

Biodiversity increases and decreases ecosystem stability

Frank Pennekamp, Mikael Pontarp, Andrea Tabi, Florian Altermatt, Roman Alther, Yves Choffat, Emanuel A Fronhofer, Pravin Ganesanandamoorthy, Aurélie Garnier, Jason Griffiths, et al.

► To cite this version:

Frank Pennekamp, Mikael Pontarp, Andrea Tabi, Florian Altermatt, Roman Alther, et al.. Biodiversity increases and decreases ecosystem stability. Nature, 2018, 563 (7729), pp.109-112. 10.1038/s41586-018-0627-8. hal-01944370

HAL Id: hal-01944370 https://hal.umontpellier.fr/hal-01944370

Submitted on 12 Dec 2018

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Biodiversity increases and decreases ecosystem stability

- 3
- 4 Frank Pennekamp^a email: frank.pennekamp@ieu.uzh.ch
- 5 Mikael Pontarp^{a, b} mikael.pontarp@ieu.uzh.ch
- 6 Andrea Tabi^a andrea.tabi^(a)ieu.uzh.ch
- 7 Florian Altermatt^{a,c} florian.altermatt@eawag.ch
- 8 Roman Alther^{a,c} roman.alther@eawag.ch
- 9 Yves Choffat^a yves.choffat@ieu.uzh.ch
- 10 Emanuel A. Fronhofer^{a,c,g} emanuel.fronhofer@umontpellier.fr
- 11 Pravin Ganesanandamoorthy^{a, c} pravin.ganesanandamoorthy@eawag.ch
- 12 Aurélie Garnier^a aurelie.garnier.research@gmail.com
- 13 Jason I. Griffiths^f jasonigriff@gmail.com
- 14 Suzanne Greene^{a,e} segreene@mit.edu
- 15 Katherine Horgan^a katherine.horgan@ieu.uzh.ch
- 16 Thomas M. Massie^a thomas.massie@outlook.com
- 17 Elvira Mächler^{a,c} elvria.maechler@eawag.ch
- 18 Gian-Marco Palamara^{a,h} gianmarco.palamara@gmail.com
- 19 Mathew Seymour^{c, d} m.seymour@bangor.ac.uk
- 20 Owen L. Petchey^a email: owen.petchey@ieu.uzh.ch
- 2122 Affiliations:
- 23 ^aDepartment of Evolutionary Biology and Environmental Studies, University of Zurich,
- 24 Winterthurerstrasse 190, 8057 Zurich, Switzerland
- ^bDepartment of Ecology and Environmental Science, Umeå University, 90187 Umeå,
- 26 Sweden
- 27 °Department of Aquatic Ecology, Eawag: Swiss Federal Institute of Aquatic Science and
- 28 Technology, Überlandstrasse 133, 8600 Dübendorf, Switzerland
- 29 ^fDepartment of Mathematics, University of Utah, Salt Lake City, UT, 84112,USA
- 30
- 31 Present addresses:
- 32 ^dMolecular Ecology and Fisheries Genetics Laboratory, School of Biological Sciences,
- 33 Environment Centre Wales Building, Bangor University, Bangor, Gwynedd, UK
- ³⁴ ^eMassachusetts Institute of Technology, 77 Massachusetts Avenue, Cambridge, MA USA
- 35 ^gISEM, Université de Montpellier, CNRS, IRD, EPHE, Montpellier, France
- 36 ^hDepartment Systems Analysis, Integrated Assessment and Modelling, Eawag: Swiss Federal
- 37 Institute of Aquatic Science and Technology, Überlandstrasse 133, 8600 Dübendorf,
- 38 Switzerland
- 39

40 This is a post-peer-review, pre-copyedit version of an article published in Nature.

- 41 The final authenticated version is available online at:
- 42 https://doi.org/10.1038/s41586-018-0627-8
- 43 44

Losses and gains in species diversity affect ecological stability¹⁻⁷ and the sustainability 45 of ecosystem functions and services^{8–13}. Experiments and models reveal positive, 46 negative, and no effects of diversity on individual components of stability such as 47 temporal variability, resistance, and resilience^{2,3,6,11,12,14}. How these stability components 48 covary is poorly appreciated¹⁵, as are diversity effects on overall ecosystem stability¹⁶, 49 conceptually akin to ecosystem multifunctionality^{17,18}. We observed how temporal 50 51 variability, resistance, and overall ecosystem stability responded to diversity (i.e. species 52 richness) in a large experiment involving 690 micro-ecosystems sampled 19 times over 53 40 days, resulting in 12939 samplings. Species richness increased temporal stability but 54 decreased resistance to warming. Thus, two stability components negatively covaried 55 along the diversity gradient. Previous biodiversity manipulation studies rarely reported 56 such negative covariation despite general predictions of negative effects of diversity on 57 individual stability components³. Integrating our findings with the ecosystem 58 multifunctionality concept revealed hump- and U-shaped effects of diversity on overall 59 ecosystem stability. That is, biodiversity can increase overall ecosystem stability when 60 biodiversity is low, and decrease it when biodiversity is high, or the opposite with a U-61 shaped relationship. Effects of diversity on ecosystem multifunctionality would also be 62 hump- or U-shaped if diversity has positive effects on some functions and negative 63 effects on others. Linking the ecosystem multifunctionality concept and ecosystem 64 stability can transform perceived effects of diversity on ecological stability and may 65 assist translation of this science into policy-relevant information.

66

67 Ecological stability consists of numerous components including temporal variability, resistance to environmental change, and rate of recovery from disturbance^{1,2,16}. Effects of 68 69 species losses and gains on these components are of considerable interest, not least due to 70 potential effects on ecosystem functioning and hence the sustainable delivery of ecosystem 71 services^{1–13}. A growing number of experimental studies reveal stabilising effects of diversity 72 on individual stability components. In particular, higher diversity often, but not always, 73 reduces temporal variability of biomass production¹³. Positive effects of diversity on 74 resistance are common, though neutral and negative effects on resistance and resilience also 75 occur^{9,13,19,20}. While assessment of individual stability components is essential, a more 76 integrative approach to ecological stability could lead to clearer conceptual understanding¹⁵ 77 and might improve policy guidance concerning ecological stability¹⁶.

78 Analogous to ecosystem multifunctionality^{17,18}, a more integrative approach considers 79 variation in multiple stability components, and the often-ignored covariation among stability 80 components. The nature of this covariation is of paramount importance, as it defines whether 81 diversity has consistent effects on multiple stability components, or whether some stability 82 components increase with diversity while others decrease. Surprisingly, the nature, 83 prevalence, and implications of negative covariation between stability components along 84 diversity gradients are almost completely overlooked, including the ensuing possibility for 85 non-monotonic effects of diversity on overall ecosystem stability.

We first describe new experimental findings of how biodiversity affects the intrinsic 86 87 stability of ecosystems and their resistance to warming. Temperature is a highly relevant 88 disturbance due to its importance for biological processes and its great variability through 89 space and time. However, our findings equally apply to and have implications for other 90 environmental changes that could result in opposing effects on stability components such as 91 flooding¹² or chemical stress²¹. We then review other evidence for negative covariation in 92 effects of diversity on stability and potential mechanisms. Finally, we analyse overall 93 ecosystem stability, a concept that embraces the covariation between stability components 94 and their weighting, and show the plausibility of previously overlooked non-monotonic 95 (hump- and U-shaped) effects of diversity on overall ecosystem stability.

