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Genetic coupling of life-history and aerobic performance in Atlantic salmon

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A better understanding of the genetic and phenotypic architecture underlying life-history variation is a longstanding aim in biology. Theories suggest energy metabolism determines life-history variation by modulating resource acquisition and allocation trade-offs, but the genetic underpinnings of the relationship and its dependence on ecological conditions have rarely been demonstrated. The strong genetic determination of age-at-maturity by two unlinked genomic regions (*vgll3* and *six6*) makes Atlantic salmon (*Salmo salar*) an ideal model to address these questions. Using more than 250 juveniles in common garden conditions, we quantified the covariation between metabolic phenotypes—standard and maximum metabolic rates (SMR and MMR), and aerobic scope (AS)—and the life-history genomic regions, and tested if food availability modulates the relationships. We found that the early maturation genotype in *vgll3* was associated with higher MMR and consequently AS. Additionally, MMR exhibited physiological epistasis; it was decreased when late maturation genotypes co-occurred in both genomic regions. Contrary to our expectation, the life-history genotypes had no effects on SMR. Furthermore, food availability had no effect on the genetic covariation, suggesting a lack of genotype-by-environment interactions. Our results provide insights on the key organismal processes that link energy use at the juvenile stage to age-at-maturity, indicating potential mechanisms by which metabolism and life-history can coevolve.

1. Introduction

Physiological processes control how life-history diversity emerges from resource allocation and acquisition trade-offs [1]. The rate of aerobic energy metabolism is a pivotal mechanism contributing to life-history variation—it modulates resource acquisition, provides cells with ATP, and constrains energy allocation to different body components and functions. Theories such as the metabolic theory of ecology and the pace-of-life syndrome theory [2,3] suggest metabolic rate covaries with life-history variation within and among species. This covariation may have a genetic basis, consequently constraining trait evolution [4], yet only a few studies have demonstrated intraspecific genetic covariation or coevolution between metabolic rate and life-history traits [5–7]. Determining whether this relation is modified by different ecological contexts (e.g. food availability) is crucial to better understand the mechanisms shaping life-history variation and demographic shifts in populations in response to environmental changes [8].

The quintessential components of energy metabolism at the organismal level (i.e. the metabolic phenotypes) are standard metabolic rate (SMR), maximum metabolic rate (MMR) and absolute aerobic scope (AS), which is the

difference between SMR and MMR [9–11]. SMR is the minimum metabolic rate of an ectothermic animal associated with self-maintenance, and therefore defines the minimal cost of living (i.e. excluding growth, digestion and locomotion). MMR defines the upper limit of aerobic performance that is functionally linked to SMR and the capacity to increase oxygen uptake and delivery beyond SMR. SMR and MMR together are the integral components of AS, which is the measure of surplus energy that can be allocated into non-maintenance functions, such as locomotion and digestion [12]. Higher AS is predicted to increase fitness via facilitating energetically demanding behaviours (such as migration, aggression, predator avoidance and prey capture) and tolerance to environmental stress [11,13–15]. However, high aerobic performance comes with costs, including maintaining a larger heart and gill surface area (associated with increased demand for osmoregulation) [12,16,17].

Allocation of energy to growth or improved condition can link metabolic phenotypes to life-history traits [18]. Life-history traits, such as the timing of maturation and migration, are determined by adaptive body-size thresholds [19–21], and metabolic phenotypes are often correlated with growth rate, albeit in a context-dependent manner [11]. Under high food availability, a high SMR in combination with high AS can increase growth rate [22], as it often correlates with traits that improve resource acquisition, such as dominance and digestive capacity [23–25]. Under low food availability, the growth benefit of high SMR or AS can be minimized (or even reversed for SMR) due to high self-maintenance costs [22,26,27]. In addition, individuals can be forced to seek new habitats or take more risks to acquire resources, exerting further fitness costs [28,29]. Hence, covariation between life-history and metabolism, whether it is genetically or environmentally driven, could be modulated by resource availability. A resource dependent change in genetic covariation (i.e. genotype-by-environment interaction, could maintain genetic variation in these traits [30–32]), but this has not been demonstrated.

In anadromous (sea-migrating) salmonids, the number of years the fish spends at sea before the first spawning (i.e. sea age-at-maturity) has a dramatic effect on its size-at-maturity [33]. Individuals spending one year at sea typically weigh 1–3 kg compared with 10–20 kg after three or more years. Increased size in late maturing individuals also translates to marked gains in reproductive investment in both sexes [33]. Earlier maturation, i.e. less time spent at sea, provides a potential fitness advantage through a higher probability of survival prior to reproduction and a shorter generation time, but comes at the expense of decreased fecundity and mating success (due to smaller size at reproduction [34–36]). The probability of early maturation at sea is also positively associated with faster growth and fat deposition in the freshwater (juvenile) stage, since the size of salmon at the onset of sea-migration has a significant influence on maturation timing at sea [19,37–40]. These relationships suggest that early maturation in salmon may be associated with higher SMR or AS via resource utilization already in early life-stages prior to sea migration [24,25]. In addition, metabolic phenotypes in the juvenile stage may explain maturation at sea through genetic correlations across life stages (e.g. [41]) if metabolic rate or aerobic performance at sea is linked to earlier maturation.

In Atlantic salmon (*Salmo salar* L. 1758), a large proportion of variation in age-at-maturity is explained by a single genomic region that encompasses the *vgll3* gene on chromosome 25 [42,43]. In addition, variation in another locus on chromosome 9, *six6*, is a strong predictor of mean age-at-maturity among populations [43] and associated with early maturation in aquaculture salmon [44]. *Vgll3* and *six6* are also associated with size-at-maturity, with the alleles conferring late maturation being associated with larger age-specific body size especially after multiple years at sea [43]. Moreover, *vgll3* is associated with precocious maturation in male salmon parr via body condition [45], emphasizing the causal effect of energy acquisition on maturation. In the last few decades, many Atlantic salmon populations have been maturing, on average, at younger ages [46], which is associated with a change in *vgll3* allele frequency in some cases [47]. Recently, a link was found between the decrease in salmon age-at-maturity and a change in prey species composition [48] (see also [49] for diet composition in relation to *six6*). These observations further highlight that genotype dependent differences in SMR or aerobic performance, which can contribute to foraging success or food assimilation (e.g. [22,50,51]), may be related to contemporary life-history evolution in Atlantic salmon.

The strong effects of the *six6* and *vgll3* genomic regions on life-history variation provide an opportunity for the genetic covariation between age-at-maturity and energy metabolism to be studied prior to maturation, i.e. at the juvenile stage, by genetic prediction. This approach makes controlled, empirical settings more feasible, as salmon require several years to reach maturation. In this study, we test if genetic covariation exists between life-history and metabolic phenotypes in juveniles; we expect that early maturation genotypes show a higher metabolic activity (SMR, MMR and AS) than late maturation genotypes under high food availability (e.g. [7]). Furthermore, because low resource availability weakens the relationship between SMR and growth [27] and may constrain aerobic performance [8], we also explore if food availability modulates the genetic covariation.

2. Material and methods

Additional experimental details are provided in the electronic supplementary material.

