








# 1 Food trade disruption after global catastrophes

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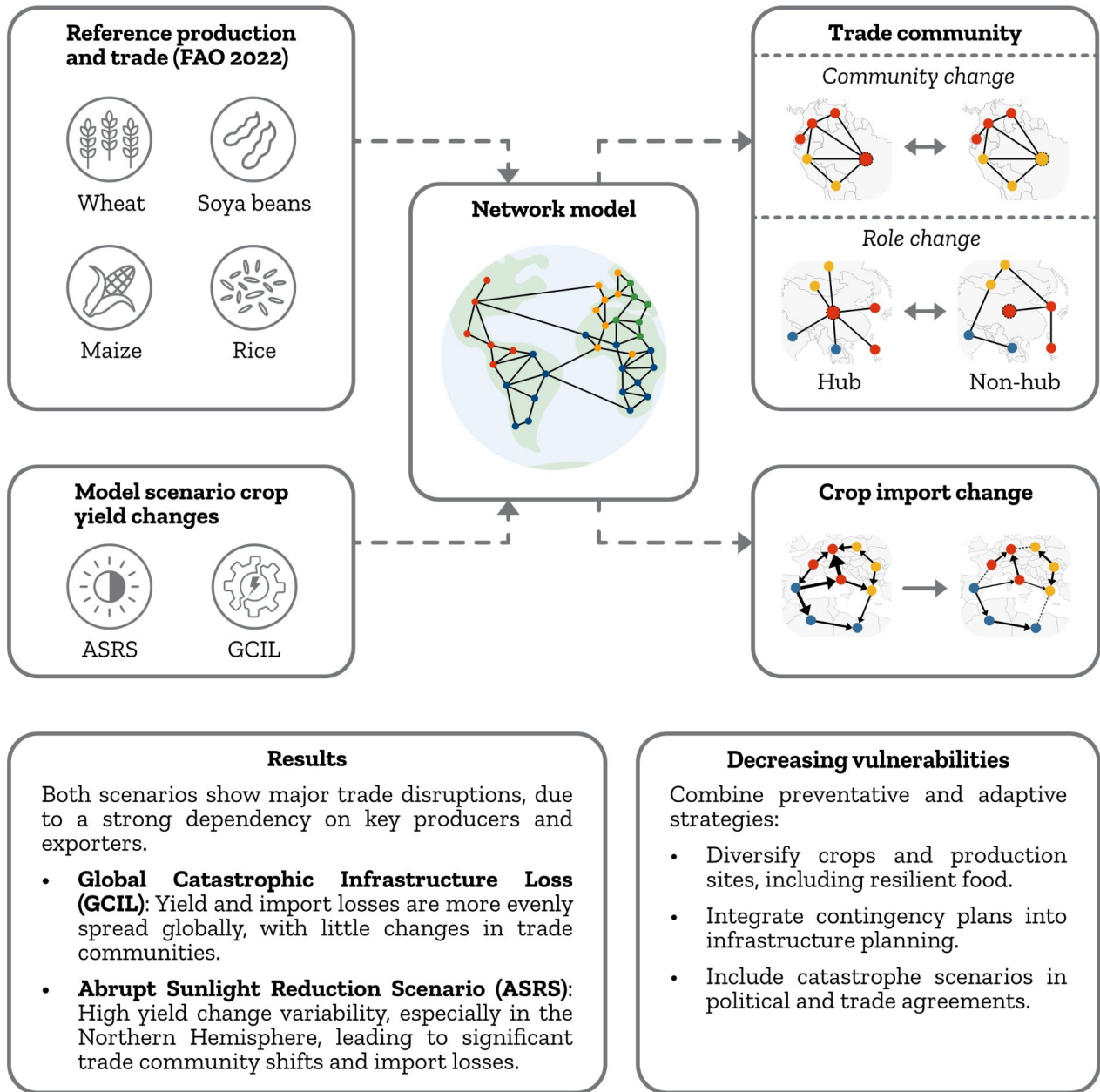
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15 **Abstract.** The global food trade system is resilient to minor disruptions but vulnerable to major ones. Major shocks can arise from  
16 global catastrophic risks, such as abrupt sunlight reduction scenarios (e.g., nuclear war) or global catastrophic infrastructure loss  
17 (e.g., due to severe geomagnetic storms or a global pandemic). We use a network model to examine how these two scenarios could  
18 impact global food trade, focusing on wheat, maize, soybeans, and rice, accounting for about 60% of global calorie intake. Our  
19 findings indicate that an abrupt sunlight reduction scenario, with soot emissions equivalent to a major nuclear war between India  
20 and Pakistan (37 Tg), could severely disrupt trade, causing most countries to lose the vast majority of their food imports (50-  
21 100 % decrease), primarily due to the main exporting countries being heavily affected. Global catastrophic infrastructure loss [with](#)  
22 [a comparable impact on yields of the same magnitude](#) as the abrupt sunlight reduction has a more homogeneous distribution of  
23 yield declines, resulting in most countries losing up to half of their food imports (25-50 % decrease). Thus, our analysis shows that  
24 both scenarios could significantly impact the food trade. However, the abrupt sunlight reduction scenario is likely more disruptive  
25 than global catastrophic infrastructure loss regarding the effects of yield reductions on food trade. This study underscores the  
26 vulnerabilities of the global food trade network to catastrophic risks and the need for enhanced preparedness.

27



## 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; Spaiser 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

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

76  
77 While the impacts of climate change and extreme events on trade have been studied more in recent years (Hedlund et al., 2022;  
78 Thang, 2024), only limited research has been conducted regarding the effects of GCIL and ASRS on food production and trade.  
79 The research that does exist assumes that trade will continue as it is now or cease completely (Hochman et al., 2022; Rivers et  
80 al., 2024a; Xia et al., 2022). These simplifications reduce the enormous complexity of how our food system might react to global  
81 catastrophic risks. While some preliminary economic research on smaller nuclear conflicts has been conducted (Hochman et al.,  
82 2022), ~~broader insight, especially into the consequences of a wider range of scenarios, is needed economic models struggle in~~  
83 ~~modelling extreme shocks, such as those associated with ASRSs or global catastrophic infrastructure loss, as they do not account~~  
84 ~~for the direct destruction, sudden and big changes, as well as loss of life and other effects of global catastrophes (Arnscheidt et~~  
85 ~~al., 2024).~~

86  
87 For an initial assessment of how global trade might evolve after such global catastrophes, we study the shifts of trade  
88 communities and trade flows caused by GCIL and ASRS in a global food trade network model (Hedlund et al., 2022). In this  
89 context, trade communities refer to groups of countries that trade extensively with one another. Understanding them and their  
90 changes allows a more targeted assessment of the disruptions caused by changes in yield. The model is intentionally simple,  
91 focusing on the direct effects of yield changes on trade without considering second-order economic aspects. Our initial analysis  
92 can serve as a foundation for future, more detailed economic assessments, while the model itself offers policymakers and  
93 scientists a practical tool to analyse the direct effects of food production shocks on global trade. Such assessments are important  
94 because they advance our understanding of how global catastrophes impact food trade, revealing the different implications of  
95 various shocks to the system. By modelling these shocks under different scenarios, we can better understand and predict changes  
96 in the global food trade system after major disruptions.

## 97 **2 Methods**

### 98 **2.1 Model setup**

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

110  
111 To detect the communities in the trade network, we used the Louvain algorithm (Blondel et al., 2008), as implemented in  
112 NetworkX (Hagberg et al., 2008). It assigns every country a trade community, i.e., a group of other countries with which said

113 country has the closest trade ties. As the Louvain algorithm is not deterministic, our model can be provided with a random seed  
114 parameter to ensure the reproducibility of the results. ¶

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

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

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

122 ¶

123 In the model, we accounted for re-exports to represent point-of-origin-to-point-of-destination trade movements, meaning that the  
124 resulting data only contain the direct trade between countries without intermediaries (more information about this in supplement  
125 section 2.2 and Hedlund et al. (2022)).

