Poor nutritional condition promotes high-risk behaviours: A systematic review and meta-analysis

Short Running Title: Condition effects on risky behaviour

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Abstract

Animal behaviour can lead to varying levels of risk, and an individual’s physical condition can alter the potential costs and benefits of undertaking risky behaviours. How risk-taking behaviour depends on condition is subject to contrasting hypotheses. The asset protection principle proposes that individuals in better condition should be more risk averse, as they have higher future reproductive potential (i.e. more to lose). The state-dependent safety hypothesis proposes that high-condition individuals that are more likely to survive and maximise the benefits of risky situations may make apparently riskier choices, as their individual risk is in fact lower. We systematically searched for studies that experimentally manipulated animals’ nutritional or energetic condition through diet treatments, and subsequently measured risk-taking behaviour in contexts relating to predation, novelty, exploration. Our meta-analysis quantified condition effects on risk-taking behaviour at both the mean and variance level. We preregistered our methods and hypotheses prior to conducting the study. Phylogenetic multilevel meta-analysis revealed that the lower nutritional condition individuals showed on average ca. 26% greater tendency towards risk than high-condition individuals (95% confidence interval: 15% – 38%; n = 126 studies, 1297 effect sizes). Meta-regressions revealed several factors influencing the overall effect, such as the experimental context used to measure risk-taking behaviour, and the life-stage when condition was manipulated. Meta-analysis of variance revealed no clear overall effect of condition on behavioural variance (on average ca. 3% decrease in variance in low- vs high-condition groups; 95% confidence interval: -8% – 3%; n = 119 studies, 1235 effect sizes), however, the experimental context was an important factor influencing the strength and direction of the variance effect. Our comprehensive systematic review and meta-analysis provide insights into the roles of state-dependency and plasticity in intraspecific behavioural variation. While heterogeneity among effect sizes was high, our results show that poor nutritional state on average increases risk-taking in ecological contexts involving predation, novelty and exploration.
Keywords: boldness, exploration, novelty, novel environment, novel object, predation, predator response, animal personality, shoaling, dietary restriction

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I. Introduction

Animals often must gamble with their lives, with behavioural decisions frequently involving trade-offs between resource acquisition, reproduction and survival. Many of those decisions have to be made in face of incomplete information or inherent stochasticity in the outcome. Some behaviours are thus inherently ‘risky’ (defined as involving high outcome variance), and promise large gains, but also the potential of large losses (Barclay, Mishra, & Sparks, 2018). The concept of risk may be applied broadly in animal ecology (e.g. participation in aggressive contests, reproductive investment decisions etc.), and is often used in contexts where the outcome is unpredictable (e.g. responses to novelty, sensu boldness; White et al., 2013) or contexts with a high relative likelihood of death (e.g. predator responses; Réale et al., 2007). When to engage in risky behaviours is an important decision in an individual’s life, and thus an important research topic in behavioural ecology. State variables, such as individual condition, can modify the costs and benefits of risk taking (Luttbeg & Sih, 2010).
State-dependency of behaviour is an important driver of among-individual variation in
behavioural traits (Sih et al., 2015; Niemelä & Dingemanse, 2018; Moiron et al. 2019), but its
specific relationship to risk taking is subject to unresolved competing hypotheses.

Individual condition, considered here as variation in nutritional or energetic state, can lead to
differences in morphological, behavioural and cognitive traits among individuals
(Borcherding & Magnhagen, 2008; Buchanan, Grindstaff, & Pravosudov, 2013; Han &
Dingemanse, 2015), which can subsequently affect risk taking in different ways. Animals in
high condition might be risk-averse, as these individuals have a lot to lose in terms of future
reproductive potential (the ‘asset-protection principle’; Ludwig & Rowe, 1990; Clark, 1994),
whereas individuals in low condition have more to gain in terms of improved condition,
elevated competitiveness, and starvation avoidance, particularly when an individual is
relatively close to their starvation threshold (Dall & Johnstone, 2002; Luttbeg & Sih, 2010;
also known as the ‘needs-based’ explanation, Barclay et al., 2018). Contrastingly, the ‘state-
dependent safety’ hypothesis (also known as the ‘ability-based’ explanation) predicts that
individuals may appear to take greater risks where they are better able to survive and
maximise the benefits of engaging in risky behaviours, as they individually experience a
lower level of risk (Barclay et al., 2018). State-dependent safety might apply if improved
condition allows greater investment in physical and/or cognitive capabilities (e.g. increased
vigour and/or ability to evade or defend against predation) that reduce the level of risk for the
individual (as in Temple, 1987).

Risk taking can depend on the current and/or past condition of an individual, and physical
condition in early life may have a disproportionate effect on risk-taking behaviour. For
example, individuals may be developmentally primed to engage in risky behaviours when
those behaviours were favoured early in life (Zimmer et al., 2017), and poor early-life
environments may drive greater risk taking in adults as a way to compensate for their poor
start (Krause & Caspers, 2016). Conversely, a favourable nutritional environment during
development in particular can increase investment in traits that improve future survival and fitness, such as cognitive ability (Buchanan et al., 2013). This might allow greater risk taking if those traits provide an advantage in certain risky contexts by altering effective risk levels, if for example high-condition individuals are better protected/less vulnerable than low-condition individuals in the same situation. Theoretical support for any one directional state-effect on risk-taking is mixed, and show that the outcome may depend on environmental conditions, such as overall resource availability or acuteness of the risk factor (Luttbeg & Sih, 2010; Engqvist, Cordes, & Reinhold, 2014). Empirical results are similarly mixed, and thus it remains unknown if there are any generally applicable effects of condition on risk-taking behaviour, or the ecological context in which any one hypothesis applies.

Regardless of the hypothesis, condition effects on risk taking are often framed as adaptive responses to variation in an individual’s future fitness expectations (as in Clark, 1994; Wolf et al., 2007). The key proposition being that decisions to take risks are related to variation in state, where an individual’s state includes all intrinsic and extrinsic factors strategically relevant for their fitness (Wolf & Weissing, 2010). State-dependent responses due to nutritional condition may have interactive effects with other state variables, such as life history-differences within- or among-species (McNamara & Houston, 1996). For example, sex is a form of state variation involving differences in reproductive roles, which may alter male and female responses to poor dietary conditions (Han & Dingemanse, 2015). In some cases, males could be more sensitive to condition due to condition-dependent sexual selection, but in other cases, females may be more sensitive to condition since they often bear a disproportionate energetic burden of reproduction (Houslay et al., 2015; English & Uller, 2016). Similarly, interspecific differences in longevity may influence behavioural responses, since long-lived species generally have a larger future reproductive asset and/or more future opportunities to improve their own condition, and thus might be less willing to display risky behaviour (Clark, 1994).
A subset of ecological contexts where variation in risk-taking behaviour can apply are those involving trade-offs between resource acquisition and (implied or direct) predation risk, which are often used in connection with the concept of ‘boldness’. For example, responses to novelty involve inherently high outcome variance, as the potential benefits and dangers of novel situations are unknown to the individual. Furthermore, greater activity or exploration increases the likelihood of both finding new resources or habitat patches, and encountering predators (Réale et al., 2007, Wohlfahrt et al., 2007). Risk taking is therefore often quantified in assays involving the presence of predators directly or via predation cues, which emphasize the risk of mortality (Moschilla, Tomkins, & Simmons, 2018). Furthermore, some studies manipulate the outcome variance of foraging-related behaviour directly (Andrews et al., 2018). Studies of risk-taking behaviour across a variety of contexts have shown different responses, for example between predator and novel object experimental setups (Carter et al., 2012), or between emergence into a novel environment and startle responses (Beckmann & Biro, 2013). As such, we expect condition effects to vary across experimental contexts. For example, state-dependent safety may be more relevant in a predator-response context, if high-condition individuals are less vulnerable to predation. Similarly, the effects of starvation avoidance may be more relevant in experimental contexts where potential food rewards are explicit, where low-condition individuals may show increased risk taking.