(a) Fish rearing and genotyping

The parental Atlantic salmon (electronic supplementary material, table S1) were first-generation hatchery brood stock (from the river Kymijoki in Finland) from Laukaa hatchery, managed by the Natural Resources Institute Finland (LUKE). In October 2019, eggs and milt were transferred to the University of Helsinki for fertilization. Full-sib families were crossed using heterozygous parents in *vgll3* and *six6* loci (*vgll3* * E/L and *six6* * E/L, where E and L refer to the alleles associated with early and late maturation, respectively). This provides offspring with all genotype combinations within each full-sib family. Feed rations were calculated assuming feed conversion efficiency of 0.8, using growth predictions by Elliott & Hurley [52]. In July 2020, fish were tagged with passive integrated transponder (PIT) tags and genotyped from fin clips.

(b) Experimental design

Food treatments were started in August 2020. Briefly, fish were fed twice a week in the low food treatment (figure 1). This was preferred over constant low ration to minimize dominance
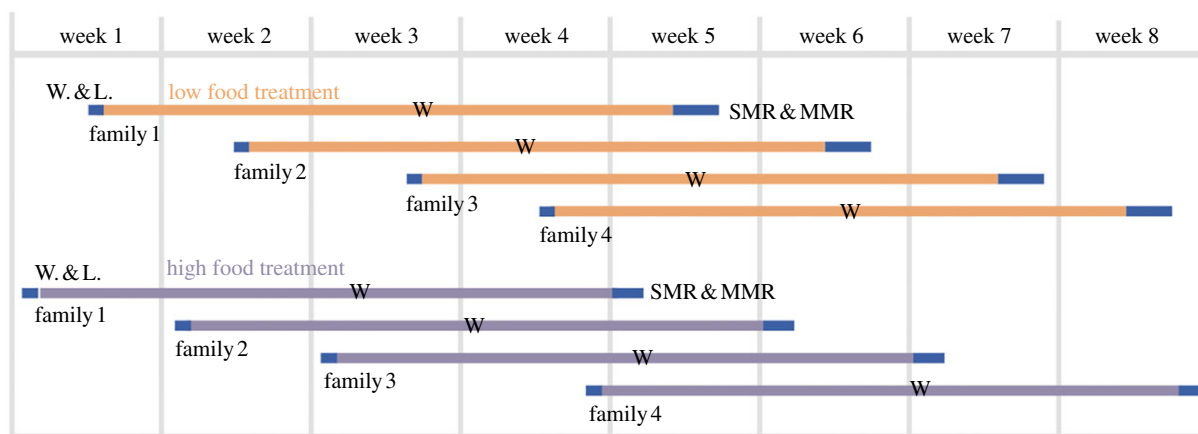
Figure 1. Timeline of the experiment. The duration of low food and high food treatments was four weeks. 'W' indicates when fish density of high food tanks was reduced to the level of low food tanks. Each horizontal line represents a separate tank. Blue blocks represent procedures (W. & L. = weight and length, and/or SMR & MMR measurements). (Online version in colour.)

hierarchies in tanks [53]. Fish in the high food treatment were fed with the total ad libitum daily ration. After 28–31 d in the treatments, 48 and 32 fish from each family in the low and high food treatment, respectively (192 and 128 in total), were measured for their SMR and MMR (figure 1), though only 290 homozygous individuals were used in the analysis (electronic supplementary material, table S2).

(c) SMR and MMR measurements

Sixteen fish at once were moved into an acclimation tank 2 days before SMR measurement (electronic supplementary material, figure S3). Each batch was from the same family and tank and balanced for genotype-sex-combinations. During acclimation, the fish were held individually, without feeding at $11°C \pm 0.1°C$ in $20 \times 20 \times 10$ cm cages, to minimize the effects of temperature fluctuations, digestion, growth and social interactions on SMR [54]. We measured SMR using intermittent flow respirometry [55–57]. The SMR measurements commenced after 42–47 h acclimation, between 11.30 and 14.30, until 8.00 the following day. Afterwards, MMR was measured using the chase method, similarly to Raby *et al.* [58], where MMR reflects increased aerobic respiration related to exercise and the oxygen debt incurred by anaerobic respiration [59,60]. Fish were then euthanized with an overdose of methanesulfonate, measured and weighed. All fish were confirmed to have immature gonads.

(d) Analysis of respirometry data

For SMR, oxygen consumption rate (MO_2, mg O_2 h^{-1}) for each linear measurement phase was derived from best-fit linear regression of dissolved oxygen concentration over time. The mean of the lowest normal distribution (MLND) was used to estimate SMR for each individual as mg O_2 h^{-1} from the extracted MO_2 slopes [10].

MMR was calculated for each individual as mg O_2 h^{-1} from the O_2 concentrations after performing background correction in *FishResp* [61]. We used two methods to identify the slope of the steepest decrease in O_2 saturation. First, we used *respR* [62] with the function *auto_rate*, fitting 1 min and 2 min windows [63] (example slope in electronic supplementary material, figure S5A). Second, slopes for MMR were extracted using a derivative of a polynomial curve fitted on each measurement (function *smooth.spline*, d.f. = 10). This is the *spline-MMR* method (electronic supplementary material, figure S5B). The slopes were then used to calculate MMR in mg O_2 h^{-1} using *Fish-Resp*-package function *calculate.MR*. MMR values calculated by the 1-min *respR* and *spline-MMR* approaches were highly

correlated (Pearson-$r = 0.98$, 95% confidence interval 0.98–0.99). We selected the spline-MMR data for further analysis.

(e) Statistical analyses

Data were analysed in R v. 3.6.2 [64]. To test for the effects of treatment and genotype on metabolic variables, we ran separate linear mixed models using SMR, MMR and AS as response variables. The response variables and body mass were $\log_{10}$-transformed to account for allometric scaling of metabolic rate, fixed effects were centred, and covariates were centred and scaled [65] (electronic supplementary material, table S3). We included treatment, *vgll3* and *six6* genotypes, and sex as fixed effects in all models. For SMR, family and measurement batch were used as random terms (models including chamber as random term were singular, and no variance was explained by chamber). For MMR and AS, family, person performing the chase test, and chamber identity were included as random terms. The order in which pairs of fish were tested for MMR each day (values 1–8) was included as a covariate for MMR and AS. To test if genotype-specific metabolic rates were affected by sex and food availability, we fitted pairwise interactions between *vgll3* and *six6* genotypes, between genotypes and treatment, and between genotypes and sex. The interaction of $\log_{10}$ body mass with treatment was included to test for potential treatment-specific allometric scaling of metabolic rate.

The full models were fitted using *lme4* v. 1.1–26 [66] with an alpha value 0.05. *P*- and *F*-test values for fixed effects were computed using type III tests with Satterthwaite's method. Pairwise differences between significant interaction terms were obtained by post hoc analysis using *emmeans* [67]. Residuals of models were confirmed to be homoscedastic and normally distributed (but the residuals of SMR between the high food and low food treatments were slightly heteroscedastic). One outlier was identified and removed from each of SMR and MMR (using *outlierTest* in package *car*, Bonferroni-corrected $p < 0.05$). The proportion of variance explained by genotypes was calculated with *partR2* [68]. Predicted means were obtained with *ggpredict* in package *ggeffects* [69]. The data were visualized using *ggplot2* v. 3.3.3 [70] and *interactions* [71]. Pearson's correlation coefficient among mass- and family-corrected SMR, MMR and AS were calculated using residuals from a mixed model.

