value Chains: A Systemic Risk  
783 Approach, *Global Perspectives*, 2, <https://doi.org/10.1525/gp.2021.27658>, 2021.

784 Isobe, H., Takahashi, T., Seki, D., and Yamashiki, Y.: Extreme Solar Flare as a Catastrophic Risk, *Journal of Disaster Research*,  
785 17, 230–236, <https://doi.org/10.20965/jdr.2022.p0230>, 2022.

786 Jaccard, P.: Etude de la distribution florale dans une portion des Alpes et du Jura, *Bulletin de la Societe Vaudoise des Sciences*  
787 *Naturelles*, 37, 547–579, <https://doi.org/10.5169/seals-266450>, 1901.

788 Jagtap, S., Trollman, H., Trollman, F., Garcia-Garcia, G., Parra-López, C., Duong, L., Martindale, W., Munekata, P. E. S.,  
789 Lorenzo, J. M., Hdaifeh, A., Hassoun, A., Salontis, K., and Afy-Shararah, M.: The Russia-Ukraine Conflict: Its Implications for  
790 the Global Food Supply Chains, *Foods*, 11, 2098, <https://doi.org/10.3390/foods11142098>, 2022.

791 Jagtap, S., Trollman, H., Trollman, F., Garcia-Garcia, G., and Martindale, W.: Surviving the storm: navigating the quadruple  
792 whammy impact on Europe’s food supply chain, *Int J of Food Sci Tech*, *ijfs*.17106, <https://doi.org/10.1111/ijfs.17106>, 2024.

793 Jahn, M.: How “Multiple Breadbasket Failure” Became a Policy Issue, *Issues in Science and Technology*, 2021.

794 Janssens, C., Havlík, P., Krisztin, T., Baker, J., Frank, S., Hasegawa, T., Leclère, D., Ohrel, S., Ragnauth, S., Schmid, E., Valin,  
795 H., Van Lipzig, N., and Maertens, M.: Global hunger and climate change adaptation through international trade, *Nat. Clim.*  
796 *Chang.*, 10, 829–835, <https://doi.org/10.1038/s41558-020-0847-4>, 2020.

797 Jehn, F. U.: Anthropocene Under Dark Skies: The Compounding Effects of Nuclear Winter and Overstepped Planetary  
798 Boundaries, in: *Intersections, Reinforcements, Cascades: Proceedings of the 2023 Stanford Existential Risks Conference*,  
799 119–132, <https://doi.org/10.25740/zb109mz2513>, 2023.

800 Jehn, F. U. and Gajewski, Ł. G.: *allfed/pytradeshifts*: Feature complete, , <https://doi.org/10.5281/ZENODO.10785381>, 2024.

801 Jehn, F. U., Dingal, F. J., Mill, A., Harrison, C., Ilin, E., Roleda, M. Y., James, S. C., and Denkenberger, D.: Seaweed as a  
802 Resilient Food Solution After a Nuclear War, *Earth’s Future*, 12, e2023EF003710, <https://doi.org/10.1029/2023EF003710>, 2024.

803 Ji, G., Zhong, H., Feukam Nzudie, H. L., Wang, P., and Tian, P.: The structure, dynamics, and vulnerability of the global food  
804 trade network, *Journal of Cleaner Production*, 434, 140439, <https://doi.org/10.1016/j.jclepro.2023.140439>, 2024.

805 Kang, H., Lee, K.-M., and Yang, J.-S.: The potential for cascading failures in the international trade network, *PLOS ONE*, 19,  
806 e0299833, <https://doi.org/10.1371/journal.pone.0299833>, 2024.

807 Karger, E., J. R., Jacobs, Z., Hickman, M., Hadshar, R., Gamin, K., Smith, T., Williams, B., McCaslin, T., and Tetlock, P.:  
808 Forecasting Existential Risks: Evidence from a Long-Run Forecasting Tournament (FRI Working Paper #1), Forecasting  
809 Research Institute, 2023.

810 Key, R., Parrado, R., Delpiazzi, E., King, R., and Bosello, F.: Potential climate-induced impacts on trade: the case of agricultural  
811 commodities and maritime chokepoints, *Journal of Shipping and Trade*, 9, 11, <https://doi.org/10.1186/s41072-024-00170-3>,  
812 2024.

813 Kivelä, M., Arenas, A., Barthelemy, M., Gleeson, J. P., Moreno, Y., and Porter, M. A.: Multilayer networks, *Journal of Complex*  
814 *Networks*, 2, 203–271, <https://doi.org/10.1093/comnet/cnu016>, 2014.