## 126 **2.2 Production and trade data**

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

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

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

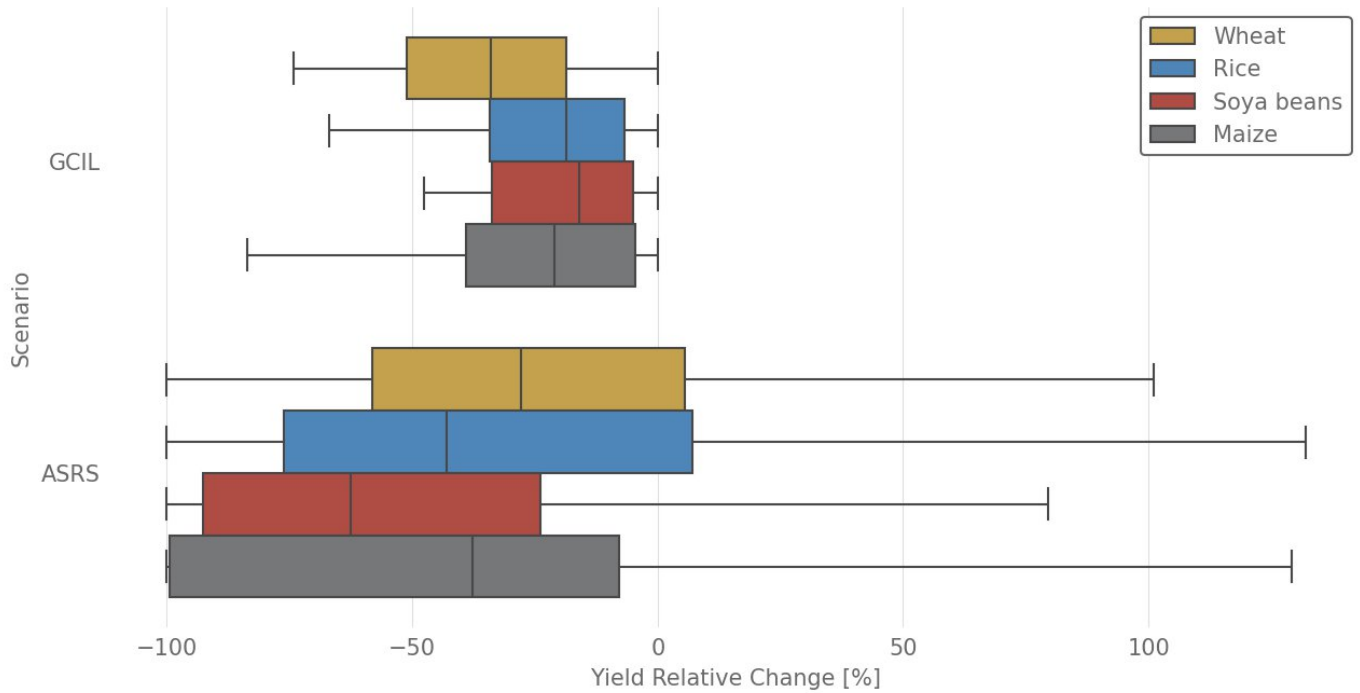
## 141 **2.3 The impact of global catastrophic risk scenarios on yields**

142 We focus on two main GCR scenarios: GCIL and ASRS (see introduction). We obtained yield losses for GCIL scenarios from  
143 Moersdorf et al. (2024). Moersdorf et. al (2024) assumed that if a GCIL happens, this will result in a global stop in the  
144 production of agricultural inputs like fuel, pesticides and fertilisers. Based on this they split their simulations into two phases.  
145 Phase 1 is the first year after GCIL with some stocks for fuel, pesticides and fertilisers remaining, while phase 2 simulates all  
146 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  
147 only available on a global (with a 5 arcmin resolution) and continental scale, we averaged the yield losses from global data for all

148 points in each country. The resulting mean values of yield reduction differ slightly from the ones stated in Moersdorf et al.  
149 (2024) because: 1) 1) Moersdorf et al., assigned weights using by pre-catastrophe productivity, but as the nuclear war data is not  
150 productivity weighted, we used Moersdorf et al's unweighted data to ensure comparability between the two scenarios. The wider  
151 yield change distribution under ASRS compared to GCIL thus reflects genuine scenario differences rather than methodological  
152 artifacts. while we did not apply any weights to ensure comparability with the nuclear war yieldclimate data and 2) 2) In our  
153 model, the connections between countries are based on the actual amount traded (corrected for re-exports). Weighting the yield  
154 changes by their productivity would thus skew the results. Also, we aggregate on country level first instead of taking a global  
155 average. The scenario by Moersdorf et al. likely would have wide ranging consequences for society beyond yield impacts, as it  
156 assumes a disruption of the industrial base. These further disruptions are not modelled here.

157  
158 For ASRSs we used the country-level nuclear war crop modelling data from Xia et al., (2022). We used nuclear war as a proxy  
159 for all ASRSs because nuclear war has the best climate model data available (Coupe et al., 2019), and the global impact on  
160 climate is possibly similar across different ASRS scenarios with similar magnitude. We used data for the third year after the  
161 nuclear war, as this represents the year with the lowest yields. To make the scenario more comparable with the GCIL scenario,  
162 we used the 37 teragram (Tg) scenario from Xia et al. (2022) as the main comparison. This is meant to simulate a nuclear war  
163 between India and Pakistan with 250 nuclear weapons of 100 kt explosive yield each and would thus equal a total of 25  
164 megatons of TNT. In this scenario, some of the smaller and hotter countries experience increases in yield due to a better climate,  
165 and the climate model used with a horizontal resolution of 2 degrees cannot resolve such small countries correctly. Thus, we  
166 limit this effect to a maximum value compared to current yields to avoid unrealistically high values (Wheat: 100 %, Rice: 132 %,   
167 Soya Beans: 79 %, Maize: 129 %). Since more accurate crop growing models are not available for nuclear war, we determine  
168 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  
169 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  
170 as a proxy for wheat in general.

171  
172 The ASRS with 37 Tg soot emissions has a median wheat yield decline similar to GCIL (Figure 1). Soya beans, maize and rice  
173 have more dissimilar ranges (Figure 1). This makes wheat the most comparable crop across the two scenarios, while also being  
174 the most traded and, therefore, our main focus; however, we also discuss the other crops and provide the figures for them in the  
175 supplement.



176  
 177 **Figure 1: Relative yield change (%) in all affected countries (combined) for both the global catastrophic infrastructure loss (GCIL) and**  
 178 **the abrupt sunlight reduction scenario (ASRS), by crop (colour). The values for GCIL yield changes are taken from Moersdorf et al.**  
 179 **(2024), and those for ASRS yield changes from Xia et al. (2022) (see Section 2.3 for details). The boxplot displays data distribution**  
 180 **using five key summary points: the minimum, first quartile, median, third quartile, and maximum. The box spans from the first to the**  
 181 **third quartile, with a line at the median. Whiskers extend to the smallest and largest values within 1.5 times the interquartile range**  
 182 **from the quartiles. Outliers are circles beyond the whiskers. This is the same for all boxplots shown in this article.**

183 **2.4 Trade communities before and after global catastrophes**

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

188 **2.4.1 Change**

189 **Jaccard distance**

190 To assess how much the trade communities of all countries have changed before and after global catastrophes we used the  
 191 Jaccard distance. This measure allows us to compare how similar/different two trade communities are. It finds the percentage of  
 192 common countries between trade communities divided by the total number of elements between them. The Jaccard *similarity*  
 193 (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  
 194 sets, and has a range from zero to one (Jaccard, 1901). The Jaccard distance ( $d_j$ ) is one minus the Jaccard similarity. Therefore,  
 195 for any given country, we can look at the set of countries that are in the same community before and after the catastrophe and  
 196 compute the Jaccard distance (dissimilarity score) for these sets.

197  
 198 Let  $A$  denote the set of community members of some country before a catastrophe and  $A'$  the set of community members of the  
 199 same country after the catastrophe. We can then define the Jaccard distance  $d_j$  as:

200

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

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

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

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 of communities.

These indices define a parameter space where different regions correspond to specific roles based on threshold values. Guimerà and Nunes Amaral identified seven node roles:

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



- **Kinless hubs** (evenly distributed links across all communities,  $P>0.75$ ).