We also evaluated alternative models, and assess parameter significance using the corrected Akaike information criterion score (AICc), an AIC score with a stronger penalty for complex models [72], using the *dredge* function from package *MuMIn* [73]. We employed model averaging if more than one model was similarly parsimonious (electronic supplementary material,

table S4), in which parameter estimates were obtained from weighted averages of all best models [74] (i.e. models with ΔAICc less than 2 compared to the most parsimonious models) using *model.avg* function (subset option = full) from package *MuMIn* [73].

3. Results

Low food treatment decreased both the specific growth rate and condition factor of the fish compared to high food treatment (electronic supplementary material, figure S6). The mean body length of fish was 70.6 ± 4.5 and 66.2 ± 4.9 mm (s.d.), and the mean body mass was 4.2 ± 0.8 and 3.3 ± 0.8 g after the high and low food treatment, respectively.

(a) Standard metabolic rate

There was no significant genotype, food availability or sex effect on SMR (table 1 and figure 2*a*). There was a marginally significant interaction effect of *six6* and food availability on SMR in the full model ($p = 0.045$) but this was non-significant in the averaged model (electronic supplementary material, table S5), and none of the pairwise contrasts were significant (the largest effect being: *six6* EE-genotype, high food versus low food, $t_{25.6} = -2.37$, $p = 0.11$). The metabolic scaling exponent, *b*, i.e. the slope of log SMR with log body mass, was marginally higher in the high food treatment (0.94, $R^2 = 0.73$) than in the low food treatment (0.87, $R^2 = 0.69$; electronic supplementary material, figure S7a; $p = 0.047$ (table 1), $p = 0.08$ (electronic supplementary material, table S5).

(b) Maximum metabolic rate

Fish with the *vgll3* early maturation genotype had a higher MMR than fish with the late maturation genotype (figure 2*a* and table 1). *Vgll3* genotype also interacted with *six6*, such that MMR was decreased when late maturation genotypes of the two loci cooccurred compared to other genotype combinations (figure 2 and table 1). The genotype effects together explained approximately 5% of the variance in MMR (electronic supplementary material, table S6). None of the treatment-genotype or sex-genotype interactions or the main effects of sex or food availability had a significant effect on MMR (table 1; electronic supplementary material, table S5). Unlike SMR, the metabolic scaling of MMR was not significantly affected by food treatment ($b = 0.86$, $R^2 = 0.76$).

(c) Aerobic scope

Fish with the *vgll3* early maturation genotype had a higher AS compared to the late maturation genotype (figure 2*a,b* and table 1, predicted means 460.2 and 440.9 mg O$_2$ kg^{-1} h^{-1} for early and late maturation genotypes, respectively). *Vgll3* was estimated to explain 1.7% of the variance in mass-corrected AS (electronic supplementary material, table S6). AS was marginally higher under low food availability than high food availability, but only in smaller fish (interaction $p = 0.037$ (table 1) and $p = 0.18$ (electronic supplementary material, table S6)); scaling exponent $b = 0.94$ ($R^2 = 0.57$) in the high food and 0.90 ($R^2 = 0.68$) in the low food treatment (electronic supplementary material, figure S7b). The *vgll3* and food treatment effects were also significant when mass adjusted SMR was included as a covariate in the model (electronic supplementary material, table S7), indicating that

the genotype effect was independent of SMR. None of the other factors had a significant effect on AS (table 1).

(d) Correlations among metabolic phenotypes

There was a positive correlation between mass- and family-corrected SMR and MMR in the high food, but not the low food treatment (electronic supplementary material, figure S8), and a very strong correlation between MMR and AS in both treatments (table 2).

4. Discussion

The timing of maturation, like many life-history traits, depends on reaching a certain body size threshold (i.e. the acquisition of sufficient energy that can be allocated for maturation processes [38,39,75,76]). In line with our hypothesis, we found that the *vgll3* early maturation genotype increased the AS of juvenile Atlantic salmon compared to the late maturation genotype. This effect was driven by a change in maximum metabolic rate (MMR), not standard metabolic rate (SMR) (nearly all variation in AS was explained by MMR in our study, table 2). A previous study showed that higher condition factor, mediated by the *vgll3* early maturation genotype, positively affected the initiation of male maturation [45]. The results presented here suggest that superior resource acquisition or assimilation via higher AS, driven by a higher MMR, is a potential mechanism by which an increased condition factor in individuals with the *vgll3* early maturation genotype could be achieved compared conspecifics with the late maturation genotype. In addition to differences in mean performance between *vgll3* genotypes, the two loci in our study exhibited physiological epistasis [77], as the cooccurrence of the late maturing genotypes in both loci was associated with lower MMR than their additive effects. The epistasis may help to maintain genetic variation under rapid adaptive responses [78,79].

The functional pathways that may explain the epistasis and the main effect of *vgll3* are not well known. However, both *six6* and *vgll3* are expressed during development and have been implicated in the control of cell fate commitment and the hypothalamus-pituitary-gonad axis in salmon [80,81]. Hence, the epistatic interaction between two genomic regions may stem from, for example, developmental canalization [82]. Addressing the causal physiological and morphological mechanisms of the link between the genomic regions and aerobic performance can shed light into the mechanisms of life-history evolution in salmon; the *vgll3* genomic region is the major genetic axis explaining variation in age-at-maturity in salmon [43], and variation in the locus is spatially divergent among populations and under rapid adaptative evolution [43,47,83,84].

Because of context-dependent covariation between metabolism and growth rate, whereby high SMR improves growth under high, but not low, resource availability [27] and because aerobic performance can be reduced under food limitation (e.g. [8]), we tested if food availability modified the genetic covariation between metabolic phenotypes and age-at-maturity. Against our predictions, there was no change in SMR or MMR due to feed restriction, nor did we find genotype-by-environment interactions, despite a strong decrease in growth rate in low food treatment. The results indicate that the different genotypes exhibit no plastic responses to food
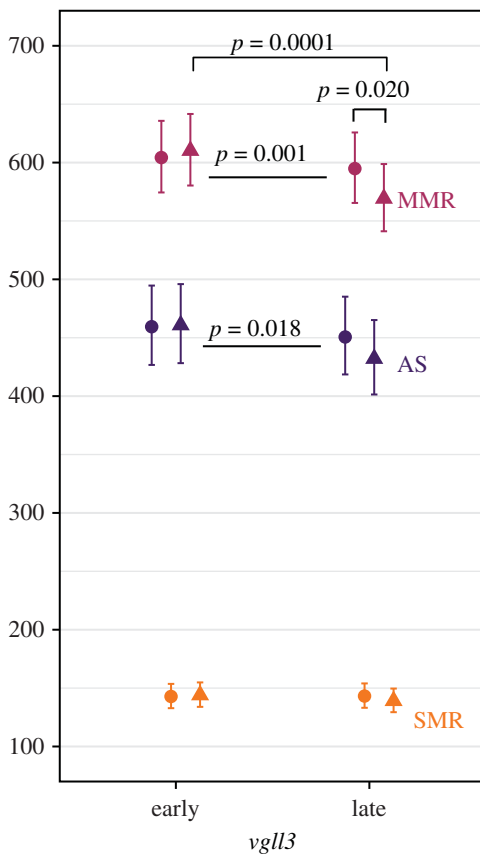
Table 1. Linear mixed models for $\log_{10}$-transformed metabolic phenotypes. All variables were centred to a mean of 0 (the category with a positive value is shown in parentheses), $\log_{10}$ body mass was scaled and centred. Significant effects shown in italics. BM, body mass, LF, low food.