96 We performed a factorial manipulation of the diversity and composition of competing 97 species (1 to 6 species, 53 unique community compositions) and temperature (six constant 98 levels, modelled as a linear predictor) in microbial communities of bacterial consumers, and 99 recorded community biomass dynamics over time. For each replicate we then calculated two 100 stability components: resistance (= [total biomass at $T^{\circ}C$ – total biomass at $15^{\circ}C$] / [$T^{\circ}C$ – 15°C] where T is the temperature of the replicate) and the temporal stability of biomass 101 102 (inverse of coefficient of variation of community biomass). While these stability indices are 103 widely used by empiricists, they should not be mistaken for mathematical definitions such as asymptotic resilience, which are more precise but also more restrictive²². 104

Increased species richness caused greater temporal stability of total biomass (figure 1a1) (linear mixed model of log inverse CV: richness effect size 0.33 with a standard error of
0.065) at all temperatures (extended data figure 1). Total biomass increased during the first
week of the experiment and then declined over the next five weeks and total biomass was
higher in more species-rich communities (figure 1-b, 1-c2, extended data table 1) (effect size
for log richness 0.05 [units of mg/mL/log(species richness) unit] with 0.0096 standard error).

111 In contrast, increased species richness decreased resistance of total biomass to 112 warming (figure 1-a2) (negative effect of log richness in a linear model, effect size of -0.006 [mg/°C/ log(species richness) unit] with a standard error of 0.0018). Richness negatively 113 114 affected resistance measured on both absolute and relative scales (extended data figure 2). 115 This effect was corroborated in analyses of total biomass by a negative interaction term 116 between temperature and richness, which persisted through the experiment except during the 117 first days (figure 1-c3) (log(richness) x temperature interaction of -0.0053 [units of 118 mg/mL/°C/log(species richness) unit] with standard error of 0.00051) despite large variation 119 in dynamics of total biomass (figure 1-b). This negative interaction reflects a stronger 120 negative effect of temperature on total biomass (i.e. lower resistance) in richer communities 121 (i.e. a richness-dependent response of total biomass to temperature).

122 Hence, temporal stability and resistance were negatively correlated across the species 123 richness gradient (figure 1-d, RMA analysis with slope = -0.009, 95% CI = -0.0178 to -124 0.0051). Niche complementarity, statistical averaging, low overall response diversity, and 125 possibly lower response diversity in more diverse communities were likely causes of the 126 opposite effects of richness on temporal stability (extended data figure 3). The two stability 127 components were, however, positively correlated within any single level of species richness 128 (figure 1-d, extended data table 2). That is, composition variation without changes in species 129 richness resulted in positively covarying temporal stability and resistance.

130

131 Next, we examined studies (including our own) measuring multiple stability 132 components across diversity gradients based on a review by Donohue et al. $(2016)^{16}$ (figure 133 2, extended data table 3 & 4). Seven of 30 comparisons show positive covariance, twenty 134 show no covariance, and three showed negative covariance. Our study for the first time 135 identifies negative covariation between resistance and temporal variability caused by intrinsic 136 dynamics only. Although infrequently reported, negative covariation is disproportionately 137 important because it complicates conclusions about and practical implications of effects of 138 diversity on stability. Furthermore, these studies may be unrepresentative of the true 139 prevalence of negative covariation, due to it being overlooked, publication bias towards 140 positive diversity-stability relationships³ or if the scale of analysis masks such covariation, 141 e.g. within richness versus across richness.

A general mechanistic understanding of why different studies find different correlations would be a major step forward. Of the 30 pairs of stability components, only seven were accompanied by quantitative analyses of mechanism for both diversity-stability relationships (extended data table 4). Response diversity was implicated in five of these seven. Indeed, response diversity has been identified as an important driver of the resilience of ecological systems^{23,24}, and correlation among effect (i.e. high biomass production) and 148 response (e.g. response to an environmental driver) traits, at least in the absence of strong 149 interspecific interactions, might predict covariance among stability components²³. None of 150 the stability is investigated and a stability of stability of stability.

150 the studies involved manipulation of mechanism.151

152 Negative covariation also raises the potential for non-monotonic effects of diversity 153 on overall ecosystem stability. Ecosystem multifunctionality is the simultaneous 154 consideration of several ecosystem functions. Similarly, overall ecosystem stability considers 155 multiple stability components simultaneously¹⁶ (see extended data figure 4 for clarification of terms and concepts). We first calculated overall ecosystem stability as the sum of 156 standardised individual ecosystem stability components²⁵. For the results of our experiment, 157 in which one stability component increases linearly with diversity, and another decreases 158 159 linearly (figure 1-a1,2), overall ecosystem stability is invariant with species richness since the 160 two standardized components perfectly balance each other.

161 We further explored the implications of the negative covariation among stability 162 components. Existing methods for calculating ecosystem multifunctionality include a 163 mathematical function for converting values of each ecosystem function into a common currency. Following others²⁶, we assumed a logistic shape conversion function, and 164 165 converted observed stability components into the common currency, with parameter values 166 chosen to ensure that the range of the data includes the lower and upper asymptote. Here we present the influence of the value of the intercept (parameter Q) of the conversion function 167 168 (figure 3a), which is similar to varying the threshold value when calculating ecosystem 169 multifunctionality with the threshold approach²⁷.

The opposing stability-diversity relationships from our experiment, coupled with 170 logistic conversion functions produced a hump-shaped relationship between overall 171 172 ecosystem stability and diversity at low values of Q, a flat relationship at intermediate values, 173 and a U-shaped relationship at higher values (figure 3b,c). Effects of richness on overall 174 ecosystem stability were weaker at the replicate level than the richness level (figure 3b) in 175 part due to relatively large variability among the replicates within richness levels and also 176 due to positive covariance of temporal stability and resistance within diversity levels (figure 177 1d).

178 Finding hump- and U-shaped diversity-stability relationships fundamentally alters the 179 importance of biodiversity, from it being unimportant (variation in biodiversity has no 180 consequence), to being state dependent. The hump-shaped and U-shaped relationships 181 occurred because measured stability components (e.g. resistance) cross the threshold of the 182 conversion function at different levels of species richness (see link to interactive figure in 183 Methods). Thus, negatively covarying stability components and non-linear conversion 184 functions can produce a non-monotonic effect of diversity on overall ecosystem stability. 185 Furthermore, we expect the same to apply to ecosystem multifunctionality when individual 186 ecosystem functions negatively covary along a diversity gradient.

187

188 The chosen mathematical conversion function and its parameter values are therefore 189 important. Ecosystem multifunctionality-diversity relationships can be negative, neutral or positive depending on the chosen threshold value²⁷. Similarly, different weightings of 190 191 ecosystem functions can alter which species and interactions are considered important for 192 multifunctionality²⁶. Other types and parameterisations of conversion functions might 193 conceivably result in other diversity-stability relationships. For instance, weighting one 194 ecosystem function (or stability component) highly, would lead to a multifunctionality / 195 overall ecosystem stability measure that is essentially univariate. We view the choice and 196 parameterization of conversion functions as an opportunity to tailor estimates of ecosystem 197 multifunctionality and overall ecosystem stability to their policy and decision making applications^{16,26,28,29}. These conversion functions and estimates of overall ecosystem stability 198 199 can then transform the perceived importance of biodiversity change for the sustainable

200 delivery of multiple ecosystem services and may assist in translating the results of scientific 201 studies into actionable information.