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

## 2.4.2 Centrality

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

## 3 Results

### 3.1 Changes in wheat trade

#### 3.1.1 Shifts in trade communities

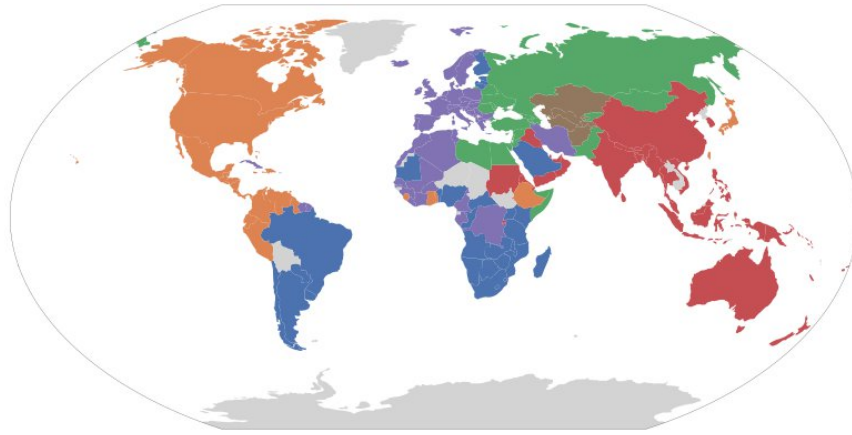
According to our modelling, the wheat trading communities (based on the Louvain algorithm, section 2.1) would evolve differently during GCIL and ASRS. This can be seen in the distribution of the trade communities globally (Figure 2), but especially in the amount of change that countries could undergo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

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

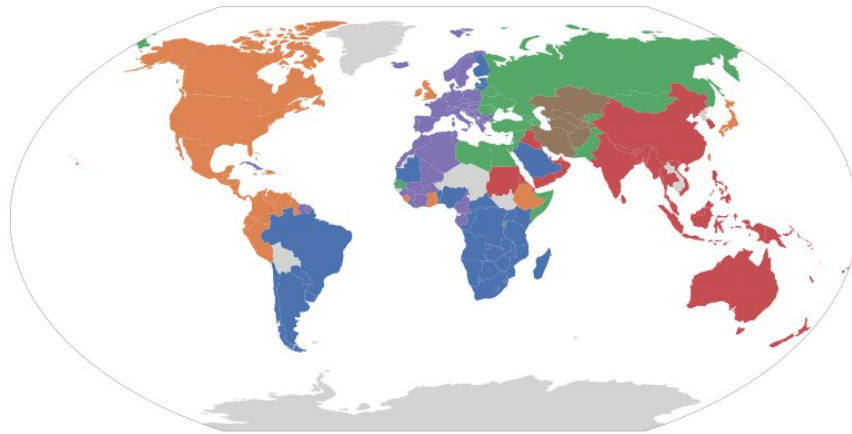
In contrast, the changes might be far more substantial in ASRSs. Nearly half of all countries could experience a shift in their trading partners, with eleven countries undergoing a complete or near-complete overhaul of their trade connections. Some countries affected are consistent with the GCIL scenario,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

The global distribution of trading communities (Figure 2) reveals that this significant shift is primarily due to the expansion of the trading community containing Russia. Today, this community comprises mainly Russia, Eastern Europe, and a portion of North Africa. In the ASRS, however, it extends across [the Balkans all of Europe and](#) most of North Africa, ~~as well as parts of~~ [South and West Africa](#).

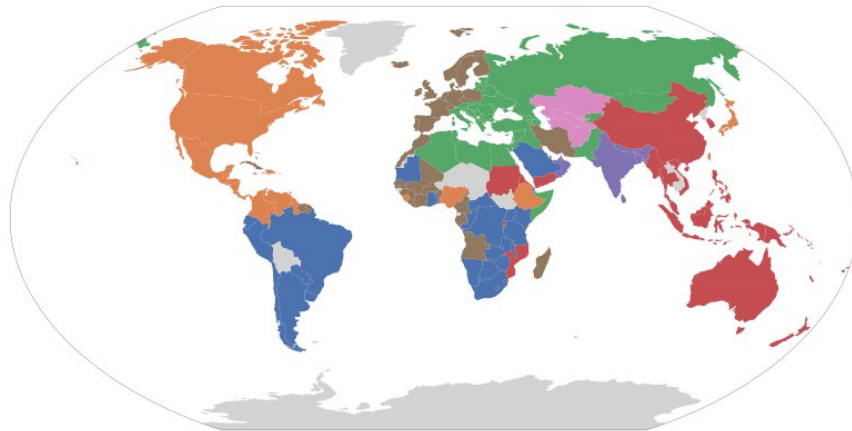
Trade communities for wheat with base year 2022



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



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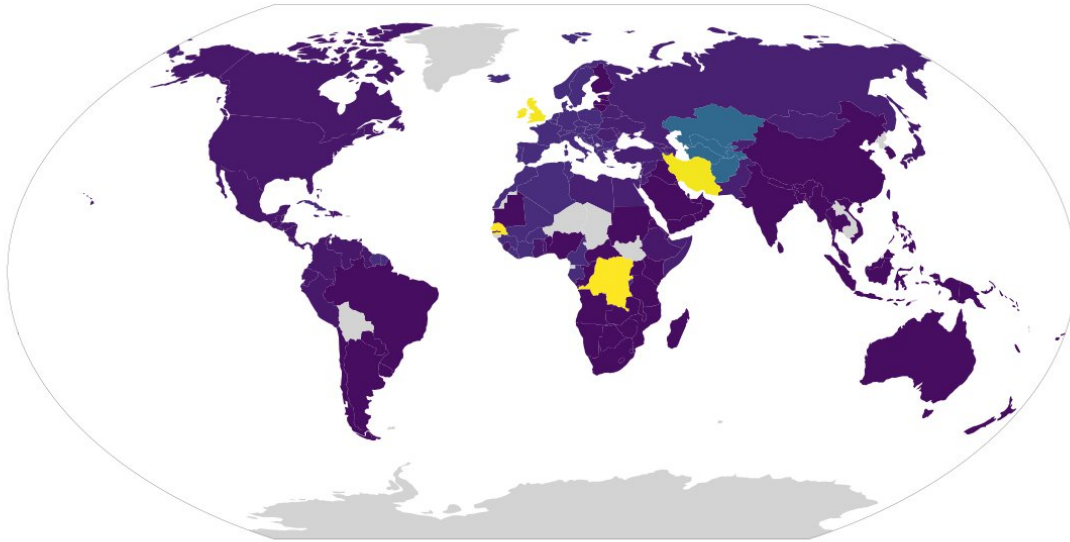
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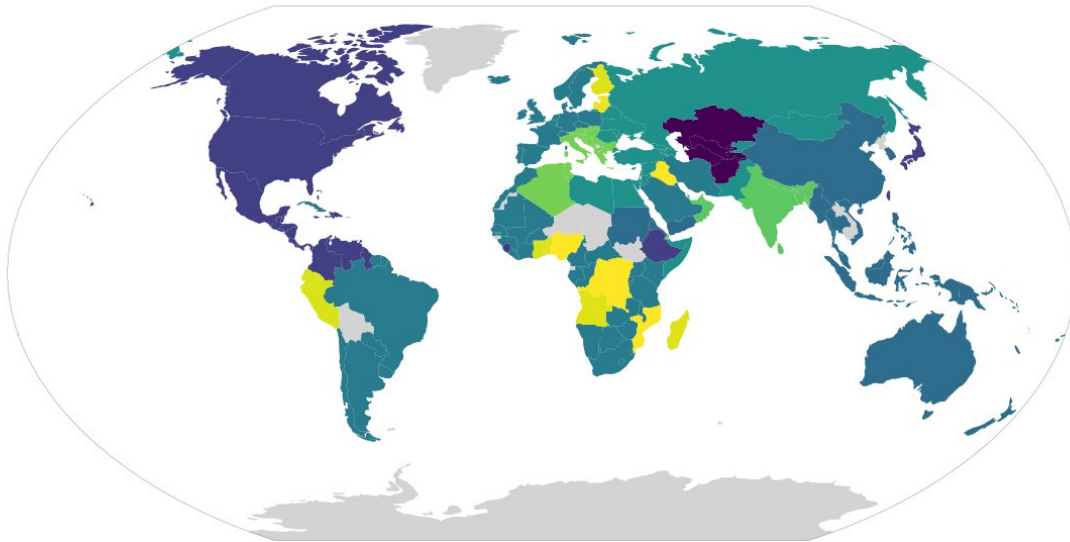
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**Figure 2: Trade communities for wheat in 2022 after yield reduction due to global catastrophic infrastructure loss as well as abrupt sunlight reduction. The colours indicate trade communities. In the GCIL scenario, despite large drops in yields,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