fixed effect	estimate	s.e.	SSq	Den DF	F	p-value
SMR						
intercept	−0.295	0.019				
treatment (LF)	0.019	0.012	0.0039	17.67	2.772	0.114
sex (male)	0.002	0.005	0.0002	248.11	0.172	0.679
Vgll3 (LL)	−0.002	0.005	0.0003	248.98	0.235	0.628
six6 (LL)	0.002	0.005	0.0002	248.11	0.172	0.679
$\log_{10}$ BM	*0.102*	*0.003*	*1.4570*	*259.04*	*1025.739*	*<0.0001*
treatment (LF):$\log_{10}$ BM	*0.013*	*0.006*	*0.0057*	*259.25*	*3.987*	*0.047*
treatment (LF):*Vgll3* (LL)	0.006	0.010	0.0006	249.25	0.410	0.523
treatment (LF):six6 (LL)	*−0.020*	*0.010*	*0.0058*	*249.21*	*4.080*	*0.044*
sex (male):*Vgll3* (LL)	0.001	0.010	0.00001	248.33	0.004	0.948
sex (male):*six6* (LL)	0.001	0.009	0.00001	247.62	0.006	0.937
Vgll3 (LL):*six6* (LL)	−0.016	0.009	0.0042	248.20	2.930	0.088
Random effect	Var (95% CI)					
Batch	0.0005 (0.0002, 0.0012)					
Family	0.0013 (0.0003, 0.0083)					
Residual	0.0014					
MMR						
intercept	0.333	0.013				
treatment (LF)	0.008	0.006	0.003	249.48	1.632	0.203
SEX (male)	−0.003	0.005	0.001	254.25	0.295	0.588
Vgll3 (LL)	*−0.017*	*0.005*	*0.017*	*252.06*	*10.429*	*0.001*
six6 (LL)	−0.006	0.005	0.002	252.45	1.371	0.243
$\log_{10}$ BM	*0.088*	*0.003*	*1.231*	*256.37*	*737.850*	*<0.0001*
test order	−0.003	0.004	0.001	13.81	0.769	0.396
treatment (LF):$\log_{10}$ BM	−0.010	0.007	0.004	255.47	2.252	0.135
treatment (LF):*Vgll3* (LL)	−0.011	0.011	0.002	253.11	1.057	0.305
treatment (LF):*six6* (LL)	−0.010	0.011	0.001	255.77	0.840	0.360
sex (male):*Vgll3* (LL)	0.016	0.010	0.004	249.17	2.474	0.117
sex (male):*six6* (LL)	0.008	0.010	0.001	251.83	0.654	0.420
Vgll3 (LL) × six6 (LL)	*−0.023*	*0.010*	*0.008*	*253.67*	*5.076*	*0.025*
Random effect	Var (95% CI)					
Chamber	0.0002 (0, 0.0005)					
Initial	0.0001 (0, 0.0012)					
Family	0.0004 (0.0001, 0.0027)					
Residual	0.0017					
AS						
intercept	0.210	0.019				
treatment (LF)	0.012	0.008	0.006	238.60	2.168	0.142
sex (male)	−0.004	0.007	0.001	243.05	0.443	0.507
Vgll3 (LL)	*−0.016*	*0.007*	*0.015*	*240.86*	*5.653*	*0.018*
six6 (LL)	−0.008	0.007	0.004	242.39	1.468	0.227
$\log_{10}$ BM	*0.089*	*0.004*	*1.183*	*245.15*	*437.482*	*<0.0001*
test order	−0.003	0.005	0.001	14.796	0.387	0.543
treatment (LF):$\log_{10}$ BM	*−0.018*	*0.009*	*0.012*	*245.12*	*4.409*	*0.037*

(Continued.)

Table 1. (*Continued.*)

fixed effect	estimate	s.e.	SSq	Den DF	*F*	*p*-value
treatment (LF):*Vgll3* (LL)	−0.016	0.014	0.003	241.59	1.231	0.268
treatment (LF):*six6* (LL)	0.001	0.014	0.000	244.70	0.002	0.964
sex (male):*Vgll3* (LL)	0.023	0.014	0.008	239.76	2.802	0.095
sex (male):*six6* (LL)	0.009	0.014	0.001	239.97	0.487	0.486
Vgll3 (LL) × *six6* (LL)	−0.020	0.014	0.006	243.02	2.086	0.150
Random effect	Var (95% CI)					
Chamber	0.0002 (0, 0.0008)					
Initial	0.0002 (0, 0.0022)					
Family	0.001 (0.0003, 0.0064)					
Residual	0.0027					

Figure 2. (*a*) Predicted means for standard metabolic rate (SMR), maximum metabolic rate (MMR) and aerobic scope (AS) in *vgll3* and *six6* early- and late-maturation genotypes with 90% confidence intervals. The means are for average body mass, treatment, and sex effects, and back transformed to linear scale. *p*-values show significant pairwise differences between genotypes for MMR on top of the points, and for the *vgll3* main effects in MMR and AS between the points (table 1). $n = 60$–71 in each genotype combination (same individuals used for all traits). (*b*) Residual aerobic scope from a linear mixed model including $\log_{10}$ aerobic scope as response, $\log_{10}$ body mass as predictor and family as random term, showing individuals (by points) in each genotype combination. (Online version in colour.)

availability. However, a stronger feed deprivation, e.g. similar to those occurring in the winter [85], could have induced a more pronounced effect on SMR [86,87]. Our low food treatment included approximately 3 days of fasting in between feeding to satiation, similar to a 'feast and famine' feeding strategy [88]. A lack of metabolic response to reduced food availability may be beneficial if it allows the individual to maximize acquisition via food assimilation during 'feasting'.

Unlike MMR (and consequently AS), SMR did not exhibit *vgll3*-linked covariation with age-at-maturity. A decoupling of SMR and MMR in relation to life-history variation was also found by Archer *et al.* [89] in resident and migratory brown trout (*Salmo trutta*), and high MMR, but not SMR, was positively selected for in Atlantic salmon under high food competition in a study by Auer *et al.* [90]. Furthermore, a lack of differences in SMR across the *vgll3* genotypes was

Table 2. Pearson's correlation coefficients between metabolic phenotypes in high food (above diagonal) and low food (below diagonal) treatments. *p*-values given in parentheses.

	rSMR	rMMR	rAbsAS
rSMR		0.28 (0.004)	0.13 (0.09)
rMMR	0.11 (0.16)		0.99 (<0.001)
rAbsAS	−0.17 (0.04)	0.96 (<0.001)	

found in our parallel study, in which the fish were smaller (mean approx. 1 g) compared to this study (mean approx. 4 g) [91]. The lack of association to the age-at-maturity loci is unexpected, as SMR, or basal metabolic rate in endotherms, has been proposed to explain life-history variation along the fast–slow axis [5,7], but see [92]. Our results suggest that the genetic control of maturation by the *vgll3* genomic region via MMR mostly involves physiological pathways that do not alter SMR simultaneously. Such pathways may be related to oxygen demand by tissues or its supply during stress and/ or exhaustive exercise. For example, structural and functional variation in the heart (i.e. cardiac output) or muscle, and mechanisms that modulate oxygen carrying capacity of the cardiovascular system might invoke changes in MMR without altering SMR [93–95]. However, our study does not rule out the possibility of the metabolic phenotypes affecting variation in age-at-maturity also phenotypically or via small-effect loci [44,96].