202

203 Methods section

204 Experimental methods

205 The experiment was conceived and designed to research biodiversity-ecosystem

206 relationships, with a focus on questions on environmental gradients and effects on temporal

- 207 changes / stability, including the one addressed in this article. We factorially manipulated
- temperature (15, 17, 19, 21, 23 and 25 °C) and species richness (1 to 6 species of
- 209 bacterivorous ciliates: Colpidium striatum, Dexiostoma campylum, Loxocephalus sp.,
- 210 Paramecium caudatum, Spirostomum teres, and Tetrahymena thermophila) in 690
- 211 microcosms (250 mL Duran bottles). There were two incubators at each temperature.
- 212 Manipulating temperature with a replicated gradient is recommended to harness the power of 212 α represented as α while still all subscripts as to test for a replicated gradient is recommended to harness the power of
- 213 a regression design, while still allowing us to test for a nonlinear temperature effect³⁰. During 214 prior testing the temperature of liquid in similar microcosms varied around the set-point
- 215 temperature by 0.1°C. Long-term ciliate cultures were kept at 15 °C, and warming usually
- 215 decreases their carrying capacities but increases growth rates³¹. For *Colpidium*, temperatures
- above 20 °C represent a stress that decreases population growth rate, whereas the other
- 218 species are still within their tolerated thermal range.

To start the experiment, ciliates were grown to their respective carrying capacities at 20°C in bottles containing 1L of bacterized medium. Throughout the experiment, medium consisted of protist pellets (Carolina Biological Supplies, Burlington, NC, USA) at a concentration of 0.055g L⁻¹ of Chalkley's medium in which the bacterium *Serratia fonticola* was grown as the resource for all ciliate consumers. Two autoclaved wheat seeds were added to each bottle for slow nutrient release³².

225 Monocultures (species richness = 1) were initiated at a density of 3 individuals mL^{-1} 226 in a total of 100mL medium. Multispecies communities (containing 2 to 6 species) were 227 initiated with a total of 40 mL ciliate culture topped up with 60 mL fresh medium (100 mL 228 culture in total). The 40mL cultures were assembled by adding a fixed fraction (i.e. 20 mL for 229 2 species, 13.33 mL for three species etc.) of each species at their specific carrying capacity, 230 adopting a substitutive design. Since the number of possible species compositions exceeded 231 the number of feasible experimental units, we used all possible compositions only for the 232 monocultures, two and six species communities. For all other levels, species compositions 233 were selected randomly from the set of all possible compositions such that all species 234 occurred the same number of times, resulting in a total of 53 different compositions. Each 235 level of species richness and composition was replicated at least twice, including an 236 additional replicate for the two and three species level, and 3 additional replicates for the six 237 species community resulting in 115 experimental units per temperature.

238 We sampled each experimental unit every day for the first 7 days, then 3 times per 239 week for the following 50 days and a final sampling 7 days later, resulting in time series of 240 27 time points over a 57-days period. We used video sampling techniques to count and measure individual ciliates in all communities³³. For sampling, microcosms were taken out of 241 242 the incubator, gently stirred to homogenize the culture and a sample was pipetted into a 243 counting chamber. The counting chamber was covered with a lid and a 5s long video was 244 taken under the microscope. The videos were subsequently processed with the R package BEMOVI³⁴. 245

We derived community biomass by summing the biovolume of all individuals of a given species in a given community and multiplying biovolume with a constant density equal to water (i.e. 1g/cm³). For each community, this resulted in a time series of community biomass. To avoid analysing monoculture time series whilst starting at a low fraction of the carrying capacity (3 individuals mL⁻¹), compared to our multi-species culture species which started between 7 and 20 % carrying capacity (depending on richness), we aligned the monoculture time series such that they started at the day where they crossed 20% of their
expected carrying capacity. Results regarding the effect of richness on temporal stability are
qualitatively robust to exclusion of monoculture data (extended data figure 1c).

255 The temporal stability of each replicate community was calculated as the inverse of 256 the coefficient of temporal variation in community biomass (mean community biomass / 257 standard deviation of community biomass). Resistance was measured for each replicate as the 258 absolute difference between total biomass in the replicate and the mean total biomass at 15°C, 259 divided by the temperature difference, with dimensions of milligrams (per ml) per degree 260 Celsius. Resistance values close to 0 indicate high resistance, whereas negative values 261 indicate lower resistance. We also quantified relative resistance as the difference between 262 total biomass in the replicate and the mean total biomass at 15°C divided by the mean total 263 biomass at 15°C (i.e. the loss proportional to the mean biomass at 15°C).

We did not quantify resilience since in a constant environment temporal variability is in principle closely related to resilience³⁵. To understand potential drivers and mechanisms, we investigated the role of statistical averaging and species asynchrony. Statistical averaging is assessed by the scaling relationship between the species mean biomass and species variance in biomass³⁶. Synchrony was assessed by a metric introduced by Gross et al. (2014)³⁷ which calculates the average correlation between the biomass of each species and the total biomass of all other species in the group. We used the R package codyn³⁸ to

271 calculate asynchrony.

272 Analyses

- 273 Analysis of total biomass used linear mixed effects models³⁹ (LMMs), with temperature
- 274 (linear, centred), species richness (log-linear), and their interaction as fixed effects (extended
- 275 data table 1a). Microcosm identity nested in composition was included as random factor to
- account for the repeated measurements. Heterogeneity of residuals was accounted for byestimating a composition-specific variance. AIC comparison consistently identified a model
- with linear temperature effect, as opposed to when temperature was a factor, as most
- 279 parsimonious (delta AIC of model with nonlinear temperature term was greater than 2 in all
- 280 comparisons). A simpler analysis of the temporal average of total biomass of each individual
- 281 microcosm to test the effect of richness, temperature, and their interaction, including a
- random effect for species composition, yielded quantitatively and qualitatively similar results
- (see extended data table 1b). Figure 1c shows the estimated effects of LMMs fitted separately
 for each day with temperature (linear, centred), species richness (log-linear), and their
- interaction as fixed effects and composition as a random effect. Analyses of temporal
- stability and resistance did not contain repeated measures and hence only accounted for
- variation in composition with a random effect. We used the R package nlme for all linear
- 288 mixed effects models⁴⁰. To calculate associations between resistance and temporal stability
- among and within richness levels, we calculated reduced major axis regressions (RMA) using
- 290 the R package $lmodel2^{41}$ as both variables potentially contain measurement error.
- 291 Significance of RMA slopes deviating from zero was assessed by one-tailed permutation
- tests. All analyses were done in R the statistical computing environment⁴².

293 Review of empirical studies

- Based on the review by Donohue et al. (2016)¹⁶ we obtained a set of studies of resilience,
- 295 resistance, and temporal variability of ecosystem functions in response to direct or indirect
- 296 experimental manipulations of diversity. Direct manipulations were defined as changing
- 297 diversity by adding different sets of species to an experimental plot, jar, or other unit,
- 298 whereas indirect manipulations induced variation in diversity via the experimental treatment,
- such as fertilization. We only included studies that performed experimental manipulations.
- To analyse whether specific mechanisms lead to covariation, we noted the type of mechanism
- 301 proposed for each of the individual diversity-stability components (extended data table 4).
- 302 Furthermore, we assessed whether a quantitative or verbal argument was provided (or the

- 303 mechanisms were not addressed at all) and synthesized the available evidence by vote
- 304 counting.