Thus far, most studies have focused on mean behavioural effects of condition (i.e. higher or lower levels of risk taking). There has, however, been growing interest in individual-level variation in recent years (Westneat, Wright, & Dingemanse, 2015), and new tools to meta-analyze variances alongside means are revealing that meta-variance effects may be both prevalent and often overlooked (Nakagawa et al., 2015). While a recent meta-analysis of variance has shown diet restriction can increase variation in longevity (Senior et al., 2017), another has shown little evidence of environmental stress (including diet restriction) effects on phenotypic behavioural variance (Sánchez-Tójar et al., 2019). Furthermore, case studies have shown increased within-individual behavioural variation in high-condition animals, via
an increased capacity to express behavioural plasticity (Royauté & Dochtermann, 2017; Royauté et al., 2019). Conversely, it is conceivable that extremely poor conditions may lead to the expression of cryptic genetic variation, and thus increased variation in state and behaviour among low-condition individuals. However, if a high-risk strategy is the only viable option for acquiring adequate resources in a poor environment, individuals (including low-condition individuals) may converge on a high-risk phenotype (Han & Dingemanse, 2017).

Overall, condition-dependent effects on the variance in risky behaviours are likely present, but currently are difficult to predict in direction and magnitude.

We here present a systematic review and meta-analysis of studies that experimentally manipulated individual nutritional or energetic condition through diet quality or quantity treatments, and independently quantified risk-taking behaviours such as exploration, and predation and novelty responses. Specifically, we address six questions, which we preregistered previous to the study (see details below):

1. Do nutritional condition manipulation treatments have an overall effect on mean risk-taking behaviour? We do not predict a clear non-zero overall effect, but instead expect high heterogeneity among effect sizes resulting from the various contexts in which risk is measured and the multiple mechanisms that may drive condition effects on risk taking.

2. Is the effect of nutritional condition on mean risk-taking behaviour context-dependent? We expect low-condition treatment groups to show increased risk-taking behaviour in both foraging and feeding contexts (starvation avoidance effect), but reduced risk-taking behaviour in predator-response contexts (state-dependent safety effect). Across the remaining contexts (e.g. novel environment exploration, novel object response), we predict high-condition treatment groups to show reduced risk-taking behaviour (asset-protection effect).

3. Does nutritional condition have differential effects on mean risk-taking behaviour in males and females? We do not predict an overall difference between males and
females, due to the high heterogeneity in sex-based ecological differentiation across species. However, sex-specific differences in behaviour are widespread, and thus should be quantified.

4. Does nutritional condition at different life stages have differential effects on mean risk-taking behaviour? We expect that early-life treatments will have a greater effect on mean risk-taking behaviour than late-life treatments, as early-life treatments may affect mean risk-taking behaviour through both developmental and state-dependent behavioural plasticity.

5. Does the life-history of a species determine how nutritional condition affects risk-taking behaviour? We expect that a species’ maximum lifespan, a key life-history measure, will influence the condition effect on risk taking. According to the asset protection principle, longer lived species should be less willing to display risky behaviour (Clark 1994).

6. Does nutritional condition affect the amount of total variation in risk-taking behaviour within high- and low-condition treatment groups? We do not predict an overall clear variance effect between high- and low-condition experimental groups, however, as for hypotheses 1 and 2, we predict variance effects to show high heterogeneity and context-dependence.

In addition to the hypotheses above, we conducted the following exploratory (i.e. not preregistered) analyses to test for an effect of: (a) manipulation type, e.g. quantity, quality or starvation treatment; (b) manipulation direction, e.g. restriction, enrichment, or combined; (c) manipulation duration relative to maximum longevity; and (d) whether study subjects were reared in the laboratory or the wild.

II. Methods

(1) Protocol
Study protocols (research questions, a priori hypotheses, search methods and planned analyses) were registered prior to data collection to enhance the objectivity of our analysis and conclusions (see preregistration at https://osf.io/xgrkz/ Moran et al., 2018). Non-preregistered analyses are hereafter labelled as exploratory. This review was conducted following PRISMA reporting guidelines (for PRISMA diagram see Supporting Information S1; Moher et al., 2009).

(2) Systematic review and data collection

Database searches were conducted in Web of Science and Scopus, with a search query designed to identify studies involving both diet manipulations (e.g. "nutrition", "calori*", "bod* condition") and risk-taking experiments (e.g. "bold", "risk", "novel", "predat") within animal behaviour and behavioural ecology (e.g. "personalit", "temperament", "behavio* type", "risk taking behavio*"); for full search strategy see Supporting Information S2).

We screened records to find original experimental studies that manipulated the condition of animals in independent treatment groups through their diet, via both dietary quantity (i.e. partial restriction, complete deprivation or enrichment) or quality treatments (e.g. protein restriction or enrichment), and including both short term and longer term manipulations up to extended periods of weeks-months. Then we screened for studies that then subjected those animals to behavioural observations in contexts relating to risk (e.g. novel environments, novel object, risk-sensitive foraging, predator response) in independent trials (for inclusion and exclusion decision trees see Supporting Information S1). Our aim was to test for adaptive condition-dependent behavioural responses in non-human animals, therefore we excluded studies using species with compromised genetic diversity and/or evolved adaptive responses (e.g. domesticated animals, laboratory breeds, genetically modified organisms; as per Kelly et al., 2018) as well as studies on humans. Studies manipulating the micronutrient content of diets, or subjecting animals to high fat diets were also excluded as the relationship between these diet manipulations and body condition is not clear and considered beyond the
scope of this review. Dietary treatments were excluded as ‘non-independent’: where the
behaviour was measured in the presence of high and low food availability, or dietary
treatments such as periods of deprivation were applied within the novel environment (i.e.
non-independence of treatments from the behavioural assay); where the dietary treatments
were coupled with additional non-dietary factors (non-independence of the diet factor within
treatments; e.g. temperature); or, the dietary treatments were applied longitudinally (within
individuals) rather than cross-sectionally (i.e. non-independence between high and low
treatments).

Both the title and abstract screening of 5453 records (post-deduplication), and the full-text
screening of 641 published papers were conducted by two authors (NPM 100%, AST 25% at
both stages) to ensure reliability. Title and abstract screening was done using Rayyan
(Ouzzani et al., 2016), from which 626 references were included for full-text screening. The
title and abstract screening resulted in 67/1377 (4.9%) conflicted decisions between
observers, confirming high inter-screener agreement. All conflicted decisions were resolved
collectively by both screeners. A few additional references that were not captured by our
search but instead identified from different sources were also included for full-text screening
(‘non-systematic’ records, n = 15). Data from five such papers were included in the final
analysis, therefore we conducted a sensitivity analysis to test the potential effects of these
additional five references by re-running the main effects models without these effect sizes,
and results remain very similar (see Supporting Information S3). Full-text screening of 641
papers resulted in 5/160 (3.1%) conflicted decisions (i.e. where one screener included a
reference, and the other excluded it), that were resolved collectively by both screeners. Full-
text screening identified 147 studies meeting the experimental design criteria for inclusion
(see https://osf.io/3tphj/ for full-text screening decision database
‘CD_FulltextScreeningDatabase.xlsx’, and Supporting Information S1 for the PRISMA
diagram and the decision tree summarizing the full-text exclusion reasons).
Data were extracted as comparisons between the low-condition groups (i.e. the treatment group for diet restriction treatments, the control group for diet enrichment treatments) and the high-condition groups (i.e. the control group for diet restriction treatments, and the treatment group for diet enrichment treatments). Extractions were conducted by NPM with data extracted from figures where necessary using the R package ‘metaDigitise’ v1.0.0 (Pick, Nakagawa, & Noble, 2019). Data required to calculate effect sizes were (a) group means and (b) estimates of uncertainty (standard error, confidence intervals) or variability (standard deviation) in combination with sample sizes (N) for the behavioural variables of interest. Full or partial extraction of relevant data was possible from the published material of 118 studies (80.2% of all studies included after full-text screening). To recover missing or partially reported data, corresponding authors of 72 studies were contacted via a standardized author correspondence email, such that 395 (29.6%) of 1334 effect sizes in the full final dataset were obtained via author correspondence. Data from 25% of included papers (37 papers) were re-extracted by an independent observer to ensure data reliability. Of 1420 re-extracted values, errors requiring correction were identified in only 6 values (0.4%) affecting only two effect sizes included in the final analyses.

