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REGULATION OF REPRODUCTIVE PROCESSES WITH 1

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DYNAMIC ENERGY BUDGETS

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31KEYWORDS

32Dynamic Energy Budget; DEB theory; Bioenergetics, Rainbow trout

33ABSTRACT

1. Linking organismal level processes to underlying suborganismal mechanisms at the molecular, cellular and organ level constitutes a major challenge for predictive ecological risk assessments. This challenge can be addressed with the simple bioenergetic models in the family of Dynamic Energy Budget (DEB), which consist of a small number of state equations quantifying universal processes, such as feeding, maintenance, development, reproduction and growth. 34 35 36 37 38 39 40

2. Motivated by the need for process-based models to evaluate the impact of endocrine disruptors on ecologically relevant endpoints, this paper develops and evaluates two general modeling modules describing demand-driven feedback mechanisms within the DEB modeling framework exerted by gonads on the allocation of resources to production of reproductive matter. 41 42 43 44 45 46

3. These modules describe iteroparous, semelparous and batch-mode reproductive strategies. The modules have a generic form with both positive and negative feedback components; species and sex specific attributes of endocrine regulation can be added without changing the core of the modules. 47 48 49 50 51

4. We demonstrate that these modules successfully describe timeresolved measurements of wet weight of body, ovaries and liver, egg diameter and plasma content of vitellogenin and estradiol in rainbow trout (Oncorynchus mykiss) by fitting these models to published and 52 53 54 55

new data, which require the estimation of less than two parameters per data type. 56 57

5. We illustrate the general applicability of the concept of demand-driven allocation of resources to reproduction by evaluating one of the modules with data on growth and seed production of an annual plant, the common bean (Phaseolis vulgaris). 58 59 60 61

Introduction 62

63Dynamic Energy Budget (DEB) theory offers a remarkably general 64 mathematical and conceptual framework for physiological ecology. Originally 65 formulated to describe growth and reproduction in animals, DEB theory now 66 describes widespread empirical patterns in metabolic behavior of a steadily 67 increasing number species (over 1,200 at the time of writing) from phyla 68from all three domains (Sousa, Domingos & Kooijman 2008; Kooijman 2010; 69 Jusup et al. 2017; AmP 2018). Its core concepts are consistent with some 70general trends in evolutionary history (Kooijman 1986; Kooijman & Troost 712007) and with the principles of thermodynamics (Sousa et al. 2010; Jusup et 72al. 2017). In addition, the theory offers a powerful framework for modeling 73 organismal response to environmental stress, notably in ecotoxicology 74(Kooijman & Bedaux 1996; Jager et al. 2014; Muller et al. 2014) and, more 75 recently, in the context of ocean acidification (Muller & Nisbet 2014; Jager, 76 Ravagnan & Dupont 2016), starvation (Gergs & Jager 2014) and crowding 77 stress (Gergs, Preuss & Palmqvist 2014). The versatility of the theory is due 78to its modular structure, through which specific attributes or 'details' of a 79 particular environment, stressor or species can be included without changing 80the core of the model. Here we follow a similar approach to accommodate 81life history strategies by which organisms allocate resources to reproduction. 82Since reproduction generally constitutes a major fraction of the total energy 83budget of an adult organism, the energetic implications of different

84 reproductive strategies and their trade-offs play a fundamental role in life 85history theory (Stearns 1992).

86An important feature of most DEB models is that resources are first 87 assimilated into somatic reserves, which are then committed to support 88 somatic, developmental and/or reproductive functions, depending on 89nutritional status and life stage. In the standard formulation of DEB (stdDEB), 90applicable to animals, the rate at which reserves are allocated to 91 reproduction depends only on the reserve density and the size of the animal 92(see Figure 1). Control mechanisms regulating the partitioning of reserves to 93favor growth over reproduction, or vice versa, are absent. Standard DEB 94ignores control mechanisms regulating the development of gonads, as the 95specifics of those mechanisms vary widely among taxa and sexes (but see Pecquerie, Petitgas and Kooijman (2009), Einarsson, Birnir and Sigurosson 96 97(2011), Augustine et al. (2012) and Llandres et al. (2015) for species or 98group specific DEB gonad loading modeling modules for anchovy, capelin, 99zebrafish and parasitic wasps, respectively). This lack of feedback simplifies 100the dynamics of resource allocation, with obvious mathematical advantages 101as a result. Yet, stdDEB quantifies reproductive output sufficiently accurately 102for many purposes, such as those that require estimates of reproductive 103output over longer time spans or those involving species that release 104gametes in a nearly continuous manner. However, it is important to consider 105feedback, e.g., mediated by endocrine regulation mechanisms, in order to 106capture the *dynamics* of gamete maturation in iteroparous and semelparous

107organisms, in which gametes mature during the later part of the 108 reproductive cycle or near the end of the life cycle, respectively. In addition, 109this kind of feedback could provide an entry to mechanistic modeling of the 110 impact of endocrine disruptors on growth and reproduction in the DEB 111framework.