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**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.**

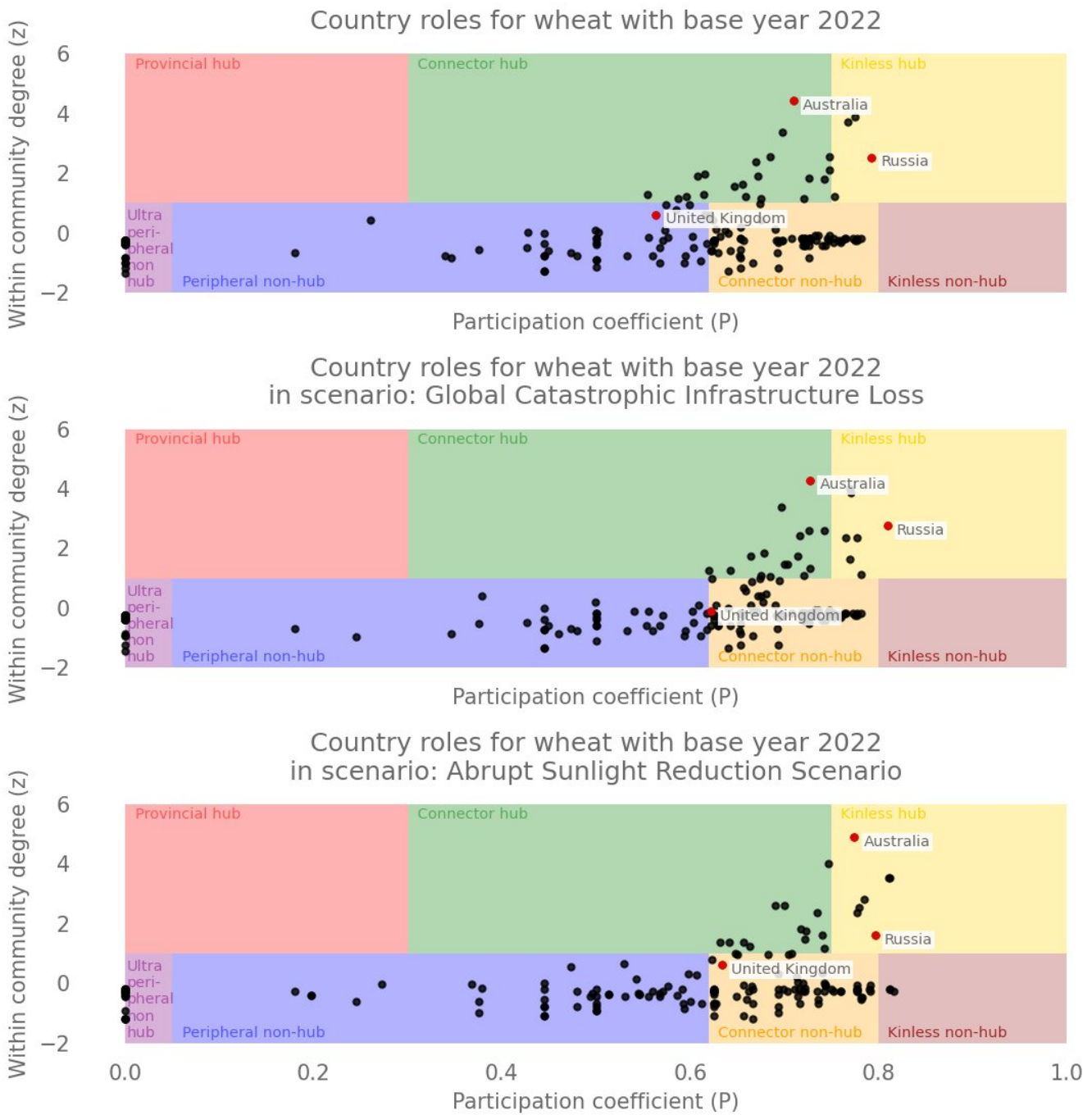
### 268 3.1.2 Community roles of countries

269 ~~In contrast to the impact we can see in the trade communities, there are no~~ Similar to the impact seen in trade communities, there  
270 ~~are~~ significant shifts in community roles ~~in the under scenarios of abrupt sunlight reduction and GCIL scenario and the default 37~~  
271 ~~Tg ASRS (Figure 4). However, we note that in larger ASRSs, country roles within trade networks do shift significantly. The 47~~  
272 ~~Tg scenario (Figure S2) reveals distinct transitions: some countries shift from non-hub connectors to peripheral non-hubs, while~~  
273 ~~others become provincial hubs. This suggests countries lose connections both within and beyond their trade communities, with~~  
274 ~~external connections most affected. Countries maintain fewer imports overall, but those remaining imports come primarily from~~  
275 ~~within their trade community. These patterns indicate a potential tipping point between 37 Tg and 47 Tg, where the system shifts~~  
276 ~~from minimal change to substantial reorganization. When comparing the current situation to a GCIL scenario, there are only~~  
277 ~~minor differences in country roles within the trade network and their communities. In the 47 Tg ASRS, some countries transition~~  
278 ~~from the role of non-hub connectors to peripheral non-hubs, and a few move into the provincial hub category. This indicates that~~  
279 ~~the larger ASRS may lead to countries losing connections both within and outside their trade community, with a more~~  
280 ~~pronounced impact on external connections. This means that the overall volume of imports decreases, but the imports that remain~~  
281 ~~are mostly from within their trade community.~~

282  
283 Another way to assess country roles in the global trade network is through in- and out-degree centrality, which identifies key  
284 importing and exporting countries (Figure S3). In-centrality remains stable across all scenarios, reflecting the overall trade  
285 volume, although total imports decrease due to reduced yields. Out-centrality experiences more significant changes. Presently,  
286 Australia has the highest out-centrality, followed by the United States, France, Canada and Russia. This order remains largely  
287 unchanged after GCIL, though Russia's out-centrality slightly surpasses that of the United States and Canada. Likely because of  
288 its ~~reduced use of less intensive~~ agricultural inputs in comparison with the other countries. The most substantial shifts occur in  
289 the ASRS, however, where Russia, Canada, and the United States experience considerable yield losses, resulting in significantly  
290 reduced out-centrality. Meanwhile, Australia maintains its top position, with France and Argentina rising to second and third  
291 place, respectively.

292  
293 ~~When we examine specific countries, we can see these changes clearly by scenario (Figure 4). Australia remains a central player~~  
294 ~~in global wheat trade in both scenarios. It is less impacted by climatic changes in ASRS and uses fewer agricultural inputs,~~  
295 ~~making it less affected by GCIL. Russia maintains its importance in GCIL but declines significantly in ASRS due to severe~~  
296 ~~climatic impacts. The United Kingdom also remains stable in GCIL but loses most of its trade connections in ASRS. These~~  
297 ~~examples show the disruptive nature of ASRS. In GCIL, most countries retain their positions in the trade network, experiencing~~  
298 ~~similar yield losses. Conversely, in ASRS, many countries lose most of their connections, while a few remain largely unaffected,~~  
299 ~~causing a major shift in the trade network.~~

300



301

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

304

305 **3.1.3 Changes in trade flows**

306 When examining the decline in imports by country, we observe greater impacts under ASRSs compared to GCIL (Figure 5).  
 307 Ukraine and Argentina, which only export wheat, remain mostly unaffected in both scenarios. Under GCIL, most countries see a  
 308 20-30% reduction in imports, with some African and European nations experiencing up to a 40-60% decrease in imports.

