Global relationships between crop diversity and nutritional stability

Charlie C. Nicholson[1*], Benjamin F. Emery[2,], Meredith T. Niles[4,5,6]

[1] Department of Entomology and Nematology, University of California, Davis, 95616, CA, USA
[2] Vermont Complex Systems Center, University of Vermont, Burlington, 05405, VT, USA
[4] Gund Institute for Environment, University of Vermont, Burlington, 05405, VT, USA
[5] Food Systems Program, University of Vermont, Burlington, 05405, VT, USA
[6] Department of Nutrition and Food Science, University of Vermont, Burlington, 05405, VT, USA
* Corresponding author.

Email: ccnicholson@ucdavis.edu

Classification

- SOCIAL SCIENCES: Environmental Sciences; Sustainability Science
- BIOLOGICAL SCIENCES: Environmental Sciences; Sustainability Science

Keywords

Agriculture, Crop diversity, Food security, Food imports, Nutrition

Author Contributions

CCN developed the framework. CCN and MTN designed the project. BFE and CCN performed analyses. CCN and MTN wrote the manuscript

Abstract

Nutritional stability – a food system's capacity to provide sufficient nutrients despite disturbance – is a critical feature of sustainable agriculture, especially in light of ongoing climate change. Yet, measuring nutritional stability has proven challenging. Addressing this challenge will help identify resilient food systems, detect shortcomings in nutrient availability, and evaluate if stability-focused interventions actually work. We develop a novel approach that uses 55 years of crop data across 184 countries to assemble over 22,000 bipartite crop-nutrient networks. We then quantify the tolerance of these networks to disturbance simulated via sequential crop loss (Fig. 1) and evaluate patterns of crop diversity and nutritional stability across countries, over time and between crop supply scenarios (imports versus in country production). We observe a positive, saturating relationship between crop diversity and nutritional stability across countries; however there is substantial variability between countries over time. Next, despite crop diversity gains since 1961, nutritional stability has remained stagnant or decreased in all regions except Asia. A decline in the average number of nutritional links per network (range: -3 to -18% across regions) and the aforementioned saturating relationship explain this counter-intuitive finding. Finally, we find that imports increase crop diversity and improve or sustain stability, indicating that nutrient availability is market exposed in many countries, particularly developing states. Although applied globally, our approach is applicable across levels of organization, from household intake to sub-national production, and provides a way forward for understanding the contributions of crop diversity to the stability of nutrients available for human consumption.

Introduction

Market volatility, land degradation, pests, and climate change make it increasingly difficult for agriculture to sustain a growing and healthy human population. In the face of these challenges, longstanding global- and national-scale policies seek to improve food security by increasing national production, leveraging international trade to counteract crop failures and supporting technological innovation (e.g. precision irrigation, drought resistant cultivars). These policies typically evaluate outcomes in terms of total yield or food calories, however expanding commitments to nutrition-sensitive agriculture (2, 3) has recognized that calories does not equate to food security. There is now increasing focus on nutritional diversity, including diet diversity and the diversity of nutrients needed to sustain a balanced diet and lead an active, healthy life (4).

With a growing focus on nutritional diversity, crop diversification is seen as a promising strategy to improve dietary diversity and nutritional status (5, 6). Nonetheless, there is mixed evidence linking crop diversity to nutritional outcomes (7), leading to calls for multilevel and systemic measurement approaches that can evaluate the influence of agrobiodiversity on the provision of nutrients at different scales (e.g., village, region, and national) (8–10). Potential nutrient adequacy (11), one advancement toward this end, incorporates into a single metric the fraction of a population potentially nourished for all nutrients by all crops. However, neither this, nor any other approach, measures the ability of agriculture to produce nutritious food through space and time in the face of chronic disturbance and acute shocks (12).

Quantifying this nutritional stability – the capacity of a food system to provide sufficient nutrients despite disturbance – can aid effective coordination and implementation of resilience planning for nutrition-sensitive agriculture. Target 3 of the Sustainable Development Goals stresses connecting crop diversity, resilient farming systems, and nutritious diets as essential components of food security (13). There are plausible mechanisms and demonstrated evidence for crop diversity contributing to human nutrition through the addition of key crops belonging to distinct nutritional functional groups (14–16). Yet, the fragility of this diversity-function relationship is unknown because a straightforward method to measure nutritional stability does not exist (1).

Here we develop a novel approach that measures the extent to which crop diversity underpins nutritional stability over space and time. Our analytical framework (**Fig. 1**) links crops and their constituent nutrients into a bipartite network and then quantifies the effect on nutrient availability in a given country when crops are removed from the network. This approach has characterized network tolerance in diverse fields, from information systems (17) to ecological networks (18), but has never been applied to better understand the stability of nutrient availability in food systems. As with most biodiverse communities, species removal order can structure loss of system function (19, 20). We therefore generate a generalized 'attack tolerance curve' (21) via permutation of removal sequence and derive our nutritional stability metric (R_N) as the area under this curve (**Fig1d, e**; see Methods). A network with more nutrient dense crops will have more nutritional links (i.e. higher degree) and thus be more robust to crop loss.