112To more accurately accommodate the alternative reproductive strategies of 113iteroparous and semelparous organisms, we develop and evaluate the 114 performance of two extensions of the standard DEB model. These extensions 115include demand-driven feedback mechanisms on gonad development, 116guided by the premise that hormones produced in the reproductive organs 117and other organs commonly mediate those feedback mechanisms. We 118 center our evaluation of model performance on a single fish species, the 119rainbow trout (Oncorynchus mykiss), due to the expansive data set on its 120growth and reproductive biology. However, we argue that the model 121 extensions are based on general principles, and therefore applicable to other 122 species. As an illustration, we discuss how simplified formalism from one of 123the model extensions can be applied to describe the growth and 124 reproductive patterns in a species very different from trout, namely the 125 common bean (Phaseolis vulgaris). Beans have a reproductive strategy 126typical for many annual plants, namely an allocation strategy that favors 127 seed production over somatic growth during the later phases of the life 128 cycle. In addition, we discuss how these extensions can be useful in 129exploring physiological mechanisms by which stressors, in particular

130endocrine disruptors, affect resource allocation, and ultimately adverse 131 outcomes to reproduction and growth.

132

133Materials and methods

134DATA SOURCES

135Three data sets about female rainbow trout (O. mykiss) were analyzed to 136evaluate model performance. The most expansive set, referred to as main 137 data set, was from Nagler et al. (2012) with additional data from Gillies et al. 138(2016), and concerns a reproductively synchronized autumn-spawning 139 population obtained from a commercial supplier (Troutlodge, Inc., Sumner, 140WA) and maintained in a temperature controlled flow-through system under 141a natural lighting regime at the Battelle Marine Science Facility (Sequim, 142WA). The main data set included time-resolved measurements of wet weight 143of body, ovaries and liver, egg diameter and plasma content of vitellogenin 144and estradiol of 58 individuals. The two supplementary data sets, SD1 and 145SD2, were more limited in scope. SD1 included time resolved measurements 146of body weight and egg mass of 12 and 9 individuals, respectively, of a 147spring spawning strain obtained from Troutlodge Inc. (Sumner, WA). SD2 148included initial and final total body and egg weights as well as weights and 149diameters of individual eggs of 16 individuals of a fall-spawning strain 150obtained from Nisqually Trout Farm (Lacey, WA). Fish of SD1 and SD2 were 151 kept in the same facility as those of the main set; see Nagler et al. (2012), 152Schultz et al. (2013) and the Supplemental Information for experimental

153detail. All sets span a single breeding cycle of approximately 11-14 months 154 starting immediately after the time of first spawning.

155The common bean, *Phaseolis vulgaris*, was used to evaluate the potential of 156the principle of demand driven resource allocation to reproduction (see next 157 section) to capture the dynamics of growth and reproduction of a species 158 wildly different from iteroparous rainbow trout; beans have a semelparous 159 reproductive strategy typical for many annual plants, namely an allocation 160strategy that favors seed production over somatic growth during the later 161phases of the life cycle. Data are from Lima et al. (2005) and include time-162resolved measurements of vegetative above ground biomass, leaf cover and 163pod biomass of 6 cultivars grown in a field setting in coastal Brazil from May 164to August (mean growing conditions: 21.2°C, 70% humidity, 6.9 h solar 165 radiation per day; 12 seeds per row meter at 0.5 m row distance; plots 166 fertilized with 2.5 g N, 4.0 g P and 4.0 g K per square meter).

167

168DYNAMIC ENERGY BUDGET THEORY

169This study uses the standard model of Dynamic Energy Budget (stdDEB) 170theory as a reference. Since Kooijman (2010) has described this theory and 171 its standard formulation in detail and several other publications provide 172 extensive summaries (Nisbet et al. 2000; Sousa, Domingos & Kooijman 2008; 173 Jusup et al. 2017), we only present features of the theory that are essential 174to evaluate the models developed in this study.

175The stdDEB formulation (see Fig. 1), describes the rates at which a 176'generalized' animal acquires resources from its environment and uses the 177energy therein for somatic and maturity maintenance, growth, maturation 178(juveniles) and reproduction (adults). A 'generalized' animal is heterotrophic, 179grows isometrically (constant shape), does not encounter conditions of stress 180(including debilitating forms of starvation), and has three life stages: 181embryonic (during which it does not feed), juvenile (feeding but no 182 reproduction) and adult. Since this study involves the adult stage only, from 183now on, all references to animals pertain to adults, unless other life stages 184are explicitly mentioned. stdDEB distinguishes three pools of biomass: 185structure, general reserve and material in the reproductive buffer. Structure 186 is defined as the biomass requiring maintenance in order to remain viable. 187The reproductive buffer contains resources tagged for reproduction 188(irreversibly, except potentially during starvation conditions). General 189 reserve is functionally defined as all other metabolizable biomass; in 190practice, general reserve typically includes conventional storage materials as 191well as compounds that are traditionally not thought of as reserve, such as 192ribosomes in excess of the minimal amount needed to ensure vitality of an 193organism of a given size (Nisbet et al. 2000). The gross biochemical 194 composition of each pool is considered to be invariant, implying that the 195 costs to produce a unit of each type of biomass and the cost to maintain a 196unit of structure are constant. The general reserve density, i.e., the ratio of 197 general reserve and structure, stabilizes in a constant food environment.

198Environmental resources are first assimilated into general reserve, which is 199subsequently committed to somatic and developmental/ reproductive 200functions, with each set of functions receiving a constant fraction K of 201committed general reserve (see Figure 1). In order to accommodate the 202changing rate of gamete development during a reproductive cycle in female 203rainbow trout, we studied two extensions to the standard model (see Figure 2041). In the first variant, the proportion of committed general reserve allocated 205to reproduction is subject to feedback regulation of the reproductive buffer, 206implying that the allocation of general reserve to reproduction is driven by 207 demand of the reproductive buffer. This variant is denoted dDEB, with the 'd' 208standing for 'demand-driven'. The second variant, a modified version of a 209capelin model by Einarsson, Birnir and Sigurosson (2011), assumes stdDEB 210but separates the reproductive buffer in pools of unspecified reproductive 211 reserve and actual reproductive matter. A gonad loading modeling module 212 describes the rate at which reproductive reserve are converted into actual 213 reproductive matter. This variant will be denoted stdDEB+, with the '+' 214 referring to the gonad loading module. Regulation of the allocation of 215 r eserves to the reproductive buffer in dDEB and of gonad loading in stdDEB+ 216are subject to endocrine control.