305 Calculating overall ecosystem stability

- 306 An interactive web page
- 307 (https://frankpennekamp.shinyapps.io/Overall_ecosystem_stability_demo/) describes the
- 308 calculation of ecosystem multifunctionality (also known as overall ecosystem functioning) or
- 309 overall ecosystem stability and illustrates the following. The calculation requires that values
- 310 of an ecosystem function (e.g. biomass production) or of a stability component
- 311 (e.g. resistance to temperature) be converted into a common currency. The threshold
- 312 approach uses a step mathematical function⁴³; the averaging approach uses a linear
- 313 mathematical function (and both equalise relative contributions of different ecosystem
- functions / stability components)²⁵; a principal component approach uses a specific linear mathematical function for each ecosystem function or stability component⁴⁴; and Slade et al.
- $(2017)^{26}$ propose step-like mathematical functions with more or less gradual changes from the
- 317 lower to higher value. The generalised logistic function (also known as the Richard's
- function) is flexible enough to give a wide range of shapes of conversion function. If x is the
- measured variable, and Y is the converted variable, the generalised logistic function is:
- 319 measured variable, and 1 is the converted variable, the generalised logistic function is. 320

$$Y = A + \frac{K - A}{(C + Qe^{-Bx})^{1/v}}$$

322

- 323 A is the lower asymptote.
- 324 K is the upper asymptote.
- B is the gradient.
- 326 v affects the symmetry, and also the value of y(0).
- 327 Q affects the value of y(0), i.e. it shifts the function horizontally.
- C is typically set to 1.
- x is a variable, here the value of the measured ecosystem function or stability component.
- 331 Overall ecosystem stability is then the sum of the standardised and converted stability

components OES = f(z(res)) + f(z(ts)), where res is the measured resistance, ts is the measured temporal stability, the function z() subtracts the mean and divides by the standard deviation, and f() is the generalised logistic function. The parameters of f() were A = -1, K = 1, B = 5, v = 1, C = 1 and Q was varied from 10^{-2} to 10^{2} . These values were chosen to produce converted stability measures that span the range A to K and to have a relatively

threshold-like change from A to K.

Standardisation prior to summation results in overall ecosystem stability with mean of zero, emphasising that the units of valuation here are arbitrary (though generally need not be). Standardisation also implies equal weights for different stability components; weighting of functions needs to be further considered and may be specified according to the specific use cases⁴⁵. Differential weightings, if desired and justified, can be incorporated into the conversions functions. Suggestions regarding the choice of conversion functions for managed systems can be found in Slade et al. 2017²⁶ and Manning et al. 2018²⁸.

345 Unimodal relationships can result from negative covariation among two stability 346 components. How does consideration of more than two components affect the unimodal 347 pattern? While the unimodal relationship is the most pronounced when equal numbers of 348 positive and negative relationships are considered, a unimodal relationship will persist as 349 long as there is at least one opposing stability component (see extended data figure 5).

350 Code availability

- 351 Code to reproduce the analyses and figures is accessible on Github
- 352 https://github.com/pennekampster/Code and data OverallEcosystemStability
- 353 (DOI: 10.5281/zenodo.1345557).
- 354
- 355 Data availability
- 356 The experimental data that support the findings of this study are available in Github
- 357 (https://github.com/pennekampster/Code_and_data_OverallEcosystemStability) with the
- 358 identifier (DOI: 10.5281/zenodo.1345557).). Source data for figures 1-3 are provided with
- the paper.

360 361	Re	ferences
362	1.	Pimm, S. L. The complexity and stability of ecosystems. <i>Nature</i> 307 , 321–326 (1984).
363	2.	McCann, K. S. The diversity-stability debate. Nature 405, 228-233 (2000).
364	3.	Ives, A. R. & Carpenter, S. R. Stability and Diversity of Ecosystems. Science 317, 58-62
365		(2007).
366	4.	Allesina, S. & Tang, S. Stability criteria for complex ecosystems. Nature 483, 205–208
367		(2012).
368	5.	Mougi, A. & Kondoh, M. Diversity of Interaction Types and Ecological Community
369		Stability. Science 337, 349–351 (2012).
370	6.	Loreau, M. & de Mazancourt, C. Biodiversity and ecosystem stability: a synthesis of
371		underlying mechanisms. Ecol. Lett. 16, 106–115 (2013).
372	7.	Grilli, J., Barabás, G., Michalska-Smith, M. J. & Allesina, S. Higher-order interactions
373		stabilize dynamics in competitive network models. Nature 548, 210–213 (2017).
374	8.	Tilman, D. & Downing, J. A. Biodiversity and stability in grasslands. <i>Nature</i> 367 , 363–
375		365 (1994).
376	9.	Pfisterer, A. B. & Schmid, B. Diversity-dependent production can decrease the stability
377		of ecosystem functioning. Nature 416, 84 (2002).
378	10	. Worm, B. et al. Impacts of Biodiversity Loss on Ocean Ecosystem Services. Science 314,
379		787–790 (2006).
380	11.	. Cardinale, B. J. et al. Biodiversity loss and its impact on humanity. Nature 486, 59-67
381		(2012).
382	12	. Wright, A. J. et al. Flooding disturbances increase resource availability and productivity
383		but reduce stability in diverse plant communities. Nat. Commun. 6, 6092 (2015).
384	13	. Isbell, F. et al. Biodiversity increases the resistance of ecosystem productivity to climate
385		extremes. Nature 526, 574–577 (2015).

386	14. Isbell, F. I., Polley, H. W. & Wilsey, B. J. Biodiversity, productivity and the temporal
387	stability of productivity: patterns and processes. Ecol. Lett. 12, 443-451 (2009).

- 388 15. Donohue, I. *et al.* On the dimensionality of ecological stability. *Ecol. Lett.* 16, 421–429
 389 (2013).
- 390 16. Donohue, I. *et al.* Navigating the complexity of ecological stability. *Ecol. Lett.* 19, 1172–
 391 1185 (2016).
- 392 17. Emmett Duffy, J., Paul Richardson, J. & Canuel, E. A. Grazer diversity effects on
 393 ecosystem functioning in seagrass beds. *Ecol. Lett.* 6, 637–645 (2003).
- 394 18. Hector, A. & Bagchi, R. Biodiversity and ecosystem multifunctionality. *Nature* 448,
 395 188–190 (2007).
- 396 19. Balvanera, P. *et al.* Quantifying the evidence for biodiversity effects on ecosystem
 397 functioning and services. *Ecol. Lett.* 9, 1146–1156 (2006).
- 398 20. Zhang, Q.-G. & Zhang, D.-Y. Resource availability and biodiversity effects on the
- 399 productivity, temporal variability and resistance of experimental algal communities.
- 400 *Oikos* **114**, 385–396 (2006).
- 401 21. Baert, J. M., De Laender, F., Sabbe, K. & Janssen, C. R. Biodiversity increases functional
- 402 and compositional resistance, but decreases resilience in phytoplankton communities.
- 403 *Ecology* **97**, 3433–3440 (2016).
- 404 22. Arnoldi, J.-F., Loreau, M. & Haegeman, B. Resilience, reactivity and variability: A
- 405 mathematical comparison of ecological stability measures. J. Theor. Biol. 389, 47–59
 406 (2016).
- 407 23. Suding, K. N. *et al.* Scaling environmental change through the community-level: a trait408 based response-and-effect framework for plants. *Glob. Change Biol.* 14, 1125–1140
 409 (2008).
- 410 24. Mori, A. S., Furukawa, T. & Sasaki, T. Response diversity determines the resilience of
 411 ecosystems to environmental change. *Biol. Rev. Camb. Philos. Soc.* 88, 349–364 (2013).