We apply this method to over 22,000 crop-nutrient networks assembled using 55 years of FAO food balance data for 184 countries and a global nutrient composition database (22). We evaluate patterns of nutritional stability – and its relationship with crop diversity and nutrient availability – across countries, over time and between two supply scenarios: food derived from production (P) and food from production and imports (PI). We then ask: (1) What is the relationship between crop diversity and nutritional stability? (2) How has crop diversity and nutritional stability

changed over time in countries and regions? (3) What regional patterns underpin differences in nutritional stability?

Results

Crop diversity begets nutritional stability with a non-linear relationship that has regional variability (**Fig. 2**). Across countries, nutritional stability increases with crop richness at significantly ($p < 0.05$) different rates between regions, generally achieving a threshold at which additional crops do not provide significant improvements in nutritional stability. (**Fig. 2, Fig. S1; Table S1**.). Across regions, gains in nutritional stability generally slow after crop-nutrient networks contain between 10 to 15 unique crops. Additional crops, beyond those present in those regions, do not provide significantly greater gain in nutrients available in those regions. This suggests there is a threshold for the extent to which adding more crop diversity into a region provides significant nutrient availability gains.

We find that crop diversity increased since 1961 for all regions except in Oceania (**Fig. 3A**, **Table S2**). Importantly, there is significant variability across the remaining four regions, including the role of imports in driving crop diversity change over time. For example, advances are pronounced in Asia and Europe, where imported crop diversity increased by 43% & 35% between 1961 and 2016, respectively. Europe also has the largest gap between imports and production, with the majority of crop diversity increases driven by imports. In other words, increases in crop diversity at the country level are not necessarily (and frequently not) the result of increased diversity of production within a country.

We also find high variability in nutritional stability between countries over time (**Fig. 3B**). For example, countries such as the United States, Brazil, and much of Europe (light blue areas) had high stability with low variability in that stability (i.e. a small range of values measured as the variance in R_N over the period 1961 – 2016), meaning there was a consistently high stability of nutrients available in the country's food supply. Conversely, other countries such as many in the Middle East, Southeast Asia, and Africa experienced low stability and large variance (light purple areas), indicating a highly variable and generally unstable supply of micronutrients available in these countries. Notably, small island developing states (SIDS), and low-income countries often experience low stability (**Fig. 4; Table S3**).

Despite finding that crop diversity is increasing, we find that nutritional stability has remained stagnant or even decreased in all regions except Asia (**Fig. 3C, Table 1**). Furthermore, across all regions, any gain in nutritional stability at the regional level over the time period are exclusively associated with a region's supply of nutrients available from production and imports, not regional production alone. For example, while Asia experienced an 8% increase in nutritional stability, this was from an increase in imports over the time period; nutritional stability from production alone actually declined by 1%. Conversely, Africa experienced a 4% decrease in nutritional stability derived from reductions in crop diversity at the production level alone, while import-based nutritional stability did not change between 1961 and 2016. European nutritional stability was both relatively high (compared to other regions) and stable across the time period. In the Americas and Oceania, we find production-based nutritional stability decreased by 7% and 4% (respectively).

We present a seemingly counter-intuitive finding: for most regions crop diversity has increased yet nutritional stability is stagnant or decreasing. The saturating relationship between crop diversity

and nutritional stability in part explains this, but changes in crop degree – the average number of
nutrients provided by each crop in a network – clarify this incongruity further (**Fig. 5**). While crop
diversity from production and imports increased for 72% of countries, 87% of these countries saw
the average crop degree decrease (**Fig. 5**, upper left), and this is associated with a decrease in
nutritional stability. In fact, crop degree has declined in all regions over the 55 year period,
regardless of supply scenario (**Fig. S2, Table S4**). Stated succinctly, even though crop diversity
has increased, there are diminishing returns on nutrient availability from adding more crops into a
network, especially since the crops being added into networks appear to provide fewer links to
nutrients not already in the food system.

108
109

Discussion

Narrowing crop diversity in the world's food supplies has been considered a potential threat to
food security (23); however, there have been few empirical studies to link crop diversity to
nutritional outcomes, especially beyond dietary intake at the household level (10). Here we
develop a novel method to link crops to specific micronutrients using a network approach and
assess the role of crop production and imports on nutritional stability outcomes in 184 countries
between 1961 and 2016. Similar to other scholars (23, 24), we find that crop diversity has
increased over time in all regions except Oceania, but that in many cases these gains are due to
imports. Despite this increase in crop diversity, nutritional stability has remained stagnant or
decreased in all regions except Asia, a trend largely attributed to our finding that recent crop
diversity gains result in fewer new nutritional links in a given food system.

The general relationship between crop diversity and nutritional stability is contextualized by
changes in crop degree, and explains why stability trends do not mirror diversification. Improving
crop diversity will always increase the size of the crop-nutrient network, but stability depends on
the number of links within this network. As in other diversity-stability relationships functional
identity matters, and declines in crop degree could reflect shifts towards networks with less
nutrient-rich crops. For example, production-based crop diversity in Senegal increased by 29%,
while crop degree dropped by 19% as the composition of its food supply shifted from staples (e.g.
millet, groundnuts, sweet potatoes) to include less nutrient dense crops (e.g. sugar cane,
watermelon, cabbage). In light of on-going homogenization of crop diversity (24), attaining the
benefits of nutritional stability will require further understanding the topology of crop-nutrient
networks.