217The derivations of the dDEB and stdDEB+ model equations in Table 1 are 218 presented in full in the Supplementary Information. Here, only the 219assumptions that are not part of stdDEB are presented and evaluated. The 220following list contains assumptions shared by and specific to both model

221 variants, though it should be stressed that reproductive matter is defined 222differently in those variants. In dDEB, reproductive matter refers to all matter 223in the reproductive buffer regardless of location in the body, whereas 224 reproductive matter roughly corresponds to gametes in stdDEB+. The 225 assumptions are:

1. At the onset of a reproductive cycle, a small fraction of somatic biomass is converted to reproductive matter, e.g., due to meiosis. General reserve and structure contribute proportionally to the initial formation of reproductive matter, and the costs of this conversion are negligible. The latter two assumptions are rather arbitrary but quantitatively insubstantial. 226 227 228 229 230 231

2. The initial density of reproductive matter is constant. This assumption maintains parameter parsimony and model simplicity. 232 233

3. An adult has a bounded capacity to carry reproductive matter. In nonstarving adults, this capacity is proportional to the amount of structural biomass, i.e., the maximum density of reproductive matter is a constant. This assumption maintains parameter parsimony and model simplicity. 234 235 236 237 238

4. dDEB only: the fraction of mobilized general reserve allocated to reproduction and maturity maintenance in adults is proportional to (1) the density of reproductive matter, and (2) the difference between the maximum and actual density of reproductive matter. The first proportionality introduces positive feedback and is based on the 239 240 241 242 243

general observation that the ovaries in fish produce estrogen, which stimulates the production of vitellogenin, the precursor of egg reserve material (Tyler & Sumpter 1996). The second proportionality provides a simple negative feedback (i.e., deceleration) mechanism that causes the accumulation of reproductive material in the gonads to slow down towards the end of a reproductive cycle. 244 245 246 247 248 249

5. stdDEB+ only: the rate at which reproductive reserves are converted to reproductive matter is proportional to (1) the density of reproductive reserves, (2) the density of reproductive matter, (3) the difference between the maximum and actual density of reproductive matter, and (4) the amount of structural biomass. The first proportionality ensures the density of reproductive reserves cannot become negative; for arguments for the two subsequent proportionalities, see previous assumption. 250 251 252 253 254 255 256 257

6. The efficiency with which reproductive reserves are converted into reproductive matter is constant. 258 259

7. Spawning requires the density of reproductive matter to exceed a threshold and, additionally, may be under the control of a time trigger or environmental factor, depending on species. 260 261 262

263LINK BETWEEN DEB QUANTITIES AND DATA

264Variables in DEB models are abstract quantities and therefore do not 265 correspond directly with measurable quantities. The mapping of DEB 266quantities onto the data analyzed in this study, including total body, ovary

267and liver wet weights, follicle diameter and plasma levels of estradiol and 268vitellogenin, is achieved through auxiliary assumptions stated in this section; 269the corresponding equations, summarized in Table 1, are derived in the 270Supplementary Information. The relationship between measurable quantities 271 pertaining to the common bean and those of a DEB model of bean growth 272and fecundity can be found in the Supplementary Information.

273In order to convert DEB mass quantities to wet weights, we use conversion 274factors from the rainbow trout entry in the DEB parameter database 275(Kooijman et al. 2017). Considering that the ovaries mainly consist of storage 276 materials in eggs, we assume the contributions of structure and general 277 reserves to the wet weight of the ovaries are negligible (to avoid confusion, 278we will use 'storage' to refer to physical materials and 'reserves' as the 279 conceptual abstraction in the context of DEB). We also assume that the 280fraction of reproductive matter that is in the ovaries is constant. 281Furthermore, we assume that reproductive matter is either in the ovaries or 282in the liver, which produces the precursors of egg storage materials. It is 283prudent to consider also including plasma vitellogenin, the precursor of egg 284storage materials. However, plasma vitellogenin levels are especially high 285just prior and after ovulation, indicating that not all plasma vitellogenin ends 286up in eggs. Furthermore, the fraction of vitellogenin in plasma is relatively 287small. Plasma contributes 2.5% to 5.5% to body wet weight in teleost fish 288(Brill et al. 1998, and references therein) and contains about 25 mg 289vitellogenin/ ml during the phase of accelerating ovary growth in a typical

290individual in this study (see figure 2F), which corresponds to only about 1.5-3.5 g vitellogenin in a 2.5 kg fish. Thus, it is reasonable to ignore the 291 292 contribution of vitellogenin to reproductive matter, though its dynamics are 293informative and are modeled later. Furthermore, we assume that the 294fractions of structure and reserves that are part of the liver are constants for 295both model variants, and, for stdDEB+, in order to retain simplicity, that the 296 amount of reproductive reserves in the liver is negligible.