309

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

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

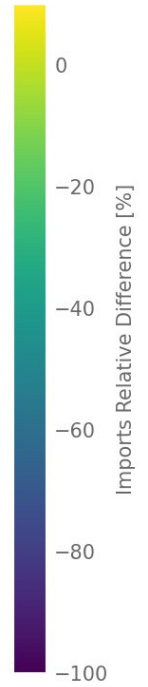
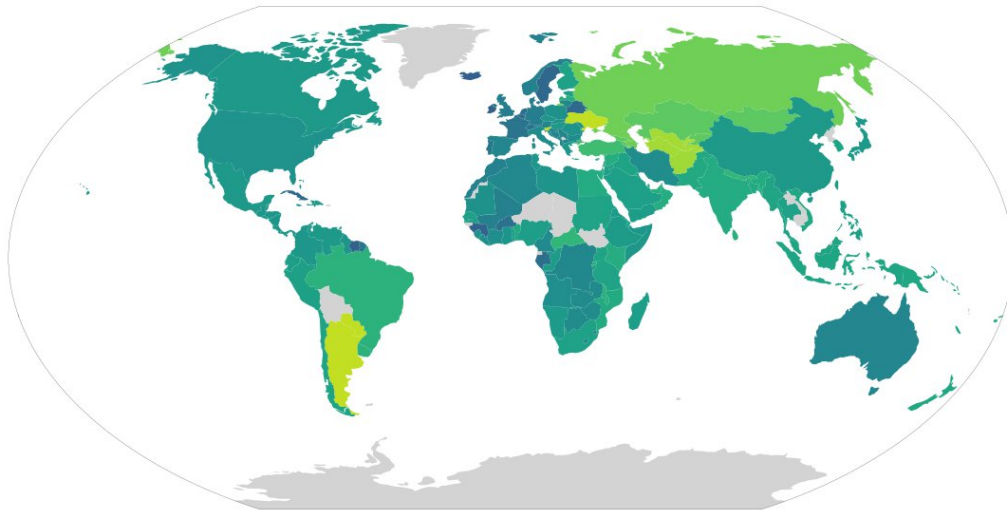
315 ¶  
316 We can also study the absolute changes in wheat imports (Figure S4). This highlights similar patterns across both GCIL and  
317 ASRS, albeit still with a higher impact in the ASRS. The strongest effects in both scenarios can be seen in China, Turkey,  
318 Indonesia and Egypt. All these countries import large amounts of wheat from Russia and Central Asian countries like  
319 Kazakhstan, which experience major yield losses in both scenarios. In particular, Turkey would experience a massive loss of  
320 wheat imports in absolute terms in an ASRS, with around 8 million tonnes of wheat imports lost.¶

321 ¶  
322 The impact to the trade network is also visible when examining remaining wheat production (Figure S5, S6). The scenarios differ  
323 markedly. The GCIL scenario (Figure S5) preserves more wheat production, particularly in Russia, where farming depends less  
324 on fertilizer than in Central Europe or the US. However, in ASRS, Russian production drops severely as lower temperatures  
325 make wheat growing nearly impossible. Australia, a major wheat exporter, continues production but at reduced levels. India  
326 produces the most wheat in ASRS, but primarily grows it for domestic use and may not become a major exporter.

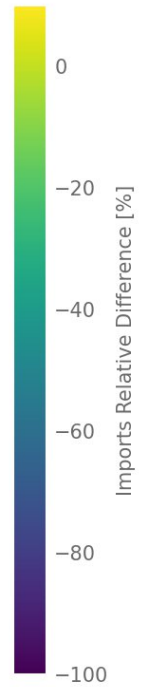
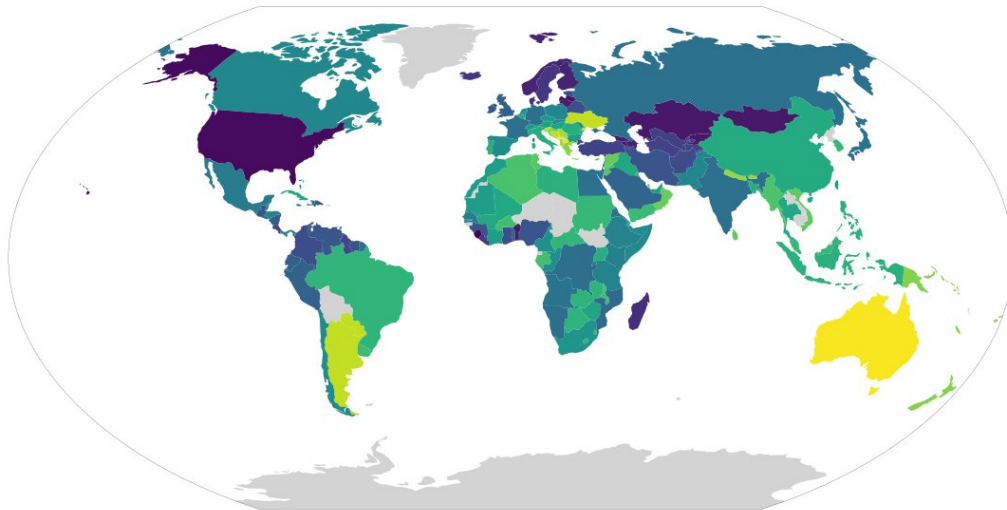
327  
328 Additionally, we performed robustness checks of our results with different metrics. The shifts in trade patterns and the  
329 heightened impact of ASRSs are also evident in other metrics, like community satisfaction and node stability. Community  
330 satisfaction gauges the proportion of a country's trade within its trade community, while node stability indicates a country's  
331 ability to replace lost trade partners. Both metrics highlight the challenges faced by nations reliant on Russia and the United  
332 States. More information on those measures is provided in the supplement (section 3.6).

333

Imports Relative Difference for wheat with base year 2022  
in scenario: Global Catastrophic Infrastructure Loss



Imports Relative Difference for wheat with base year 2022  
in scenario: Abrupt Sunlight Reduction Scenario



334

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

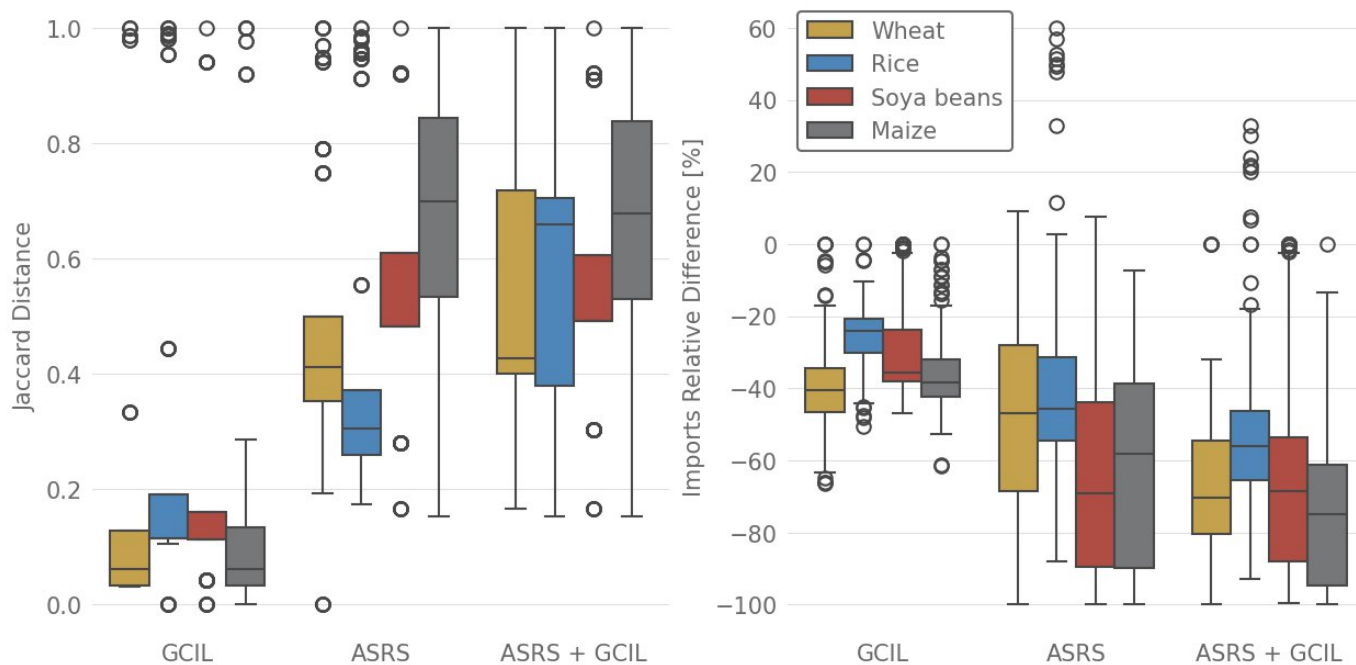
### 337 **3.2 Trends in rice, maize and soya beans**

#### 338 **3.2.1 Overall pattern and comparison across scenarios**

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

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

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



351  
352 **Figure 76: Jaccard Distance and reduction in imports, for each country and crop, for Global Catastrophic Infrastructure Loss (GCIL),**  
353 **and the Abrupt Sunlight Reduction Scenario (ASRS).**

### 355 3.2.2 Rice

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

### 365 3.2.3 Maize

366 In GCIL, the impact on maize is evenly spread worldwide. However, under ASRSs, there's a stark contrast between the Northern  
367 and Southern Hemispheres. Nearly all Northern Hemisphere countries lose most or all of their maize imports, while in the  
368 Southern Hemisphere, South America, much of Africa, and Southeast Asia maintain some imports, mainly from less affected



369 regions like South America. Country roles are similarly affected as in wheat, but many countries switch to the connector non-  
 370 hub role, likely as most countries experience low trade volumes overall. The maize trade network in ASRSs exhibits low stability  
 371 and is heavily affected by the removal of the other major exporters, after the United States' decline in importance due to yield  
 372 reductions. See supplement section 4.2 for the figures showing the trends.