These results have many important implications for our understanding of nutritional outcomes
and their relationship to crop diversity. First, while our work reaffirms the existing body of
research demonstrating that crop diversity is important for agricultural sustainability, it does so at
a national scale. Much of the existing work in understanding crop diversity and its links to
nutritional outcomes are at a field or landscape level (10). Our work answers recent calls (9) to
explore crop diversity and nutritional outcomes at a larger scale through a country level analysis
and incorporates both production and imports, the latter of which has been significantly important
for driving an increase in the types of crops available in a given country over time.

Second, our work adds novelty by establishing a functional relationship between crop diversity
and nutritional stability, and suggests that this non-linear relationship has important implications
for thinking about the types of crops grown or imported in a given region and their relationship to
food availability. A foundation shared by ecology and nutrition is that diversity can improve

long-term functioning of complex biological systems (25, 26). Like other ecological diversity-
resilience relationships, we observe that diversity loss can result in rapid non-monotonic loss of
function (27). In countries where diversity is already low, our results indicate that crop failures,
either through production failure or an inability to import such crops, could lead to rapid
reductions in nutrient availability within a country. Moreover, multiple failures of highly
important regional crops, as might occur during a drought or other extreme event, could have
catastrophic nutritional impact. Such countries are thus vulnerable to a variety of potential global
challenges both ecological (e.g. climate change) and economic (e.g. trade wars).
156
Third, that nutritional stability is stagnant or decreased over time in all regions but Asia highlights
that increasing crop diversity – at least at the national level – does not necessarily lead to better
nutritional outcomes. Instead, the wide variability in nutritional stability across countries, despite
the overall increase in crop diversity, highlights clear vulnerabilities both across regions and
within regions. Africa has the greatest inter-regional variability, demonstrating that in some cases
neighboring countries have very different capacities and stabilities of crop nutrients in their food
supply chain in any given year. This variability is likely driven by multiple factors including the
capacity of a country to trade (28), and in country food availability as a result of war or
political/social unrest (29–31), or climate induced disasters (32).
166
Finally, the important role of imports in our findings for many regions highlights that nutritional
stability is market exposed. While trade can positively affect food security (33), it can also hinder
nutrition efforts (34), and could be a vulnerability if imports comprise a significant portion of
nutrients for a given population. Countries with a high reliance on imports are thus subject to
trade wars, market shifts, and price shocks that can occur for a variety of reasons. Such countries
may therefore be more likely to experience increased variability in the future, especially as
climate change is expected to affect both agricultural production and markets and trade (35).
174
The use of these results could help inform high-level discussions within countries and regions
about the key crops for a given place, and their availability via import versus domestic
production. Scenario development using our metric could help identify country-specific crop
additions that would maximize nutritional stability improvements. Our approach could be used to
identify potential tradeoffs in production and import outcomes, at least as it relates to the
availability of a given amount of nutrients in a certain place. Such potential applications also
highlight the importance of identifying several caveats and important limitations. First, while we
are addressing the nutrients available in a given country in a given time, we are not equating this
with food security. This "availability" is only one component of food security, with access,
utilization, and stability being other critical pillars (FAO 2008). Thus, even though nutritional
stability is generally high in most regions and remained stagnant (or increased in Asia), this does
not mean that people are not food insecure. Adequate food and nutritional security comprises
much more than the factors captured in our analysis, which provides a relative measure of
nutrient availability not an absolute metric of adequacy. Furthermore, there are recognized
shortcomings with the existing FAO data, especially in many low-income countries (36) and we
recognize that data quality is a challenge with this dataset in some cases. Nevertheless, to our
knowledge, it is the best available data of its kind and scale available, so we utilize it knowing
there are many opportunities to improve this work moving forward.
193
Despite these caveats, this work advances a novel method to assess the relationship between crop
diversity and nutritional stability globally over the past fifty years. Future research could expand
this work in multiple ways by combining crop nutrient availability data with nutritional intake

