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1	Stoichiometric constraints modulate temperature and nutrient effects on biomass
2	distribution and community stability
3	
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19	

#### 20 Abstract

21 Temperature and nutrients are two of the most important drivers of global change. Both can modify 22 the elemental composition (i.e. stoichiometry) of primary producers and consumers. Yet their combined effect on the stoichiometry, dynamics, and stability of ecological communities remains 23 24 largely unexplored. To fill this gap, we extended the Rosenzweig-MacArthur consumer-resource 25 model by including thermal dependencies, nutrient dynamics, and stoichiometric constraints on 26 both the primary producer and the consumer. We found that stoichiometric and nutrient 27 conservation constraints dampen the paradox of enrichment and increased persistence at high 28 nutrient levels. Nevertheless, stoichiometric constraints also reduced consumer persistence at 29 extreme temperatures. Finally, we also found that stoichiometric constraints and nutrient dynamics 30 can strongly influence biomass distribution across trophic levels by modulating consumer 31 assimilation efficiency and resource growth rates along the environmental gradients. In the Rosenzweig-MacArthur model, consumer biomass exceeded resource biomass for most parameter 32 33 values whereas, in the stoichiometric model, consumer biomass was strongly reduced and 34 sometimes lower than resource biomass. Our findings highlight the importance of accounting for 35 stoichiometric constraints as they can mediate the temperature and nutrient impact on the dynamics 36 and functioning of ecological communities.

37

#### 39 Introduction

40 Temperature and nutrients regulate many biological processes, including species geographical 41 distribution, primary production, species interactions, and energy and material fluxes (Falkowski 42 et al., 1998; Enquist et al., 1999; Elser et al., 2007; Thomas et al., 2017). They are at the core of 43 several ecological theories. While temperature is a fundamental component of metabolic scaling 44 theory (Brown et al., 2004), nutrients are at the core of resource competition theory (Tilman, 1982) and ecological stoichiometry (i.e. the element composition of organisms) theory (Sterner & Elser, 45 46 2002). Cross et al. (2015) suggested that a better understanding of the interactions between 47 temperature and nutrients is crucial for developing realistic predictions about ecological responses 48 to multiple drivers of global change, including climate warming and elevated nutrient supply. 49 Nutrients can modulate the effects of warming on communities directly by altering primary 50 production, and/or indirectly by changing the elemental composition of primary producers. 51 Conversely, thermal effects on trophic interaction strengths (i.e. the *per capita* effect of predators 52 on prey population densities) and on consumer energetic efficiencies (i.e. ingestion relative to 53 metabolic demand) depend on both the quantity and quality of their resources. While Cross et al. 54 (2015) provided a road map on how to investigate the combined effects of temperature and nutrients 55 on ecological processes, we still lack an integrative theory to better understand how the links 56 between stoichiometry, nutrient enrichment, and temperature influence the dynamics and stability 57 of multispecies communities. Such a theory will allow us to understand how and when 58 stoichiometric variation modulates the consequences of single and combined components of global 59 change on trophic interactions, community dynamics, and ecosystem functioning.

60

62 Predicting the effects of global warming and nutrient changes on ecosystems is challenging as 63 species are embedded within communities of multiple interacting species (Petchey *et al.*, 1999; 64 Tylianakis et al., 2008; Montoya & Raffaelli, 2010; Gilbert et al., 2014). Increased resource availability (hereafter: enrichment) and warming can jointly affect food-web stability and structure 65 66 by modifying the strength of trophic interactions (O'Connor et al., 2009; Binzer et al., 2012; 67 Kratina et al., 2012; Sentis et al., 2014; Binzer et al., 2016, Synodinos et al. 2021). Enrichment typically increases energy flux from resources to higher trophic levels which often leads to the 68 69 well-known paradox of enrichment where the amplitude of population fluctuations increase with 70 nutrients, leading to extinctions at high nutrient concentrations (Rosenzweig, 1971; Rip & 71 McCann, 2011; Gilbert et al., 2014). Nevertheless, most consumer species become less efficient at 72 processing matter and energy at warmer temperatures as their metabolic rates often increase faster 73 with temperature than their feeding rates (Vucic-Pestic et al., 2011; Fussmann et al., 2014; Iles, 74 2014). This reduction of energetic efficiency lessens energy flow between trophic levels and can 75 hence stabilizes food-web dynamics by reducing population fluctuations as long as interaction 76 strength decreases faster with warming than maximal energetic efficiency (Synodinos et al., 2021). 77 As a result, mild warming may alleviate the paradox of enrichment by decreasing interaction 78 strength and consumer energetic efficiency (Binzer et al., 2012; Sentis et al. 2017; Synodinos et 79 al., 2021).

80

The theoretical expectations and results described above have already improved our ability to understand and predict the effects of temperature and enrichment on food webs (Boit *et al.*, 2012; Tabi *et al.*, 2019). However, most previous studies using metabolic scaling theory assumed that nutrient enrichment lead to an increase in resource carrying capacity without influencing resource 85 elemental composition (Vasseur & McCann, 2005; Binzer et al., 2012; Gilbert et al., 2014; Binzer 86 et al., 2016; Sentis et al., 2017). Yet nutrient enrichment effects are more complex. The elemental 87 composition of primary producers is likely to be altered, in response to the supplies of energy and 88 materials relative to their growth and nutrient intake rates (Rastetter *et al.*, 1997; Robert W. Sterner 89 et al., 1997; Finkel et al., 2009). This, in turn, can affect the dynamics of the producer population 90 and the herbivores feeding on it. For instance, previous modelling studies showed that introducing 91 stoichiometric heterogeneity in predator-prey population dynamic models can dampen the negative 92 effect of nutrient enrichment on system persistence by reducing population biomass fluctuations 93 (Andersen, 1997; Loladze et al., 2000; Andersen et al., 2004; Elser et al., 2012). More generally, 94 the stoichiometric flexibility of primary producers, in particular the flexibility in carbon to nutrient 95 ratios (e.g. C:N or C:P), has important implications for animal feeding behaviour (White, 1993), 96 consumer population stability (White, 1993; Sterner & Hessen, 1994; Hessen et al., 2002), 97 community structure (Andersen, 1997), and ecosystem processes such as biogeochemical cycling 98 (Andersen, 1997; Hessen et al., 2004).