(3) Effect size calculation
We analysed mean effects using the log response ratio of group means (‘lnRR’; Hedges, Gurevitch, & Curtis, 1999), instead of Cohen’s D or Hedge’s g, as lnRR is less sensitive to heteroscedasticity. Variance effects were analyzed using the log coefficient of variation ratio (‘lnCVR’), as this effect size, unlike log ratio of variances (‘lnVR’), is less sensitive to potential mean-variance correlations (Nakagawa et al., 2015). Both ratios were calculated using low condition over high condition, such that a positive effect size represents higher risk taking or larger variance in risk taking in low-condition animals, respectively (effect sizes calculated via R package ‘metafor’ version v2.1-0, Viechtbauer, 2010). To maintain consistent directionality, effect sizes were reversed for a subset of lnRR effect sizes where lower values reflected higher risk behaviours (e.g. ‘latency to emerge from a shelter’,
‘distance from a predator’ etc.). Since lnCVR directionality is independent of the mean, sign reversals were not required. To assess if our choice of effect sizes affected our conclusions, main effects analyses were also run using alternate effect sizes for mean (standardised mean difference with heteroscedasticity correction ‘SMDH’; Bonett, 2009), and variance (lnVR; Nakagawa et al., 2015). Conclusions remained robust (see Supporting Information S4 for details).

(4) Data analysis - main effects models

Two multilevel intercept-only meta-analytic models were run for each effect size, testing for a general effect of condition treatments on risk-taking behaviour at a mean and variance level (using the function ‘rma.mv’ from the R package ‘metafor’ v2.1-0, Viechtbauer, 2010). Phylogenetic and non-phylogenetic models were run to investigate whether non-independence due to the degree of relatedness between species influenced both the overall effects and their level of uncertainty. Phylogenetic relatedness were estimated based on existing phylogenies and taxonomic information from the Open Tree of Life, and any polytomies were resolved by randomization (Hinchliff et al., 2015; via R package ‘rotl’ v3.0.7; Michonneau, Brown, & Winter, 2016; for the final phylogenetic tree see Supporting Information S5). Branch lengths were estimated using Grafen’s method (Grafen, 1989; via R package ‘ape’ v5.3; Paradis & Schliep, 2019), and were used to construct a phylogenetic variance-covariance relatedness matrix.

In addition to phylogeny, we included other random effects in our models to account for non-independence due to the use of the same species across studies (SpeciesID), multiple effect sizes taken from the same study (StudyID), and multiple effect sizes taken from the same experimental group of animals within the same behavioural experiment (ExperimentalID). A unit level random effect (EffectID) was also included as a measure of residual heterogeneity. For a subset of effect sizes, an experimental group was compared to multiple treatment groups (i.e. shared-control non-independence). Sampling variances were modeled as
variance-covariance matrices that accounted for correlated sampling variances due to the shared group designs, and were constructed following Lajeunesse (2011; for estimation methods see Supporting Information S4).

A subset of studies used a crossed factorial experimental design by applying an additional treatment factor (e.g. diet x temperature treatments; juvenile x adult dietary treatments etc.). To avoid including variance associated with the additional treatment factor in our analysis, we combined groups across the treatment factor that was not of interest to us (e.g. low condition/low temperature and low condition/high temperature). Groups were combined by calculating marginalised means and SDs (following equations for pooled means and SDs from Pick et al., 2019).

For main effects models, we investigated total, residual and random effect specific relative heterogeneity by calculating ‘$I^2$’ values (Nakagawa & Santos, 2012, via R package v0.0.0.9000 ‘MetaAidR’, Noble, 2019), and estimated absolute heterogeneity ‘$Q$’. For moderator models, we calculated the percentage of heterogeneity explained by the inclusion of moderators ‘$R^2_{\text{marginal}}$’ (i.e. as the estimated percentage decrease in heterogeneity between the moderator model and the non-moderator model), the residual heterogeneity ‘$Q_e$’, and moderator specific heterogeneity ‘$Q_M$’ (via R package ‘metafor’ v2.1-0, Viechtbauer, 2010). Where applicable, estimates are presented with 95% confidence intervals in square brackets (hereafter simply refer to as ‘confidence interval’).

(5) Data analysis - hypothesis testing models

All hypotheses were tested using phylogenetic multilevel meta-regression models for both InRR and InCVR including random effects as above (for detailed descriptions of all moderators used for hypothesis testing models see Supporting Information S6).
First, we included a categorical moderator (‘RiskContext’) to test if effects were context-dependent by classifying behavioural variables by both the functional context of the experiment (e.g. assays involving predators or predator cues, novel objects, novel environments etc.; Luttbeg & Sih, 2010) and the specific behavioural measurements (e.g. activity levels, areas explored, willingness to feed and forage, shoaling tendencies etc.; for descriptions of all categories see Supporting Information S6). Second, a categorical moderator (‘Sex’) tested for differences between male and female experimental groups. Effect sizes were calculated separately for males and females where sufficient data was available, otherwise effect sizes were categorized as mixed (i.e. groups including both sexes), or unknown (i.e. no information about the sex of study subjects). Third, a categorical moderator (‘ManipLifeStage’) tested for an effect of life-stage at the time of the treatments, with the level of maturity during diet manipulations categorised as juvenile, adult, both (i.e. for treatments spanning both periods), or unknown/mixed. If the paper did not present sufficient information to determine the subject’s life-stage, this was inferred from the available information (e.g. age, average length, weight etc.). If life-stage could not be reasonably inferred or if groups may have included both juvenile and adult individuals, these were classed together as mixed/unknown. Since treatments in juveniles may have been imposed a longer time before behavioural testing (e.g. early-life diet treatments with adult behavioural testing) relative to adult diet treatments, life-stage models also included the time between condition treatment(s) and behavioural experiments relative to the species maximum longevity as a continuous moderator (‘RelativeTimeFromTreatment.C’). Finally, to assess the role of life-history variation among species, we separately tested for effects of maximum lifespan (‘MaxLongevity.C’) and the natural logarithm of maximum lifespan (‘lnMaxLongevity.C’) as continuous moderators. Log transformed lifespan was used to better captures the variability in lifespan between species, as estimates for included species were heavily biased towards short lifespans. Lifespan estimates were obtained from online databases (AnAge, genomics.senescence.info; FishBase, fishbase.se, Animal Diversity Web, animaldiversity.org; Longevity Records, demogr.mpg.de/longevityrecords). If no
estimates were available, *ad hoc* searches for lifespan estimates from primary literature were conducted via *Google Scholar*. Where available, sex-specific and wild/captive-specific longevity estimates were used. Continuous moderators were z-transformed to aid interpretation (Schielzeth, 2010).

(6) Data analysis - publication bias tests

Several meta-regression models were used to assess our lnRR dataset for evidence of publication bias (for all included moderators and descriptions see Supporting Information S6).

First, the precision of each effect was included as a moderator, calculated as the square root of the inverse sampling variance (‘*Precision*’, a variant of an Egger’s regression based on Nakagawa & Santos, 2012), to test for small-study bias. Next, time-lag bias was tested using the year of publication as a continuous moderator (‘*Year.C*’), where a commonly observed trend is a decrease in effect size over time (Jennions & Møler, 2002; Sánchez-Tójar et al., 2018). For both the precision and time-lag models, a limited dataset excluding effect sizes obtained through author correspondence was used so that we were specifically testing for effects of publication bias in published material. Finally, using the full dataset, we used a categorical moderator to test whether effect sizes were larger in studies with partial or incomplete reporting of results (‘*EffectSizesFromPublication*’, i.e. complete, partial or none; where none refers to studies where all effect sizes had to be obtained via author correspondence). In addition, funnel plots were produced using lnRR and precision for a visual assessment of funnel asymmetry (Nakagawa & Santos, 2012; for plots see Supporting Information S7). As there appeared to be some evidence of publication bias, we also calculated fail-safe N to test the robustness of our results (function ‘fsn’, R package ‘metafor’ v2.1-0, Viechtbauer, 2010; see Supporting Information S7). Publication bias tests were not conducted for lnCVR, as the overwhelming majority of papers were focused on effects at the
mean behavioural level, with very few testing for effects on behavioural variance, so we did
not expect publication bias on lnCVR.