297This leaves the follicle diameter and estradiol and vitellogenin plasma levels 298as the experimental quantities that need to be related to DEB variables. In 299order to relate the mean diameter of a follicle to reproductive matter, we 300assume that follicles are perfect spheres and that the specific gravity of 301biomass equals unity. Estradiol is produced by the ovaries and regulates the 302flow of vitellogenin to the ovaries. Accordingly, we link the gonad loading 303 module of stdDEB+ and the reproduction flux in dDEB to the plasma 304 estradiol concentration assuming simple proportionality.

305To model the dynamics of plasma vitellogenin, we assume that the volume of 306plasma is proportional to the amount of structural biomass, and that the rate 307at which vitellogenin is cleared from plasma is proportional to the amount of 308structural biomass (e.g., by structural mass in the ovaries). Furthermore, for 309dDEB, we assume that the rate at which vitellogenin is released into the 310blood stream is proportional to the rate at which somatic reserves are 311allocated to reproduction. For stdDEB+, we assume that the rate at which

312vitellogenin is released into the blood stream is proportional to the rate at 313 which reproductive reserves are allocated to reproductive matter.

314

315PARAMETERIZATION

316

317In the evaluation of model performance with trout data, the values of some 318or all parameters in Table 2 were fixed, depending on the information 319 content of the data and on the purpose of the analysis (see legend to Figure 3204 for information about parameter values regarding the analysis of bean 321 data). The main data set was used to parameterize the model variants; 322 subsequently, this parameterization was used to predict the observations in 323the supplementary data sets SD1 and SD2 (with one exception - see next 324section). However, not all parameters were estimable from the main data set 325due to a lack of information about, e.g., elemental biomass composition and 326some conversion efficiencies, and therefore had to be fixed; similar values 327were used for fixed parameters that occur in both model variants. The values 328of eight fixed parameters, as marked in Table 2c, were taken or calculated 329 from the rainbow trout entry in the DEB parameter database (Kooijman et al. 3302017). Among those was the somatic maintenance rate parameter, which 331 could not be estimated as it strongly covaried with other parameters, notably 332the general reserve turnover rate. Since the value of the somatic 333 maintenance rate parameter is relatively invariant across species (Kooijman

3342010), it was fixed at the value in the DEB parameter database, while the 335 latter was treated as a free parameter.

336The reasoning for the remaining five fixed values is as follows. First, the 337value for the scaled food density was set at 0.9, which is close to its 338 maximum of 1.0, as the fish were well fed. Second, according to the 339parameter database, maturity maintenance costs would have been an 340insubstantial fraction of the total energy budget of the fishes and were 341therefore ignored. Third, the initial density of reproductive reserve in 342stdDEB+ was assumed negligible, since there was no information available 343that could be used to identify the reproductive reserve pool as a pool 344 separate from general reserve and reproductive matter in this model variant 345(in contrast, this parameter could be estimated for dDEB - see Table 2d). 346This assumption is supported by the fact the fish had recently matured and 347were stripped before the experiment. Fourth, the maximum density of 348 reproductive matter in stdDEB+ strongly covaried with other parameters and 349was therefore fixed; it was identical to the density of reproductive matter in 350a female of ultimate size at optimal conditions after one year according to 351the parameter database. Fifth, the conversion efficiency of reproductive 352 reserves to reproductive matter in stdDEB+ was set at unity, implying that 353all the conversion overheads were subsumed in the conversion of general 354 into reproductive reserve.

355Free parameters were estimated by maximizing likelihood considering all 356data types in a set simultaneously, while assuming that discrepancies

357between data and model predictions were due to normally distributed 358homoscedastic error in the data. These estimations were done with a modified version of the BYOM platform coded in Matlab 360(www.debtox.info/byom). Confidence intervals were estimated from the 361likelihood profile of each parameter. Universally suitable goodness-of-fit 362 measures are lacking for nonlinear models (see e.g. Shcherbakov et al. 3632013), which problem was compounded by the composite nature of the trout 364data sets analyzed in this study. Therefore, in the analysis of trout data sets, 365in addition to likelihood values, two goodness-of-fit measures were used to 359

366evaluate model performance: the symmetric mean scaled error, \overline{SMScE}_{i} , and 367the model efficiency, ME - see Supplemental Information for equations. 368

369Results

370The dDEB and stdDEB+ models are relatively parameter sparse. The dDEB 371 model needed 21 parameters, of which 12 were estimated, to describe the 372patterns in the main data set by Gillies et al. (2016), including total body, 373ovaries, total body less ovaries and liver wet weight, mean follicle diameter 374and vitellogenin and estradiol plasma content. The stdDEB+ model required 375two more parameters, 23 in total, of which 11 could be estimated from the 376main data set. Thus, on average, less than two parameters were estimated 377 from each data type.

378Despite this relative parameter sparseness, both models fit the trends in the 379 main data set well (see Figure 2 and Table 2). The fits to the weight and 380follicle diameter data are virtually indistinguishable between the two models 381(see Figure 2A-E). The goodness-of-fit measures are also similar for the two 382 models (see Table 2). In addition, the estimated values for the general

383 reserve turnover rate k_{F} , the only free core DEB parameter, are statistically 384indistinguishable at the 95% level (see Table 2d), though the value implied 385by the parameters published in the DEB parameter database for rainbow 386trout (Kooijman et al. 2017) is about 10-20% lower (2.92 10^{-3} day⁻¹ at 11° C). 387More divergence in model performance is seen in the predictions of plasma 388vitellogenin and estradiol contents, notably during the last third of the 389reproductive cycle (see Figure 2F-G). The peaks of those plasma contents in 390this period are substantially better described by dDEB than by stdDEB+, as 391the latter cannot capture the drop in plasma vitellogenin and estradiol levels 392near the end of the reproductive cycle. The goodness-of-fit measures for 393those plasma contents also favor dDEB over stdDEB+ (see Table 3). In 394addition, the overall goodness-of-fit measures point to dDEB as the superior 395 model. The AIC criterion also points to dDEB as the preferable model, since 396the log likelihood of dDEB is 21.9 higher than that of stdDEB+, which is a 397large difference, especially given that dDEB has only one more free 398 parameter than stdDEB+.