99

100 Previous theoretical and empirical studies reported that stoichiometric variations can have a strong 101 influence on the stability of consumer-resource interactions (Andersen, 1997; Andersen et al., 102 2004; Diehl et al., 2005; Elser et al., 2012). For instance, populations of crustacean Daphnia 103 feeding on low quality (i.e. low nutrient: carbon ratio) algae cannot persist even when resource 104 quantity is not a limiting factor (Elser et al., 2007). Consumer extinction is explained by the fact 105 that the consumer assimilation efficiency is, for most organisms, a function of resource quality 106 (Elser et al., 2000). When resource quality is low, the consumers assimilate only few nutrients 107 relative to the biomass they ingest, which limits their growth and reproduction (Elser *et al.*, 2000; 108 Elser *et al.*, 2012). Temporal variations in resource quality can stabilize the system by weakening 109 interaction strength and dampening population fluctuations (Andersen et al., 2004; Diehl et al., 110 2005) but see (Loladze et al., 2000; Elser et al., 2012). However, it remains unclear whether and 111 how temporal variations in the elemental composition of primary producers and consumers can 112 modulate the effects of temperature and nutrients on important community features such as stability 113 and biomass distribution across trophic levels. Previous studies indicated that the spatial and 114 temporal intraspecific variations in the elemental composition of primary producers are expected 115 to increase in response to global change drivers such as temperature,  $CO_2$ , and nutrient availability 116 (Bezemer & Jones, 1998; Woods et al., 2003; Finkel et al., 2009). This increased variation can be 117 of importance for both primary producer and consumer populations as the growth rate of primary 118 producers is well known to depend on their elemental composition (Droop, 1974) as is the 119 assimilation efficiency of the consumers (Sterner & Elser, 2002).

120

121 Altogether, previous studies indicated that both temperature and stoichiometric variations can have 122 important effects on species interactions and community dynamics (Andersen et al., 2004; Diehl 123 et al., 2005; Fussmann et al., 2014; Binzer et al., 2016; Sentis et al., 2017, Synodinos et al., 2021). 124 However, the effects of temperature and nutrient stoichiometry on food web dynamics and stability 125 have only been studied in isolation. Recent theory by Uszko et al. (2017) showed that considering 126 nutrient dynamics can help to better understand the influence of temperature on consumer-resource 127 population dynamics and resource carrying capacity. Nutrient conservation (i.e. mass balance) 128 constrains the dynamics of both the resource and consumer populations that fluctuate less than in 129 other models not considering nutrient dynamics and conservation. Nevertheless, Uszko et al. 130 (2017) considered that the elemental composition of both the resource and the consumer are 131 constant and independent of temperature and nutrient dynamics. This contrasts with the empirical 132 observation that resource elemental composition is flexible and can vary with both temperature and 133 nutrient dynamics (Droop, 1974; Elser et al., 2000; Woods et al., 2003). Here we thus focused on 134 the combined effects of temperature and nutrients on the stoichiometry of primary producers and 135 how this affects community stability and biomass distribution across trophic levels in a consumer-136 resource system. Understanding the determinants of stability and biomass distribution has been at 137 the core of ecology for a long time (Elton (1927), Lindeman (1942)). Recent theory aims at 138 explaining empirical observations of trophic pyramids (i.e. population biomass decreases with 139 trophic levels), inverted trophic pyramids (i.e. population biomass increases with trophic levels), 140 trophic cascades and the link between biomass distribution and stability (McCauley *et al.*, 2018; 141 Barbier & Loreau, 2019).

142

143 Here, we used the Rosenzweig-MacArthur model as a baseline non-stoichiometric model because 144 this model is one of the most studied models used to investigate the effects of temperature and 145 nutrient enrichment on community dynamics (Vasseur & McCann, 2005; Binzer et al., 2012; 146 Fussmann et al., 2014; Sentis et al., 2017; Synodinos et al., 2021). Inspired by previous 147 temperature-independent stoichiometric consumer-resource models (Andersen, 1997; Andersen et 148 al., 2004; Diehl et al., 2005), we then extended the Rosenzweig-MacArthur model to account for 149 nutrient dynamics with nutrient conservation, as well as for the simultaneous dependence of 150 community dynamics on temperature and flexible resource stoichiometry. Our objective here was 151 not to develop a complex and very realistic stoichiometric model that would include additional 152 important abiotic and biotic features such as light intensity (Diehl, 2007) or compensatory feeding 153 (Cruz-Rivera & Hay, 2000). Instead, we aimed at introducing two fundamental stoichiometric 154 features (i.e. stoichiometric flexibility and stoichiometric imbalance) and investigate how these 155 stoichiometric considerations can change predictions of the Rosenzweig-MacArthur model. We 156 thus used our extended Rosenzweig-MacArthur model to predict the effects of warming and 157 nutrient enrichment on population dynamics and biomass distribution across trophic levels and 158 compared these predictions with the predictions of the nonstoichiometric Rosenzweig-MacArthur 159 model. We particularly addressed two questions: (i) How do stoichiometric constraints modulate 160 the effects of enrichment and warming on community stability and persistence? and (ii) How do 161 stoichiometric constraints modulate the effects of enrichment and warming on biomass distribution 162 across multiple trophic levels?

163

#### 164 Methods: Population dynamic models

#### 165 The Rosenzweig-MacArthur (RM) model.

166 Rates of change of the consumer and resource biomass densities  $\acute{C}$  and  $\acute{R}$  depend on their respective 167 biomass densities C and R (g.m<sup>-3</sup>):

168 
$$\dot{R} = r\left(1 - \frac{R}{K}\right)R - \frac{aR}{1 + ahR}C$$
(1)

169 
$$\dot{C} = \left(e\frac{aR}{1+ahR} - m\right)C$$
(2)

The population growth rate of the resource is given by the logistic equation where r is the resource maximum growth rate and K is the resource carrying capacity. The population growth rate of the consumer is equal to its feeding rate multiplied by its assimilation efficiency e (i.e. the fraction of resource biomass converted into consumer biomass) minus a loss term associated to metabolic losses m. The feeding rate of the consumer C depends on the density of its resource R and follows a Holling type II functional response, with consumer-resource attack rate a and handling time h. 176

177 In the RM model, consumer and resource population growth rates are only limited by nutrient or 178 resource density. Nutrient enrichment is assumed to increase resource carrying capacity, which 179 often leads to the well-known paradox of enrichment where populations fluctuates up to extinctions 180 (Rosenzweig, 1971). Nevertheless, this model neither considers nutrient dynamics nor temporal 181 variations of resource stoichiometry and their consequences on population dynamics. Because 182 nutrient conservation is not guaranteed in the RM model, the nutrient content of the resource and 183 consumer populations can exceed the total amount of nutrient in the system and thus violate mass 184 balance, which can lead to unrealistically large population fluctuations (Andersen et al. 2004). To 185 circumvent these limitations of the RM model, we extended it to better consider nutrient dynamics 186 (and nutrient conservation), resource stoichiometry and the way they can affect resource and 187 consumer population dynamics.