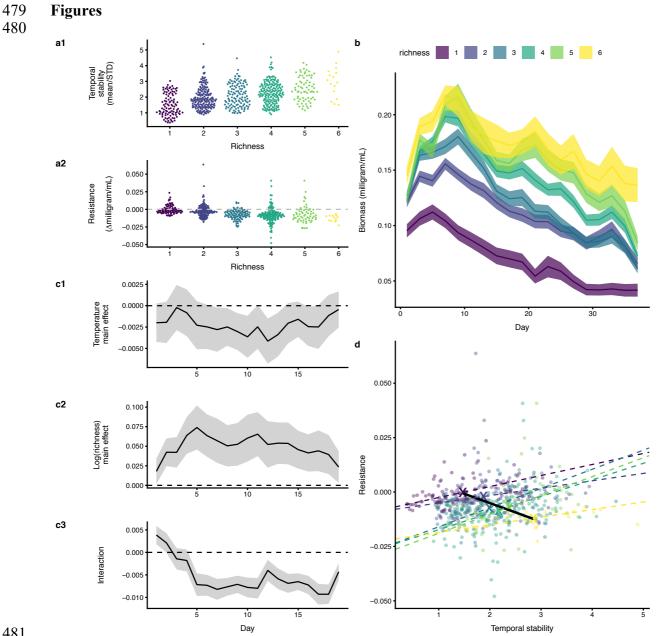
- 412 25. Maestre, F. T. et al. Plant species richness and ecosystem multifunctionality in global
- 413 drylands. *Science* **335**, 214–218 (2012).
- 414 26. Slade, E. M. et al. The importance of species identity and interactions for
- 415 multifunctionality depends on how ecosystem functions are valued. *Ecology* 98, 2626–
 416 2639 (2017).
- 417 27. Gamfeldt, L. & Roger, F. Revisiting the biodiversity–ecosystem multifunctionality
 418 relationship. *Nat. Ecol. Evol.* 1, s41559–017 (2017).
- 419 28. Manning, P. et al. Redefining Ecosystem Multifunctionality. Nat. Ecol. Evol. (2018).
- 420 29. Armsworth, P. R. & Roughgarden, J. E. The economic value of ecological stability. *Proc.*
- 421 Natl. Acad. Sci. 100, 7147–7151 (2003).
- 422 30. Cottingham, K. L., Lennon, J. T. & Brown, B. L. Knowing when to draw the line:
- 423 designing more informative ecological experiments. *Front. Ecol. Environ.* 3, 145–152
 424 (2005).
- 425 31. Leary, D. J. & Petchey, O. L. Testing a biological mechanism of the insurance hypothesis
 426 in experimental aquatic communities. *J. Anim. Ecol.* 78, 1143–1151 (2009).
- 427 32. Altermatt, F. *et al.* Big answers from small worlds: a user's guide for protist microcosms
- 428 as a model system in ecology and evolution. *Methods Ecol. Evol.* **6**, 218–231 (2015).
- 429 33. Pennekamp, F. et al. Dynamic species classification of microorganisms across time,
- 430 abiotic and biotic environments—A sliding window approach. *PLOS ONE* 12, e0176682
 431 (2017).
- 432 34. Pennekamp, F., Schtickzelle, N. & Petchey, O. L. BEMOVI, software for extracting
- 433 behavior and morphology from videos, illustrated with analyses of microbes. *Ecol. Evol.*434 5, 2584–2595 (2015).
- 435 35. May, R. M. Stability and complexity in model ecosystems. *Monogr. Popul. Biol.* 6, 1–
 436 235 (1973).

- 437 36. Tilman, D., Lehman, C. L. & Bristow, C. E. Diversity-stability relationships: statistical
- 438 inevitability or ecological consequence? *Am. Nat.* **151,** 277–282 (1998).
- 439 37. Gross, K. *et al.* Species richness and the temporal stability of biomass production: a new
 440 analysis of recent biodiversity experiments. *Am. Nat.* 183, 1–12 (2014).
- 38. Hallett, L. M. *et al.* codyn: An r package of community dynamics metrics. *Methods Ecol.*
- 442 *Evol.* **7**, 1146–1151 (2016).
- 39. Schmid, B., Baruffol, M., Wang, Z. & Niklaus, P. A. A guide to analyzing biodiversity
 experiments. *J. Plant Ecol.* 10, 91–110 (2017).
- 445 40. Pinheiro, J., Bates, D., DebRoy, S., Sarkar, D. & R Core Team. nlme: Linear and
- 446 *Nonlinear Mixed Effects Models.* (2018).
- 447 41. Legendre, P. *Imodel2: Model II Regression*. (2018).
- 448 42. R Core Team. *R: A language and environment for statistical computing*. (R Foundation
 449 for Statistical Computing, 2018).
- 450 43. Byrnes, J. E. K. et al. Investigating the relationship between biodiversity and ecosystem
- 451 multifunctionality: challenges and solutions. *Methods Ecol. Evol.* 5, 111–124 (2014).
- 452 44. Antiqueira, P. A. P., Petchey, O. L. & Romero, G. Q. Warming and top predator loss
- 453 drive ecosystem multifunctionality. *Ecol. Lett.* **21**, 72–82 (2018).
- 454 45. Gamfeldt, L., Hillebrand, H. & Jonsson, P. R. Multiple functions increase the importance
- 455 of biodiversity for overall ecosystem functioning. *Ecology* **89**, 1223–1231 (2008).
- 456

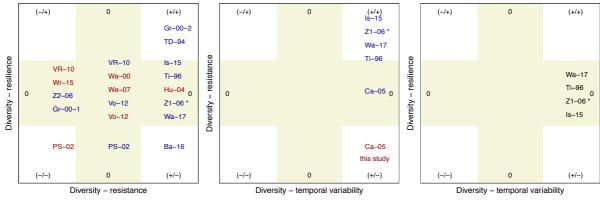
457 Acknowledgements

- 458 Frederik De Laender and Bernhard Schmid provided valuable feedback on previous drafts of
- 459 the article. Ian Donohue kindly donated the list of publications from his 2016 review paper.
- 460 The University of Zurich Research Priority Programme on Global Change and Biodiversity461 supported this research. Furthermore, funding came from the Swiss National Science
- 462 Foundation (grant PP00P3 150698 to FA, and 31003A 159498 to OLP).
- 463
- 464 **Author Contributions**:
- 465 Conceived study: OP, FP, FA
- 466 Designed experiment: OP, FP, MS, EAF, FA, GMP, TMM, MP
- 467 Led experiment: FP
- 468 Performed experiment: all, except JG, AT
- 469 Prepared data: FP, OP, JG

- Analysed data: FP, OP, MP, AT, MS Wrote the first draft: FP, OP
- Contributed to revisions of the manuscript: all
- **Competing interests:** The authors declare no competing interests.
- Correspondence and requests for materials should be addressed to F.P. or O.P.