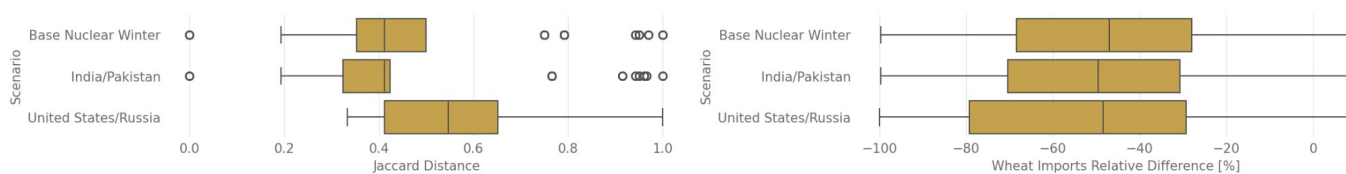
### 373 3.2.4 Soya beans

374 Regarding soya beans, there is a shift in the distribution of affected countries compared to wheat. Many African countries remain  
 375 relatively unaffected, primarily due to their low trade volumes. Under GCIL, most countries face a similar reduction, roughly 20-  
 376 40%, in imports. In ASRSs, the patterns resemble those of wheat, except for South-East Asia and, Oceania, and Argentina. These  
 377 regions still receive wheat imports from Australia and each other, but their soya bean imports mainly come from the United  
 378 States, resulting in a decline. This trend is reflected in trade communities, which remain mostly stable for GCIL but converge  
 379 into two primary and one minor communities for ASRSs. Soya bean export is heavily concentrated in the United States, so a  
 380 sharp yield decline there disrupts trade communities significantly. Only countries importing soya beans from Brazil maintain  
 381 higher import levels, and the trade community with Brazil stays very stable. Similarly to wheat, the role of countries in their  
 382 communities shifts, with most staying the same for GCIL but losing much connectivity in ASRSs. Another notable deviation  
 383 from wheat patterns lies in network vulnerability to node removal. With only two major exporters, the United States and Brazil,  
 384 if the United States is already affected by yield reduction, the network becomes less stable, experiencing another shock when  
 385 Brazil is removed. See supplement section 4.3 for the figures showing these trends.

## 386 3.3 Comparison of nuclear war scenarios

### 387 3.3.1 Impact of removing countries

388 The ASRS data is based on nuclear war simulations. To explore these further, we simulated the removal of Russia/United States  
 389 and Pakistan/India from the 37 Tg scenario (Figure 87). We compared these scenarios with the ASRS that includes all countries  
 390 and the wheat trade of today. The findings reveal that while removing these countries affects both trade communities and overall  
 391 imports (Figure 87), the effect of the yield reduction due to abrupt sunlight reduction is already so big that the removal of those  
 392 countries is negligible. Thus, if countries involved in a nuclear conflict were to cease as trading partners due to the destruction of  
 393 their territories, it would cause additional disruptions to the global trade network beyond those due to the yield reductions, but  
 394 only marginally and for a subset of countries.



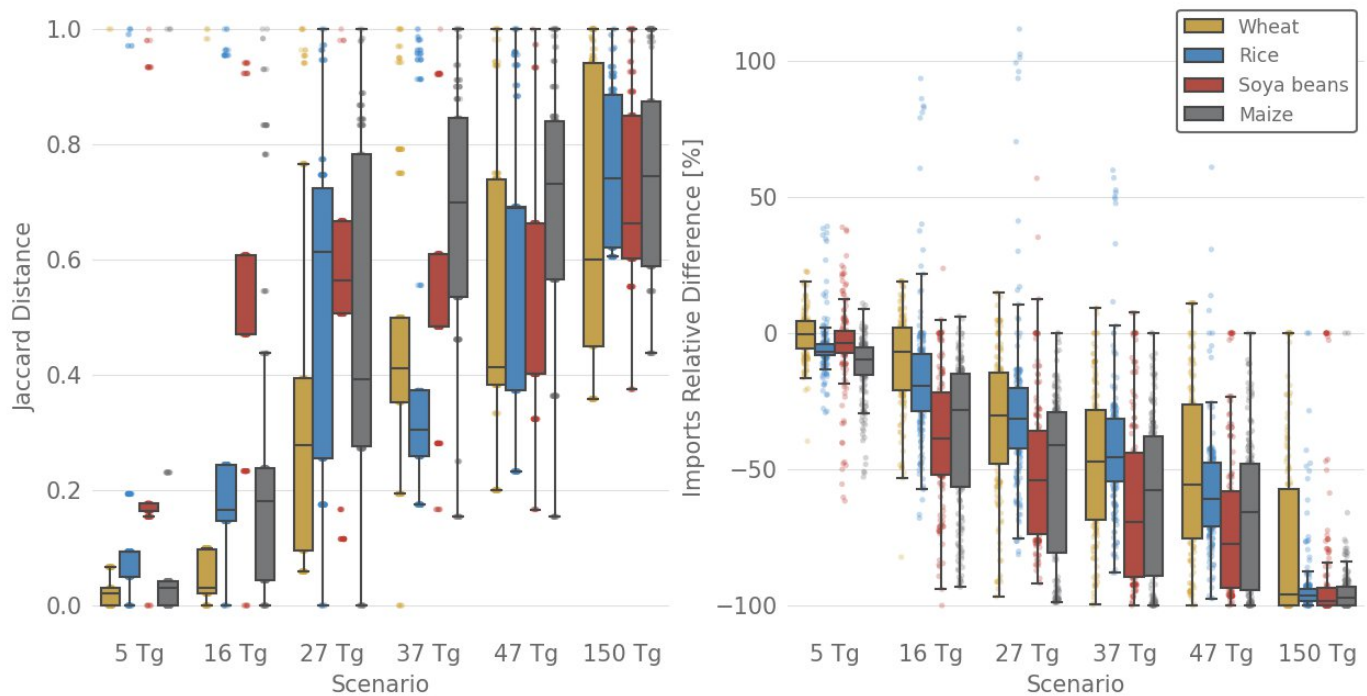
395 **Figure 87: Country removal impact on wheat imports in nuclear war scenarios.**  
 396

397 When simulating the gradual removal of nodes, the results indicate that removing random nodes causes a slow but steady decline  
 398 in network stability. In contrast, specifically targeting the most active exporting nodes results in a rapid decline in stability,  
 399 leading to network collapse after removing 10-20% of these crucial nodes. Further details can be found in Supplement Section  
 400 3.7.

### 401 3.3.2 Impact of soot emission magnitude after nuclear exchange

402 Assessments of the impacts for nuclear war scenarios of different magnitudes (Figure 98) show a consistent pattern across all  
 403 crops analysed. While the most significant impacts can be seen in the worst nuclear war with 150 Tg of soot emitted (nuclear  
 404 winter), the effects would already be quite severe at 37 Tg (nuclear fall). The 37 Tg scenario engenders a substantial of about  
 405 60 % import loss, suggesting that trade would be massively impacted in the 37 Tg case. However, for most countries, food  
 406 imports would have ceased almost entirely in a 150 Tg scenario. In addition, even at merely 5 Tg, some countries could  
 407 experience a 50 % loss of maize, and at 16 Tg, a considerable number of countries have import reductions from 40 % to almost  
 408 100 % across all crops.