815 Kornhuber, K., Lesk, C., Schleussner, C. F., Jägermeyr, J., Pfleiderer, P., and Horton, R. M.: Risks of synchronized low yields  
816 are underestimated in climate and crop model projections, *Nat Commun*, 14, 3528, <https://doi.org/10.1038/s41467-023-38906-7>,  
817 2023.

818 Linkov, I., Galaitsi, S. E., Trump, B. D., Pinigina, E., Rand, K., Cline, E. H., and Kitsak, M.: Are civilizations destined to  
819 collapse? Lessons from the Mediterranean Bronze Age, *Global Environmental Change*, 84, 102792,  
820 <https://doi.org/10.1016/j.gloenvcha.2023.102792>, 2024.

821 Ma, J., Li, M., and Li, H.-J.: Robustness of the International Wheat Trade Network, *IEEE Transactions on Network Science and*  
822 *Engineering*, 1–11, <https://doi.org/10.1109/TNSE.2023.3283251>, 2023.

823 Mamonova, N., Wengle, S., and Dankevych, V.: Queen of the fields in wartime: What can Ukrainian corn tell us about the  
824 resilience of the global food system?, *The Journal of Peasant Studies*, 50, 2513–2538,  
825 <https://doi.org/10.1080/03066150.2023.2255568>, 2023.

826 Mani, L., Tzachor, A., and Cole, P.: Global catastrophic risk from lower magnitude volcanic eruptions, *Nat Commun*, 12, 4756,  
827 <https://doi.org/10.1038/s41467-021-25021-8>, 2021.

828 McLaughlin, C., Shi, Y., Viswanathan, V., Leonard, L., Sawers, R. J., Kemanian, A., and Lasky, J. R.: Maladaptation in cereal  
829 crop landraces following a soot-producing climate catastrophe, <https://doi.org/10.1101/2024.05.18.594591>, 21 May 2024.

830 Méjean, A., Collins-Sowah, P., Guivarch, C., Piontek, F., Soergel, B., and Taconet, N.: Climate change impacts increase  
831 economic inequality: evidence from a systematic literature review, *Environ. Res. Lett.*, 19, 043003, <https://doi.org/10.1088/1748-9326/ad376e>, 2024.

833 Miller, S. J., Dee, L. E., Hayden, M. T., Jarrett, U., Carrico, A. R., Brauman, K. A., and Aceves-Bueno, E.: Telecoupled systems  
834 are rewired by risks, *Nat Sustain*, 1–8, <https://doi.org/10.1038/s41893-024-01273-2>, 2024.

835 Moersdorf, J., Rivers, M., Denkenberger, D., Breuer, L., and Jehn, F. U.: The Fragile State of Industrial Agriculture: Estimating  
836 Crop Yield Reductions in a Global Catastrophic Infrastructure Loss Scenario, *Global Challenges*, 8, 2300206,  
837 <https://doi.org/10.1002/gch2.202300206>, 2024.

838 Myers, S., Fanzo, J., Wiebe, K., Huybers, P., and Smith, M.: Current guidance underestimates risk of global environmental  
839 change to food security, *BMJ*, 378, e071533, <https://doi.org/10.1136/bmj-2022-071533>, 2022.

840 Nelson, D., Turchin, A., and Denkenberger, D.: Wood Gasification: A Promising Strategy to Extend Fuel Reserves after Global  
841 Catastrophic Electricity Loss, *Biomass*, 4, 610–624, <https://doi.org/10.3390/biomass4020033>, 2024.

842 Nyström, M., Jouffray, J.-B., Norström, A. V., Crona, B., Søgaaard Jørgensen, P., Carpenter, S. R., Bodin, Ö., Galaz, V., and  
843 Folke, C.: Anatomy and resilience of the global production ecosystem, *Nature*, 575, 98–108, [https://doi.org/10.1038/s41586-019-](https://doi.org/10.1038/s41586-019-1712-3)  
844 1712-3, 2019.

845 Ogie, R. I.: Cyber Security Incidents on Critical Infrastructure and Industrial Networks, in: *Proceedings of the 9th International*  
846 *Conference on Computer and Automation Engineering*, New York, NY, USA, 254–258,  
847 <https://doi.org/10.1145/3057039.3057076>, 2017.

848 Paluch, R., Gajewski, Ł. G., Suchecki, K., and Hołyst, J. A.: Impact of interactions between layers on source localization in  
849 multilayer networks, *Physica A: Statistical Mechanics and its Applications*, 582, 126238,  
850 <https://doi.org/10.1016/j.physa.2021.126238>, 2021.

851 Pham, A., García Martínez, J. B., Brynych, V., Stormbjorne, R., Pearce, J. M., and Denkenberger, D. C.: Nutrition in Abrupt  
852 Sunlight Reduction Scenarios: Envisioning Feasible Balanced Diets on Resilient Foods, *Nutrients*, 14, 492,  
853 <https://doi.org/10.3390/nu14030492>, 2022.