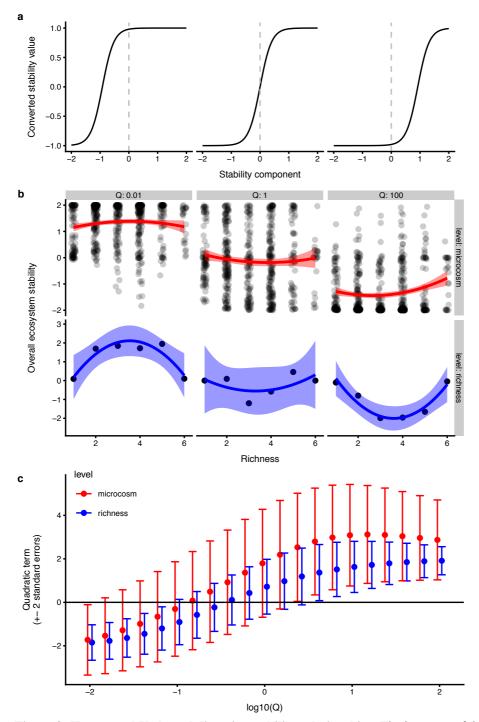
481 Day Temporal stability
482 Figure 1: Biomass and stability in experimental microbial communities. Richness increased temporal stability (a1, n=681 independent microcosms), but decreased resistance (a2, n = 567 independent microcosms).
484 Average (±1 standard error of the mean) biomass for each diversity level (b, n = 12939 microcosm x day combinations). Temporal dynamics of effect sizes (and 95% confidence intervals) of a linear mixed effects model of total biomass showed on average a negative effect of temperature (c1), a positive effect of diversity (c2), and a more negative effect of temperature at higher richness (persistent negative interaction term, c3) (n = 681 independent microcosms per day). Resistance and temporal stability (n = 567 independent microcosms) covaried negatively across richness (d, solid line through centroid of each richness level, n = 6 independent richness levels) but positively within richness levels (dashed lines).



492Diversity - resistanceDiversity - temporal variabilityDiversity - temporal variability493Figure 2: Positive, negative and neutral relationships among resistance, resilience and temporal

494 variability in empirical studies with diversity manipulation. 30 bivariate relationships were reported by 17

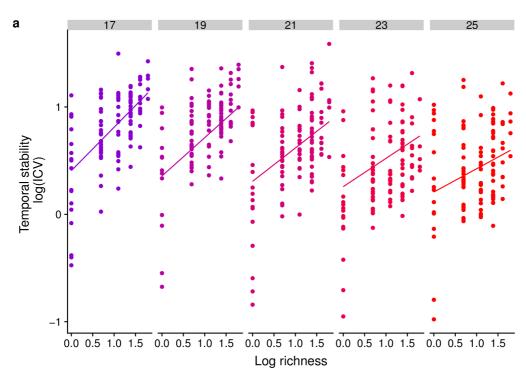
- independent studies (in addition to this study). Detailed information about individual studies (e.g. code VR-09)
- 496 is provided in extended data table 3 & 4. Beige regions indicate no covariation. Relative positions within
- 497 regions are arbitrary and do not indicate relative strengths of relationships. Different colours indicate the effect 498 of diversity on absolute (red) or relative resistance (blue), whereas temporal stability and resilience are shown in
- 498 of diversity on absolute (red) or relative resistance (blue), whereas temporal stability and resilience are shown in 499 black.
- 500



501 502 503 504 505 506 507 Figure 3: Hump- and U-shaped diversity-stability relationships. The intercept of the generalised logistic to convert measured stability components into a common currency varies with parameter Q (a). The nonaggregated (n = 567 independent microcosms) or aggregated (n = 6 richness levels) data exhibits hump- to flatto U-shaped diversity-stability relationships as Q varies. Lines show the fit of a quadratic model and the 95% confidence interval (bands). (b). The variation from hump-shaped to U-shaped relationship depends smoothly on Q, i.e. the position of the threshold (quantified by the quadratic term of a regression with mean (dot) and 508 509 510 95% confidence intervals (bars)) (c).

511 **Extended data figures:**

512



b

	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value
Intercept	0.3320	0.1897	0.4744	0.07250	626	4.58	0.000006
Richness	0.3335	0.2038	0.4631	0.06459	51	5.16	0.000004
Temperature	-0.0249	-0.0360	-0.0137	0.00569	626	-4.37	0.000015
Rich. x temp.	-0.0234	-0.0334	-0.0134	0.00508	626	-4.61	0.000005

CI = 95% confidence interval, DF = degrees of freedom, Std.Error = Standard error of the estimate

С

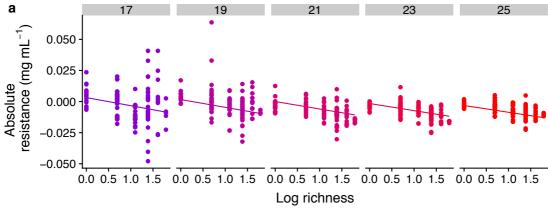
					-			
	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value	
Intercept	0.3937	0.2163	0.57118	0.09032	531	4.36	0.000016	
Richness	0.2842	0.1319	0.43648	0.07562	45	3.76	0.000491	
Temperature	-0.0317	-0.0506	-0.01272	0.00965	531	-3.28	0.001095	
Rich. x temp0.0181 -0.0337 -0.00247 0.00794 531 -2.28 0.023283								
CI = 95% confider	nce interval, D	F = degrees of	freedom, Std.E	rror = Standar	d error	of the estima	ate	

513 514 Extended data figure 1: Richness increased temporal stability across temperatures. a) The 515 stabilizing effect of richness was present across all temperatures, although temperature has a negative effect on mean stability. b) Result table for linear, mixed effects model of log 516 richness, temperature and their interaction on temporal stability supporting the stabilizing 517 518 effects of richness and the negative effect of temperature on temporal stability (n = 681519 independent microcosms). c) Result table for the same analysis as b) but without the

520 monocultures. Results are qualitatively the same, indicating that the relationship between

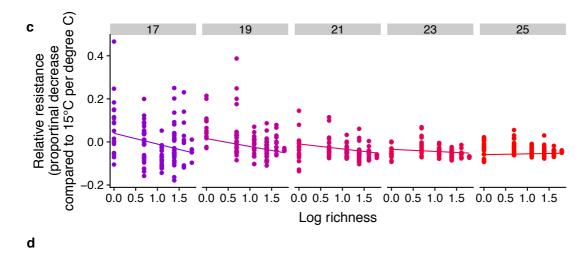
521 richness and temporal stability is not only driven by the monocultures (n = 580 independent

522 microcosms).



b

	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value
Intercept	0.000812	-0.003137	0.004760	0.002010	512	0.404	0.686407
Richness	-0.006115	-0.009710	-0.002520	0.001791	51	-3.415	0.001258
Temperature	-0.000796	-0.001215	-0.000376	0.000214	512	-3.726	0.000216
Rich. x temp.	0.000133	-0.000241	0.000507	0.000190	512	0.697	0.486295



	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value
Intercept	0.00284	-0.02027	0.02595	0.01176	512	0.241	0.80951
Richness	-0.02980	-0.05085	-0.00875	0.01048	51	-2.842	0.00643
Temperature	-0.01241	-0.01544	-0.00937	0.00155	512	-8.023	< 0.000001
Rich. x temp.	0.00670	0.00399	0.00941	0.00138	512	4.858	0.000002

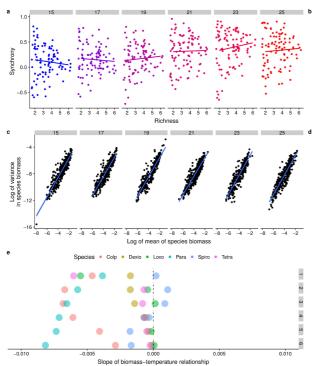
523 524

Extended data figure 2: The effect of richness on absolute and proportional resistance. a, c) 525 Richness decreased resistance, regardless whether its measured on the absolute or

proportional scale. b, d) Result tables of linear, mixed effects model of richness, temperature 526

527 and their interaction on absolute and proportional richness (n = 567 independent

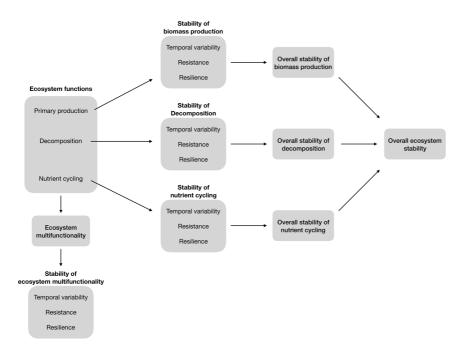
528 microcosms).