399Although cultivation conditions were roughly similar among the three 400experiments, the fish in the supplementary data sets SD1 and SD2 grew 401more vigorously than those in the main data set. This can be clearly seen in 402Fig. 3A, which shows that the model predictions by dDEB and stdDEB+ with 403the parameters estimated from the main data set (bottom two curves) 404underestimate growth of fish in set SD1. The predictions are greatly 405improved, however, by adjusting the general reserve turnover rate parameter. Increasing this value by 25% (dDEB) or 20% (stdDEB+) yields 406 407 curves that are virtually indistinguishable and represent the growth data 408well. Similarly, with the value of the general reserve turnover rate parameter 409from the main data set, both models estimate the predictions of end weights 410in data set SD2 about 25-30% lower than actually observed. Also with this 411data set, satisfactory estimates of final body weights are obtained by 412increasing the value of the general reserve turnover rate parameter with 35% (dDEB) or 20% (stdDEB+) (results not shown). 413

414The analysis of reproductive data from SD1 and SD2 comes with two 415 caveats. First, the exact moment of spawning in these experiments is 416unknown. This hinders the comparison of model predictions of reproductive 417endpoints with observed values, as the former depend strongly on timing, 418given the relatively steep increase in ovary weight during the final weeks of 419the reproductive cycle (*cf.* Fig 2C). Second, the models predict the weight of 420ovaries, whereas the data report egg mass. With these caveats in mind, we 421take the census time to be 355 days into the reproductive cycle and assume

422the final weight of the ovaries equals that of eggs. Then, with the reserve 423turnover rate from the main data set, the models overestimate the 424reproductive effort in data set SD1 by about a third (see Table 3). With the 425general reserve turnover rate adjusted (see above), this overestimation 426increases to 45-70%, though the gonadosomatic index (GSI) remains 427relatively unaffected as body masses are also predicted higher. Relative to 428data set SD2, the models underestimate reproduction 25-30%, assuming 429general reserve turnover rates estimated from the main data set. With those 430estimates adjusted as before, underestimates shrink to 2% and 20% for 431dDEB and stdDEB+, respectively, while predicted GSI values change 432 relatively little. The models predict reproductive effort at day 355 as a 433function of total body mass about similarly, considering the scatter in the 434data (see Fig. 3B). With general reserve turnover rates adjusted, the 435measured mean mass and diameter of single eggs in data set SD2, 105.7 436(\pm 14.5) mg and 5.54 (\pm 0.36) mm, respectively, are close to the values 437predicted by dDEB (93.3 mg and 5.62 mm, respectively), whereas the 438predictions by stdDEB+ differ more (65.3 mg and 4.96 mm, respectively).

439

440**Discussion**

441We have formulated and evaluated two models of feedback control on the 442production of reproductive matter. The models provide a key to 443quantitatively connecting molecular level processes to organismal 444 performance, a major challenge in biology. In particular, they describe

445growth and reproduction as processes subject to hormonal regulation, and 446thus provide a link between detailed physiologically-based models about the 447endocrine system (see e.g. Gillies et al. 2016) to the DEB modeling 448 framework.

449Important strengths of DEB include its generality and relative simplicity. The 450core dynamics of the standard DEB model for a healthy animal consist of 451only three state equations and involve universal processes, such as feeding, 452 maintenance, development, reproduction and growth, with similarly general 453formulae relating these processes to measurable rates, such as respiration, 454waste and heat production. The additional equations required for modeling 455particular species and context specific measurable quantities (e.g., Equations 4568-19 in Table 1) are somewhat narrower in applicability, but still have 457considerable generality. For example, we would expect these equations to be 458applicable to most fishes, albeit with species-specific values for their 459 parameters.

460Our representation of demand-driven energy allocation to the production of 461 reproductive matter focuses on a general dynamic mechanism, namely 462feedback control of gonads. We used this mechanism to develop two 463extensions of the standard DEB model, stdDEB+ and dDEB (see Figure 1). 464These extensions share the feature that, depending on the nutritional state 465of an adult, growth may occur concurrently with the accumulation of 466reproductive matter; this contrasts with other simple models, often used in 467optimality arguments, in which an adult commits either resources to growth

468or to reproduction at any given time (see e.g. Cohen 1971; Quince et al. 4692008). However, a dDEB organism may cease to grow, and may even shrink, 470while it continues to allocate resources to reproduction (see below). We 471evaluated these extensions in depth with data on a single fish species, i.e., 472rainbow trout, due to the availability of extensive, time-resolved information 473on whole organism performance as well as on suborganismal processes 474 related to the endocrine system.