188

### 189 The Stoichiometric Rosenzweig-MacArthur (SRM) model.

We derived a stoichiometric extension of the Rosenzweig-MacArthur consumer-resource model with additional stoichiometric and temperature dependencies of several biological rates. We considered two stoichiometric constraints: one on the resource population growth rate, and the other on the consumer assimilation efficiency (see below for more details). These stoichiometric constraints have been observed for several consumer-resource pairs suggesting that they are core components of species growth and interactions (Sterner & Elser, 2002).

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197

#### 199 Stoichiometric constraint on the resource population growth rate

Inspired by previous stoichiometric models (Andersen, 1997; Loladze *et al.*, 2000; Andersen *et al.*, 201 2004; Diehl *et al.*, 2005), we extended the RM model by considering explicit nutrient dynamics 202 and nutrient effects on resource population growth rate. The system is assumed to be closed for 203 nutrients. Thus, nutrient supply originates exclusively from biomass excretion and 204 remineralization. We assume that free nutrients are taken up very quickly relative to the dynamics 205 of the consumer and resource biomasses, as is often the case in aquatic systems. The total amount 206 of nutrients in the system ( $N_{tot}$ ) is then a measure of nutrient enrichment.

207

208 In contrast to the very high plasticity in C:N or C:P exhibited by autotrophs, heterotrophs regulate 209 elemental composition within narrower bounds, even when consuming food with large variation in 210 elemental composition (Andersen & Hessen, 1991; Sterner & Hessen, 1994; Andersen, 1997; Elser 211 et al., 2000). In other words, the elemental homeostasis is much stronger for consumers compared 212 to primary producers. We thus assumed the nutrient quota (i.e. the nutrient to carbon ratio) of the 213 consumer  $Q_{\rm C}$  to be conserved whereas the one of the resource  $Q_{\rm R}$  is flexible over time with the 214 only constraint that  $Q_R > Q_{min}$ . As in the RM model, rates of change of the consumer and resource biomass densities  $\acute{C}$  and  $\acute{R}$  depend on their respective carbon biomass densities C and R (gC.m<sup>-3</sup>), 215 216 except that the resource population growth rate follows the Droop equation (Droop, 1974) given by  $r(1-Q_{\min}/Q_R)R$  and is now limited by  $Q_R$  relative to the minimum nutrient quota  $Q_{\min}$ : 217

$$218 N_{tot} = Q_R R + Q_C C (3)$$

219 
$$\dot{R} = r \left( 1 - \frac{Qmin}{Q_R} \right) R - \frac{aR}{1 + ahR} C$$
(4)

220 
$$\dot{C} = \left(e\frac{aR}{1+ahR} - m\right)C$$
(5)

From the nutrient conservation equation (eqn 3) we obtain that  $Q_R = \frac{N_{tot} - Q_C C}{R}$ . The intuitive interpretation is that the resource nutrient quota  $Q_R$  changes instantaneously with the density of the resource population *R* and with the concentration of the nutrient stored in the consumer biomass  $Q_C C$ , to maintain nutrient balance (see Text S1 for details). This contrasts with the RM model where nutrient conservation is not guaranteed and where the sum of the resource and consumer population's nutrient content can exceed the total amount of nutrient in the system.

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#### 228 Stoichiometric constraint on the consumer population growth rate

229 In the RM model, the growth rate of the consumer population only depends on resource density. In 230 other words, the RM model assumes that resource stoichiometry is not limiting and conversion 231 efficiency e is often taken for a consumer feeding on a high quality resource (Yodzis & Innes, 232 1992; Binzer et al., 2012; Fussmann et al., 2014; Uszko et al., 2017). However, conversion 233 efficiency can be much lower when the resource is of poor quality (i.e. when there is a 234 stoichiometric unbalance between the consumer and the resource nutrient: carbon ratio) (Elser et 235 al., 2000; Elser et al., 2007). We relaxed this assumption of the RM model by making the 236 population growth rate of the consumer dependent on both resource quality (i.e. nutrient quota) and 237 quantity (i.e. biomass density). In the SRM model, consumer production is also limited by resource 238 quality as the consumer assimilation efficiency e is a saturating function of resource nutrient quota 239  $Q_{R}$ :

$$240 e(Q_R) = e_{max} \frac{Q_R}{Q_R + Q_C} (6)$$

The intuitive interpretation of eqn. 6 is that resource quality is not a limiting factor for consumer growth as long as the nutrient content of the resource is superior to the nutrient content of the consumer (i.e.  $Q_R > Q_C$ ). In other words,  $e(Q_R)$  is proportional to  $Q_R$  for  $Q_R << Q_C$  and is at its maximum ( $e_{max}$ ) for  $Q_R >> Q_C$ . The later scenario corresponds to the assumption of the RM model where conversion efficiency is taken for a high-quality resource and thus  $e = e_{max}$ . By replacing eby  $e(Q_R)$  in eqn. 5, we obtain the SRM model.

247

#### 248 **Temperature dependence of model parameters**

249 To investigate the effect of temperature and stoichiometric constraints on consumer-resource 250 dynamics, we next extended the RM and SRM models described above by adding thermal 251 dependencies of the parameters. Following Uszko et al. (2017), we assumed that the total amount 252 of nutrient  $N_{\text{tot}}$ , the maximum food conversion efficiency  $e_{\text{max}}$ , and fixed stoichiometric traits ( $Q_{\text{C}}$ ) 253 are independent of temperature, as there is no evidence of systematic temperature dependence for 254 any of them (Peters, 1983; Ahlgren, 1987; Borer et al., 2013; Yvon-Durocher et al., 2015). Rate of 255 maintenance respiration and natural background mortality *m* typically increases exponentially with 256 temperature (Fig. S1a and b). We thus used the Arrhenius equation to describe the effect of 257 temperature T (in Kelvin) on m:

258 
$$m(T) = m_0 e^{\frac{-E_m}{kT}}$$
 (7)

where  $m_0$  is a parameter-specific constant calculated at temperature of 0°C (= 273.15 K). The temperature dependence is characterized by the respective activation energy  $E_m$  (eV) and the Boltzmann constant  $k=8.62 \times 10^{-5}$  eVK<sup>-1</sup>. As the temperature dependencies of resource intrinsic growth rate r and functional response parameters (a, 1/h) are often unimodal rather than exponential (Englund *et al.*, 2011; Rall *et al.*, 2012; Sentis *et al.*, 2012; Thomas *et al.*, 2012; Synodinos *et al.*, 2021), we used Gaussian functions for r and a and an inverted Gaussian function for h:

266 
$$g(T) = g_0 e^{\pm \frac{(T - Topt)^2}{2s^2}}$$
 (8)

where  $T_{opt}$  is the temperature at which the rate *g* reaches its minimum or maximum, *s* is the function width and  $g_0$  is a parameter-specific constant calculated at  $T_{opt}$ . The minus-sign corresponds to Gaussian functions and the plus-sign to inverted Gaussian functions.