(7) Data analysis - exploratory models
Additional exploratory analyses (i.e. not preregistered) were included to test if differences in
the experimental designs of included studies influenced the results of both lnRR and lnCVR
(for moderators and descriptions see Supporting Information S6).

We tested a categorical moderator based on the differing types of diet manipulation
included in our analysis ('ManipType'). This included quantity (where the amount of food
ration/food access differed between groups), starvation (where one group was entirely
deprived of food for an extended period), quality (where the nutritional content of food
differed between groups) or combined (where both quality and quantity was manipulated in
the same treatment group). Since our main models compared low- versus high-condition
treatments, we also explored potential effects of this by including a categorical moderator
('ManipDirection'). This categorised treatments as restriction (where low-condition groups
were restricted relative to high condition/control groups), supplementation (where high
condition groups were enriched relative to low-condition/control groups), and dual (where
both the low-condition group was restricted and the high condition group was enriched from
standard conditions). To explore how the duration of diet treatments influenced the outcome,
a continuous moderator ('RelativeManipDuration.C') was defined as the time that the
treatment was applied as a proportion of the maximum lifespan of the species. Finally, the
influence of the source of the study subjects was tested using via a categorical moderator
('WildLabRear', wild, laboratory, commercial or mixed).

III. Results
(1) Main effects models

Intercept-only models showed a significant positive effect for lnRR, with the mean estimate corresponding to a 26% increase in risk-taking behaviour in low-condition animals compared to high-condition animals (non-phylogenetic method: lnRR = 0.23 [0.14 – 0.32], phylogenetic method: lnRR = 0.23 [0.09 – 0.38]; Table 1, Figure 1). For lnCVR, the overall estimate was small, negative and the confidence intervals overlapped zero substantially (lnCVR = -0.03 [-0.09 – 0.03]; Table 1, Figure 1). As phylogeny failed to resolve any heterogeneity in lnCVR, the estimates from the phylogenetic and non-phylogenetic models were identical.

(2) Hypothesis testing models

The magnitude of the lnRR was influenced by the experimental context, with the RiskContext moderator explaining a large amount of heterogeneity among effect sizes ($R^2_{marginal} = 12.03\%$; Table 2). Although most context-specific confidence intervals overlapped with zero, all the mean estimates were positive (Table 4). The highest estimates were found for behaviours relating to feeding under predation (lnRR = 0.75 [0.53 – 0.97]), feeding in a novel environment (lnRR = 0.36 [0.20 – 0.52]), and shoaling in a novel environment (lnRR = 0.36 [0.06 – 0.67]; Table 4; Fig 2A). The risk context also explained a large amount of heterogeneity in lnCVR ($R^2_{marginal} = 10.22\%$; Table 3), and the confidence intervals of some context-specific effects did not overlap with zero, including refuge use in a novel environment (lnCVR = 0.18 [0.04 – 0.31]), feeding in a novel environment (lnCVR = -0.16 [-0.25 – -0.07]), and, dispersal/migration decisions (lnCVR = -0.49 [-0.86 – -0.11]; Table 5; Fig 2B), showing a reduction in total variance in low- vs. high-condition treatments in those specific risk contexts.

Sex appeared to have some effect on lnRR (Table 2), but there was no evidence for an effect on lnCVR (Table 3). The lnRR estimates were positive but the confidence intervals slightly overlapped with zero for both females (lnRR = 0.15 [-0.03 – 0.33]) and males (lnRR =0.12 [-0.06 – 0.30]), while effects were strongest for mixed (lnRR = 0.34 [0.06 – 0.61]) and
unknown sex groups (lnRR = 0.29 [0.14 – 0.44]; Fig 2C). Life-stage also influenced lnRR (Table 2), and less clearly also lnCVR (although this model showed a particularly high $R^2_{\text{marginal}} = 16.64$, Table 3). Life-stage specific estimates for lnRR were lowest and overlapping zero in adult treatments (lnRR = 0.12 [-0.06 – 0.30]), and strongest for treatments that spanned both the juvenile and the adult life stage (lnRR = 0.45 [0.17 – 0.73]; Table 4; Fig 2E). Life-stage effects on lnCVR showed a negative estimate for juvenile treatments (lnCVR = -0.08 [-0.16 – 0.00]), and a positive effect, i.e. an increase in behavioural variance in low-condition treatments, when treatments spanned both the juvenile and the adult life stage (lnCVR = 0.18 [0.01 – 0.34]; Table 5; Fig 2F). Untransformed maximum lifespan did not appear to influence lnRR (0.00 [-0.08 – 0.09]). However, log-transformed lifespan showed a positive lnRR effect, with its confidence intervals only slightly overlapping with zero (0.15 [-0.01 – 0.30]; Table 2, 4), although this moderator did not appear to explain any heterogeneity ($R^2_{\text{marginal}} = 0.00\%$; Table 2). Neither lifespan estimate appeared to have a clear effect on lnCVR, however, these moderators explained a substantial amount of heterogeneity ($R^2_{\text{marginal}} = 13.81\%, 13.14\%$ respectively; Table 3, 5).

(3) Publication bias tests

Funnel plots showed some potential evidence of asymmetry (for plots and fail-safe N calculations see Supporting Information S7). The estimated effect of Precision on lnRR was negative and the confidence intervals slightly overlapped with zero (-0.002 [-0.005 – 0.000]; Table 2, 4), while $R^2_{\text{marginal}}$ was comparably high (7.81%; Table 2), showing some potential evidence of small-study bias. There was also possible evidence of time-lag bias in published data, with effect sizes appearing to trend slightly downwards over time but the confidence intervals overlapped with zero (-0.05 [-0.14 – 0.05]; Table 2, 4), while $R^2_{\text{marginal}}$ was again relatively high (8.18%; Table 2). Last, effects calculated from papers where effect sizes could be partially (lnRR = 0.26 [0.07 – 0.63]) or completely (lnRR = 0.24 [0.09 – 0.40]) calculated from the publicly available material were relatively large (Fig 3), whereas the effect from papers where effect sizes could only be obtained through author correspondence
were small and the confidence intervals overlapped with zero (lnRR = 0.10 [-0.16 – 0.35]), however, $R^2_{marginal}$ was zero for this moderator (Table 2). This difference suggests that non-reported results might be biased towards inconclusive (likely statistically non-significant) results.

(4) Exploratory models

There was limited evidence that either the type or direction of diet manipulation influenced lnRR with all diet types and directional treatments, respectively, showing positive mean estimates, and no heterogeneity explained by either of those moderators ($R^2_{marginal}$ = 0.00; Table 2, 4; Fig 4A, 4C). The effect of the duration of diet treatments on lnRR was almost zero too (Table 2, 4). There a small amount of heterogeneity explained by the rearing environment of the experimental subjects ($R^2_{marginal}$ = 1.44%; Table 2, 4), with effect sizes from laboratory reared animals being the smallest (lnRR = 0.13 [-0.03 – 0.30]), and effect sizes from wild reared animals being the largest (lnRR = 0.32 [0.16 – 0.48]; Fig 4E).

Both the type and direction of diet manipulation did not appear to influence lnCVR substantially, whereas the duration of diet treatments had a small positive effect on behavioural variance (0.05 [0.00 – 0.10]), and explained a substantial amount of heterogeneity ($R^2_{marginal}$ = 16.17%; Table 3, 5; Fig 4B, 4D). There was limited evidence that rearing environment influenced lnCVR, with less than 1% of heterogeneity explained by this moderator (Table 3, 5; Fig 4F).