475Our models describe the production of biomass and reproductive matter in 476female rainbow trout in the three data sets analyzed here about equally well 477(see Fig. 2A-D, 3 and Table 3). Values of the core DEB parameter quantifying 478the rate of general reserve turnover estimated from these data sets differ 47920-35% from each other, and they are 10-55% higher than the value 480published in the DEB parameter database (Kooijman et al. 2017), though are 481rather similar in dDEB and stdDEB+ (see Table 2d). Rainbow trout are a 482remarkably adaptable species with a long history of domestication and wide 483geographic distribution, existing as both anadromous and land locked 484varieties and have a relatively high level of genetic variation among different 485populations (Maccrimmon 1971; Hershberger 1992). Thus, it is not surprising 486that the general reserve turnover rate parameter varies among strains. The 487dDEB variant performs better in describing the dynamics of plasma estradiol 488and vitellogenin contents as well as the development of individual eggs (see 489Fig. 2E-G), and overall dDEB fits the main data set significantly better than 490stdDEB+, as judged from likelihood values (see Table 3). While the types of

491data best described by dDEB are of relatively minor importance to whole 492organism performance, their consideration reflects conceptual differences 493between model variants with important implications.

494The major conceptual difference between dDEB and stdDEB+ lies in the 495timing of (somatic) reserve allocation to reproduction. In stdDEB+, a well-fed 496adult allocates a constant fraction of mobilized reserves to reproduction plus 497 maturity maintenance throughout the reproductive cycle and grows at a rate 498that is independent of the size of the reproductive buffer. This contrasts with 499the dynamic allocation of reserves in dDEB, in which the allocation is under 500the control of the size of the reproductive buffer relative to that of the 501animal. Consequently, this allocation can vary a great deal over a 502 reproductive cycle (see Fig. S1 in the Supporting Information). Concurrently, 503growth follows an opposite trend. In a constant environment, dDEB predicts 504that most of the growth of a species with a seasonal reproduction pattern 505occurs before the gonads start developing substantially, whereas growth in 506stdDEB+ is of the von Bertalanffy type. Consequently, size data could 507discriminate between the two models. Unfortunately, the total body weight 508 measurements analyzed in this study contain too much scatter to be of much 509help. Length measures typically are relatively precise and could therefore be 510used to evaluate the merits of dDEB and stdDEB+. It should be noted, 511though, that dDEB reduces to stdDEB in a hypothetical adult animal that 512releases gametes nearly continuously, as the density of the reproductive 513buffer would be almost constant.

514Both dDEB and stdDEB+ predict the growth of the gonads occurs primarily 515during the later parts of the reproductive cycle, which is a common 516observation for synchronous annually spawning fishes like rainbow trout 517(Tyler & Sumpter 1996) as well as many marine invertebrates, 518notwithstanding the time-invariant fraction of reserves being allocated to 519reproduction in the latter model variant. In stdDEB+, this is made possible by 520separating the reproductive buffer into two sequential pools, of which the 521first, reproductive reserves, receives somatic reserves according to the 522kappa rule of standard DEB, whereas the second containing actual 523reproductive matter (e.g., eggs) exerts positive and negative feedback 524 control on the rate at which it is being filled with reserves from the first pool 525(see Equations 10-11 and Fig. 1). A potentially unrealistic consequence of 526 separating the reproductive buffer into two pools is that although the gonad 527pool may be completely emptied during spawning, an animal following 528stdDEB+ may be left with a substantial amount of reproductive reserves at 529the time of spawning. Indeed, in stdDEB+ parameterized with the main data 530set, a three year old female rainbow trout releases only a little over 50% of 531the total amount of somatic reserves allocated to reproduction at spawning, 532 despite its negligible reproductive buffer at the beginning of the reproductive 533cycle (see Figure S2 in the Supporting Information). In addition, stdDEB+ 534 recognizes two reserve pools, reproductive and somatic, with different 535dynamics; this begs the question how an animal following stdDEB+ would be

536able to tell apart those reserve pools, given their likely large overlap in 537 chemical nature and storage location.

538A particular characteristic of dDEB is that reproduction can induce starvation 539symptoms, even when environmental resources are abundant. Due to the 540demand driven positive feedback of the reproductive buffer on reserve 541allocation in dDEB, the energy flow to the somatic branch may become 542insufficient to meet somatic maintenance demands. At that point, an 543organism has several options (Kooijman 2010). For instance, it could 544increase the reserve mobilization rate, give maintenance requirements 545priority over reproduction, reabsorb reproductive matter, skimp on 546 maintenance, or use structural biomass as an energy source to meet 547 maintenance, i.e., shrink. All these options may be realistic, depending on 548the life history strategy of the organism. For instance, reabsorption of 549gonads under stress conditions occurs in parasitoid wasps (Richard & Casas 2009; Richard & Casas 2012), bivalves (Gosling 2003) and fishes (Schreck, 550 551Contreras-Sanchez & Fitzpatrick 2001), among other groups. Here we 552allowed structural biomass to be recycled for maintenance purposes, but did 553so in a provisional manner (the thermodynamic implications of shrinking are 554rather intricate and fall beyond the scope of this paper). This mechanism of 555structure recycling may be of use to describe the degeneration of structures 556and the loss in vitality before and after spawning in semelparous fishes, such 557as species of eel and salmon.