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271

#### 272 Model parameterisation and simulations

To parameterise the models we assumed the resource and consumer species to be a unicellular freshwater algae and a *Daphnia* grazer, respectively. The choice for this system was motivated by the good characterization of both the stoichiometric parameters and thermal dependencies for this system (Andersen, 1997; Uszko *et al.*, 2017). Uszko *et al.* (2017) recently estimated the thermal dependencies for biological rates of the green algae *Monoraphidium minutum* and the grazer *Daphnia hyalina*. We thus used their estimates of stoichiometric parameters and thermal dependencies (See Table S1 and Fig. S1 for further details).

280

281 To investigate the individual and combined effects of enrichment, warming, and stoichiometric 282 constraints, we varied temperature (401 values ranging from 0 to 40°C by 0.1°C) and total amount of nutrients (parameter N<sub>tot</sub> in eqn. 10; 60 values ranging from 0.001 to 0.06 gP.m<sup>-3</sup> by 0.001 gP.m<sup>-</sup> 283 284 <sup>3</sup>, overlapping with reported mean phosphorus concentration in European peri-alpine lakes 285 (Anneville et al., 2005)). For the RM model, we used the minimum nutrient quota to convert 286 nutrients into resource (i.e.  $K = N_{tot}/Q_{min}$ ). This implies that carrying capacity is independent of 287 temperature which is expected for closed, nutrient-limited systems (Uszko et al., 2017) although 288 more experimental evidence are needed to verify this assumption (but see Bernhardt et al. 2018). 289 We then simulated the consumer-resource dynamics for 1000 days to enable the system to reach 290 an attractor (either an equilibrium point or a limit cycle) before we assessed the final state. 291 Therefore, for each model, we simulated 24060 combinations of environmental conditions (401 292 temperatures by 60 nutrient concentrations). Initial biomass density of each species was set to 0.98 293 times its equilibrium density in the two-species system (calculated by solving for the two-species 294 equilibrium, using either eqns 1-2 for model RM or eqns 3-5 for model SRM). The value of 0.98 295 was chosen to be (1) close enough to equilibria to avoid extinctions caused solely by transient 296 dynamics and (2) not exactly the equilibrium value to probe the stability of the equilibrium. 297 Additionally, at the end of each simulation, we used the last 100 time steps to calculate to 298 coefficient of variation CV (standard deviation divided by mean population density) of each species 299 to assess if populations were fluctuating or at equilibrium (CV is zero when the system is at 300 equilibrium and positive when populations fluctuate). Any population falling below the extinction threshold of 10<sup>-9</sup> g.m<sup>-3</sup> during the simulations was deemed extinct and its biomass set to zero to 301 302 exclude ecologically unrealistic low biomass densities. For each model, we calculated system 303 persistence as the percentage of simulations with the two species remaining extant at the end of the 304 simulations. We also calculated system persistence without considering the extinction threshold to 305 assess the proportion of extinctions that are driven by population fluctuations resulting in 306 unrealistic low biomass densities.

To reveal the dynamic effects of the stoichiometric constraints, we calculated the values of assimilation efficiencies and carrying capacities predicted by the SRM model for each temperaturenutrient scenario and used these effective parameter values to replace the values of parameters eand K in the RM model for each temperature-nutrient scenario. The objective of using these effective parameter values was to disentangle the static effect of stoichiometric constraints (i.e. changing the average parameter values of consumer assimilation efficiency and of the resource carrying capacity) from their population dynamical effects. Population dynamics were simulated with R version 3.4.3 (R Development Core Team, 2017) using the "deSolve" package (Soetaert *et al.*, 2012) with an absolute error tolerance of  $10^{-10}$  and a relative error tolerance of  $10^{-6}$ .

#### 317 **Results**

#### 318 Stability: population fluctuations and persistence

Considering stoichiometric constraints and nutrient dynamics dampened the paradox of enrichment, reducing fluctuations at high nutrient levels and hence increasing persistence. However, the persistence of the consumer at low and high temperatures was reduced in the SRM model compared to the RM model. As a result, the overall effect of stoichiometric constraints and nutrient dynamics on stability depends on their relative influence on population fluctuations versus consumer persistence. In the two following paragraphs, we explain in more detail these results and highlight key differences between the outcomes from RM and SRM models.

326

The RM model predicts that increasing nutrient concentration is strongly destabilizing: the system shifts from a stable equilibrium point to limit cycles (i.e. the system crosses a Hopf bifurcation, Fig. 1: consumer CV goes from zero to positive values when nutrient concentration increases). This destabilizing effect is known as the paradox of enrichment (Rosenzweig 1971). As population biomass fluctuations (i.e. cycle amplitude, represented by increasing CV values in Fig. 1a) increase with nutrient concentration, minimal population densities are very low at high nutrient concentrations leading to the extinction of the resource once the extinction threshold is crossed and 334 then the consumer goes to extinction as it cannot survive without resource (Fig. 1a). In the range 335 of temperatures where the consumer persists, warming does not have a strong influence on the 336 nutrient concentration at which the system shifts from the stable equilibrium point to limit cycles (CV>0 in Fg.1), although this qualitative shift is absent at very high temperatures (i.e. 32°C) when 337 338 the consumer is close to extinction. Warming enhances the persistence of the consumer-resource 339 system at high nutrient concentrations (Fig. 1c). This is explain by the fact that, at low temperatures, 340 when the system starts fluctuating with increasing nutrient concentration, the fluctuations are so 341 large that they systematically lead to extinction (Fig. 1a). At higher temperatures, the system 342 fluctuates but the fluctuation amplitudes are smaller which thus dampens extinctions driven by the 343 paradox of enrichment. However, very warm and cold temperatures cause the extinction of the 344 consumer (see below for the mechanisms underlying extinctions), releasing resources from topdown control. Overall, we found that, without considering the extinction threshold of 10<sup>-9</sup> g.m<sup>-3</sup> 345 346 (see Model parametrisation and simulations), both the consumer and the resource can persist in 347 74% of the temperature-nutrient concentration scenarios (i.e. black + orange areas in Fig 1c). 348 Nevertheless, when considering the extinction threshold, they persist in only 21% of the 349 temperature-nutrient scenarios (i.e. black area in Fig. 1c). In other words, comparing the model 350 simulations with and without extinction threshold revealed that, in the RM model, extinctions are 351 mostly driven by population fluctuations leading to very low biomass densities at which the 352 population is at risk of extinction.