The differences in aerobic performance between life-history genotypes may have arisen due to correlated selection mediated by resource acquisition: higher AS, which was associated with early maturation, could enable higher feeding capacity [50] and improve foraging efficiency, for example, via a shorter searching time of prey [11,85]. Furthermore, salmon in the wild are increasingly experiencing higher than optimal temperatures due to climate change [97], and MMR is typically less plastic than SMR in response to environmental temperature [15]. Thus, our results indicate a potentially important advantage for individuals carrying the early maturation genotype under global warming, which could be mediated by a higher appetite [98]. This advantage may also extend to higher survival during spawning migration if the genotype effect on AS persists across life-stages [13,99,100]. It can also be relevant to survival of salmon after spawning, and thereby repeated spawning (iteroparity), which is co-inherited with the same *vgll3* genotype as early maturation [101]. However, the lack of sex differences in metabolic phenotypes in our study both across and within age-at-maturity genotypes, suggests that sex-dependent life-history variation in salmon [33] is not reflected in metabolic rates during the juvenile stage (see also [91]).

Although a causal relationship between metabolism at the juvenile stage and age at maturity would not be surprising, pleiotropy or linkage provide alternatives for the basis of the observed association. For example, the *vgll3* gene encodes a transcription cofactor associated with cell fate commitment and is expressed in many tissues [80,102], thus it is probably pleiotropic with multiple independent functions. Similarly, several polymorphic loci with putative functional variation are co-localized (i.e. linked) in the genomic region [43]. Finally, stage-dependent genetic

correlations in metabolism may obscure the time point that the trait influences the life-history variation. For instance, higher AS can induce maturation via facilitating the size attained both in the freshwater and at sea (e.g. [19,39]), but whether AS genetically covaries across life stages is yet to be explored.

The presence of genetic covariation between AS at the juvenile stage and age-at-maturity at the *vgll3* genomic region suggests potential for multi-trait evolution across life-stages, whereby selection acting on either trait would alter the phenotypic variation of the other [4,103]. For example, if natural selection towards later age-at-maturity increases the frequency of late maturing allele, this would constrain the AS of juveniles in the population, even if that may be a suboptimal phenotype. On the other hand, genetic covariation may help to maintain optimal trait variation in age-at-maturity, by limiting potentially maladaptive environmentally induced (i.e. plastic) variation in age-at-maturity (e.g. [104,105]). For example, river geophysical properties are important determinants of the optimal age structure at maturity, whereby populations in smaller tributaries have a younger, and populations in large, fast-flowing rivers have an older age structure [33]. Forecasting age-at-maturity from aerobic performance at earlier stages (e.g. via improved growth [22,51]) would result in maladaptive age structure if the covariation was explained entirely by environmental effects. However, our study was aimed at measuring the statistical association, and does not provide estimates using a classical quantitative genetic framework (i.e. we did not quantify environmental sources of variation, or variation due to technical or other genetic effects) or measure evolutionary change. Therefore, partitioning biologically meaningful covariation and quantifying the correlated response to selection were beyond the scope of this study [106].

Understanding the physiological basis of life-history variation in different life-stages and environmental conditions can provide insights into the factors driving life-history evolution, and hence, better predictions of the responses of populations to environmental changes. Wild salmon populations have declined in recent decades, with a concomitant decrease in the frequency of late maturing individuals [47,97]. Our study used an eco-physiological approach to identify a potentially adaptive phenotype relating genetic variation and age-at-maturity in salmon and suggests that evolution towards an earlier age-at-maturity can cause correlated selection towards increased MMR and AS. In conclusion, the integration of age-at-maturity and aerobic performance in the early life-stages via simple genetic mechanisms, as shown in this study, could contribute to the diversification of ecotypes within species.

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References

1. Ricklefs RE, Wikelski M. 2002 The physiology/life-history nexus. Trends Ecol. Evol. 17, 462–468. (doi:10.1016/s0169-5347(02)02578-8)

2. Brown JH, Gillooly JF, Allen AP, Savage VM, West GB. 2004 Toward a metabolic theory of ecology. Ecology 85, 1771–1789. (doi:10.1890/03-9000)

3. Dammhahn M, Dingemanse NJ, Niemelä PT, Reale D. 2018 Pace-of-life syndromes: a framework for the adaptive integration of behaviour, physiology and life history. Behav. Ecol. Sociobiol. 72, 1–8. (doi:10.1007/s00265-018-2473-y)

4. Roff DA. 1997 Evolutionary quantitative genetics. New York, NY: Chapman and Hall.

5. Boratynski Z, Koskela E, Mappes T, Schroderus E. 2013 Quantitative genetics and fitness effects of basal metabolism. Evol. Ecol. 27, 301–314. (doi:10.1007/s10682-012-9590-2)

6. Krams IA et al. 2017 Metabolic rate associates with, but does not generate covariation between, behaviours in western stutter-trilling crickets, Gryllus integer. Proc. R. Soc. B 284, 20162481. (doi:10.1098/rspb.2016.2481)

7. Auer SK, Dick CA, Metcalfe NB, Reznick DN. 2018 Metabolic rate evolves rapidly and in parallel with the pace of life history. Nat. Commun. 9, 1–6. (doi:10.1038/s41467-017-02514-z)

8. Lailvaux SP, Husak JF. 2014 The life-history of whole-organism performance. Q. Rev. Biol. 89, 285–318. (doi:10.1086/678567)

9. Fry FEJ. 1947 Effects of the environment on animal activity. Biol. Ser. 55, 1–62.

10. Chabot D, Steffensen JF, Farrell AP. 2016 The determination of standard metabolic rate in fishes. J. Fish Biol. 88, 81–121. (doi:10.1111/jfb.12845)

11. Metcalfe NB, Van Leeuwen TE, Killen SS. 2016 Does individual variation in metabolic phenotype predict fish behaviour and performance? J. Fish Biol. 88, 298–321. (doi:10.1111/jfb.12699)

12. Farrell AP. 2009 Environment, antecedents and climate change: lessons from the study of temperature physiology and river migration of salmonids. J. Exp. Biol. 212, 3771–3780. (doi:10.1242/jeb.023671)

13. Clark TD, Jeffries KM, Hinch SG, Farrell AP. 2011 Exceptional aerobic scope and cardiovascular performance of pink salmon (Oncorhynchus gorbuscha) may underlie resilience in a warming climate. J. Exp. Biol. 214, 3074–3081. (doi:10.1242/jeb.060517)

14. Eliason EJ, Wilson SM, Farrell AP, Cooke SJ, Hinch SG. 2013 Low cardiac and aerobic scope in a coastal population of sockeye salmon Oncorhynchus nerka with a short upriver migration. J. Fish Biol. 82, 2104–2112. (doi:10.1111/jfb.12120)

15. Sandblom E et al. 2016 Physiological constraints to climate warming in fish follow principles of plastic floors and concrete ceilings. Nat. Commun. 7, 1–8. (doi:10.1038/ncomms11447)

16. Killen SS, Glazier DS, Rezende EL, Clark TD, Atkinson D, Willener AST, Halsey LG. 2016 Ecological influences and morphological correlates of resting and maximal metabolic rates across teleost fish species. Am. Nat. 187, 592–606. (doi:10.1086/685893)

17. Priede IG. 1985 Metabolic scope in fishes. In Fish energetics (eds P Tytler, P Calow), pp. 33–64. Dordrecht, The Netherlands: Springer.