data to better assess whether available nutrients in the supply chain are making their way into
household consumption. This would more completely link crop diversity with food and
nutritional security outcomes, rather than just food availability as this work as done. Furthermore,
our novel network method could be advanced by exploring the importance of certain crops for a
given country or region. Finally, with climate change expected to affect the yields of many
globally important crops (37) and potentially cause multiple crop failures at once (32), this type
of analysis could advance our understanding of food system vulnerability to specific crop failures,
and provide guidance on climate adaptation efforts or crop diversification strategies to safeguard
against climate change. Resilience is now a central paradigm in many sectors – humanitarian aid,
disaster risk reduction, climate change adaptation, social protection. Erosion of biological
diversity typically leads to loss of ecosystem functioning and services, likewise loss of crop
diversity may to lead to potentially drastic shifts in nutritional stability. Together this, and future
analyses, have the potential to direct the protection or restoration of crop diversity so as to best
support nutrient availability that is stable given current and future challenges.
211
212
213 **Materials and Methods**
214
Data sources
Crop composition information came from annual national-level food balance data for years 1961
to 2016 obtained from the FAOSTAT database (Food and Agriculture Organization of the United
Nations; http://fao-stat.fao.org/) for 201 nations that comprise >95% of the global population. We
focus our analysis on yield of crop plants (hg/ha) available for human consumption. To explore
the impact of international trade on nutrient stability, we created two food supply scenarios by
evaluating national yield separately as its production (P) and production plus imports (PI)
components. We linked annual national crop lists for each scenario with crop-specific nutrient
data reported in the Global Expanded Nutrient Supply (GENuS) database (Smith et al 2016), a
global dataset of nutrient supply of reports the amount of nutrient available per 100g of edible
food for 23 nutrients across 225 food categories. We focused on 17 nutrients (calcium,
carbohydrates, copper, fiber, folate, iron, magnesium, niacin, phosphorous, potassium, protein,
riboflavin, thiamin, vitamin A, vitamin B6, vitamin C, and Zinc), omitting calories, fats, water,
ash and refuse. Linking these two datasets results in 22,400 crop-nutrient networks, one for each
country, year, and food supply scenario. However, the nutritional contribution of some foods is
often negligible, yet because links in our bipartite crop-nutrient networks are treated as binary
these would still be considered as contributing to stability. To overcome this minimum quantities
problem (38), we calculated the proportion daily value per serving (PDV) for each country for
each food as: $PDV = 100*N*((Pop/Prod)/DV)$, where N is the grams of nutrient per 100 grams of
food (as provided by GENuS), Pop is a country's population (United Nations, 2019), Prod is the
amount of that food produced (FAOSTAT, 2019) and DV is the daily value of the nutrient in
grams (FDA, 2019). We used PDVs as link weights in each network and then define a threshold
on the computed link weight, below which we consider the link as nonexistent. To do this we
compute the stability for each country in 2016 at values of the threshold cutoff ranging from 0 to
0.7. The mean of the stabilities decreases monotonically as a function of this cutoff, but the
variance shows a plateau from 0 to 0.1 before decreasing. In order to magnify between-country
variation while removing negligible links, we choose the threshold of 0.1, at the upper end of this
plateau. Data availability varied among global regions and missing data values were more
common in the early years of the data set, we therefore removed countries that did not have at
least 20 years reported (N = 17 countries), leaving 184 countries and 19,044 networks.
245

Formalizing nutritional stability (R_N)
Our algorithm processes networks through permutation of crop species removal order to generate
an attack tolerance curve and uses the area under this curve as a coefficient of stability. Consider
the set of M crops $C = \{c_1, \cdots, c_M\}$ and the set of P nutrients $N = \{n_1, \cdots, n_P\}$. We begin by
constructing a bipartite graph, G, with left nodes consisting of the set C and right nodes consisting
of the set N. Place an edge between c_i (the i-th crop) and n_j (the j-th nutrient) if crop c_i contains
nutrient n_j . Given a subset of crops $D \subset C$, define $nbr(D)$ to be the neighborhood of D, meaning
the subset of N that is connected by an edge to an element of D. Finally, given a set A, let $|A|$ be
the number of elements in A (thus $|N(D)|$ is the number of elements that D is connected to). Now
consider the following sampling procedure
256   1. Initialize $C^0 = C$
257   2. At step k, remove one randomly selected crop from C^{k-1}, leaving the set C^k , which
258   contains $M - k$ crops. Let $X(k) = |nbr(C^k)|$, the number of nutrients connected to C^k .
259   3. Repeat Step 2 M times (i.e., until there are no more crops left).
260
Thus, at each iteration the procedure described above removes a randomly selected crop and all of
its edges from G, and counts the number of nutrients that remain connected to rest of the graph.
The procedure yields a trajectory $X(k)$, with k ranging from 1 to M, which we will call a sample.
Repeating this procedure multiple times yields multiple samples $X_i(k)$. We generate an average
attack tolerance and take its integral as our coefficient of nutritional stability (R_N). This unitless
measure represents how robust a food system is to the sequential elimination of crops.
267
Statistical analyses
To explore the relationship between crop diversity and nutritional stability (Fig. 2) we compared
three non-linear functional forms (saturating, logarithmic, exponential) using non-linear mixed
effects models with crop diversity as a fixed effect and country as a random effect. A saturating
function ($\alpha * x/(\beta + x)$) was selected based on AIC (Table S5) and used for subsequent analyses.
We used separate non-linear models for each region to explore the relationship between
nutritional stability and crop diversity averaged across years for each country. We test whether
response curves for this relationship are different between regions by extracting individual
country regression coefficients for the model parameters (asymptote, α; doubling time, β) from
non-linear mixed effects models with country as a random effect. We then used linear models to
compare parameter coefficient values between regions (Fig S1; Table S6). We conducted these
analyses using production scenario crop information, but results did not qualitatively change if
imports were included. To explore changes in crop diversity, nutritional stability and average
crop degree over time (Fig. 3) we used separate linear mixed effects models for each region with
an interaction between supply scenario and year as fixed effects, a random intercept for scenario
nested in country, and a first order autoregressive (AR-1) correlation structure to account for the
temporal dependency structure of our data. Finally, to explore how nutritional stability differs
between macroeconomic factors (Fig. 4; small island state status, development status) we
averaged crop diversity across years for each country and used linear mixed effects models with
an interaction between macroeconomic status and scenario as fixed effects and country as a
random effect. We performed all analyses in R (v 3.6.1) using 'nlme' (39) and 'lme4' (40) for
statistical modeling.
290
291
292