353

In contrast, the SRM model shows that increasing nutrient concentrations causes fewer fluctuations than those observed for the RM model (Fig. 1b, d). This is because: (1) more nutrients are needed to shift the system from a stable equilibrium point to limit cycles—the system can indeed persist

357 without fluctuations (i.e. CV = 0) up to 0.02 gP.m<sup>-3</sup> whereas it was only up to 0.0005 gP.m<sup>-3</sup> in the 358 RM model—and (2) when the system fluctuates (i.e. CV > 0), the amplitude of the fluctuations is 359 smaller in the SRM than in the RM model. As a result, stoichiometric and nutrient conservation 360 constraints dampen the amplitude of population fluctuations (i.e. the paradox of enrichment) and 361 hence increase system persistence at high nutrient levels. While the qualitative effect of 362 temperature is similar to that observed in the RM model, the thermal thresholds for consumer 363 persistence are reduced at low and high temperatures in the SRM predictions (Fig. 1b, d). 364 Moreover, thermal thresholds remain almost constant along the nutrient gradient in the RM model, 365 whereas in the SRM model they depend on nutrient concentration, with a smaller thermal range at 366 low nutrient levels compared to high nutrient levels (Fig. 1b, d). The consumer is thus more likely 367 to go extinct at low nutrient concentrations and extreme temperatures in the SRM model than in 368 the RM model. Overall, system persistence for the SRM model was 44% without considering the 369 extinction threshold and 37% when considering it. In other words, comparing the model 370 simulations with and without extinction threshold revealed that, in the SRM model, few extinctions 371 are driven by population fluctuations leading to very low biomass densities. We thus conclude that 372 the RM model predicts larger population fluctuations leading to high probabilities of populations 373 extinctions in comparison to the SRM model.

374

#### 375 Biomass distribution

We next compared the predictions of both models for consumer-resource biomass ratios along the temperature and nutrient gradients (Fig. 2). We found that the RM model systematically predicts biomass ratios > 1 (i.e. consumer biomass is larger than resource biomass). In contrast, the SRM model predicts biomass ratios both > or < than 1 depending on temperature and nutrient levels. The 380 RM model predicts that, as soon as the consumer can persist, its population biomass density always 381 exceeds the resource population biomass density (Fig. 2). With the SRM model, the biomass ratios 382 are below one at low nutrient levels (Fig. 2). However, at medium and high nutrient levels, the 383 ratios are above one as soon as the consumer can persist. We found qualitatively similar results 384 when considering unstable equilibrium points (Fig. S2). Finally, we showed that, for equivalent 385 parameter values, the RM model predicts biomass ratio that are superior or equal to the ones 386 predicted by the SRM model (text S2). This difference between the two models is independent of 387 the shape and position of the temperature function used to parametrise the models.

388

#### 389 Mechanisms underlying stability and biomass distribution patterns

390 Here, we detail the mechanisms underlying the stability and biomass distribution patterns to better 391 understand how and when stoichiometric constraints modulate the effects of temperature and 392 nutrients on consumer-resource dynamics. The first mechanism corresponds to the effect of 393 stoichiometric constraints on the consumer energetic efficiency that determines the consumer 394 persistence at extreme low and high temperatures. The second mechanism relates to the influence 395 of the stoichiometric constraints on population dynamical feedback that explains why the 396 stoichiometric model predicts more stability at high nutrient levels compared to the non-397 stoichiometric model.

398

#### 399 *Consumer energetic efficiency*

400 The persistence of the consumer at low and high temperatures is driven by the energetic efficiency
401 *EE* of the consumer (i.e. its feeding rate relative to metabolic losses) calculated as follows:

402 
$$EE = \frac{ef(R^*)}{m}$$
 (9)

Where  $f(R^*)$  is the functional response of the consumer at resource density  $R^*$  (i.e. the resource equilibrium density in absence of the consumer). We recall that the assimilation efficiency *e* is a function of resource quality  $Q_R$  in the SRM model whereas it is assumed constant in the RM model. The intuitive interpretation of eqn. 9 is that *EE* should be above one for the consumer population to grow and persist. *EE* is equivalent to the invasion rate of the consumer into a system with resource only.

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410 To better understand the influence of stoichiometric constraints on consumer persistence, we thus 411 investigated differences in the RM and SRM model predictions regarding the consumer energetic 412 efficiency *EE* along the temperature gradient at two nutrient concentrations (Fig. 3). For both 413 models and the parameter values we used, energetic efficiency at equilibrium has a hump-shaped 414 relationship with temperature with maximal efficiency values at medium temperatures. While this 415 unimodal shape is conserved across nutrient levels and models, the RM model systematically 416 predicts higher consumer energetic efficiency values than the SRM model because consumer 417 assimilation efficiency is lower in the SRM than in the RM model (Fig. S3). As a result, the 418 temperatures at which energetic efficiency falls below one and drives consumers extinct are more 419 extreme in the RM model compared to the SRM model (Fig. 3). In other words, energetic efficiency 420 is above one for a narrower thermal range in the SRM model. Moreover, energetic efficiency 421 remains just above one for most temperatures at low nutrient levels which might suggest greater 422 sensitivity to perturbations affecting consumer feeding gains or metabolic losses.

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#### 425 Dynamical feedbacks due to the stoichiometric constraints

426 The second mechanism by which stoichiometric constraints influence consumer-resource stability 427 and biomass distribution are the dynamical feedbacks due to stoichiometric and nutrient 428 conservation constraints on the resource population growth rate and on the consumer energetic 429 efficiency. In the SRM model, the growth rate of the resource population depends on both the total 430 nutrient load and the consumer population density as  $Q_{\rm R} = (N_{\rm tot} - Q_{\rm C}C)/R$ . In other words, when 431 consumer population increases, this decreases resource population growth by reducing both 432 resource density (through predation) and quality (through nutrient mass balance) leading to a 433 negative feedback on consumer population growth rate imposed by both nutrient conservation and 434 flexible stoichiometry. In contrast, for the RM model, the negative consumer feedback is only 435 driven by the reduction in resource density as resource quality and nutrient dynamics are not 436 considered. In addition to this first dynamical feedback, there is a second dynamical feedback as the consumer population growth rate also depends on  $Q_R$  and thus on its own biomass density. 437 438 Thus, also this second negative feedback loop limits the consumer population growth rate when its 439 density increases. Altogether, dynamical feedbacks reduce strongly the amplitude of population 440 fluctuations, which in turn increases resource and consumer persistence.