854 Puma, M. J., Bose, S., Chon, S. Y., and Cook, B. I.: Assessing the evolving fragility of the global food system, *Environ. Res.*  
855 *Lett.*, 10, 024007, <https://doi.org/10.1088/1748-9326/10/2/024007>, 2015.

856 Quante, L., Willner, S. N., Otto, C., and Levermann, A.: Global economic impact of weather variability on the rich and the poor,  
857 *Nat Sustain*, 7, 1419–1428, <https://doi.org/10.1038/s41893-024-01430-7>, 2024.

858 Rampino, M. R.: Supereruptions as a Threat to Civilizations on Earth-like Planets, *Icarus*, 156, 562–569,  
859 <https://doi.org/10.1006/icar.2001.6808>, 2002.

860 Reis, T. N. P. dos, Meyfroidt, P., zu Ermgassen, E. K. H. J., West, C., Gardner, T., Bager, S., Croft, S., Lathuillière, M. J., and  
861 Godar, J.: Understanding the Stickiness of Commodity Supply Chains Is Key to Improving Their Sustainability, *One Earth*, 3,  
862 100–115, <https://doi.org/10.1016/j.oneear.2020.06.012>, 2020.

863 Rivers, M., Hinge, M., Rassool, K., Blouin, S., Jehn, F. U., García Martínez, J. B., Amaral Grilo, V., Jaeck, V., Tieman, R.,  
864 Mulhall, J., Butt, T., and Denkenberger, D.: Food System Adaptation and Maintaining Trade Could Mitigate Global Famine in  
865 Abrupt Sunlight Reduction Scenarios, <https://doi.org/10.5281/zenodo.11484350>, 5 June 2024a.

866 Rivers, M., Gajewski, Ł. G., and Denkenberger, D.: Global transformer overheating from geomagnetic storms,  
867 <https://doi.org/10.48550/arXiv.2403.18070>, 26 March 2024b.

868 Rivington, M., Bailey, R., Benton, T., Challinor, A., Elliott, J., Gustafson, D., Hiller, B., Jones, A., Jahn, M., Kent, C., Lewis, K.,  
869 Meacham, T., Robson, D., Tiffin, R., and Wuebbles, D.: Extreme weather and resilience of the global food system - Synthesis  
870 Report, Foreign and Commonwealth Office / UK Science and Innovation Network / Global Food Security, 2015.

871 Rougier, J., Sparks, R. S. J., Cashman, K. V., and Brown, S. K.: The global magnitude–frequency relationship for large explosive  
872 volcanic eruptions, *Earth and Planetary Science Letters*, 482, 621–629, <https://doi.org/10.1016/j.epsl.2017.11.015>, 2018.

873 Sandström, V., Huan-Niemi, E., Niemi, J., and Kummu, M.: Dependency on imported agricultural inputs - global trade patterns  
874 and recent trends, *Environ. Res.: Food Syst.*, <https://doi.org/10.1088/2976-601X/ad325e>, 2024.

875 Savary, S., Akter, S., Almekinders, C., Harris, J., Korsten, L., Rötter, R., Waddington, S., and Watson, D.: Mapping disruption  
876 and resilience mechanisms in food systems, *Food Sec.*, 12, 695–717, <https://doi.org/10.1007/s12571-020-01093-0>, 2020.

877 Scherrer, K. J. N., Harrison, C. S., Heneghan, R. F., Galbraith, E., Bardeen, C. G., Coupe, J., Jägermeyr, J., Lovenduski, N. S.,  
878 Luna, A., Robock, A., Stevens, J., Stevenson, S., Toon, O. B., and Xia, L.: Marine wild-capture fisheries after nuclear war,  
879 *Proceedings of the National Academy of Sciences*, 117, 29748–29758, <https://doi.org/10.1073/pnas.2008256117>, 2020.

880 Spaiser, V., Juhola, S., Constantino, S. M., Guo, W., Watson, T., Sillmann, J., Craparo, A., Basel, A., Bruun, J. T.,  
881 Krishnamurthy, K., Scheffran, J., Pinho, P., Okpara, U. T., Donges, J. F., Bhowmik, A., Yasseri, T., Safrá de Campos, R.,  
882 Cumming, G. S., Chenet, H., Krampe, F., and Abrams, J. F.: Negative Social Tipping Dynamics Resulting from and Reinforcing  
883 Earth System Destabilisation, <https://doi.org/10.5194/egusphere-2023-1475>, 4 September 2023.