18. Thorpe JE. 2007 Maturation responses of salmonids to changing developmental opportunities. Marine Ecol. Progress Ser. 335, 285–288. (doi:10.3354/meps335285)

19. Salminen M. 1997 Relationships between smolt size, postsmolt growth and sea age at maturity in Atlantic salmon ranched in the Baltic Sea. J. Appl. Ichthyol. -Zeitschrift Fur Angewandte Ichthyologie 13, 121–130. (doi:10.1111/j.1439-0426.1997.tb00111.x)

20. Theriault V, Garant D, Bernatchez L, Dodson JJ. 2007 Heritability of life-history tactics and genetic correlation with body size in a natural population of brook charr (Salvelinus fontinalis). J. Evol. Biol. 20, 2266–2277. (doi:10.1111/j.1420-9101.2007.01417.x)

21. Roff DA. 1994 The evolution of dimorphic traits: predicting the genetic correlation between environments. Genetics 136, 395–401.

22. Auer SK, Salin K, Rudolf AM, Anderson GJ, Metcalfe NB. 2015 The optimal combination of standard metabolic rate and aerobic scope for somatic growth depends on food availability. Funct. Ecol. 29, 479–486. (doi:10.1111/1365-2435.12396)

23. Niemelä PT, Dingemanse NJ. 2018 Meta-analysis reveals weak associations between intrinsic state and personality. Proc. R. Soc. B 285, 20172823. (doi:10.1098/rspb.2017.2823)

24. Mathot KJ, Dingemanse NJ, Nakagawa S. 2019 The covariance between metabolic rate and behaviour varies across behaviours and thermal types: meta-analytic insights. Biol. Rev. 94, 1056–1074. (doi:10.1111/brv.12491)

25. Rosenfeld J, Richards J, Allen D, Van Leeuwen T, Monnet G. 2020 Adaptive trade-offs in fish energetics and physiology: insights from adaptive differentiation among juvenile salmonids. Can. J. Fish. Aquat. Sci. 77, 1243–1255. (doi:10.1139/cjfas-2019-0350)

26. Zeng LQ, Zhang A-J, Killen SS, Cao Z-D, Wang Y-X, Fu S-J. 2017 Standard metabolic rate predicts growth trajectory of juvenile Chinese crucian carp (Carassius auratus) under changing food availability. Biol. Open 6, 1305–1309. (doi:10.1242/bio.025452)

27. Auer SK, Solowey JR, Rajesh S, Rezende EL. 2020 Energetic mechanisms for coping with changes in resource availability. Biol. Lett. 16, 20200580. (doi:10.1098/rsbl.2020.0580)

28. Killen SS, Marras S, McKenzie DJ. 2011 Fuel, fasting, fear: routine metabolic rate and food deprivation exert synergistic effects on risk-taking in individual juvenile European sea bass. J. Anim. Ecol. 80, 1024–1033. (doi:10.1111/j.1365-2656.2011.01844.x)

29. Auer SK et al. 2020 Metabolic rate interacts with resource availability to determine individual variation in microhabitat use in the wild. Am. Nat. 196, 132–144. (doi:10.1086/709479)

30. Hughes KA, Rodd FH, Reznick DN. 2005 Genetic and environmental effects on secondary sex traits in guppies (Poecilia reticulata). J. Evol. Biol. 18, 35–45. (doi:10.1111/j.1420-9101.2004.00806.x)

31. Gutteling EW, Doroszuk A, Riksen JAG, Prokop Z, Reszka J, Kammenga JE. 2007 Environmental influence on the genetic correlations between life-history traits in Caenorhabditis elegans. Heredity 98, 206–213. (doi:10.1038/sj.hdy.6800929)

32. Gillespie JH, Turelli M. 1989 Genotype-environment interactions and the maintenance of polygenic variation. Genetics 121, 129–138.

33. Fleming IA, Einum S. 2011 Reproductive Ecology: A Tale of Two Sexes. In Atlantic salmon ecology, pp. 33–65.

34. Mobley KB et al. 2021 Maturation in Atlantic salmon (Salmo salar, Salmonidae): a synthesis of ecological, genetic, and molecular processes. Rev. Fish Biol. Fish. 31, 523–571. (doi:10.1007/s11160-021-09656-w)

35. Hendry AP, Berg OK, Quinn TP. 1999 Condition dependence and adaptation-by-time: breeding date, life history, and energy allocation within a population of salmon. *Oikos* **85**, 499–514. (doi:10.2307/3546699)

36. Stearns SC. 2000 Life history evolution: successes, limitations, and prospects. *Naturwissenschaften* **87**, 476–486. (doi:10.1007/s001140050763)

37. Friedland KD, Haas RE. 1996 Marine post-smolt growth and age at maturity of Atlantic salmon. *J. Fish Biol.* **48**, 1–15. (doi:10.1111/j.1095-8649.1996.tb01414.x)

38. Skilbrei OT. 1989 Relationships between smolt length and growth and maturation in the sea of individually tagged Atlantic salmon (*Salmo salar*). *Aquaculture* **83**, 95–108. (doi:10.1016/0044-8486(89)90064-1)

39. Hutchings JA, Jones MEB. 1998 Life history variation and growth rate thresholds for maturity in Atlantic salmon, *Salmo salar*. *Can. J. Fish. Aquat. Sci.* **55**, 22–47. (doi:10.1139/cjfas-55-S1-22)

40. Tréhin C, Rivot E, Lamireau L, Meslier L, Besnard A-L, Gregory SD, Nevoux M. 2021 Growth during the first summer at sea modulates sex-specific maturation schedule in Atlantic salmon. *Can. J. Fish. Aquat. Sci.* **78**, 659–669. (doi:10.1139/cjfas-2020-0236)

41. Moore MP, Martin RA. 2019 On the evolution of carry-over effects. *J. Anim. Ecol.* **88**, 1832–1844. (doi:10.1111/1365-2656.13081)

42. Ayllon F et al. 2015 The vgll3 locus controls age at maturity in wild and domesticated Atlantic salmon (*Salmo salar* L.) males. *PLoS Genet.* **11**, e1005628. (doi:10.1371/journal.pgen.1005628)

43. Barson NJ et al. 2015 Sex-dependent dominance at a single locus maintains variation in age at maturity in salmon. *Nature* **528**, 405–408. (doi:10.1038/nature16062)

44. Sinclair-Waters M, Ødegård J, Korsvoll SA, Moen T, Lien S, Primmer CR, Barson NJ. 2020 Beyond large-effect loci: large-scale GWAS reveals a mixed large-effect and polygenic architecture for age at maturity of Atlantic salmon. *Genet. Sel. Evol.* **52**, 9. (doi:10.1186/s12711-020-0529-8)

45. Debes PV, Piavchenko N, Ruokolainen A, Ovaskainen O, Moustakas-Verho JE, Parre N, Aykanat T, Erkinaro J, Primmer CR. 2021 Polygenic and major-locus contributions to sexual maturation timing in Atlantic salmon. *Mol. Ecol.* **30**, 4505–4519. (doi:10.1111/mec.16062)