558In addition, this recycling mechanism is relevant for species with marked 559biomass turnover processes, such as holometabolous insects and annual 560plants. In the pupa stage, holometabolous insects degrade most tissues and 561build new structures. Without demand-driven feedback mechanisms and 562implied recycling mechanisms for structural biomass, such as in dDEB, the 563 modeling of holometabolous insects within a DEB context is cumbersome 564(Llandres et al. 2015). Many annual plants feature strategies in which 565vegetative structures wither while seed mass is still increasing. The common 566bean, P. vulgaris, for instance, clearly displays this pattern (see e.g. Lima et 567al. 2005). In order to illustrate the ability of dDEB to capture this pattern, we 568used a stripped-down dDEB model without reserves, added an empirical 569relationship describing the dynamics of relative leaf cover (see Figure 4A) 570and a simple standard model describing photosynthesis as a function of leaf 571 cover (see Supplemental Information for a full description of the model). This 572 modified dDEB model describes the dynamic allocation of resources to above 573ground vegetative biomass and reproductive matter in this particular data 574set quite well (see Fig. 4B). It should be noted that the apparent relocation of 575structural biomass to seeds is due to an indirect mechanism: structural 576biomass is metabolized to meet the maintenance demands of the remaining 577structure, while an increasing fraction of photosynthate is invested in seed 578 production.

579Our models are designed to serve as pivots connecting Adverse Outcome 580Pathways (AOP) for endocrine disruptors to processes at ecological levels of

581organization. AOPs conceptualize the transfer of information from molecular 582to organismal levels of organization as the first step in scaling up to inform 583ecological risk assessment (Ankley *et al.* 2010). Starting with one or more 584 molecular initiating events, i.e., perturbations caused by a chemical stressor, 585AOP models quantify the impacts of that stressor on molecular, cellular 586and/or organ-level processes. However, these models currently lack the 587ability to further these impacts to projections of those adverse effects on 588individual growth, reproduction, and survival, which are in the realm of the 589DEB modeling framework. Thus, the AOP framework could provide the 590 mechanistic basis for modeling toxic effects within the DEB modeling 591framework, and thereby opening the door to process-based risk assessments 592in ecotoxicology (Murphy et al. 2018).

593In conclusion, by including gonadal feedback control on energy allocation to 594 reproduction and somatic processes we obtain three major benefits. Firstly, 595through this mechanism, the formation of reproductive matter can take on a 596 marked seasonal, semelparous or batch-mode pattern with a minimum of 597 mathematical complexity. Secondly, it facilitates the modeling of growth and 598 reproduction as processes subjected to endocrine regulation, that is, it 599enables a connection between organismal and suborganismal level 600processes. Thirdly, since the control variable, i.e., the density of reproductive 601 matter, has a generic form, species and sex specific attributes of endocrine 602 regulation can be added without changing the core of the model. We 603 anticipate that this mechanism, and our two model extensions that follow

604from it, will provide a gateway for incorporating molecular-level mechanisms 605of endocrine disruption into organismal-level models of individual 606 performance, such as those in the DEB framework.

607

Authors' contributions 608

609All authors conceived the ideas. EM, KL and RN developed the models. IS 610collected the data. EM analyzed models and data and led the writing of the 611 manuscript. All authors contributed critically to the drafts and gave final 612approval for publication.

Data accessibility 613

available from the Dryad Digital Repository: 615 <https://doi.org/10.5061/dryad.58j9r88> (Muller *et al.* 2018). 614

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759Table 1. Equations.

DEB Model Expressions

$$
stdDEB + (1 + mE + mF + mG) dM MV / dW
$$
 (9)

Ovary wet weight, W_0

dDEB ^k OVmFd^M M^V /d^W (10)

$$
stdDEB + \qquad \kappa_{ov} m_{\rm g} d_{\rm M} M_{\rm v} / d_{\rm W} \tag{11}
$$

Liver wet weight, W_{L}

dDEB $(p+m)(1-k_{\text{OV}}) d_{\text{M}} M_{\text{V}}/d_{\text{W}}$ (12)

 $stdDEB+$ $(p+m_G)(1-k_{OV})d_MM_V/d_W$ (13)

$$
stdDEB+ d_{\tau}k_{RE}m_{ke}m_{\tau}(m_{Em}-m_{s})-(k_{\tau}+j_{v})V_{\tau}
$$
\n(19)

760

761

Table 2. Parameters and variables used in the analysis of the main set of 762 763rainbow trout data. (a) Dynamic model quantities; (b) Experimental 764 variables; (c) fixed parameters; (d) estimated parameters.

765 (a) Dynamic model quantities

766

767(b) Experimental variables

768

769 (c) Fixed parameters $(T=11^{\circ}C)$

770* Add my Pet' DEB parameter data base (Kooijman et al. 2017)

771** Parameterization section in Materials and Methods

772^{***} Free parameter in dDEB - see Table 2d

774(d) Estimated parameters

776* Fixed parameter in stdDEB+ - see Table 2c.

Data Figur dDEB $\ln \mathcal{L} = -1551.2$ **stdDEB+** $\ln \mathcal{L} = -1573.1$ **type e** σ SMScE ME σ SMScE ME 0.592 0.55 $\overline{14.2}$ ng 0.724 0.363 E2 2G 11.8 ng. ml^{-1} ml^{-1} 6 2E 0.445 mm 0.162 0.90 0.491 mm 0.298 0.885 L_F 6 2A 170 g 0.077 0.81 157 g 0.071 0.838 $W_{\!\scriptscriptstyle R}$ 1 2D 5.66 g 0.176 0.44 5.70 g 0.175 0.439 W_{i} 9 2B 44.7 g 0.304 0.87 47.7 g 0.298 0.853 W_{α} $\frac{1}{0.07}$ 2C 305 g 0.109 299 g 0.107 0.110 $\mathsf{W}_{\!_\mathcal{B}}$ - $\mathsf{W}_{\!_O}$ 1 2F 35.2 mg 0.542 0.64 40.3 mg 0.635 0.534 V_{τ} ml^{-1} 4 ml^{-1} Overall ² 0.280 0.61 0.309 0.575 5

778 T **able 3.** Statistics of model fits to Main data set 1 .

779¹A perfect fit implies $S\!M\!S\!c\!E$ =0 and $M\!E$ =1.