Acknowledgments
We are thankful to Taylor Ricketts and Molly Brown for providing thoughtful comments and feedback on earlier versions of the manuscript. We would like to thank Rishidev Chaudhuri for comments and helping formalize our nutritional network approach. Funding for this work was provided by the USDA Hatch Program (grant number VT-H02303) given to MTN. The authors declare no conflicts of interest.

Data Accessibility
The data presented in this manuscript will be made publicly available through Figshare (doi: xx.xxxx/xxxx [to be completed before final submission]).

References

1. C. Béné, D. Headey, L. Haddad, K. von Grebmer, Is resilience a useful concept in the context of food security and nutrition programmes? Some conceptual and practical considerations. *Food Secur.* **8**, 123–138 (2016).

2. M. T. Ruel, H. Alderman, Nutrition-sensitive interventions and programmes: How can they help to accelerate progress in improving maternal and child nutrition? *Lancet* **382**, 536–551 (2013).

3. R. DeFries, *et al.*, Metrics for land-scarce agriculture. *Science (80-.).* **349**, 238–240 (2015).

4. R. DeFries, *et al.*, Synergies and trade-offs for sustainable agriculture: Nutritional yields and climate-resilience for cereal crops in Central India. *Glob. Food Sec.* **11**, 44–53 (2016).

5. J. Fanzo, D. Hunter, T. Borelli, F. Mattei, *Diversifying food and diets: Using agricultural biodiversity to improve nutrition and health* (2013) https:/doi.org/10.4324/9780203127261.

6. A. D. Jones, Critical review of the emerging research evidence on agricultural biodiversity, diet diversity, and nutritional status in low- and middle-income countries. *Nutr. Rev.* (2017) https:/doi.org/10.1093/nutrit/nux040.

7. K. T. Sibhatu, M. Qaim, Review: The association between production diversity, diets, and nutrition in smallholder farm households. *Food Policy* (2018) https:/doi.org/10.1016/j.foodpol.2018.04.013.

8. J. C. Fanzo, Decisive Decisions on Production Compared with Market Strategies to Improve Diets in Rural Africa. *J. Nutr.* (2017) https:/doi.org/10.3945/jn.116.241703.

9. R. Remans, F. A. J. DeClerck, G. Kennedy, J. Fanzo, Expanding the view on the production and dietary diversity link: Scale, function, and change over time. *Proc. Natl. Acad. Sci. U. S. A.* (2015) https:/doi.org/10.1073/pnas.1518531112.

10. D. Tobin, K. Jones, B. C. Thiede, Does crop diversity at the village level influence child nutrition security? Evidence from 11 sub-Saharan African countries. *Popul. Environ.*

335   (2019) https:/doi.org/10.1007/s11111-019-00327-4.

336 11. S. A. Wood, Nutritional functional trait diversity of crops in south-eastern Senegal. *J.*
337   *Appl. Ecol.* **55**, 81–91 (2018).

338 12. J. M. Bullock, *et al.*, Resilience and food security: rethinking an ecological concept. *J.*
339   *Ecol.* **105**, 880–884 (2017).

340 13. United Nations, Transforming our world: the 2030 Agenda for Sustainable Development.
341   United Nations Sustainable knowledge platform. *Sustain. Dev. Goals* (2015)
342   https:/doi.org/https://sustainabledevelopment.un.org/post2015/transformingourworld.

343 14. B. G. Luckett, F. A. J. DeClerck, J. Fanzo, A. R. Mundorf, D. Rose, Application of the
344   Nutrition Functional Diversity indicator to assess food system contributions to dietary
345   diversity and sustainable diets of Malawian households. *Public Health Nutr.* (2015)
346   https:/doi.org/10.1017/S136898001500169X.

347 15. F. A. J. DeClerck, F. Jessica, P. Cheryl, R. Roseline, Ecological approaches to human
348   nutrition. *Food Nutr. Bull.* **32**, S41–S50 (2011).

349 16. R. Remans, S. A. Wood, N. Saha, T. L. Anderman, R. S. DeFries, Measuring nutritional
350   diversity of national food supplies. *Glob. Food Sec.* **3**, 174–182 (2014).

351 17. R. Albert, H. Jeong, A.-L. Barabási, Error and attack tolerance of complex networks.
352   *Nature* **406**, 378–382 (2000).