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



412  
 413 **Figure 98: Relative change in imports and Jaccard distance for the four primary crops and nuclear war climate changes resulting from**  
 414 **the emission of 5, 16, 27, 37, 47, and 150 teragrams (Tg) of soot across all countries. Coloured points represent individual countries.**

415 **4 Discussion**

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

421  
 422 Our results show clear differences between the effects of ASRSs and GCIL on food trade. The scenarios have different effects on  
 423 how much trade communities are disrupted, the decrease in overall imports, and the roles of countries within their trade  
 424 communities. Across all these measures, ASRSs lead to much larger disruption than GCIL, even for a similar net global yield  
 425 loss. This is due to the way these global catastrophes play out and the spatial distribution of their effects.

426

427 When both scenarios are combined to simulate the co-occurrence of both kinds of catastrophes, the impacts increase and result in  
428 food import losses in the range of 70-100 % for many countries. The impact on yields (Figure 1) and trade (Figure 76) change for  
429 both catastrophes in a similar way. We can see that the median losses are similar for both trade and production. However, while  
430 there are still countries that will likely see little impact on their food production by direct effects of the catastrophes, almost all  
431 countries experience a major loss in their food imports. This is due to the countries that are least affected are usually not major  
432 exporters. For GCIL the least affected countries are those that have very low input agriculture, which is usually also not very  
433 productive, while for ASRS the positive effects are mostly in countries which are too warm now for most agriculture.

434

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

440

441 On the other hand, for ASRSs, we see a much larger split between which countries are more or less affected. Generally, the  
442 higher the country's latitude, the more it is affected (Coupe et al., 2019). In addition, countries in the Northern Hemisphere are  
443 affected more overall. This is partly because nuclear war would most likely occur in the Northern Hemisphere (Coupe et al.,  
444 2019), resulting in somewhat lower soot concentrations in the Southern Hemisphere. In addition, the Southern Hemisphere has  
445 more land closer to the equator and more ocean (which acts as a temperature buffer), suggesting that the Northern Hemisphere  
446 may still be more affected even for ASRSs that do not involve nuclear war. These factors may lead to an especially large yield  
447 decline in the United States, Canada, Central and Northern Europe and Russia. These are all major food exporters, particularly  
448 for wheat or, in the case of the United States, for all major crops. This loss of exports from the major exporting countries  
449 cascades across the whole system. For all crops we can see significant changes in trade communities and large declines in the  
450 amount of imported food. This is especially true for maize, as maize is not very cold tolerant and, therefore, especially vulnerable  
451 to drops in temperature.

452

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

#### 460 **4.1 Implications for the food system**

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

467 were less concentrated in the United States. Other studies of the current trade network show a high dependency on other major  
468 exporters, such as Brazil and Russia (Ji et al., 2024). Our results also indicate that a disturbance of these nodes is very  
469 significant. For instance, Australia would be the last remaining major exporter of wheat during an ASRS (Figure S3). Therefore,  
470 if this country were to stop exporting, the wheat trade would effectively end globally.

471  
472 We know that complex networks become more susceptible to perturbations as they get more centralised (Wiliński et al., 2013),  
473 and the food system is getting increasingly centralised and concentrated (Clapp, 2023). This means that if we do not alter our  
474 approaches to food trading, we will get more and more vulnerable to major shocks and the kinds of scenarios we have described.  
475 There are some indications that this global concentration of trade might be beginning to change (Kang et al., 2024; Mamonova et  
476 al., 2023), as more countries rethink how they handle food and trade more generally. Whether these trends continue depends on  
477 how the geopolitical situation develops in the coming years and decades.

478  
479 Our research also shows that the ASRS has a much wider range of effects. Some countries could even increase imports, as their  
480 neighbouring countries profit from the changed climate (e.g. more precipitation and cooler temperatures in some semi-arid  
481 regions), while others could lose all of their incoming food products. This means that recommendations to tackle these scenarios  
482 have to be tailored to specific countries, as there can be no approach that applies to all countries. For GCIL, more general  
483 recommendations might be possible, as all countries are affected similarly.

484  
485 Recent studies have compiled lists of countries that have experienced substantial food import shocks in the past (Zhang et al.,  
486 2023b). While there is some overlap between these countries and the ones affected the most here, it also becomes clear that  
487 especially the Central European countries, as well as the major exporting countries, have not experienced large food import  
488 shocks on that scale in modern history. This indicates that these countries have no experience with import shocks and are  
489 possibly less prepared to handle the scenarios described in this study. ¶

490 ¶  
491 [Shifting dietary patterns could also help decrease vulnerability to trade disruptions. A move toward plant-based diets with more  
492 locally-produced fruits, vegetables, and legumes could reduce dependence on international grain trade, as much of the currently  
493 traded grain \(especially soya beans and maize\) is used for animal feed rather than direct human consumption. This conversion of  
494 grain to animal products is inefficient from an energy perspective. However, this strategy presents a trade-off: while reducing  
495 animal feed imports would decrease trade dependencies, it might also reduce the system's overall flexibility. Current livestock  
496 systems, despite their inefficiencies, create a buffer by maintaining large stores of grain that could be redirected to human  
497 consumption during crises. Additionally, ruminants can digest cellulose that humans cannot, potentially providing an additional  
498 food source during catastrophes. Therefore, while dietary shifts toward plant-based foods could improve local food security  
499 under normal conditions, maintaining some animal agriculture may provide valuable system redundancy for extreme scenarios.  
500 The optimal balance likely varies by region based on local agricultural conditions and trade relationships.](#)¶

501 ¶  
502 [Similarly, more strategic use of agricultural land could enhance resilience. Currently, significant agricultural capacity is devoted  
503 to non-food purposes – particularly biofuel production and crops like tobacco. While biofuel crops are often heavily subsidized,  
504 transitioning to electric vehicles would be more energy efficient and free up land for food production. The land used for tobacco  
505 cultivation could be repurposed for food crops, providing dual benefits of improved food security and public health.  
506 Additionally, reducing food waste, which accounts for approximately one-third of food production in many developed countries,  
507 represents a readily available opportunity to build resilience \(Alexander et al., 2017\). However, as with dietary shifts, these  
508 changes present trade-offs. Some biofuel infrastructure could potentially be repurposed to produce food during crises, similar to](#)

509 [how breweries can be converted to produce sugar from cellulose \(Throup et al., 2022\). Moreover, maintaining diverse](#)  
510 [agricultural systems and processing capabilities, even for non-food crops, helps preserve farming knowledge and infrastructure](#)  
511 [that could be valuable during catastrophes. The key is finding a balance between efficient land use under normal conditions and](#)  
512 [maintaining adaptable agricultural systems that can respond to major disruptions.](#)

## 513 4.2 Study Limitations

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

- 517 - We only looked at the direct effects of yield reduction on trade flows and did not consider any additional adaptations.  
518 For example, it seems likely that many countries would introduce export bans if their own yields dropped significantly,  
519 worsening the overall situation. This means that our study can be seen as the minimal amount of change that one can  
520 expect to happen after global catastrophes by the yield changes alone, barring the introduction of resilient food  
521 adaptations to counter the loss of yields (Pham et al., 2022). Further research is needed to understand how societies  
522 might react to the effects explained here.
- 523 - GCIL also includes the assumption that we lose the majority of our mechanisation and transportation. This is not  
524 modelled in this study, but plausibly could have major implications beyond the impact on yields and make it more  
525 catastrophic than ASRSs. A GCIL would disrupt fossil fuel production, hampering international trade. However,  
526 possible interventions include retrofitting ships to be wind powered (Abdelkhalik et al., 2016) or wood gasification to  
527 replace fossil fuels (Nelson et al., 2024).
- 528 - We studied the four major food crops in isolation to understand what effects might play out on that level. However, the  
529 food system also consists of other parts like fisheries or livestock. While those are also predicted to decline after a  
530 global catastrophe (Scherrer et al., 2020; Xia et al., 2022), it remains unclear how the totality of all food trade might be  
531 affected by global catastrophes. Livestock would be more strongly affected than major crops because it mostly depends  
532 on them; whereas fisheries, while less affected than crops, make up a small percentage of global caloric requirements  
533 (<2%).
- 534 - Additional layers of interaction from non-food products through social dynamics to economic policies could be  
535 considered in a multi-layer network model, which has been shown to be impactful and effective in other scientific  
536 disciplines (De Domenico, 2023; Kivelä et al., 2014; Paluch et al., 2021).
- 537 - We treated nuclear war simulations as a proxy for large size impact over the land and super volcanic eruptions. While  
538 this is a reasonable assumption, the results might end up very different, especially if the impact/eruption happens in the  
539 Southern Hemisphere, as nuclear war scenarios usually only involve the Northern Hemisphere, as there are no nuclear  
540 weapon states in the Southern Hemisphere. ~~Although~~ Although these extreme events all produce large amounts of aerosols [in](#)  
541 [the upper in upper](#) atmosphere, which block sunlight and cause significant cooling, the compositions of the aerosols  
542 differ. This results in variations in the duration of the cooling and some climate impacts. Recent research indicates that a  
543 simulated volcanic winter shows similar trends (Enger et al., 2024) to previous studies on nuclear winter (Coupe et al.,  
544 2019), although volcanic winters are likely to be shorter in duration. Additionally, there is a possibility that multiple  
545 mid-sized volcanic eruptions could occur simultaneously, releasing enough sulphate aerosol to cool the Earth  
546 significantly.
- 547 - Even for such a relatively well-studied global catastrophe as nuclear winter, there is still much we do not understand.  
548 For example, the work of Coupe et al., (2023) suggests that nuclear winter can paradoxically lead to a decrease in

549 Antarctic sea ice despite global cooling. As our understanding of global catastrophic risks increases, we may see shifts  
550 in our expected effects on the food system.