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To reveal the dynamic effects of the stoichiometric constraints, we calculated the values of assimilation efficiencies and carrying capacities predicted by the SRM model for each temperaturenutrient scenario (Fig. S3) and used these effective parameter values to replace the values of parameters e and K in the RM model for each temperature-nutrient scenario. In other words, we calculated average values of e and K in the dynamic SRM model and used them as constant input parameters in the RM model. The objective of using these effective parameter values was to

448 disentangle the static effect of stoichiometric constraints (i.e. changing the average parameter 449 values of consumer assimilation efficiency and of the resource carrying capacity) from their 450 population dynamical effect (i.e. the two dynamical feedback described above). We thus simulated 451 population dynamics along the temperature-nutrient gradient using the RM model with these 452 effective parameters; referred hereafter as effective RM model (Fig. 4). Comparing predictions 453 from the RM, effective RM, and SRM models allowed to disentangle the static stoichiometric 454 effects when going from the RM to the effective RM predictions (Fig. 4, panels a to b) from the 455 dynamical stoichiometric effects when going from the effective RM to the SRM predictions (Fig. 456 4, panels b to c). In other words, the RM and effective RM only differ in their parameter values 457 because the effective RM takes into account the effect of stoichiometric constraints on the average 458 parameter values. On the other hand, the effective RM and SRM have similar parameter values but 459 different population dynamics, which helps understanding the dynamical feedback induced by 460 stoichiometric constraints.

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462 We found that, at low nutrient concentrations, population fluctuations and consumer persistence 463 predicted by the effective RM model agreed with predictions of the SRM model. However, the 464 system shifted from a stable equilibrium point to a limit cycle at lower nutrient concentrations for 465 the effective RM model than for the SRM model. This suggests that more nutrients are needed to 466 destabilize the system with the SRM model. Moreover, the effective RM model predicts ampler 467 population fluctuations than the SRM model. As a result, the effective RM predicts high extinction 468 rates at high nutrient concentrations compared to the SRM model. Overall, we found that the 469 effective RM model cannot fully reproduce the dynamics predicted by the SRM, which indicates that including stoichiometric constraints in the RM model involves more than only changingparameter values.

472

#### 473 **Discussion**

474 Temperature and nutrient enrichment are two of the most important drivers of global change 475 (Nelson, 2005). However, most research on the effects of temperature and nutrients on community 476 dynamics assumes that the elemental composition of primary producers and consumers are constant 477 and independent of changes on energy and material fluxes (Binzer et al., 2012; Boit et al., 2012; 478 Amarasekare & Coutinho, 2014; Gilbert et al., 2014; Amarasekare, 2015; Binzer et al., 2016; 479 Gilarranz et al., 2016; Synodinos et al., 2021). Yet, the elemental composition of primary producers 480 is known to be flexible, which can have important consequences for community dynamics and 481 ecosystem processes (Elser *et al.*, 2000). We have shown how stoichiometric constraints that 482 account for flexible stoichiometry can affect predictions on how temperature and nutrients 483 influence community stability and biomass distribution across trophic levels. We thus argue that 484 considering stoichiometric constraints is an important step toward a better understanding of the 485 effects of global change on ecosystems.

486

#### 487 Stoichiometric constraints and temperature can dampen the paradox of enrichment

We showed that both stoichiometric constraints and temperature dampen the negative effect of nutrient enrichment on consumer-resource fluctuations and increase system persistence at high nutrient levels. Temperature effects are driven by physiological mechanisms. In agreement with previous empirical studies, our model parametrization reflects the observation that metabolic loss 492 rates increase faster with warming than consumer feeding rates (Vucic-Pestic et al., 2011; Sentis 493 et al., 2012; Fussmann et al., 2014; Iles, 2014). Consumers are thereby less energetically efficient 494 at higher temperatures which stabilizes food-web dynamics as long as interaction strength 495 decreases faster with warming than maximal energetic efficiency (Synodinos et al. 2021). In 496 contrast, as reported for previous stoichiometric models (Andersen, 1997; Loladze et al., 2000; 497 Andersen et al., 2004; Diehl et al., 2005), the effect of stoichiometric constraints is mainly linked 498 to two mechanisms: a shift in the position of the Hopf bifurcation and negative dynamical 499 feedbacks of the consumer and resource on their population growth rates. Both resources and 500 consumers are composed of the same essential elements (N, P, and C), which implies that the sum 501 of essential elements contained in the resource and consumer biomasses cannot exceed the total 502 amount of essential elements present in the system (i.e. mass balance is satisfied). As a result, when 503 consumer or resource population biomass increases, it reduces the pool of free nutrients available 504 for the growth of the resource population and thus limits large population fluctuations (as shown 505 in Andersen et al. 2004). Therefore, more nutrients are needed to shift the system from a stable 506 equilibrium to population cycles. In other words, as reported in previous theoretical studies 507 (Andersen, 1997; Loladze et al., 2000; Andersen et al., 2004; Diehl et al., 2005), the paradox of 508 enrichment is displaced to higher nutrient concentrations (i.e., the position of the Hopf bifurcation 509 is shifted to higher nutrient levels. In contrast, the RM model does not take into account the storage 510 of nutrients in both the resource and consumer biomasses (i.e. the carrying capacity only depends 511 on the total nutrient load). Mass balance is not guaranteed which implies that consumer and 512 resource populations can reach high equilibrium biomasses and fluctuate strongly as they are not 513 limited by nutrient conservation. Less enrichment is thus required to shift the system from a stable 514 equilibrium point to limit cycles. Our findings are similar to previous modelling studies comparing stoichiometric and non-stoichiometric models (Andersen, 1997; Loladze et al., 2000; Andersen et
al., 2004; Diehl et al., 2005), and reinforce the call to better account for nutrient conservation and
stoichiometry when addressing the impact of nutrient enrichment on population dynamics.

518

519 We found two dynamic effects that correspond to negative dynamical feedbacks of the consumer 520 and the resource on themselves. The first one is imposed by nutrient conservation: when consumer 521 population increases, it decreases the population growth rate of the resource by limiting nutrient 522 availability, diminishing resource biomass, which, in turn, decreases the consumer population 523 growth rate. Conversely, when the resource biomass increases, this decreases the nutrient content 524 of the resource, which, in turn, limits the growth rates of both the resource and consumer 525 populations. These stoichiometric negative feedback loops strongly decrease the amplitude of 526 population fluctuations and thus dampen the paradox of enrichment. Interestingly, our comparisons 527 of the RM, effective RM and SRM model predictions indicate that the dynamical effects contribute 528 more to the reduction of fluctuations than the static effects: population fluctuations are large in the 529 effective RM model accounting for the static effect only, whereas they are much smaller in SRM 530 model accounting for both static and dynamical effects (Fig. 4). This implies that the impact of 531 stoichiometric constraints on community dynamics goes beyond a simple modification of 532 parameter values and encompass more complex population feedbacks between the consumer and 533 the resource. To disentangle the relative contribution of flexible stoichiometry versus nutrient 534 dynamics for population fluctuations, it would be interesting to compare the RM and the SRM 535 models to a model where both resource and consumer stoichiometry are fixed but where nutrient 536 dynamics are considered and mass balance is guaranteed (e.g. Uszko et al (2017)).