353 18. J. A. Dunne, R. J. Williams, N. D. Martinez, Network structure and biodiversity loss in
354   food webs: Robustness increases with connectance. *Ecol. Lett.* **5**, 558–567 (2002).

355 19. T. H. Larsen, N. M. Williams, C. Kremen, Extinction order and altered community
356   structure rapidly disrupt ecosystem functioning. *Ecol. Lett.* **8**, 538–547 (2005).

357 20. E. S. Zavaleta, K. B. Hulvey, Realistic species losses disproportionately reduce grassland
358   resistance to biological invaders. *Science (80-.).* **306**, 1175–1177 (2004).

359 21. E. Burgos, *et al.*, Why nestedness in mutualistic networks? *J. Theor. Biol.* **249**, 307–313
360   (2007).

361 22. M. R. Smith, R. Micha, C. D. Golden, D. Mozaffarian, S. S. Myers, Global Expanded
362   Nutrient Supply (GENuS) Model: A New Method for Estimating the Global Dietary
363   Supply of Nutrients. *PLoS One* **11**, e0146976 (2016).

364 23. D. Renard, D. Tilman, National food production stabilized by crop diversity. *Nature*
365   (2019) https:/doi.org/10.1038/s41586-019-1316-y.

366 24. C. K. Khoury, *et al.*, Increasing homogeneity in global food supplies and the implications
367   for food security. *Proc. Natl. Acad. Sci.* **111**, 4001–4006 (2014).

368 25. S. Naeem, Species redundancy and ecossystem reliability. *Conserv. Biol.* **12**, 39–45
369   (1998).

370 26. N. Steyn, J. Nel, G. Nantel, G. Kennedy, D. Labadarios, Food variety and dietary diversity

scores in children: are they good indicators of dietary adequacy? *Public Health Nutr.*
(2006) https:/doi.org/10.1079/phn2005912.

27.   B. J. Cardinale, *et al.*, Biodiversity loss and its impact on humanity. *Nature* **486**, 59–67
(2012).

28.   J. Dithmer, A. Abdulai, Does trade openness contribute to food security? A dynamic panel
analysis. *Food Policy* (2017) https:/doi.org/10.1016/j.foodpol.2017.04.008.

29.   C. S. Hendrix, H. J. Brinkman, Food insecurity and conflict dynamics: Causal linkages
and complex feedbacks. *Stability* (2013) https:/doi.org/10.5334/sta.bm.

30.   P. White, War and food security in Eritrea and Ethiopia, 1998-2000. *Disasters* (2005)
https:/doi.org/10.1111/j.0361-3666.2005.00286.x.

31.   L. Smets, H. A. Tusiime, R. Renard, Food aid and household food security in a conflict
situation: Empirical evidence from Northern Uganda. *Food Policy* (2013)
https:/doi.org/10.1016/j.foodpol.2013.07.005.

32.   F. Gaupp, J. Hall, S. Hochrainer-Stigler, S. Dadson, Changing risks of simultaneous
global breadbasket failure. *Nat. Clim. Chang.* (2020) https:/doi.org/10.1038/s41558-019-
0600-z.

33.   S. A. Wood, M. R. Smith, J. Fanzo, R. Remans, R. S. Defries, Trade and the equitability
of global food nutrient distribution. *Nat. Sustain.* **1**, 34–37 (2018).

34.   S. Friel, A. Schram, B. Townsend, The nexus between international trade, food systems,
malnutrition and climate change. *Nat. Food* (2020) https:/doi.org/10.1038/s43016-019-
0014-0.

35.   J. S. Baker, *et al.*, Evaluating the effects of climate change on US agricultural systems:
Sensitivity to regional impact and trade expansion scenarios. *Environ. Res. Lett.* (2018)
https:/doi.org/10.1088/1748-9326/aac1c2.

36.   C. Carletto, D. Jolliffe, R. Banerjee, From Tragedy to Renaissance: Improving
Agricultural Data for Better Policies. *J. Dev. Stud.* (2015)
https:/doi.org/10.1080/00220388.2014.968140.

37.   C. Zhao, *et al.*, Temperature increase reduces global yields of major crops in four
independent estimates. *Proc. Natl. Acad. Sci. U. S. A.* (2017)
https:/doi.org/10.1073/pnas.1701762114.

38.   M. Arimond, M. T. Ruel, Dietary Diversity Is Associated with Child Nutritional Status:
Evidence from 11 Demographic and Health Surveys. *J. Nutr.* (2004)
https:/doi.org/10.1093/jn/134.10.2579.

39.   J. C. Pinheiro, D. M. Bates, The NLME package. *October* (2010).

40.   D. Bates, M. M chler, B. Bolker, S. Walker, Fitting Linear Mixed-Effects Models Using
lme4. *J. Stat. Softw.* **67**, 1–48 (2015).