46. Chaput G. 2012 Overview of the status of Atlantic salmon (*Salmo salar*) in the North Atlantic and trends in marine mortality. *Ices J. Mar. Sci.* **69**, 1538–1548. (doi:10.1093/icesjms/fss013)

47. Czorlich Y, Aykanat T, Erkinaro J, Orell P, Primmer CR. 2018 Rapid sex-specific evolution of age at maturity is shaped by genetic architecture in Atlantic salmon. *Nat. Ecol. Evol.* **2**, 1800–1807. (doi:10.1038/s41559-018-0681-5)

48. Czorlich Y, Aykanat T, Erkinaro J, Orell P, Primmer C. 2021 Evolution in salmon life-history induced by direct and indirect effects of fishing. bioRxiv. (doi:10.1101/2021.01.08.425869)

49. Aykanat T et al. 2020 Life-history genomic regions explain differences in Atlantic salmon marine diet specialization. *J. Anim. Ecol.* **89**, 2677–2691. (doi:10.1111/1365-2656.13324)

50. Auer SK, Salin K, Anderson GJ, Metcalfe NB. 2015 Aerobic scope explains individual variation in feeding capacity. *Biol. Lett.* **11**, 20150793. (doi:10.1098/rsbl.2015.0793)

51. Norin T, Clark TD. 2017 Fish face a trade-off between 'eating big' for growth efficiency and 'eating small' to retain aerobic capacity. *Biol. Lett.* **13**, 20170298. (doi:10.1098/rsbl.2017.0298)

52. Elliott JM, Hurley MA. 1997 A functional model for maximum growth of Atlantic salmon parr, *Salmo salar*, from two populations in northwest England. *Funct. Ecol.* **11**, 592–603. (doi:10.1046/j.1365-2435.1997.00130.x)

53. Ward AJW, Webster MM, Hart PJB. 2006 Intraspecific food competition in fishes. *Fish Fish.* **7**, 231–261. (doi:10.1111/j.1467-2979.2006.00224.x)

54. Rosenfeld J, Van Leeuwen T, Richards J, Allen D. 2015 Relationship between growth and standard metabolic rate: measurement artefacts and implications for habitat use and life-history adaptation in salmonids. *J. Anim. Ecol.* **84**, 4–20. (doi:10.1111/1365-2656.12260)

55. Forstner H. 1983 *An automated multiple-chamber intermittent-flow respirometer*, pp. 111–126. Berlin, Germany: Springer.

56. Svendsen MBS, Bushnell PG, Steffensen JF. 2016 Design and setup of intermittent-flow respirometry system for aquatic organisms. *J. Fish Biol.* **88**, 26–50. (doi:10.1111/jfb.12797)

57. Killen SS et al. 2021 Guidelines for reporting methods to estimate metabolic rates by aquatic intermittent-flow respirometry. *J. Exp. Biol.* **224**, jeb242522. (doi:10.1242/jeb.242522)

58. Raby GD, Doherty CLJ, Mokdad A, Pitcher TE, Fisk AT. 2020 Post-exercise respirometry underestimates maximum metabolic rate in juvenile salmon. *Conserv. Physiol.* **8**, coaa063. (doi:10.1093/conphys/coaa063)

59. Brett JR. 1964 The respiratory metabolism and swimming performance of young sockeye salmon. *J. Fish. Res. Board Canada* **21**, 1183–1226. (doi:10.1139/f64-103)

60. Norin T, Clark TD. 2016 Measurement and relevance of maximum metabolic rate in fishes. *J. Fish Biol.* **88**, 122–151. (doi:10.1111/jfb.12796)

61. Morozov S, McCairns RJS, Merilä J. 2019 FishResp: R package and GUI application for analysis of aquatic respirometry data. *Conserv. Physiol.* **7**, coz003. (doi:10.1093/conphys/coz003)

62. Harianto J, Carey N, Byrne M. 2019 respR-An R package for the manipulation and analysis of respirometry data. *Methods Ecol. Evol.* **10**, 912–920. (doi:10.1111/2041-210x.13162)

63. Little AG et al. 2020 Maxed out: optimizing accuracy, precision, and power for field measures of maximum metabolic rate in fishes. *Physiol. Biochem. Zool.* **93**, 243–254. (doi:10.1086/708673)

64. R Core Team. 2019 *R: a language and environment for statistical computing*. Vienna, Austria: R Foundation for Statistical Computing.

65. Schielzeth H. 2010 Simple means to improve the interpretability of regression coefficients. *Methods Ecol. Evol.* **1**, 103–113. (doi:10.1111/j.2041-210X.2010.00012.x)

66. Bates D, Machler M, Bolker BM, Walker SC. 2015 Fitting linear mixed-effects models using lme4. *J. Stat. Softw.* **67**, 1–48. (doi:10.18637/jss.v067.i01)

67. Lenth RV. 2020 emmeans. R package version 1.5.3.

68. Stoffel MA, Nakagawa S, Schielzeth H. 2021 partR2: partitioning R-2 in generalized linear mixed models. *Peerj* **9**, e11414. (doi:10.7717/peerj.11414)

69. Lüdecke D. et al. 2018 ggeffects. See https://strengejacke.github.io/ggeffects.

70. Wickham H. 2009 *gplot2: elegant graphics for data analysis*, pp. 1–212. New York, NY: Springer-Verlag. (doi:10.1007/978-0-387-98141-3)

71. Long JA. 2019 interactions: comprehensive, user-friendly toolkit for probing interactions. See https://rdrr.io/github/jacob-long/interactions.

72. Hurvich CM, Tsai CL. 1989 Regression and time-series model selection in small samples. *Biometrika* **76**, 297–307. (doi:10.1093/biomet/76.2.297)

73. Bartoń K. 2020 MuMIn: Multi-Model Inference. R package version 1.43.17. See https://CRAN.R-project.org/package=MuMIn.

74. Johnson JB, Omland KS. 2004 Model selection in ecology and evolution. *Trends Ecol. Evol.* **19**, 101–108. (doi:10.1016/j.tree.2003.10.013)

75. Weimerskirch H. 1992 Reproductive effort in long-lived birds – age-specific patterns of condition, reproduction and survival in the Wandering albatross. *Oikos* **64**, 464–473. (doi:10.2307/3545162)

76. Cheung CC, Thornton JE, Kuijper JL, Weigle DS, Clifton DK, Steiner RA. 1997 Leptin is a metabolic gate for the onset of puberty in the female rat. *Endocrinology* **138**, 855–858. (doi:10.1210/en.138.2.855)

77. Cheverud JM, Routman EJ. 1995 Epistasis and its contribution to genetic variance components. *Genetics* **139**, 1455–1461. (doi:10.1093/genetics/139.3.1455)

78. Cheverud JM, Routman EJ. 1996 Epistasis as a source of increased additive genetic variance at population bottlenecks. *Evolution* **50**, 1042–1051. (doi:10.2307/2410645)

79. Merilä J, Sheldon BC. 1999 Genetic architecture of fitness and nonfitness traits: empirical patterns and development of ideas. *Heredity* **83**, 103–109. (doi:10.1046/j.1365-2540.1999.00585.x)