780

Table 3. Measured and predicted body and egg masses supplementary data 781 782 sets on day 355.

785**Figure 1.** Conceptual representations of the standard DEB (stdDEB) model 786 for healthy adults and of two types of modifications, dDEB and stdDEB+. 787stdDEB (Nisbet et al. 2000; Kooijman 2010; Jusup et al. 2017) describes the 788 rates at which an adult animal acquires food, assimilates the energy and 789 nutrients therein into general reserves, and allocates those reserves to 790 somatic and maturity maintenance, growth and reproduction; this allocation 791 is defined as catabolism. A fixed fraction K of the catabolic flux is allocated 792to somatic maintenance and growth. Somatic and maturity maintenance are 793demand-driven processes and take priority over growth and reproduction; all 794 other processes in stdDEB are supply-driven. In dDEB, stdDEB is modified to 795 include positive and negative feedback of the reproductive buffer on the 796allocation of the catabolic flux. Thus, in dDEB, reproduction is a demand797driven process with a variable fraction λ of the catabolic flux allocated to 798 maturity maintenance and reproduction. stdDEB+ separates the 799 reproductive buffer in two pools: reproductive reserves and actual 800reproductive matter (gonads). The rate at which reproductive reserves are 801 converted into reproductive matter depends on the densities of reproductive 802 reserve and reproductive matter, implying that gonad loading is a demand-803driven process. Solid arrows represent energy and material fluxes; broken 804 arrows represent feedback mechanisms; boxes represent state variables; 805 modifications of dDEB and stdDEB+ relative to stdDEB are presented in black 806while communalities are shown in grey. Note that DEB processes and 807 quantities are abstractions; auxiliary rules are required to relate them to 808experimental quantities - see Table 1.

Figure 2. Model fits of dDEB (solid line) and stdDEB+ (dashed line) to main 813 814data set with rainbow trout (symbols), including (A) total body wet weight; 815(B) total body wet weight less wet weight of ovaries; (C) wet weight of 816ovaries; (D) wet weight of liver; (E) mean diameter of maturing follicles 817(mean per fish); (F) plasma vitellogenin content; and (G) plasma estradiol 818 content. Measurements denoted 'x' in Panel A were used to calculate 819 corresponding data in Panel B and were therefore omitted in the fitting 820procedure. Error bars denote standard deviations ($n = 3$ or 4). Parameter 821 estimates are given in Table 1d and goodness-of-fit measures in Table 2. 822Data from Nagler et al. (2012) and Gillies et al. (2016).

Figure 3. The ability of dDEB and stdDEB+ parameterized with values 825 826estimated from the main data set (see Fig. 2 and Table 1d) to predict 827production in rainbow trout was evaluated with supplementary data set SD1 828(A) and set SD2 (B). (A) With the estimated parameter values, both dDEB 829(dotted curve) and stdDEB+ (dot-dashed curve) underestimated the gain in 830weight in set SD1 (circles). Predictions are greatly improved by increasing 831the reserve turnover rate by 25% (dDEB, solid curve) or 20% (stdDEB+, 832broken curve) relative to the value estimated from the main data set. (B) 833dDEB (solid curve, reserve turnover rate 35% higher than the one in the 834 main data set) and stdDEB+ (broken curve, reserve turnover rate 20% 835higher than the one in the main set) predict measured total egg mass versus 836body weight (symbols) from data set SD2 about equally well. 837

839

Figure 4. Application of a simplified version of dDEB to production in the 840 841common bean, Phaseolis vulgaris. (A) An empirical third degree polynomial 842describes the dynamics of the leaf area index, defined as the total green leaf 843surface are per unit area ground cover, an important determinant of the

photosynthetic capacity ($\mathsf{P}_{\!1}$ =30.5 min 1 , $p_2 = 5.2 \text{ min}^{-2}$, $p_3 = 0.08 \text{ min}^{-3}$ 844photosynthetic capacity $({}^{\nu_1}$ ^{-50.5}11111, ${}^{\nu_2}$ ^{-5.211111}, ${}^{\nu_3}$ ^{--6.0011111}). (B) The 845simplified dDEB model fits above ground vegetative biomass (open circles, 846solid curve) and pod mass (closed circles, dotted curve) with mean bean 847 mass as the initial amount of structural biomass, observed mean time of first 848flowering (34 d) as starting point of photosynthate allocation to reproduction,

849 $m_{\!\scriptscriptstyle \! F}$ =0.01 $\,$ and negligible losses in converting photosynthate into vegetative 850and reproductive biomass. Parameter estimates (with 95% confidence 851intervals) are λ_{m} = 0.52 (0.30-0.87), $m_{\hbox{\scriptsize $fm}}$ = 1.09 (0.95-1.24), $\dot{J}_{\hbox{\scriptsize $M}}$ = 0.08 (0.03-8520.16) d⁻¹ and C= 0.12 (0.08-0.17); \int_{Pm}^{∞} 65.2 g dry weight m⁻² d⁻¹⁻ based on 853the net photosynthesis rate estimated by Sale (1975). Data are from Lima et 854al. (2005) and represent the means of four replicates of six cultivars grown 855from large seeds. See Supplemental Information for model description.