Figures and Tables

Figure 1 A new approach for understanding nutritional stability. The robustness of networks, such as a crop-nutrient network (a), can be gauged through assembling attack tolerance curves (b & c), whereby as nodes (i.e. crops) are eliminated constituent nutrients are lost. Of course, removal order matters (b vs c), therefore permuted combinations of crop loss generate average loss curves (see Methods). We use the area under this curve as our measure of nutritional stability (R_N). Crop-nutrient networks with few redundant connections are susceptible to crop removal and can experience rapid nutrient loss (e.g. Maldives; d), whereas countries with a diverse food supply are more robust (e.g. China; e). This unitless metric is generalizable across different levels of aggregation (e.g. individual, household, community, national food system).

424
425
426   **Figure 2** Nutritional stability increased non-linearly with crop diversity. Individual curve's
427   doubling-time parameter (β) differed significantly from each other (Fig. S1; Table S1), indicating
428   region-dependent responses to crop diversity. Each point is a country average across all years.
429

430
Figure 3 Global patterns and trends of crop diversity and nutrient stability. Crop diversity
increased over the 1961-2016 period for production plus imports (blue) and production alone
(black) scenarios (a). Nutritional stability (R_N) varies over space (b) and through time (c).
Bivariate choropleth maps depict R_N mean and variance across years. For example, light purple
countries experienced low average nutritional stability and high year-to-year variation. Countries
filled with dark gray lacked sufficient data. Trends in R_N (c) varied between region and depended
on supply scenario (Table 1), with Africa and Asia exhibiting diverging responses for food
supplies derived from production plus imports (blue) and production alone (black). Trend lines
depict means ± 95% confidence intervals. The percentage change over the 55 year period for each
regions and scenario is in the lower left of each panel.
441

442
443

444
445 **Figure 4** National factors related to nutritional stability. Developing states (a, light gray) and
446 small island states (b, light gray) often experienced low nutritional stability. Each point is a
447 country's nutritional stability averaged across years (black: production alone; blue: production
448 plus imports). Small island states are included in developing states. See Table S3 for statistical
449 comparisons, * $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$.
450

 Figure 5 Country-level changes in crop diversity, degree and nutritional stability. Each point
depicts the difference between a country's last (2016) and first annual record of crop diversity
(i.e., number of crops) and crop degree (i.e., average number of nutrients connected to each crop)
for production (P; white circles) and production plus imports (PI; gray circles). The response
surface projects the expected change in nutritional stability as predicted by simple linear
regression ($R_N \sim$ diversity + degree).

461 **Table 1** Nutritional stability trends over time. Results are from region-specific linear mixed
462 effects model with an interaction between supply scenario (production + imports v. production
463 alone) and year as fixed-effects, country nested in scenario as random effects and an
464 autoregressive correlation structure to account for temporal autocorrelation. Values are model
465 coefficients with standard error in parentheses.
466

	Nutritional stability				
	Africa	Americas	Asia	Europe	Oceania
Supply Scenario	-1.360^{**}	$-1.391^{.}$	-2.197^{*}	-0.394	-0.452
	(0.412)	(0.816)	(0.822)	(0.306)	(0.548)
Year	-0.0004^{**}	-0.001^{**}	-0.0003	0.0002^{*}	-0.001^{**}
	(0.0001)	(0.0003)	(0.0003)	(0.0001)	(0.0002)
Supply Scenario × Year	0.001^{***}	$0.001^{.}$	0.001^{**}	0.0002	0.0002
	(0.0002)	(0.0004)	(0.0004)	(0.0002)	(0.0003)
Observations	5,546	4,217	4,577	3,076	1,628
Log Likelihood	13,648.020	9,471.530	9,456.794	7,189.631	4,561.369
Note:	* $p<0.05$; ** $p<0.01$; *** $p<0.001$				

467
468
469

472 **TABLES**

474 **Table S1.** Parameter estimates testing for a non-linear relationship between nutritional stability
475 and crops diversity. Curves fit with a saturating function (a * x/(b + x)). This functional form
476 was selected after multiple model comparison (**Table S5**). Individual models were fit for each
477 region. For details on regional differences see **Figure S1** and **Table S6**. Values are model
478 coefficients with standard error in parentheses.

	Nutritional stability ~ a*x/(b + x)				
	Africa	Americas	Asia	Europe	Oceania
a	1.097^{***}	1.335^{***}	1.131^{***}	0.995^{***}	0.979^{***}
	(0.042)	(0.099)	(0.052)	(0.015)	(0.027)
b	4.965^{***}	12.574^{***}	6.738^{***}	2.353^{***}	2.263^{***}
	(0.814)	(2.486)	(1.085)	(0.366)	(0.299)
Observations	50	39	46	33	15
Log Likelihood	67.642	39.932	56.713	71.664	26.597
Note:	. $p<0.1$; * $p<0.05$; ** $p<0.01$; *** $p<0.001$				

Table S2. Crop diversity trends over time. Results are from region-specific linear mixed effects
model with an interaction between scenario and year as fixed-effects, country nested in scenario
as random effects and an autoregressive correlation structure (i.e., time-lag correlation) to account
for temporal autocorrelation. Values are model coefficients with standard error in parentheses.