80. Kurko J, Debes PV, House AH, Aykanat T, Erkinaro J, Primmer CR. 2020 Transcription profiles of age-at-maturity-associated genes suggest cell fate commitment regulation as a key factor in the Atlantic Salmon maturation process. *G3-Genes Genomes Genet.* **10**, 235–246. (doi:10.1534/g3.119.400882)

81. Moustakas-Verho JE, Kurko J, House AH, Erkinaro J, Debes P, Primmer CR. 2020 Developmental expression patterns of six6: a gene linked with

spawning ecotypes in Atlantic salmon. *Gene Expr. Patterns* **38**, 119149. (doi:10.1016/j.gep.2020.119149)

82. Debat V, David P. 2001 Mapping phenotypes: canalization, plasticity and developmental stability. *Trends Ecol. Evol.* **16**, 555–561. (doi:10.1016/s0169-5347(01)02266-2)

83. Pritchard VL, Mäkinen H, Vähä JP, Erkinaro J, Orell P, Primmer CR. 2018 Genomic signatures of fine-scale local selection in Atlantic salmon suggest involvement of sexual maturation, energy homeostasis and immune defence-related genes. *Mol. Ecol.* **27**, 2560–2575. (doi:10.1111/mec.14705)

84. Zueva KJ, Lumme J, Veselov AE, Primmer CR, Pritchard VL. 2021 Population genomics reveals repeated signals of adaptive divergence in the Atlantic salmon of north-eastern Europe. *J. Evol. Biol.* **34**, 866–878. (doi:10.1111/jeb.13732)

85. Johansen M, Erkinaro J, Amundsen PA. 2011 The when, what and where of freshwater feeding. In *Atlantic salmon ecology* (eds Ø Aas, S Einum, A Klemetsen, J Skurdal), pp. 89–114. Wiley-Blackwell.

86. O'Connor KI, Taylor AC, Metcalfe NB. 2000 The stability of standard metabolic rate during a period of food deprivation in juvenile Atlantic salmon. *J. Fish Biol.* **57**, 41–51. (doi:10.1006/jfbi.2000.1280)

87. Auer SK, Salin K, Rudolf AM, Anderson GJ, Metcalfe NB. 2015 Flexibility in metabolic rate confers a growth advantage under changing food availability. *J. Anim. Ecol.* **84**, 1405–1411. (doi:10.1111/1365-2656.12384)

88. Armstrong JB, Schindler DE. 2011 Excess digestive capacity in predators reflects a life of feast and famine. *Nature* **476**, 84–87. (doi:10.1038/nature10240)

89. Archer LC, Hutton SA, Harman L, Poole WR, Gargan P, Mcginnity P, Cooke S. 2020 Metabolic traits in brown trout (*Salmo trutta*) vary in response to food restriction and intrinsic factors. *Conserv. Physiol.* **8**, coaa096. (doi:10.1093/conphys/coaa096)

90. Auer SK *et al.* 2018 Nutrients from salmon parents alter selection pressures on their offspring. *Ecol. Lett.* **21**, 287–295. (doi:10.1111/ele.12894)

91. Åsheim ER, Prokkola JM, Morozov S, Aykanat T, Primmer CR. In press. Standard metabolic rate does not associate with age-at-maturity genotype in juvenile Atlantic salmon. *Ecol. Evol.* **00**, 1–14. (https://doi.org/10.1002/ece3.8408)

92. Polverino G, Santostefano F, Diaz-Gil C, Mehner T. 2018 Ecological conditions drive pace-of-life syndromes by shaping relationships between life history, physiology and behaviour in two populations of Eastern mosquitofish. *Sci. Rep.* **8**, 14673. (doi:10.1038/s41598-018-33047-0)

93. Harter TS, Zanuzzo FS, Supuran CT, Gamperl AK, Brauner CJ. 2019 Functional support for a novel mechanism that enhances tissue oxygen extraction in a teleost fish. *Proc. R. Soc. B* **286**, 20190339. (doi:10.1098/rspb.2019.0339)

94. Nikinmaa M, Berenbrink M, Brauner CJ. 2019 Regulation of erythrocyte function: multiple evolutionary solutions for respiratory gas transport and its regulation in fish. *Acta Physiol.* **227**, e13299. (doi:10.1111/apha.13299)

95. McArley TJ, Morgenroth D, Zena LA, Ekstrom AT, Sandblom E. 2021 Normoxic limitation of maximal oxygen consumption rate, aerobic scope and cardiac performance in exhaustively exercised rainbow trout (*Oncorhynchus mykiss*). *J. Exp. Biol.* **224**, jeb242614. (doi:10.1242/jeb.242614)

96. Johnston SE, Orell P, Pritchard VL, Kent MP, Lien Ş, Niemelä E, Erkinaro J, Primmer CR. 2014 Genome-wide SNP analysis reveals a genetic basis for sea-age variation in a wild population of Atlantic salmon (*Salmo salar*). *Mol. Ecol.* **23**, 3452–3468. (doi:10.1111/mec.12832)

97. Friedland KD, MacLean JC, Hansen LP, Peyronnet AJ, Karlsson L, Reddin DG, Ó Maoiléidigh N, McCarthy JL. 2009 The recruitment of Atlantic salmon in Europe. *Ices J. Mar. Sci.* **66**, 289–304. (doi:10.1093/icesjms/fsn210)

98. Jutfelt F, Norin T, Åsheim ER, Rowsey LE, Andreassen AH, Morgan R, Clark TD, Speers-Roesch B. 2021 'Aerobic scope protection' reduces ectotherm growth under warming. *Funct. Ecol.* **35**, 1397–1407. (doi:10.1111/1365-2435.13811)

99. Eliason EJ *et al.* 2011 Differences in thermal tolerance among sockeye salmon populations. *Science* **332**, 109–112. (doi:10.1126/science.1199158)

100. Mottola G, Kristensen T, Anttila K. 2020 Compromised thermal tolerance of cardiovascular capacity in upstream migrating Arctic char and brown trout-are hot summers threatening migrating salmonids? *Conserv. Physiol.* **8**, coaa101. (doi:10.1093/conphys/coaa101)

101. Aykanat T *et al.* 2019 Co-inheritance of sea age at maturity and iteroparity in the Atlantic salmon vgll3 genomic region. *J. Evol. Biol.* **32**, 343–355. (doi:10.1111/jeb.13418)

102. Verta JP *et al.* 2020 Cis-regulatory differences in isoform expression associate with life history strategy variation in Atlantic salmon. *PLoS Genet.* **16**, e1009055. (doi:10.1371/journal.pgen.1009055)

103. Lande R. 1979 Quantitative genetic analysis of multivariate evolution, applied to brain: body size allometry. *Evolution* **33**, 402–416. (doi:10.2307/2407630)

104. De Jong G. 1999 Unpredictable selection in a structured population leads to local genetic differentiation in evolved reaction norms. *J. Evol. Biol.* **12**, 839–851. (doi:10.1046/j.1420-9101.1999.00118.x)

105. Tufto J. 2000 The evolution of plasticity and nonplastic spatial and temporal adaptations in the presence of imperfect environmental cues. *Am. Nat.* **156**, 121–130. (doi:10.1086/303381)

106. Sgro CM, Hoffmann AA. 2004 Genetic correlations, tradeoffs and environmental variation. *Heredity* **93**, 241–248. (doi:10.1038/sj.hdy.6800532)