IV. Discussion

Despite our expectations, we found a convincing directional effect on mean risk-taking behaviour, where individuals subject to low condition dietary treatments are more likely to show high-risk behaviour in a range of contexts involving predation and novelty. This condition-dependency may be caused by increased risk aversion in higher-condition
individuals due to their greater reproductive expectations (an interpretation consistent with
the asset-protection principle applying to the context of nutritional condition and predation-
novelty based risk), or by increased risk preference in low-condition animals due to their
elevated danger of starvation (a starvation avoidance mechanism; Luttbeg & Sih, 2010).
These adaptive interpretations contrast with a recent meta-analysis showing that riskier
behavioural types had higher survival in the wild (Moiron, Laskowski, & Niemelä, 2020),
which may highlight a distinction between behavioural variation due to personality trait
differences and due to state-dependent effects. Nonetheless, our result is consistent with the
idea of a trade-off between the potential benefits of high outcome-variance behaviours (e.g.
accessing resources) and the potential costs (e.g. predation or starvation), which animals
balance based on their current or past nutritional state (Ludwig & Rowe, 1990; Clark, 1994;
McNamara & Houston, 1996).

Although our overall effect was relatively strong, there was high heterogeneity in lnRR effect
sizes with a large proportion (>20%) related to among-species differences. Variation among
species, however, was only minimally related to their shared ancestry, with phylogeny only
accounting for a very small proportion of heterogeneity (3%). It would be interesting to know
if condition-dependence of risk-taking behaviour also applies to humans (Wilson et al., 1994;
Gosling, 2008), but the large amount of context-specificity might suggest that the effect
might vary between contexts. The high heterogeneity among effect sizes is also evident from
the wide prediction intervals estimated, and the substantial heterogeneity among studies and
experiments. Since theory predicts that state-dependent effects on risk taking vary in
strength and direction with factors such as life history traits (Clark, 1994; McNamara &
Houston, 1996) and/or local environmental/ecological conditions (Luttbeg & Sih, 2010), such
a pattern of variation among species, studies and experiments was to be expected. Critically,
given the high heterogeneity, our overall effect does not preclude the opposite pattern being
applicable in certain systems. Also, our findings focus on nutritional state in contexts often
involving direct or indirect predation risk, so state-dependent safety may be more directly
applicable when considering types of state variables that provide a more direct advantage in reducing predation risk (e.g. defensive traits), or in risk-taking contexts where physical condition provides a clearer advantage (e.g. intraspecific contests).

The experimental context of risk-taking behaviour was the most explanatory of lnRR moderators, revealing that the effect of condition in certain contexts was clear and particularly strong, such as those involving feeding. This is consistent with studies showing that the choice of experiment used to measure risk taking is important to the outcome, and that different risk-taking behaviours can show divergent patterns of individual-level variation (e.g. Carter et al., 2012). The concept of a ‘risky’ behaviour can be applied to a broad range of circumstances, as shown by the range of behavioural variables included here, and ‘risk-taking’ can refer to a suite of potentially independent behaviours. A risk context that was particularly strongly affected was shoaling behaviour in a novel environment (and, with less certainty, shoaling when exposed to a predator). Whether decisions to venture from a group can be considered a risk-taking behaviour or boldness trait has been disputed, partly due to overlap with sociability traits (Toms, Echevarria, & Jouandot, 2010), but our findings are consistent with these decisions being related to risk taking as a trade-off between resource acquisition and group safety. Contrastingly, the estimated effect was highly uncertain and close to zero for refuge emergence into a novel environment, a commonly used variable to measure bold-exploratory personalities. Studies have shown refuge emergence to be unrelated to within-species variation in other risk-taking behaviours (e.g. startle responses in Pomacentrus spp., Beckmann & Biro, 2013; or novel object tests in Chlamydogobius eremius, Moran et al., 2016), such that the relationship between refuge emergence and risk taking remains unclear.

Sex effects on lnRR did not show evidence of male-female differences, with both male- and female-specific effects being relatively small and similar to each other. It has been suggested that different reproductive roles may lead to sex-specific responses to diet
variation (Han & Dingemanse, 2015), but there does not appear to be a generalizable
direction to this effect. Life-stage effects did show evidence that treatments in juvenile stages
had strong and positive effects, while effects in adults were less clear. The effects of life-
age and sex may be interrelated in a way that was not originally anticipated, as the strong
effect in unknown sex groups may be related to an overrepresentation of juveniles in that
category. Whereas studies where sex was identifiable may have been more likely to involve
adult treatments groups, with both sex-specific and adult-specific estimates being smaller.
The influence of longevity was ambiguous, but ongoing theoretical support for asset
protection to be sensitive to life-history traits (e.g. iteroparous vs. semelparous reproductive
strategies; Luttbeg et al., 2020) suggests that a more focused analysis incorporating life-
history differences is warranted, particularly in relation to reproductive traits.

Our exploratory analyses revealed a few key patterns in condition-dependent behavioural
responses, and the suitability of our methodology. Modelling studies have suggested there
may be non-linearity in state-dependent phenotypic responses in risk-taking behaviour, due
to potential factors such as inconstant correlations between condition and reproductive value
(Clark, 1994; McNamara & Houston, 1996; Luttbeg & Sih, 2010). While not directly testing
this, evidence of a non-linear effect of condition and risk taking was not detected in the
analysis of diet manipulation direction. Effects were similar for each group (i.e. reduced vs.
standard condition; standard vs. enriched condition, reduced vs. enriched condition),
supporting a more constant directional effect of condition on mean risk taking, and
suggesting that our methodology of pooling these designs together for analysis was sound.
Similarly, the mean effect estimate was positive across all classes of diet treatment analysed
(e.g. quality, quantity etc.), such that pooling these experiments was unlikely to influence
results. Finally, wild-reared animals did show the largest effect of treatment on mean risk
taking (and also a particularly strong negative effect on behavioural variation), suggesting
that these animals might be either more sensitive to imposed dietary manipulations or more
responsive to predator-based risk due to past experiences in the wild.
Contrasting with overall mean effects, support for an overall effect of condition on behavioural variation was limited, with only a small, slightly negative and rather uncertain overall lnCVR estimate. This contrasts with the expectation that poor condition may increase phenotypic variability (e.g. by exposing cryptic genetic variation), but agrees with a recent meta-analysis showing that environmental stress does not seem to influence variation in behavioural traits across species (Sánchez-Tójar et al., 2019). Heterogeneity was generally lower in lnCVR models relative to lnRR ones, which is likely because variance effect sizes are generally associated with larger sampling variances (Sánchez-Tójar et al., 2019). Variance meta-analyses are expected to be more data hungry, although this is unlikely to be the cause of the overall weak lnCVR effect found in our study given the large dataset used.

Variation in behaviour was sensitive to the experimental context of risk-taking behaviour, with variation in both the strength and direction of context-specific effects. In particular, variance in feeding behaviour within novel environments was far lower in low-condition groups, providing some evidence that being highly motivated to feed in this context is an optimum phenotype for individuals in poor energetic state. In contrast, variation in refuge use in a novel environment was higher in low-condition groups, which may be evidence of the opposite (complementary) pattern where high refuge use is a preferred strategy for high condition individuals. Effects of life stage on behavioural variation are consistent with recent empirical evidence suggesting that developmental diet is related to phenotypic plasticity and personality development (see examples in Royauté & Dochtermann 2017; Kelleher et al. 2019). Buchanan, Grindstaff, & Pravosudov (2013) suggested that poor condition during early life stages may reduce an individual's capacity to express behavioural plasticity. This is potentially consistent with our finding of reduced behavioural variation in groups subject to low-condition treatments as juveniles, while the effect in adults heavily overlapped with zero. We also found that treatments that spanned juvenile and adult life stages (often longer term, chronic diet restriction treatments) had a positive effect on behavioural variation. Similarly,
the duration of diet treatments had a positive effect on behavioural variation, consistent with
the proposition that extremely poor diet conditions can expose cryptic genetic and
phenotypic variation (Han & Dingemanse, 2017). Nonetheless, identifying mechanisms from
unpartitioned phenotypic variance remains challenging, as the proposed mechanisms for
effects on variability in risk-taking behaviour often apply specifically to among- or within-
individual levels (Han & Dingemanse, 2015).

A pertinent question in behavioural ecology is whether phenotypic variation is primarily within
or among individuals (Westneat, Wright, & Dingemanse, 2015). Any effects on the variance
as estimated in our meta-analysis (and more generally in most meta-analysis using lnCVR)
may arise from either source. Individuals might become more variable in their behaviour in
response to some treatment (or some environmental effect) as a form of behavioural bet-
hedging or reduce accuracy of performance (i.e. within-individual level). Alternatively,
individuals might differ in their average responses to changes in conditions if they have
intrinsically different reaction norms (i.e. among-individual level). Only repeated
measurements per individual would help to separate the two variance components.