538 Overall, these results demonstrate that considering both flexible stoichiometry and temperature can 539 synergistically dampen the paradox of enrichment by two different mechanisms: population 540 dynamic feedbacks and physiological constraints. Our consumer-resource model is simplified compared to natural communities composed of numerous species. Moreover, in natural systems, a 541 542 large amount of nutrient can be stored in abiotic and slow biotic pools that have long turnover times 543 which, in turn, can influence the population dynamics. In particular, the amplitude of the population 544 fluctuations is expected to be smaller as abiotic pools can buffer the population feedback. 545 Nevertheless, considering the nutrient held in slow abiotic or biotic pools would not change the 546 equilibrium densities of primary producers and grazer if nutrients are released in the environment 547 proportionally to their concentration stored in the abiotic pool (Menge *et al.*, 2012). Moreover, the 548 predictions of the stoichiometric model fit with empirical observations. In eutrophic lakes and 549 experimental mesocosms, populations can persist at relatively high nutrient concentrations even if 550 fertilisation enhance population fluctuations (O'Connor et al., 2009; Boit et al., 2012; Kratina et 551 al., 2012), as our stoichiometric model predicts. In contrast, the Rosenzweig-MacArthur model 552 does not guarantee mass balance and tends to produce very large population fluctuations and 553 extinctions at low nutrient concentrations which can explain why these predictions are not well 554 supported by empirical observations (McAllister *et al.*, 1972; Jensen & Ginzburg, 2005).

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556 *Effects of stoichiometric and nutrient constraints on system persistence across environmental* 557 gradients

558 While stoichiometric and nutrient conservation constraints dampen the paradox of enrichment and 559 thus increase persistence at high nutrient levels, they also reduce the persistence of the consumer 560 at low and high temperatures. Stoichiometric constraints affect the thermal thresholds for consumer 561 extinctions. Consumers can only persist over a narrower range of intermediate temperatures when 562 they are constrained by stoichiometry. This is due to the reduced biomass assimilation of the 563 consumer at low and high temperatures that, in turn, decreases its energetic efficiency and thus 564 fastens consumer extinction. In our stoichiometric model, the reduction of biomass assimilation 565 efficiency emerges from the effect of temperature on resource quality: extreme high and low 566 temperatures decrease resource quality and thus less resource biomass can be converted in 567 consumer biomass at these temperatures. The emergence of a thermal dependency for assimilation 568 efficiency contrasts with previous theoretical studies that used the RM model and assumed that the 569 assimilation efficiency is temperature independent as resource quality is assumed constant (Binzer 570 et al., 2012; Gilbert et al., 2014; Sentis et al., 2017; Uszko et al., 2017). In the SRM model, the 571 thermal dependency of the consumer assimilation efficiency is fully driven by the change in the 572 resource stoichiometry induced by temperature. The SRM model thus predicts an additional 573 mechanism by which temperature can influence trophic interactions: temperature changes resource 574 stoichiometry, which, in turn, impacts the consumer assimilation efficiency and its population 575 growth rate. This prediction matches with empirical results showing that primary producer 576 stoichiometric composition can change with temperature (Woods et al., 2003) and that consumer 577 assimilation efficiency is sensitive to resource stoichiometric composition (Andersen, 1997; Elser 578 et al., 2000). To sum up, the overall effect of stoichiometric and nutrient conservation constraints 579 on system persistence thus depends on the temperature range considered and on their relative 580 influence on population fluctuations versus consumer persistence.

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#### 584 Effects of stoichiometric constraints on biomass distribution

585 We found that stoichiometric constraints can modulate the effects of temperature and nutrients on 586 biomass distribution across trophic levels. In the Rosenzweig-MacArthur model, biomass ratios 587 are above one for almost all temperatures or nutrient levels as the biomass produced by the resource 588 is efficiently transferred to the consumer level consistently along the environmental gradients. This 589 finding agrees with theoretical studies reporting that Lotka-Volterra and RM models predict 590 biomass ratios above one and fail to reproduce biomass pyramids for a substantial region of 591 parameter values (Jonsson, 2017; Barbier & Loreau, 2019). However, in nature, consumer-resource 592 biomass ratios are often below one (McCauley & Kalff, 1981; Del Giorgio & Gasol, 1995; 593 McCauley et al., 1999; Irigoien et al., 2004) suggesting that additional mechanisms should be 594 included to better understand and predict biomass distribution patterns in natural food webs. Our 595 stoichiometric model agrees with experimental observations. It predicts that, at low nutrient concentrations (i.e. < 0.01 gP.m<sup>-3</sup>), the biomass ratio never exceeds one along the entire 596 597 temperature gradient. This is observed in oligotrophic aquatic systems where primary production 598 is too low to sustain high consumer populations (O'Connor et al., 2009). In addition, we also found 599 that increasing nutrient levels decreased the temperature ranges within which biomass ratio is 600 below one. This corresponds to results from manipulated nutrient concentrations and temperature 601 in aquatic mesocosms, where zooplankton to phytoplankton biomass ratio only exceeds one in the 602 enriched mesocosms at medium or warm temperatures (i.e. 27°C) (O'Connor et al. 2009). This 603 suggests that the models with stoichiometric constraints and nutrient conservation better reproduce 604 the biomass patterns observed in experimental and natural systems. Nevertheless, further 605 experiments investigating the links between stoichiometric flexibility and consumer-resource dynamics are needed to determine if these stoichiometric mechanisms are underlying patterns ofbiomass distribution in nature.