551 - Trade is only a part of the global value chain, and if we look at the whole value chain, we can expect many more  
552 disruptions (Ibrahim et al., 2021). ¶

553 — The modeling of Xia et al. (2022), which we use to calculate yield reductions during a nuclear war, assumes the usage  
554 of spring wheat. However, during an ASRS, wheat producers could switch to winter wheat, which is more resistant to  
555 cooler temperatures and frost and generally has slightly higher yields than spring wheat. Therefore, the wheat yields in  
556 this study are potentially underestimated. ¶

557 - We only consider the global aspects of the catastrophes. However, there are a variety of plausible scenarios where  
558 regional effects could have global repercussions. For example, the food system has several so-called choke points  
559 (Bailey and Wellesley, 2017; Key et al., 2024; Wellesley et al., 2017), where much food trade is funnelled through a  
560 small geographic area. Some of these choke points are near volcanoes and could be severely affected by eruptions  
561 (Mani et al., 2021). Should these choke points close in the aftermath of a global catastrophe, the disruption of the food  
562 system would further increase.

### 563 **4.3 Comparison to climate change**

564 The model employed in this study was originally developed to study the effects of climate change on food trade (Hedlund et al.,  
565 2022). We can see that the impact of a rather severe climate change scenario based on RCP 8.5 has considerably lower effects  
566 than the catastrophes explored here and even results in an increase in imports for almost all countries (Figure S25). For all crops  
567 the trade communities stay mostly the same, while they would be much more disrupted in our scenarios. A similar pattern holds  
568 up for all crops considered. These differences are likely due to the different magnitudes of the catastrophes considered. For RCP  
569 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  
570 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  
571 war. (Xia et al., 2022). Therefore, the ASRS considered here not only has the larger temperature change, but also in a much  
572 shorter time period. Also, in the case of climate change, the countries that will be more affected are those closer to the equator  
573 (Frame et al., 2017). Since the main exporting countries are mostly at higher latitudes, they will be less affected by climate  
574 change, contributing to a more stable food trade in comparison to the scenarios we explored.

### 575 **4.4 Gaining a deeper understanding of how global catastrophes impact the food system.**

#### 576 **4.4.1 Research gaps**

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

584 - Understanding how global catastrophic risk might affect different parts of the global population by socio-demographic  
585 metrics. We know that climate impacts are felt differently depending on how rich the country is (Quante et al., 2024)  
586 and also increase wealth inequality (Méjean et al., 2024). Therefore, it is likely that these differences also exist as a  
587 consequence of global catastrophes.

- 588 - While there is little research on the effects of the dependency on very few food trading companies (Clapp, 2023), there  
589 is none when it comes to the question of how this might affect the outcomes of global catastrophic risk scenarios.
- 590 - There exists some research that acknowledges the potential cascading effects and systemic risk of an ASRSs, like  
591 nuclear war, for instance, recent summaries by Green (2024) or Glomseth (2024), but for many of the events that could  
592 cause GCIL, we know only very little about the potential cascading effects. Beyond this However, even sophisticated  
593 modeling efforts like Xia et al. (2022) have limitations - they did not account for several factors that could further  
594 impact agriculture after nuclear war, such as changes in irrigation water availability, increased surface ozone levels,  
595 ultraviolet light damage, effects on pollinators, and killing frost risk. For many of the events that could cause GCIL, we  
596 know even less about the potential cascading effects and systemic risks.
- 597 - We need more understanding of the effects of catastrophes like geomagnetic storms and how the loss of industrial  
598 inputs might affect agriculture. There is some global research on the direct effects (Cliver et al., 2022; Isobe et al., 2022;  
599 Rivers et al., 2024b) but less on the indirect effects, especially on agriculture (Moersdorf et al., 2024). There are some  
600 recent research studies which explore similar effects yet do not frame it in regards to global catastrophic risk but instead  
601 as a general disruption in the trade of industrial inputs for agriculture (Ahvo et al., 2023; Sandström et al., 2024).
- 602 - There is a good chance that catastrophes will not happen in isolation but interact with each other and existing  
603 vulnerabilities. An example is the possible interaction between nuclear winter and planetary boundaries (Jehn, 2023) or  
604 termination shock caused by civilization collapse (Baum et al., 2013). These are only two of the possible interactions,  
605 and many others are entirely unexplored (for example, having a major geomagnetic storm during a pandemic). .
- 606 - Our food system is not reliant on the food trade network alone but on a highly complex supply chain with many  
607 interacting goods and services (Ibrahim et al., 2021), also consisting of many non-food items. It would be valuable to  
608 understand how these might react to the scenarios described in this manuscript. There has been some work to study this  
609 for current conditions (Deteix et al., 2024), but not with a focus on global catastrophes.
- 610 - We do not know what might happen after the initial effects play out, as this paper only describes the minimal amount of  
611 change that is expected to happen due to the yield changes alone. However, if we look into history, we can see that such  
612 disruptions of trade networks can have massive consequences. If they unravel the whole network, countries lose access  
613 to many goods they need, leading to internal problems and possibly collapse, as happened in the Late Bronze Age  
614 (Linkov et al., 2024). Important insights could be gained here by applying insights from quantitative history to the last  
615 100 years, as proposed by Hoyer et al. (2024). This could be built upon by using historical worst cases and using them  
616 as downward counterfactuals to create more realistic and comprehensive scenarios (Woo, 2019).

617

618 Furthermore, all those research topics that need further exploration and studies like ours should be regularly re-assessed. As the  
619 Russian invasion of Ukraine has shown, major disruptions in the food network can and are likely to happen again (Miller et al.,  
620 2024). They reshuffle existing trade connections, making research like this less accurate as time passes.

#### 621 **4.4.2 Decreasing vulnerability to global hazards**

622 Since the global food system is vulnerable to major disruptions, it is of high priority to decrease these vulnerabilities. Myers et al.  
623 (2022) suggest a list of interventions that could decrease the vulnerability of agriculture to climate change. Some of these  
624 suggestions would also help here, like having more diverse crops to ensure flexibility with respect to climate conditions or  
625 strengthening international trade agreements to ensure that the flow of food is stable. This also ties in with the criticism of  
626 concentration in the food system by Clapp (2023). These concentrations on all levels of the food system increase the risks of  
627 collapse and need to be decreased, especially for the safety of people in net food importing countries (Yıldırım and Önen, 2024).

#### 628 4.4.3 Increasing resilience after a global catastrophe

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

#### 653 5 Conclusion

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

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



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

#### 671 **Author contributions**

672 Conceptualization: FUJ, JH

673 Data curation: FUJ, LGG

674 Formal analysis: FUJ, LGG

675 Funding acquisition: DD

676 Investigation: FUJ, LGG

677 Methodology: FUJ, LGG

678 Project administration: FUJ

679 Software: FUJ, LGG

680 Supervision: FUJ

681 Validation: FUJ, LGG, CWA

682 Visualization: FUJ, LGG

683 Writing-original draft: FUJ

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

#### 685 **Data and code availability**

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

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

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

689 The model code (with additional documentation) can be found at: <https://github.com/allfed/pytradeshifts> (Jehn and Gajewski,  
690 2024).

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695

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699

#### 700 **Competing interest**

701 The authors declare no competing interest.

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