However, this type of data is usually not available in the literature (Niemelä and Dingemanse
2018). Future studies should focus on the relative importance of within- vs. among-individual
variance in the variance effects identified in our study.

Considered together, our publication bias analyses suggest there may be some limited
influence on the overall results. Time-lag analysis showed that effect sizes might be
decreasing over time, while precision analysis showed a small negative effect, both of which
can be signs of publication bias toward a positive effect (Jennions & Møler, 2002; Jennions
et al., 2013). Moreover, effect sizes obtained from author correspondence where no data
could be extracted from published material showed the lowest and most uncertain effect,
suggesting preferential publication of positive effects. Intriguingly, publication bias appears to
be present even where there are competing hypotheses, with positive effect hypotheses
(e.g. the asset protection principle) potentially seemingly preferred. We avoided methods to compensate for bias (e.g. trim and fill) as these can perform poorly in high heterogeneity datasets (Moreno et al., 2009). Instead, we advise caution when interpreting our results, and ecological meta-analyses in general, given the ubiquity publication bias effects in the literature.

V. Conclusions

(I) The overall evidence of diet and thus nutritional condition effects on risk-taking behaviour in the literature is clear, as low-condition individuals appear willing to on average take greater risks in ecological contexts relating to predation risk and novelty.

(II) While condition-dependency appears to have broad relevance across the animal kingdom, the strength and certainty of this effect may be somewhat overstated due to publication bias and large heterogeneity among effect sizes.

(III) Furthermore, the effect is strongly context-dependent, at both the mean and the variance level, suggesting that the specific ecological (and experimental) factors of any context must be considered when studying risk-taking behaviour.

(IV) Overall, there appears to be complex and nuanced effects of diet and condition on behavioural variance warranting further empirical study. Future research should focus on separating among- and within-individual variance effects of individual condition.

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VII. Authorship

NPM: Conceptualization, Investigation, Methodology, Data curation, Formal analysis, Project administration, Software, Validation, Visualization, Writing - original draft, Writing - review & editing. AST: Conceptualization, Investigation, Methodology, Data collection, Software, Validation, Writing - review & editing. HS: Conceptualization, Funding acquisition, Writing - review & editing. KR: Conceptualization, Funding acquisition, Supervision, Writing - review & editing.

VIII. Data Accessibility

All data and code used (including data processing, preparation, analysis and presentation) are available at the Open Science Framework (https://osf.io/3tphj/, doi: 10.17605/OSF.IO/3TPHJ).

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Table 1: Main effects models estimates, with random effect specific heterogeneity estimates \((\hat{r})\) expressed as percentages, and \(Q\)-test for absolute heterogeneity among effect sizes \((Q)\).

Square brackets represent 95% confidence intervals. Round brackets represent 95% prediction intervals, i.e. the range in which 95% of future or unknown effects are likely to fall.

Positive log response ratio (\(\lnRR\)) and log coefficient of variation ratio (\(\lnCVR\)) effects represent higher either risk taking or variance in risk taking in low-condition animals, respectively.

| Effect size          | \(k\) | Mean effect | \(\hat{r}_{\text{Experiment ID}}\) (%) | \(\hat{r}_{\text{Study ID}}\) (%) | \(\hat{r}_{\text{Species ID}}\) (%) | \(\hat{r}_{\text{Phylogeny}}\) (%) | \(\hat{r}_{\text{Effect ID}}\) (%) | \(\hat{r}_{\text{Total}}\) (%) | \(Q\)     |
|----------------------|-------|-------------|----------------------------------------|----------------------------------|----------------------------------|----------------------------------|----------------------------------|---------------------------------|----------|
| \(\lnRR\) (non-phylo) | 1297  | 0.23        | [20.3, 17.1 - 23.5]                    | [7.9, 6.1 - 9.8]                 | [23.2, 18.6 - 28.3]              | -                                | -                                | [45.9, 42.1 - 49.8]              | [98.0, 97.8 - 98.1] | 25864.30 |
|                      |       | [0.14, 0.32] | [0.90, 1.36]                           |                                 |                                  |                                  |                                  |                                 |          |
| \(\lnRR\) (phylo)    | 1297  | 0.23        | [19.9, 17.0 - 23.0]                    | [7.9, 6.0 - 9.8]                 | [21.7, 17.1 - 26.7]              | 3.4                              | 45.3                             | [41.7, 49.2]                   | [98.0, 97.9 - 98.2] | 25864.30 |
|                      |       | [0.09, 0.38] | [0.91, 1.37]                           |                                 |                                  |                                  |                                  |                                 |          |
| \(\lnCVR\) (non-phylo) | 1235 | -0.03       | [11.6, 9.8 - 13.5]                     | [21.6, 17.5 - 26.1]             | [0.0, 0.0]                       | -                                | 28.0                             | [25.9, 25.9 - 30.2]            | [61.2, 58.8 - 63.6] | 2543.32 |
|                      |       | [-0.09, 0.03] | [-0.78, 0.72]                         |                                 |                                  |                                  |                                  |                                 |          |
| \(\lnCVR\) (phylo)   | 1235  | -0.03       | [11.5, 9.7 - 13.5]                     | [21.6, 17.3 - 26.0]             | [0.0, 0.0]                       | [0.0, 0.0]                       | 28.1                             | [25.9, 25.9 - 30.2]            | [61.1, 58.8 - 63.6] | 2543.32 |
|                      |       | [-0.09, 0.03] | [-0.78, 0.72]                         |                                 |                                  |                                  |                                  |                                 |          |
Table 2: Hypothesis testing, publication bias and exploratory moderators for log response ratio (lnRR) models, with Q-test for residual heterogeneity ($Q_e$), moderator explained heterogeneity ($Q_m$), and the estimated percentage of heterogeneity explained by the moderators ($R^2_{marginal}$). Note, where $R^2_{marginal}$ estimates were negative, the value was set to zero. Numbers preceding hypotheses refer to the a priori hypotheses as laid out in the introduction.

| Hypothesis (model) | Effect size | $k$ | Moderator(s) | $Q_e$ (residual) | $Q_m$ (moderator) | $R^2_{marginal}$ (%) |
|--------------------|-------------|-----|--------------|------------------|-------------------|---------------------|
| Hyp. 2. Context-dependency of risk (rr.Full.h2) | InRR | 1297 | RiskContext | 14657.13 p < 0.0001 | 79.42 *** p < 0.0001 | 12.03 |
| Hyp. 3. Sex difference in risk taking (rr.Full.h3) | InRR | 1297 | Sex | 24006.28 p < 0.0001 | 15.92 ** p = 0.0031 | 0.53 |
| Hyp. 4. Effects across life stages (rr.Full.h4) | InRR | 1214 | ManipLifeStage + RelativeTimeFromTreatment.C | 16753.8 p < 0.0001 | 21.2 *** p = 0.0007 | 0.00 |
| Hyp. 5(i). Life-history effects (rr.Full.h5.i) | InRR | 1214 | MaxLongevity.C | 23933.71 p < 0.0001 | 0.00 p = 0.9651 | 0.00 |
| Hyp. 5(ii). Life-history effects (rr.Full.h5.ii) | InRR | 1214 | InMaxLongevity.C | 22654.52 p < 0.0001 | 3.46 p = 0.0628 | 0.00 |
| Publication bias 1 (rr.Full.pub1) | InRR | 908 | Precision | 13245.28 p < 0.0001 | 2.81 p = 0.0938 | 7.81 |
| Publication bias 2 (rr.Full.pub2) | InRR | 908 | Year.C | 21211.43 p < 0.0001 | 0.97 p = 0.3254 | 8.18 |
| Publication bias 3 (rr.Full.pub3) | InRR | 1297 | EffectSizesFromPublication | 23269.07 p < 0.0001 | 11.43 * p = 0.0096 | 0.00 |
| Exp a. Effect of manipulation type (rr.Full.exp.a) | InRR | 1297 | ManipType | 22616.48 p < 0.0001 | 8.24 p = 0.0833 | 0.00 |
| Exp b. Effect of manipulation direction (rr.Full.exp.b) | InRR | 1297 | ManipDirection | 20399.67 p < 0.0001 | 10.26 * p = 0.0165 | 0.00 |
| Exp c. Effect of manipulation duration (rr.Full.exp.c) | InRR | 1214 | RelativeManipDuration.C | 24024.39 p < 0.0001 | 0.06 p = 0.8007 | 0.00 |
| Exp d. Effect of rearing environment (rr.Full.exp.d) | InRR | 1297 | WildLabRear | 22799.97 p < 0.0001 | 16.57 ** p = 0.0023 | 1.44 |
Table 3: Hypothesis testing, publication bias and exploratory moderators for log coefficient of variation ratio (lnCVR) models, with $Q$-test for residual heterogeneity ($Q_e$), moderator explained heterogeneity ($Q_m$), and the estimated percentage of heterogeneity explained by the moderators ($R^2_{marginal}$). Note, where $R^2_{marginal}$ estimates were negative, the value was set to zero.