	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value
Intercept	0.18940	-0.04873	0.4275	0.12122	531	1.562	0.1188
Richness	0.00666	-0.01117	0.0245	0.00908	531	0.734	0.4635
Temperature	0.04958	-0.15500	0.2542	0.10157	45	0.488	0.6278
Rich. x temp.	0.01821	0.00354	0.0329	0.00747	531	2.439	0.0151
CI = 95% confide	nce interval. D	F = decrees of	freedom Std F	rror - Standar	d error	of the estim	ate

	Estimate	CI (lower)	CI (upper)	Std. Error	t value	Pr(> t)
Intercept	-2.58856	-2.69474	-2.48238	0.0541	-47.810	< 0.000001
Log species mean biomass	1.44577	1.41858	1.47297	0.0139	104.261	< 0.000001
Temperature	0.04280	0.01195	0.07366	0.0157	2.721	0.00657
Log species mean biomass x temperature	0.00139	-0.00645	0.00923	0.0040	0.348	0.72772
CI = 95% confidence interval. DF = degrees of freed	om. Std.Error	 Standard error 	or of the estimat	e		

530 531 **Extended data figure 3:** Niche complementarity and low response diversity were likely caused negative covariance of stability components. Niche complementarity and the resulting 532 533 increase in total biomass with richness tended to increase temporal stability (figure 1 in main 534 text). a, b) We found little evidence for an effect of population asynchrony on temporal 535 stability (linear mixed effects model with composition as random effect and log richness and 536 temperature as fixed effects; n = 681 independent microcosms). c, d) In contrast, statistical 537 averaging contributed to stabilization (linear regression between mean species biomass and 538 the variance of species biomass; n = 2077 species mean-variance biomass observations). e) 539 Low response diversity was inferred because the biomass of most species decreased or was 540 unaffected by temperature (linear regression between temperature and species biomass; n = 541 972 species biomass x temperature observations). Consequently, when there were more 542 species, there was greater total biomass, greater temporal stability, but greater biomass loss 543 with temperature increase. Thus, niche complementarity (i.e. effect diversity) likely caused a 544 positive effect of diversity on temporal stability but in the absence of high response diversity, 545 also had a negative effect of diversity on resistance. However, this explanation cannot apply 546 within richness levels, where there was positive covariance among stability components. 547



548

549 **Extended data figure 4:** Overview of terms and the overall ecosystem stability concept.

550 Measured ecosystem functions (left most upper box) can each have multiple components of

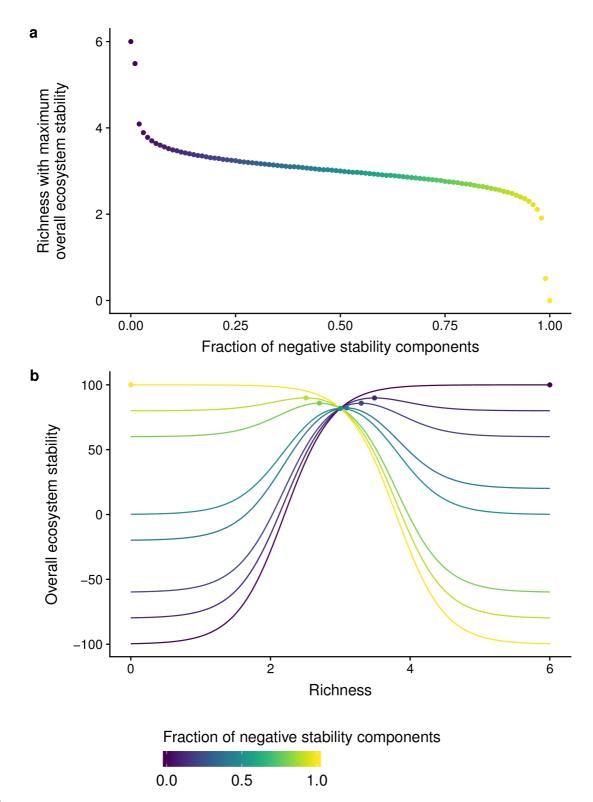
stability (e.g. temporal variability, resistance, and resilience of biomass production), which can each be combined into a measure of overall stability. When, as in our study, there is only

553 one ecosystem function, this overall stability of a specific function is also the overall

554 ecosystem stability. In studies of more than one ecosystem function, the overall stability of

555 several functions could be combined to give overall ecosystem stability. Alternatively, one

556 could first calculate ecosystem multifunctionality, and then measure its stability components.





558 559 Extended data figure 5: The effect of aggregating more than two stability components into 560 overall ecosystem stability. The fraction of stability components with negative sign influences whether or not a unimodal pattern will result for a total of 100 stability 561 components. a) A unimodal relationship between diversity and OES will result if at least one 562 563 stability component is negative. b) However, the strength of the pattern depends on the relative balance of positive and negative relationships. 564

566 Extended data table captions:

567

	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value
Intercept	0.07286	0.05182	0.093890	0.010731	12258	6.79	< 0.000001
Richness	0.04977	0.03058	0.068958	0.009559	51	5.21	0.000003
Temperature	-0.00205	-0.00315	-0.000948	0.000561	626	-3.65	0.000281
Rich. x temp.	-0.00534	-0.00634	-0.004345	0.000507	626	-10.54	< 0.000001

CI = 95% confidence interval, DF = degrees of freedom, Std.Error = Standard error of the estimate

b

а

	Estimate	CI (lower)	CI (upper)	Std.Error	DF	t-value	p-value
Intercept	0.07323	0.05206	0.094407	0.010782	626	6.79	< 0.000001
Richness	0.04959	0.03031	0.068872	0.009603	51	5.16	0.000004
Temperature	-0.00220	-0.00351	-0.000892	0.000667	626	-3.30	0.00102
Rich. x temp.	-0.00547	-0.00664	-0.004296	0.000596	626	-9.18	< 0.000001

CI = 95% confidence interval, DF = degrees of freedom, Std.Error = Standard error of the estimate

568 569 **Extended data table 1:** Richness increased while temperature decreases biomass production.

a) Result table for the linear, mixed effects model of richness, temperature and their

571 interaction on the temporal dynamics of biomass (n = 12939 microcosm x day combinations).

b) Result table for linear, mixed effects model of richness, temperature and their interaction

with the time series aggregated to the average biomass for each microcosm (n = 681 independent microcosms).

richness	Intercept	Slope	P-perm (1-tailed)	N
1	-0.00760617400951036	0.00506367932695562	0.01	84
2	-0.00843466390669893	0.00339781153057296	0.02	148
3	-0.0247877029765781	0.00880722901581866	0.04	100
4	-0.0249323685277533	0.00762486814778502	0.01	150
5	-0.027634192253533	0.0086737833813624	0.01	60
6	-0.0220565962582076	0.00349526254239345	0.04	25

P-perm = permutation-based p value

577

Extended data table 2: Reduced major axis regression revealed positive relationships between temporal stability and resistance within each of the six richness levels (p value based

on one-tailed permutation tests; N = sample size for each richness level).