608

609 Implications of our findings for global change

610 Temperature and nutrients do not act in isolation from each other. Climate warming, for example, 611 causes stronger water stratification, which, in turn, can limit nutrient cycling (Sarmiento et al., 612 2004; Tranvik et al., 2009). Environmental policies such as the European water framework 613 directive (i.e. Directive 2000/60/EC of the European Parliament and of the Council establishing a 614 framework for the Community action in the field of water policy) effectively reduces input of 615 nutrients in aquatic ecosystems (Anneville et al., 2005) while the climate keeps warming. With 616 these two phenomena, water will often be warmer and contain fewer nutrients in aquatic systems. 617 Our models consistently predict that warmer temperatures should stabilise consumer-resource 618 dynamics but, if temperature further increases, the consumer goes extinct as energetic efficiency 619 decreases with warming. Moreover, we found that stoichiometric constraints can reduce this 620 thermal extinction threshold (i.e. the consumer persists in a narrower thermal range), especially at 621 low nutrient levels. Our stoichiometric model thus suggests that decreasing nutrient concentrations 622 alongside warmer temperatures should fasten the extinction of consumer populations. This 623 prediction matches empirical observations of consumer extinctions at warm temperatures in 624 oligotrophic aquatic systems (Petchey et al., 1999; O'Connor et al., 2009). Altogether, these results 625 indicate that considering stoichiometric constraints can be of importance for the management of 626 nutrient inputs and the conservation of natural populations and communities under climate change.

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629 Knowledge of how temperature and nutrient simultaneously influence the elemental composition 630 of primary producers and consumers is crucial to better understand and predict the effects of global 631 change on species interactions, community dynamics and fluxes of energy and material within and 632 among ecosystems. Here we showed that stoichiometric and nutrient conservation constraints 633 dampen the negative effect of enrichment on stability by reducing population fluctuations through 634 population dynamics feedbacks. However, stoichiometric constraints also decrease consumer 635 energetic efficiency, which increases consumer extinction risk at extreme temperatures and low nutrient concentrations. Finally, stoichiometric constraints can reverse biomass distribution across 636 637 trophic levels by modulating consumer efficiency and resource population growth rate along the 638 temperature and nutrient gradients. Overall, our study suggests that accounting for stoichiometric 639 constraints can strongly influence our understanding of how global change drivers impact 640 important features of ecological communities such as stability and biomass distribution patterns. It 641 opens new perspectives and possible crossing-over with prior studies. For instance, it would be 642 interesting to assess the effects of other stoichiometric mechanisms (e.g. changes in the element 643 limiting growth (Daufresne & Loreau 2001), the role of multiple limiting nutrients (Cherif & 644 Loreau 2010) or of the microbial decomposer loop (Cherif & Loreau 2009)) on food web stability 645 and its response to nutrient enrichment, as well as the effects of other global change drivers 646 associated with altered biochemical cycles such as the increase in carbon dioxide and the 647 acidification of marine ecosystems.

648

649 Speculations

650 Our study provides a first step in the exploration of the consequences of stoichiometric constraints651 and temperature on ecological communities. It also calls for a better investigation of the factors

652 determining the carrying capacity of biological systems in order to design mathematical models 653 that can capture well community structure and dynamics. Is the carrying capacity of most 654 ecological systems limited by nutrients or by other factors such as space or light? This is an 655 important and, to our knowledge, unanswered question. Our study and others before us (e.g. 656 (Andersen et al. 2004) indicate that the popular Rosenzweig-MacArthur model is not appropriate 657 to model nutrient limited systems as this model often violates the principle of mass conservation 658 popularized by the French chemist Antoine Lavoisier with the famous quote "rien ne se perd, rien 659 ne crée, tout se transforme" ("Nothing is lost, nothing creates, everything transforms"). The 660 Rosenzweig-MacArthur model should be limited to systems where nutrients are not the main limit 661 for carrying capacity. We speculate these biological systems are seldom: food quantity and quality 662 is a major determinant of growth, from cells to populations. This would call into question the 663 numerous studies using the Rosenzweig-MacArthur model to assess to impacts of nutrient 664 enrichment and other global change drivers on community dynamics and stability. We would thus 665 need a different standard model that satisfies mass conservation and accounts for important nutrient 666 feedbacks between consumers and resources. These nutrient-based models already exists (e.g. 667 Daufresne & Loreau 2001; Andersen et al. 2004; Uszko et al. 2017, our study) and it is now time 668 to use them in global change ecology, especially when investigating nutrient enrichment.

669

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679	R codes are available on Dryad (https://doi.org/10.5061/dryad.msbcc2fv2).
680	
681	Recommendation by Peer community In Ecology
682	This manuscript has not been submitted for publication in another journal, but has already been
683	peer-reviewed and recommended by Peer community In Ecology. Two reviewers (anonymous)
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690	A.S., B.H., and J.M.M. conceived the study. B.H. and A.S. developed and analysed the models.
691	JMM obtained the funding and managed the project. A.S. wrote the first draft of the manuscript.
692	All authors contributed substantially to revisions.
693	
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#### 878 Figure legends

879 **Figure. 1.** Population fluctuations (consumer biomass coefficient of variation; panels a and b) and 880 species persistence (number of species; panels c and d) across the temperature (y axis) and nutrient 881 (x axis) gradients as predicted by the Rosenzweig-MacArthur (RM; panels a and c) and by the 882 Stoichiometric Rosenzweig-MacArthur (SRM; panels b and d) models. In panels a and b, 883 coefficient of variation (hereafter CV) represents fluctuation amplitudes. CV is null when the 884 system is at equilibrium and positive when populations fluctuate. In panels a and b, the white colour 885 corresponds to the temperature-nutrient scenario for which the consumer has gone extinct whereas 886 the orange to red to dark red represent population fluctuations of increasing amplitude. In panels c 887 and d, in black: both consumer and resource persist; in red: only the resource persists; in orange: 888 none persists. Resource biomass CV is not shown; it is qualitatively similar to the consumer 889 biomass CV as resource and consumer biomass fluctuation are strongly coupled.

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**Figure. 2.** Consumer-resource biomass ratio along the temperature gradient for the Rosenzweig-MacArthur (RM, green lines) and the Stoichiometric Rosenzweig-MacArthur (SRM, black lines) models at three nutrient concentrations (0.008, 0.02, and 0.032 gP.m<sup>-3</sup>). In each panel, the dotted line represents biomass ratio of one; i.e. the biomass densities of the resource and the consumer are equal. Biomass values shown at equilibrium points. For unstable equilibrium points (i.e. limit cycles), see Fig. S2.

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Figure. 3. Consumer energetic efficiency along the temperature gradient for the RosenzweigMacArthur (RM, in green) and the Stoichiometric Rosenzweig-MacArthur (SRM, in black) models

at two nutrient concentrations (0.008 and 0.02 gP/m<sup>3</sup>). In each panel, the dotted line represents
energetic efficiency equal to one.

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Figure. 4. Population fluctuations (consumer biomass coefficient of variation) across the
temperature (y axis) and nutrient (x axis) gradients as predicted by the Rosenzweig-MacArthur
(RM; panel *a*), the RM with effective parameters (panel *b*), and the Stoichiometric RosenzweigMacArthur (SRM; panel *c*) models.









#### Figure 4