| Hypothesis (model) | Effect size | $k$ | Moderator(s) | $Q_e$ (residual) | $Q_m$ (moderator) | $R^2_{marginal}$ (%) |
|--------------------|-------------|-----|--------------|------------------|-------------------|---------------------|
| Hyp. 2. Context-dependency of risk (cvr.Full.h2) | lnCVR | 1235 | RiskContext | 2460.98 (p < 0.0001) | 38.4 *** (p = 0.0002) | 10.22 |
| Hyp. 3. Sex difference in risk taking (cvr.Full.h3) | lnCVR | 1235 | Sex | 2520.5 (p < 0.0001) | 5.9 (p = 0.2066) | 2.44 |
| Hyp. 4. Effects across life stages (cvr.Full.h4) | lnCVR | 1153 | ManipLifeStage + RelativeTimeFromTreatment.C | 2158.2 (p < 0.0001) | 9.5 (p = 0.0908) | 16.64 |
| Hyp. 5(i). Life-history effects (cvr.Full.h5.i) | lnCVR | 1153 | MaxLongevity.C | 2185.53 (p < 0.0001) | 1.41 (p = 0.2348) | 13.81 |
| Hyp. 5(ii). Life-history effects (cvr.Full.h5.ii) | lnCVR | 1153 | lnMaxLongevity.C | 2187.91 (p < 0.0001) | 0.34 (p = 0.5615) | 13.14 |
| Exp a. Effect of manipulation type (cvr.Full.exp.a) | lnCVR | 1235 | ManipType | 2535.9 (p < 0.0001) | 3.1 (p = 0.5406) | 0.00 |
| Exp b. Effect of manipulation direction (cvr.Full.exp.b) | lnCVR | 1235 | ManipDirection | 2541.4 (p < 0.0001) | 2.23 (p = 0.5266) | 0.00 |
| Exp c. Effect of manipulation duration (cvr.Full.exp.c) | lnCVR | 1153 | RelativeManipDuration.C | 2182.57 (p < 0.0001) | 4.59 * (p = 0.0322) | 16.17 |
| Exp d. Effect of rearing environment (cvr.Full.exp.d) | lnCVR | 1235 | WildLabRear | 2514.93 (p < 0.0001) | 4.6 (p = 0.3312) | 0.86 |
bias, and exploratory models, with 95% confidence intervals. $k$ shows the number of effect sizes, and $n_{\text{study}}$ shows the number of studies. Bold estimates correspond to confidence intervals that do not overlap zero. Note that models with categorical moderators were run as no-intercept models for ease of interpretation.

| Hypothesis (model) | Moderator(s) | Level | $k$ | $n_{\text{study}}$ | Estimate |
|--------------------|--------------|-------|-----|---------------------|----------|
| Hyp. 2. Context-dependency of risk (rr.Full.h2) | RiskContext | novelenvironment_activity | 248 | 46 | 0.09 [-0.06, 0.25] |
| | | novelenvironment_exploration | 153 | 33 | 0.11 [-0.05, 0.28] |
| | | novelenvironment_feeding | 331 | 37 | 0.36 [0.20, 0.52] |
| | | novelenvironment_lightdarktest | 26 | 6 | 0.20 [-0.11, 0.52] |
| | | novelenvironment_refugeemergence | 39 | 7 | 0.03 [-0.23, 0.30] |
| | | novelenvironment_refugeuse | 75 | 16 | 0.22 [0.03, 0.42] |
| | | novelenvironment_shoaling | 29 | 5 | 0.36 [0.06, 0.67] |
| | | novelobject_response | 92 | 11 | 0.18 [-0.04, 0.41] |
| | | predation_feeding | 81 | 14 | 0.75 [0.53, 0.97] |
| | | predation_response | 172 | 34 | 0.19 [0.02, 0.36] |
| | | predation_shoaling | 20 | 4 | 0.28 [-0.04, 0.61] |
| | | dispersalmigration | 15 | 6 | 0.03 [-0.38, 0.45] |
| | | other | 16 | 5 | 0.23 [-0.16, 0.61] |
| Hyp. 3. Sex difference in risk taking (rr.Full.h3) | Sex | female | 421 | 39 | 0.15 [-0.03, 0.33] |
| | | male | 291 | 37 | 0.12 [-0.06, 0.30] |
| | | mixed | 120 | 14 | 0.34 [0.06, 0.61] |
| | | unknown | 465 | 61 | 0.29 [0.13, 0.44] |
| Hyp. 4. Effects across life stages (rr.Full.h4) | ManipLifeStage | adult | 423 | 48 | 0.12 [-0.06, 0.30] |
| | | both | 179 | 8 | 0.45 [0.17, 0.73] |
| | | juvenile | 601 | 66 | 0.30 [0.14, 0.46] |
| | | unknown/mixed | 94 | 11 | 0.40 [0.11, 0.69] |
| | | RelativeTimeFromTreatment.C (covariate) | - | - | 0.01 [-0.03, 0.06] |
| Hyp. 5(i). Life-history effects (rr.Full.h5.i) | MaxLongevity.C | intercept | - | - | 0.26 [0.15, 0.36] |
| | | (covariate) | - | - | 0.00 [-0.08, 0.09] |
| Hyp. 5(ii). Life-history effects (rr.Full.h5.ii) | lnMaxLongevity.C | intercept | - | - | 0.22 [0.02, 0.43] |
| | | (covariate) | - | - | 0.15 [-0.01, 0.30] |
| Publication bias 1 | Precision | intercept | - | - | 0.28 [0.08, 0.49] |
| | | (rr.Full.pub1) (covariate) | - | - | 0.00 [-0.01, 0.00] |
| Publication bias 2 | Year.C | intercept | - | - | 0.26 [0.07, 0.44] |
| | | (rr.Full.pub2) (covariate) | - | - | -0.05 [-0.14, 0.05] |
| Publication bias 3 | EffectSizesFromPublication | no | 130 | 13 | 0.10 [-0.16, 0.35] |
| | | partial | 360 | 31 | 0.26 [0.07, 0.45] |
| | | yes | 807 | 82 | 0.24 [0.09, 0.40] |
| Exp a. Effect of manipulation type (rr.Full.exp.a) | ManipType | combined | 24 | 4 | 0.27 [-0.08, 0.62] |
| | | quality | 248 | 18 | 0.35 [0.07, 0.63] |
| | | quantity | 390 | 50 | 0.30 [0.07, 0.53] |
| | | starvation | 635 | 59 | 0.19 [0.04, 0.41] |
| Exp b. Effect of manipulation direction (rr.Full.exp.b) | ManipDirection | dual | 60 | 7 | 0.30 [-0.06, 0.66] |
| | | restrict | 1170 | 112 | 0.23 [0.09, 0.38] |
| | | supplement | 67 | 9 | 0.20 [-0.04, 0.44] |
| Exp c. Effect of manipulation duration (rr.Full.exp.c) | RelativeManipDuration.C | intercept | - | - | 0.25 [0.16, 0.35] |
| | | (covariate) | - | - | -0.01 [-0.07, 0.05] |
| Exp d. Effect of rearing environment (rr.Full.exp.d) | WildLabRear | commercial | 139 | 12 | 0.25 [-0.02, 0.52] |
| | | lab | 711 | 58 | 0.13 [-0.03, 0.03] |
| | | mixed | 15 | 1 | 0.21 [-0.5, 0.93] |
| | | wild | 432 | 57 | 0.32 [0.16, 0.46] |
Table 5: Parameter estimates for log coefficient of variation ratio (lnCVR) hypothesis testing, and exploratory models, with 95% confidence intervals. \( k \) shows the number of effect sizes, and \( n_{study} \) shows the number of studies. Bold estimates correspond to confidence intervals that do not overlap zero. Note that models with categorical moderators were run as no-intercept models for ease of interpretation.