884 Tabor, C. R., Bardeen, C. G., Otto-Bliesner, B. L., Garcia, R. R., and Toon, O. B.: Causes and Climatic Consequences of the  
885 Impact Winter at the Cretaceous-Paleogene Boundary, *Geophysical Research Letters*, 47, e60121,  
886 <https://doi.org/10.1029/2019GL085572>, 2020.

887 Thang, D. N.: How do regional extreme events shape supply-chain trade?, *Int Econ Econ Policy*, 21, 117–149,  
888 <https://doi.org/10.1007/s10368-023-00582-9>, 2024.

889 Throup, J., García Martínez, J. B., Bals, B., Cates, J., Pearce, J. M., and Denkenberger, D. C.: Rapid repurposing of pulp and  
890 paper mills, biorefineries, and breweries for lignocellulosic sugar production in global food catastrophes, *Food and Bioproducts*  
891 *Processing*, 131, 22–39, <https://doi.org/10.1016/j.fbp.2021.10.012>, 2022.

892 Toon, O. B., Robock, A., and Turco, R. P.: Environmental consequences of nuclear war, *Physics today*, 61, 37, 2008.

893 Tukey, J. W.: *Exploratory Data Analysis*, 1st edition., Pearson, Reading, Mass, 503 pp., 1977.

894 Wang, X., Ma, L., Yan, S., Chen, X., and Grove, A.: Trade for Food Security: The Stability of Global Agricultural Trade  
895 Networks, *Foods*, 12, 271, <https://doi.org/10.3390/foods12020271>, 2023.

896 Wellesley, L., Preston, F., Lehne, J., and Bailey, R.: Chokepoints in global food trade: Assessing the risk, *Research in*  
897 *Transportation Business & Management*, 25, 15–28, <https://doi.org/10.1016/j.rtbm.2017.07.007>, 2017.

898 White, S.: *The climate of rebellion in the early modern Ottoman Empire*, Cambridge University Press, Cambridge, 2013.

899 Wieland, A. and Durach, C. F.: Two perspectives on supply chain resilience, *Journal of Business Logistics*, 42, 315–322,  
900 <https://doi.org/10.1111/jbl.12271>, 2021.

901 Wiliński, M., Sienkiewicz, A., Gubiec, T., Kutner, R., and Struzik, Z. R.: Structural and topological phase transitions on the  
902 German Stock Exchange, *Physica A: Statistical Mechanics and its Applications*, 392, 5963–5973,  
903 <https://doi.org/10.1016/j.physa.2013.07.064>, 2013.

904 Wilson, C.: *High Altitude Electromagnetic Pulse (HEMP) and High Power Microwave (HPM) Devices: Threat Assessments*,  
905 *Defense Technical Information Center*, 26 pp., 2008.



906 Woo, G.: Downward Counterfactual Search for Extreme Events, *Front. Earth Sci.*, 7, <https://doi.org/10.3389/feart.2019.00340>,  
907 2019.

908 World Economic Forum: The global risks report 2023, 18th ed., World Economic Forum, Geneva, 2023.

909 Xia, L., Robock, A., Scherrer, K., Harrison, C. S., Bodirsky, B. L., Weindl, I., Jägermeyr, J., Bardeen, C. G., Toon, O. B., and  
910 Heneghan, R.: Global food insecurity and famine from reduced crop, marine fishery and livestock production due to climate  
911 disruption from nuclear war soot injection, *Nat Food*, 1–11, <https://doi.org/10.1038/s43016-022-00573-0>, 2022.

912 Yıldırım, C. and Önen, H. G.: Vulnerabilities of the neoliberal global food system: The Russia–Ukraine War and COVID-19,  
913 *Journal of Agrarian Change*, n/a, e12601, <https://doi.org/10.1111/joac.12601>, 2024.

914 Zhang, J., Wu, T., Li, L., Furtado, K., Xin, X., Xie, C., Zheng, M., Zhao, H., and Zhou, Y.: Constraint on regional land surface  
915 air temperature projections in CMIP6 multi-model ensemble, *npj Clim Atmos Sci*, 6, 1–9, [https://doi.org/10.1038/s41612-023-](https://doi.org/10.1038/s41612-023-00410-6)  
916 00410-6, 2023a.

917 Zhang, Y.-T., Nguyen, D. K., and Zhou, W.-X.: Spatiotemporal characteristics of agricultural food import shocks,  
918 <https://doi.org/10.48550/arXiv.2303.00919>, 1 March 2023b.

919 Zhang, Y.-T., Li, M.-Y., and Zhou, W.-X.: Impact of the Russia-Ukraine conflict on the international staple agrifood trade  
920 networks, <https://doi.org/10.48550/arXiv.2403.12496>, 19 March 2024.

921

922