Code	Description	Abbreviated reference
Ca-05	Diversity manipulation of grassland plants (Portuguese BIODEPTH site) analysing temporal variability and resistance to natural perturbation (frost and low precipitation) of biomass production.	Caldeira et al (2005) Oikos, 110, 115
PS-02	Diversity manipulation of grassland plants (Swiss BIODEPTH site) analysing resistance and resilience of biomass production to experimental drought perturbation.	Pfister & Schmid (2002) Nature, 416, 84
Z2-06	Diversity manipulation of unicellular algae analysing resistance and resilience of biomass production to experimental cold perturbation.	Zhang & Zhang (2006) Oikos, 112, 218
Z1-06 *	Diversity manipulation of unicellular algae analysing temporal variability and resistance to experimental cold perturbation of biomass production.	Zhang & Zhang (2006) Oikos, 114, 385
ls–15	Forty-six diversity manipulations of grassland plants analysing temporal variability, resistance, and resilience to natural perturbations (drought and wet events) of biomass production.	Isbeli et al (2015) Nature, 526, 574
VR-09	Diversity manipulation of plants, analysing resistance and resilience to natural perturbation (drought) of biomass production.	Van Ruijven & Berendse (2009) Journal of Ecology, 98, 81
Ba-16	Diversity manipulation of diatoms, analysing resistance and resilience to perturbation (chemical stressor, atrazine) of biomass production.	Baert et al (2016) Ecology, 97, 3433
Gr-00-1	Diversity manipulation by fumigation of soil microorganisms, analysing resilience to perturbation (heating) and resistance to perturbation (heating) of respiration.	Griffiths et al (2000) Oikos, 2, 279
Gr-00-2	Diversity manipulation by fumigation of soil microorganisms, analysing resilience to perturbation (chemical stress) and resistance to perturbation (chemical stress, CuSO4) of respiration.	Griffiths et al (2000) Oikos, 2, 279
TD-94	Diversity manipulation by fertilisation of plants, analysing resistance and resilience of biomass production to natural perturbation (drought).	Timan & Downing (1994) Nature, 367, 363
Wa-00	Diversity manipulation of plants, analysing resistance and resilience to experimental perturbation (drought) of various ecosystem functions.	Wardle et al (2000) Oikos, 98, 11
Hu–04	Analysis of diverse studies of marine ecosystems, some of which examined diversity-resistance and some diversity-resilience relationships for various ecosystem functions.	Hughes & Stachwicz (2004), PNAS, 101, 24
Vo-12	Diversity manipulation of grassland plants. Analysis of resistance and resilience of biomass production to drought perturbation.	Vogel et al (2012) PLoS One, 7, e36992
Wr-15	Diversity manipulation of grassland plants. Analysis of temporal variability and resilience of biomass production to drought perturbation.	Wright et al (2015) Nature Communications, 6, 6092
We-07	Manipulation of soil microbial diversity by serial dilution. Analysis of resistance and resilience of denitrification and nitrite oxidisation to a temperature perturbation.	Wertz et al (2007) Environmental Microbiology, 9, 2211
Wa-17	Diversity manipulation of grassland plants. Analysis of temporal variability and resilience of biomass production to drought perturbation.	Waag et al (2017) Ecology, DOI: 10.1002/ecy.2003
Ti–96	Diversity manipulation of grassland plants via fertilisation. Analysis of temporal variability, resilience and resistance of biomass production to natural drought perturbation.	Tilman (1996) Ecology, 77, 350

582

Extended data table 3: Studies used for literature survey (figure 2). The individual study code, a short description and the abbreviated reference for each study is shown. * denotes studies that examine intrinsic stability alongside response to a disturbance.

Code	diversity_stability_relation	direction	mechanism1	mechanism2	evidence1	evidence2	scale
Ba-16	resistance/resilience	positive/negative	low response diversity	selection effect	signals of mechanisms	signals of mechanisms	relative
Ca-05	resistance/temporal_var	negative/positive	selection effect	increased evenness, synchrony, statistical averaging	signals of mechanisms	signals of mechanisms	absolute
Ca-05	resistance/temporal_var	neutral/positive	selection effect	increased evenness, synchrony, statistical averaging	signals of mechanisms	signals of mechanisms	relative
Gr-00-1	resistance/resilience	negative/neutral	not stated	not stated	missing	missing	relative
Gr-00-2	resistance/resilience	positive/positive	insurance effect	insurance effect	verbal	verbal	relative
Hu-04	resistance/resilience	positive/neutral	trade-offs in response diversity among stressors	not a sampling effect	signals of mechanisms	signals of mechanisms	absolute
ls-15	resistance/resilience	positive/neutral	not stated	not stated	missing	missing	relative
ls-15	resistance/temporal_var	positive/positive	not stated	not stated	missing	missing	relative
ls-15	resilience/temporal_var	neutral/positive	not stated	not stated	missing	missing	
PS-02	resistance/resilience	negative/negative	high response diversity; evenness	species composition	signals of mechanisms	signals of mechanisms	absolute
PS-02	resistance/resilience	neutral/negative	high response diversity; evenness	species composition	signals of mechanisms	signals of mechanisms	relative
TD-94	resistance/resilience	positive/positive	high response diversity	not stated	verbal	missing	relative
Ti-96	resistance/resilience	positive/neutral	high response diversity	not stated	verbal	missing	relative
Ti-96	resistance/temporal_var	positive/positive	high response diversity	high response diversity; competition	verbal	signals of mechanisms	relative
Ti-96	resilience/temporal_var	neutral/positive	not stated	high response diversity; competition	missing	signals of mechanisms	
Vo-12	resistance/resilience	neutral/neutral	not stated	not stated	missing	missing	absolute
Vo-12	resistance/resilience	neutral/neutral	not stated	not stated	missing	missing	relative
VR-09	resistance/resilience	negative/neutral	high response diversity	not stated	signals of mechanisms	missing	absolute
VR-09	resistance/resilience	neutral/neutral	high response diversity	not stated	signals of mechanisms	missing	relative
Wa-00	resistance/resilience	neutral/neutral	species composition	species composition	verbal	verbal	absolute
Wa-17	resistance/resilience	positive/neutral	insurance effect	insurance effect	verbal	verbal	relative
Wa-17	resistance/temporal_var	positive/positive	insurance effect	insurance effect	verbal	verbal	relative
Wa-17	resilience/temporal_var	neutral/positive	insurance effect	insurance effect	verbal	verbal	
We-07	resistance/resilience	neutral/neutral	high response diversity	high response diversity	verbal	verbal	absolute
Wr-15	resistance/resilience	negative/neutral	selection effect; community composition	not stated	signals of mechanisms	missing	absolute
Z1-06 *	resistance/resilience	positive/neutral	high response diversity	not stated	verbal	missing	relative
Z1-06 *	resistance/temporal_var	positive/positive	high response diversity	statistical averaging	verbal	signals of mechanisms	relative
Z1-06 *	resilience/temporal_var	neutral/positive	not stated	statistical averaging	missing	signals of mechanisms	
Z2-06	resistance/resilience	negative/neutral	low response diversity	not stated	signals of mechanisms	missing	contrology.

586 587 **Extended data table 4:** Putative mechanisms and type of evidence reported for all bivariate diversity-stability relationships.