| Hypothesis (model) | Moderator(s) | Level | \( k \) | \( n_{study} \) | Estimate |
|--------------------|--------------|-------|--------|----------------|----------|
| Hyp. 2. Context-dependency of risk (cvr.Full.h2) | RiskContext | novelenvironment_activity | 248 | 46 | 0.02 [0.06, 0.11] |
| | | novelenvironment_exploration | 153 | 33 | -0.05 [0.15, 0.05] |
| | | novelenvironment_feeding | 312 | 34 | -0.16 [-0.25, -0.07] |
| | | novelenvironment_lightdarktest | 24 | 5 | -0.09 [-0.35, 0.16] |
| | | novelenvironment_refugeeemergence | 39 | 7 | 0.04 [-0.18, 0.25] |
| | | novelenvironment_refugeuse | 75 | 16 | 0.18 [0.04, 0.31] |
| | | novelenvironment_shoaling | 29 | 5 | 0.01 [-0.25, 0.26] |
| | | novelobject_response | 88 | 10 | -0.09 [-0.35, 0.16] |
| | | predation_feeding | 61 | 13 | -0.01 [-0.21, 0.18] |
| | | predation_response | 167 | 33 | 0.02 [-0.08, 0.13] |
| | | predation_shoaling | 20 | 4 | 0.01 [-0.24, 0.26] |
| | | dispersalcumulation | 13 | 6 | -0.49 [-0.86, -0.11] |
| | | other | 6 | 3 | 0.59 [0.16, 1.02] |
| Hyp. 3. Sex difference in risk taking (cvr.Full.h3) | Sex | female | 401 | 38 | 0.05 [-0.05, 0.15] |
| | | male | 276 | 37 | 0.03 [-0.08, 0.14] |
| | | mixed | 117 | 13 | -0.09 [-0.28, 0.09] |
| | | unknown | 441 | 56 | -0.08 [-0.17, 0.00] |
| Hyp. 4. Effects across life stages (cvr.Full.h4) | ManipLifeStage | adult | 402 | 45 | 0.00 [-0.10, 0.09] |
| | | both | 116 | 7 | 0.18 [0.01, 0.34] |
| | | juvenile | 578 | 63 | -0.08 [-0.16, 0.00] |
| | | unknown/mixed | 89 | 11 | -0.02 [-0.21, 0.16] |
| | | RelativeTimeFromTreatment.C | intercept | - | - | -0.03 [-0.09, 0.03] |
| | | (covariate) | - | - | -0.03 [-0.08, 0.02] |
| Hyp. 5(i). Life-history effects (cvr.Full.h5.i) | MaxLongevity.C | intercept | - | - | -0.03 [-0.09, 0.03] |
| | | (covariate) | - | - | -0.03 [-0.08, 0.02] |
| Hyp. 5(ii). Life-history effects (cvr.Full.h5.ii) | lnMaxLongevity.C | intercept | - | - | -0.03 [-0.09, 0.03] |
| | | (covariate) | - | - | -0.03 [-0.08, 0.02] |
| Exp a. Effect of manipulation type (cvr.Full.exp.a) | ManipType | combined | 24 | 4 | 0.07 [-0.21, 0.35] |
| | | quality | 246 | 18 | 0.05 [-0.09, 0.18] |
| | | quantity | 363 | 48 | -0.07 [-0.16, 0.03] |
| | | starvation | 602 | 54 | -0.04 [-0.12, 0.05] |
| Exp b. Effect of manipulation duration (cvr.Full.exp.b) | ManipDirection | dual | 60 | 7 | 0.11 [-0.14, 0.35] |
| | | restrict | 1116 | 106 | -0.04 [-0.10, 0.03] |
| | | supplement | 59 | 8 | -0.06 [-0.27, 0.14] |
| Exp c. Effect of manipulation duration (cvr.Full.exp.c) | RelativeManipDuration.C | intercept | - | - | -0.03 [-0.08, 0.03] |
| | | (covariate) | - | - | -0.05 [0.00, 0.10] |
| Exp d. Effect of rearing environment (cvr.Full.exp.d) | WildLabRear | commercial | 127 | 11 | -0.02 [-0.21, 0.17] |
| | | lab | 679 | 54 | 0.02 [-0.06, 0.11] |
| | | mixed | 15 | 1 | 0.10 [-0.41, 0.62] |
| | | wild | 414 | 55 | -0.09 [-0.18, 0.00] |
Figure Legends

Fig. 1 Higher mean risk taking in low-condition compared to high-condition animals, but similar behavioural variation between them. Phylogenetic (black circles) and non-phylogenetic (white circles) meta-analytic means for log response ratio (lnRR) and log coefficient of variation ratio (lnCVR) with 95% confidence intervals. The number of effect sizes used in each model is k.

Fig. 2 Category-specific estimates for log response ratio (lnRR) and log coefficient of variation ratio (lnCVR) with meta-regression models testing the effect of (A, B) the experimental context for risk-taking behaviour; (C, D) sex of study subjects; and (E, F) life-stage of study subjects during the diet manipulation treatments. lnRR effects are presented on the left (A, C, D) and lnCVR on the right (B, D, F). The areas of the blue shaded circles are proportional to the number of effect sizes k used and bars represent 95% confidence intervals. A positive effect shows higher risk taking or higher variance in risk taking in low-condition animals, respectively.

Fig. 3 Category-specific estimates based on the degree that log response ratio (lnRR) effect sizes could be extracted from published material. Fully reported effect sizes are from papers where all effect sizes could be extracted from published material, partially reported effect sizes are from papers where some effect sizes could be extracted but additional effect sizes could be obtained from authors (therefore includes effect sizes from published material and author correspondence), and not reported effect sizes are those that could only be calculated from data obtained through author correspondence. The areas of the green shaded circles are proportional to the number of effect sizes k used and bars represent 95% confidence intervals. A positive effect shows higher risk taking and higher variance in risk taking in low-condition animals.
Fig 4  Category-specific estimates for log response ratio (lnRR) and log coefficient of variation ratio (lnCVR) meta-regression models for effect of (A, B) the type of diet manipulation; (C, D) the direction of the diet manipulation; and (E, F) the rearing environment of the experimental subjects. lnRR effects are presented on the left (A, C, D) frames and lnCVR on the right (B, D, F). The areas of the orange shaded circles are proportional to the number of effect sizes $k$ used, and bars represent 95% confidence intervals. A positive effect shows higher risk taking and higher variance in risk taking in low-condition animals, respectively.
Fig. 1

Effects on average behaviour

\[
\ln R (\text{phylo})
\quad k = 1267
\]

\[
\ln R (\text{non-phylo})
\quad k = 1267
\]

Effects on variance in behaviour

\[
\ln CVR (\text{phylo})
\quad k = 1235
\]

\[
\ln CVR (\text{non-phylo})
\quad k = 1235
\]
Fig. 3

Effect Size, lnRR

Fully reported
k = 807

Partially reported
k = 360

Not reported
k = 130
Fig 4.

A. Effect size, lnRR
- Combined: k = 24
- Quality: k = 246
- Quantity: k = 390
- Starvation: k = 635

B. Effect size, lnCVR
- k = 24
- k = 246
- k = 363
- k = 602

C. Dual
- k = 60
- Restriction: k = 1170
- Supplementation: k = 677

D. k = 59
- k = 1116

E. Commercial: k = 139
- Laboratory: k = 711
- Mixed: k = 15
- Wild: k = 432

F. k = 127
- k = 679
- k = 15
- k = 414