	Crop diversity				
	Africa	Americas	Asia	Europe	Oceania
Scenario	-59.466[*]	-75.863[*]	-130.076[*]	-219.441[***]	-31.948
	(23.739)	(33.737)	(50.347)	(36.733)	(35.314)
Year	0.047[***]	0.060[***]	0.071[***]	0.047[***]	0.018
	(0.008)	(0.012)	(0.018)	(0.013)	(0.013)
Scenario × Year	0.030[*]	0.039[*]	0.067[**]	0.113[***]	0.017
	(0.012)	(0.017)	(0.025)	(0.018)	(0.018)
Observations	5,546	4,217	4,577	3,076	1,628
Log Likelihood	-8,590.126	-7,130.420	-8,581.432	-5,601.669	-2,366.630
Note:		. $p<0.1$; * $p<0.05$; ** $p<0.01$; *** $p<0.001$			

491 **Table S3** Macroeconomic factors drive nutritional stability differences. Differences (estimate
492 standard error in parentheses) in nutritional stability between (a) developing and non-developing
493 countries and (b) small island developing states (SIDS). Results are from separate linear mixed
494 effects models with an interaction between macroeconomic status and scenario as fixed-effects
495 and country as a random effect.

(a)

scenario	contrast	estimate	SE	df	t ratio	p value
P	Developing - Non-developing	-0.0674	0.0265	363	-2.541	0.0115
PI	Developing - Non-developing	-0.0693	0.0265	363	-2.612	0.0094

(b)

scenario	contrast	estimate	SE	df	t ratio	p value
P	Non-SIDS - SIDS	0.133	0.0257	363	5.172	< 0.0001
PI	Non-SIDS - SIDS	0.129	0.0255	363	5.083	< 0.0001

498 **Table S4** Crop degree trends over time. Results are from region-specific linear mixed effects
499 model with an interaction between scenario and year as fixed-effects, country nested in scenario
500 as random effects and an autoregressive correlation structure (i.e., time-lag correlation) to account
501 for temporal autocorrelation. Values are model coefficients with standard error in parentheses.

	Crop degree				
	Africa	Americas	Asia	Europe	Oceania
Scenario	-4.168	-8.645	-0.841	35.491***	-9.692
	(7.645)	(9.645)	(10.455)	(5.733)	(16.069)
Year	-0.022***	-0.024***	-0.020***	-0.011***	-0.015**
	(0.003)	(0.003)	(0.004)	(0.002)	(0.006)
Scenario × Year	0.002	0.004	0.0005	-0.018***	0.005
	(0.004)	(0.005)	(0.005)	(0.003)	(0.008)
Observations	5,546	4,217	4,577	3,076	1,628
Log Likelihood	-4,488.510	-2,567.151	-3,741.008	-2,694.267	-1,716.429
Note:		. p<0.1; * p<0.05; ** p<0.01; *** p<0.001			

505 **Table S5** Comparing the relationship between nutritional stability and crop diversity using three
506 saturating model forms. Based on AIC scores the saturating function $\alpha * x/(\beta + x)$ was used in
507 subsequent analyses (Fig. 2; Table S1 & Table S6). Values are model coefficients with standard
508 error in parentheses.

	Nutritional stability		
	$\alpha + \beta * \log(x)$	$\alpha * x/(\beta + x)$	$\alpha * \exp(\beta * x)$
b	0.161^{***}	5.126^{***}	0.017^{***}
	(0.004)	(0.462)	(0.001)
a	0.234^{***}	1.085^{***}	0.592^{***}
	(0.002)	(0.023)	(0.017)
Observations	183	183	183
Log Likelihood	184.529	200.093	125.840
AIC	-359.059	-390.186	-241.679
Note:	. $p<0.1$; * $p<0.05$; ** $p<0.01$; *** $p<0.001$		

513 **Table S6** Comparing differences in parameter estimates for region-specific relationship between
514 nutrient stability and crop diversity (Africa is contrast reference). Curves fit with a saturating
515 function ($\alpha * x/(\beta + x)$) via non-linear mixed effects models (see Methods) and coefficient values
516 were extracted from random effects for each country. Values are model coefficients with standard
517 error in parentheses.

	Saturating function parameter	
	α	β
Americas	0.238***	7.619***
	(0.00000)	(0.100)
Asia	0.034***	1.762***
	(0.00000)	(0.095)
Europe	-0.102***	-2.602***
	(0.00000)	(0.105)
Oceania	-0.118***	-2.692***
	(0.00000)	(0.137)
Observations	183	183

Note: . $p<0.1$; * $p<0.05$; ** $p<0.01$; *** $p<0.001$

FIGURES

Figure S1. Slope parameter estimates from non-linear mixed effects models relating crop
diversity and nutritional stability (see Fig. 1). Curves fit with a saturating function ($\alpha * x/(\beta + x)$)
via non-linear mixed effects models (**Table S6**; see Methods) and coefficient values were
extracted from random effects for each country. Points depict the average ± sd across countries
for the α parameter (a) and β parameter (b).

Figure S2. Average degree of crops in crop-nutrient networks decreased over time. Only Europe exhibited scenario-dependent differences, with production plus imports (blue) decreasing more than production alone (black), see **Table S4** for statistics. Trend lines depict means ± 95% confidence intervals.

537
538
539